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COUNCIL OF THE CITY OF PHILADELPHIA SPECIAL COMMITTEE ON POVERTY REDUCTION AND PREVENTION

Room 400, City Hall Philadelphia, Pennsylvania Thursday, October 10, 2019 3:13 p.m.

PRESENT:

COUNCILWOMAN MARIA D. QUINONES-SANCHEZ
COUNCIL PRESIDENT DARRELL L. CLARKE
COUNCILWOMAN JANNIE L. BLACKWELL
COUNCILMAN ALLAN DOMB
COUNCILWOMAN BLONDELL REYNOLDS BROWN
EVA GLADSTEIN, Deputy Managing Director
of Health and Human Services
SHARMAIN MATLOCK-TURNER, President and
CEO, Urban Affairs Coalition
MEL WELLS, President and CEO, One Day At
A Time

RESOLUTION 190239 - Resolution authorizing the creation of a "Special Committee on Poverty Reduction and Prevention" to hold hearings to propose and implement actionable policies and programs that substantively prevent and alleviate poverty in every Philadelphia neighborhood.

Page 2 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. COUNCILWOMAN SANCHEZ: afternoon, everyone. I want to thank 3 everyone who is here today for the first 4 5 meeting of the Special Committee on 6 Poverty Reduction and Prevention. I want to acknowledge Council President Clarke, who will open our 8 9 hearing. COUNCIL PRESIDENT CLARKE: 10 Thank you very much, Councilwoman. 11 12 And good afternoon, everyone, here today and those of you that may be 13 14 watching on Channel 64 or 65. I just 15 want to say that from my perspective, 16 this is a very exciting day for a lot of 17 reasons. Obviously this whole issue about poverty has been something that's 18 been discussed for a lot of years, 19 20 actually literally decades as it relates 21 to the City of Philadelphia, but I think something special and something a little 22 23 different will happen as a result of this particular format and these individuals 2.4 who are -- I want to welcome and I want 25

Page 3 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. to thank personally Eva Gladstein, my good friend and colleague Councilwoman 3 4 Quinones-Sanchez, Sharmain 5 Matlock-Turner, and Mel Wells for their 6 willingness to serve as Co-Chairs. will have a significant number of individuals that will participate, and I 8 9 want to say thank you all to them. So in recent days, I've 10 11 actually been calling this initiative --12 I actually did an interview, I think, with Al Dia yesterday and I talked about 13 14 moonshot, right? And people kind of go, 15 moonshot? So to be honest with you, I 16 saw a documentary the other day when I'm flipping channels and it talked about 17 18 moonshot, where we actually sent a person to the moon, and the intent of that 19 20 moonshot at that time was to uplift 21 humanity. And, you know, in some 22 respects, this obviously and that term 23 was relevant, because we're talking about uplifting literally one out of every four 2.4 25 persons in the City of Philadelphia out

Page 4 1 10/10/19 - POVERTY REDUCTION - RES. 190239 of poverty. That is a significant 2. number, I think a little less than 3 4 400,000 people. 5 So while it took a gargantuan effort on behalf of the United States and 6 7 all the other participants to get us to the moon, similarly it will take a 8 9 significant level of participation and effort to move those individuals out of 10 poverty and prevent other people from 11 12 slipping back into poverty. But if we're all together and if we have the same 13 14 drive, the same commitment as they had 15 when people decided to go to the moon, I 16 genuinely think that we'll be able to get 17 there. 18 So I just want to say how 19 important this is, as you all know, and I know I'm kind of speaking to the choir in 20 21 terms of the people that are here, but I just wanted to give you a sense of how 22 23 important this is to us. So to my colleagues, thank you 2.4 25 all very much for your willingness, and

Page 5 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. we look forward to providing whatever 3 level of support and to continue the 4 theme of moving the City forward and 5 losing that very, very unfavorable 6 designation as being the poorest big city 7 in the country. That's something that, frankly, we no longer want to be known 8 9 as. So thank you, Madam Co-Chair 10 11 and the members of this illustrious panel 12 up here. COUNCILWOMAN SANCHEZ: 13 Thank 14 you, Council President Clarke. I'm going to just try to 15 16 facilitate and work through this meeting. 17 I want to thank Council President Clarke and all the Council colleagues who, in 18 one way or another over at least the 12 19 20 years that I've been here in Council, 21 have worked really aggressively on this issue, but really pulling it together and 22 23 creating not only this Committee but a process where we can put together some of 2.4 25 the best minds and empowering us to work

Page 6 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. through not just the regular approach but 3 really bold and decisive action. 4 I want to give an opportunity to all of the Co-Chairs of the Committee 5 6 to say some opening words, but I want to 7 acknowledge the membership of the central committee and all of the people that are 8 9 participating in our subcommittees for lending of their time, their energy, 10 11 their expertise, many of them providing some of their staff and their staff hours 12 to help us pull this together. 13 14 So the Committee members, if 15 you haven't done so already, we have 16 assigned seats with your names on the 17 front, but I wanted to take a moment and acknowledge the full Committee members 18 and then I will provide an opportunity 19 20 for our Co-Chairs to say some opening 2.1 words. Patrick Clancy, the President 22 23 and CEO of Philadelphia Works; obviously our Council colleague Councilman Allan 2.4 25 Domb; Bill Golderer, President and CEO of

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2	United Way of Greater Philadelphia and	
3	Southern New Jersey; Dr. Donald Guy	
4	Generals, the President of Community	
5	College of Philadelphia; Otis Hackney,	
6	the Chief Education Officer for the	
7	Mayor's Office of Education; Sidney	
8	Hargro, President of Philanthropy Network	
9	of Greater Philadelphia.	
10	Please stand if I call your	
11	name and you're here so that folks know	
12	who you are.	
13	Maura Hesdon, General Manager	
14	of Shoemaker Construction; Dr. William	
15	Hite, Jr., Superintendent. He's not	
16	here, but he sends his regrets, and I	
17	know he has some of his staff and his	
18	team here. Sheila Ireland, Executive	
19	Director of the Office of Workforce	
20	Development; Dr. Roberta Rehner Iversen,	
21	Associate Professor of Social Policy and	
22	Practice at the University of	
23	Pennsylvania; Dr. Judith Levine,	
24	Associate Professor of Sociology and	
25	Director of Public Policy at Temple;	

Page 8 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. Mitch Little, Executive Director of the 3 Mayor's Office of Community Development; Yvette Nunez, Vice President of Civic 4 Affairs at the Chamber of Commerce; Mecca 5 6 Robinson, Executive Director of Forget Me Knot Youth Services; Dr. Theresa Singleton, Senior Vice President, 8 9 Community Affairs Officer at the Federal Reserve Bank of Philadelphia; and of 10 11 course our own Chris Woods, Executive Vice President of District Council 1199C. 12 Along with this illustrious 13 14 group of folks, again, some of the best 15 minds in the City, we have a one-page fact sheet with all of the members of the 16 subcommittee, and I invite anyone who is 17 18 interested in seeing the entire list, and I want to personally, on behalf of 19 20 Council President and all the Co-Chairs, 21 thank every single one. We have been charged with 22 23 something -- to produce a document by the 2.4 end of the year. So this is not a 25 meeting to meet, but really a process to

Page 9 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. create some definitive action steps. So with that, I want to provide 3 Eva Gladstein, who heads the Mayor's 4 5 cabinet on, I quess, health and human 6 services, to say some opening words. MS. GLADSTEIN: Thank you, Councilwoman. Thank you, Council 8 9 President Clarke. I'm really honored to be here and very pleased to be appointed 10 one of the Co-Chairs of this Committee. 11 I've spent a lot of years 12 looking at this subject and thinking 13 14 about it and trying to take action, and I 15 really think this is an opportunity, with 16 City Council calling us all to action, 17 really to take the reports and the work 18 that many people sitting here have been engaged in for many years and really 19 20 bring it up to a new level. And so I 21 really welcome that. I'm here also as Co-Chair of a 22 23 I want to give a subcommittee. shout-out. Many of the members of that 2.4 25 Subcommittee on the Social Safety Net are

Page 10 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. here today. We've met several times with 3 that subcommittee already and we've been working hard at looking at our own 4 5 practice, our own experience, people who 6 are using services, people who are providing services, people with other experiences in terms of what our best 8 9 ideas are around the social safety net and how to bring them to this Committee 10 11 for action, short, medium, and long-term 12 ideas. And then we only have a minute, 13 14 but the other thing I want to say is that 15 we have been working particularly in this 16 Administration but in many Administrations making many investments, 17 and yet they haven't had the results that 18 we all want to see, and many of those 19 20 folks who are doing those investments are 2.1 in the room and I think are energized by this work. 22 23 We've been looking at financial 2.4 empowerment, access to benefits, housing, 25 and certainly trying to serve people who

Page 11 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. are without homes, who are homeless, the 3 workforce system and in particular how we can help people who are disconnected from 4 5 the workforce system enter that system, 6 be it gradually if they have barriers 7 towards that, and obviously education investments at every stage of folks' 8 9 life. So I think we've been 10 11 investing. We're doing good work, and we 12 had a slight decrease in the number of people as measured who are living in 13 14 poverty, but we have much more work to 15 do, and I'm really hopeful that this 16 formation can get us there and we can be 17 finished in December and start taking --18 be acting all along, but start taking some new actions as of the new year. 19 20 Thank you. 2.1 COUNCILWOMAN SANCHEZ: Thank 22 you. 23 Our other Co-Chair in heading the Subcommittee on Jobs and Education is 2.4 25 Sharmain Matlock-Turner.

Page 12 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. MS. MATLOCK-TURNER: Thank you very much. It's a pleasure to be here 3 this afternoon. I really want to thank 4 5 you for all of you being here and being a part of this critical conversation. 6 certainly is an issue that all of us feel 7 very deeply about and how it's affecting 8 9 our city. I really want to thank City 10 11 Council, Council President Clarke, along 12 with Councilmembers Quinones-Sanchez, Councilman Domb, and so many others who 13 14 have been a part of saying that we need 15 to figure out how we can come together to 16 really work on this issue. So I'm excited to be working with Maria, with 17 Eva, with my friend here, Mel Wells, who 18 you're going to hear from shortly, on 19 20 this Special Committee to Reduce Poverty, and I think all of us in the end want to 21 22 prevent poverty. 23 Most of you, I think, know a little bit about the Urban Affairs 2.4 25 Coalition. We have been in the community

Page 13 1 10/10/19 - POVERTY REDUCTION - RES. 190239 for 50 years. We have over 74 groups, 2. 3 non-profits and social entrepreneurs who 4 are a part of the UAC family of programs 5 who are working every day on critical 6 issues that are facing our community, 7 ending homelessness, finding paths and bridges to self-sufficiency, supporting 8 9 quality education, and helping young people plan their futures and so much 10 11 more. We believe in a multifaceted 12 approach to achieving self-sufficiency, 13 14 and we have over the years seen friends, family, and neighbors succeed and move on 15 16 to productive and fulfilling lives. 17 However, it is no secret that 18 the City of Philadelphia has almost 400,000 of our 1.6 million citizens 19 20 living below the poverty line. They are 21 children, they are adults, they are workers, they are the disabled, they are 22 23 our seniors. Only with recognizing the 2.4 25 unique needs of each person will we be

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2	able to craft the solutions necessary to		
3	begin the process to end poverty. The		
4	work will require the collaboration of		
5	the partners in this room representing		
6	government, the private sector, the		
7	public sector, foundations, and each of		
8	us as individual citizens. And we are		
9	not starting with a blank slate. I		
10	really want to also thank City Council		
11	for the 2018 report Narrowing the Gap:		
12	Strategies to Alleviate Poverty and		
13	Prevent Poverty in Philadelphia. This		
14	report, which there's a copy here if you		
15	haven't seen it, lays out 27 strategies		
16	to begin our conversation and		
17	deliberations. The Council states in the		
18	report that we think this is a good		
19	beginning but, and I quote, we do not		
20	presume that these ideas are		
21	comprehensive solutions to poverty nor do		
22	we believe that they are the only		
23	solutions.		
24	However, I do believe that this		
25	report is a good place for us to start to		

Page 15 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. specifically tackle the issue of poverty in Philadelphia. As the Co-Chair, along 3 4 with my friend Mel Wells of One Day At A 5 Time, we are co-chairing the Jobs and Education Subcommittee, and we're looking 6 forward to having a great conversation 7 around the issues that are laid out in 8 9 the report. Whether we're looking at the skills gap, establishing a living wage, 10 11 or making sure that workers have the kind 12 of protections that they need in the workplace, we're going to personally be 13 14 engaged in helping to identify with you 15 some of the key strategies that are going 16 to get us to where we need to be. 17 You see the list in the report 18 of the people who are working as a part 19 of our Jobs and Education Subcommittee. We have a terrific list of business 20 leaders, community leaders, labor, and 21 individuals who are going to be tackling 22 23 those suggestions and certainly hearing from all of you with additional ones, and 2.4 25 we hope to come out, Councilwoman

Page 16 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Sanchez, with three to five 3 recommendations for Council to consider as you get ready for your budget process 4 for next year. 5 6 Thank you all very much. COUNCILWOMAN SANCHEZ: And our next Co-Chair brings a lot of energy and 8 9 he's like let's not talk about it, let's be about it, Mel Wells. 10 11 MR. WELLS: Good afternoon. Τ 12 would like to thank Council for having me here today. To keep it real for 13 14 everybody, I can't believe that I'm 15 actually sitting up here in this room. 16 I'm extremely grateful. Over 37 years 17 ago, me and my family, we sat in the welfare office trying to figure out how 18 will we get out of poverty with our 19 20 company One Day At A Time. Now we're located in London, India, and also 21 Cambodia serving over 56,000 people. 22 23 my job is to make sure that the people in the community voices are heard and that 2.4 25 they can cross the bridge of resources to

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 1
         get them out of poverty.
 3
                   So my job here is to keep it
 4
         real for the people that's in the
 5
         community who need to come out of
 6
         poverty. So thank you for having me,
 7
         everyone.
                   COUNCILWOMAN SANCHEZ: Thank
 8
 9
         you.
                   I don't know if Councilman
10
11
         Domb, I want to provide you an
12
         opportunity -- I also want to acknowledge
         that Councilwoman Jannie Blackwell has
13
14
         joined us. All of you know the level of
         advocacy that the Councilwoman has had as
15
16
         it relates to housing and homelessness in
17
         the City.
18
                   Thank you, Councilwoman, for
19
         joining us.
                   Councilman Domb.
20
                   COUNCILMAN DOMB: Thank you.
2.1
22
         first want to thank my colleagues,
         Council President Darrell Clarke and
23
         Councilmember Maria Quinones-Sanchez, for
2.4
25
         overseeing this Special Committee, along
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Page 18 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. with the additional Co-Chairs, Sharmain 3 Matlock-Turner, Eva Gladstein, and Mel 4 Wells. 5 It's important that we have 6 industry experts and community service providers joining us on this Committee, because they see firsthand how poverty 8 9 impacts our neighborhoods. I commend their participation today and, more 10 11 importantly, their ongoing commitment to 12 serving low-income Philadelphians. While I anticipate some of the 13 14 conversation today will be focused on 15 programs and initiatives that are in 16 place by both the City and outside organizations, I want to raise two 17 18 important factors that we really need to keep in mind when discussing poverty in 19 20 Philadelphia. One of those factors -- Council 2.1 President, you know what I'm going to 22 23 say -- is the Earned Income Tax Credit, but there's also -- you're going to hear 2.4 25 from, I believe, Pauline Abernathy from

Page 19 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Benefits Data Trust later in this meeting 3 today. We still leave on the table -and the charts are on the table in the 4 5 front -- \$450 million in federal and 6 state monies that could go to people in Philadelphia, 450 million over about five 7 different programs, whether it's Earned 8 9 Income Tax, SNAP, medical assistance, PACENET, property tax, and rent relief. 10 11 We have to figure out a better way to get 12 that money into the hands of Philadelphians. 13 14 The second issue -- and this is 15 really, I think, a major issue -- that 16 there has been an unfair burden on our 17 poor in this city that we have yet to 18 In a recent analysis, tackle. Philadelphia taxes its poor residents 19 20 more than any other large city in the 21 country. We tax higher than -- in any 50 states or the 50 largest cities, we tax 22 23 people the highest. For a family in Philadelphia 2.4 25 earning 25,000 a year, their tax burden

Page 20 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. is 4,500, or 18 percent, the highest in the country. Other cities range from 8 3 to 13 percent. The bulk of the tax 4 5 burden comes from their income, the wage tax, which accounts for almost 40 percent 6 7 of the total tax burden. Philadelphia needs to provide a fair system and remove 8 9 one of the largest barriers that keeps our poor from ever improving their 10 11 financial situation. 12 It is with high hopes that the bill that Councilwoman Sanchez and 13 14 Councilwoman Blackwell and I introduced 15 the other week will provide a wage tax 16 refund for low-income residents as part 17 of this Committee's discussions. bill could affect 60,000 households with 18 an average refund of \$700 per household 19 and potentially 100,000 people. The cost 20 21 could be \$40 to \$42 million. And people say, how are we going to pay for it? The 22 23 income tax, the wage tax increase per year, just average inflation, is about 81 2.4 25 to 82 million. Half of our increase just

Page 21 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. from inflation could pay for this 3 program. 4 So we all look forward to these 5 discussions, ideas, and results that come 6 from today's hearing, and I thank you for 7 coordinating this and organizing it. Thank you very much. 8 9 COUNCILWOMAN SANCHEZ: Thank 10 you. 11 So we gather here today to hear 12 from you, Philadelphia, about why and how each and every City employee can and must 13 14 work to improve, as I like to say, income 15 access and opportunity to improve the 16 quality of life for those who suffer and 17 live in poverty. We need to hear from 18 you, and that's why this task force is committed to our very condensed schedule 19 20 through the end of the year, and we hope 21 to be able to influence next year's 22 budget, our Five Year Plan, and our 23 ten-year projections in a very 2.4 intentional way. 25 If you haven't already done so,

Page 22 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. please sign up to testify at the table at the front, as I mentioned earlier. I 4 will begin -- I am going to call our 5 first panel of witnesses. 6 The framing of who lives in 7 poverty is hugely important. I want to thank Council President Clarke for 8 9 allowing this Committee access to The Reinvestment Fund, who has been working 10 11 with us and will work with us through 12 this process. So we will begin with the 13 14 presentation entitled "Defining Poverty in Philadelphia" by Ira Goldstein from 15 16 The Reinvestment Fund, so if he can come 17 forward. And Octavia Howell from Pew will then follow. 18 19 (Witnesses approached witness 20 table.) 2.1 DR. GOLDSTEIN: Well, good afternoon and thank you for the 22 23 opportunity to come here. COUNCILWOMAN SANCHEZ: 2.4 25 into the mic and say your name for the

Page 23 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. record. I'm sorry. DR. GOLDSTEIN: Sure. Thank 3 you for the opportunity to come here and 4 5 give you the results of some of the work that we've done to try and begin to set 6 the stage for what the poverty climate is 7 here in the City of Philadelphia. 8 9 I'm joined by my colleague from Reinvestment Fund, Emily Dowdall, and our 10 11 friend from the Philadelphia Research Initiative at Pew Trust, Octavia Howell. 12 What we're going to do is 13 14 really split the presentation up in a way 15 that gives you some basic sort of 16 environmental data about what poverty is 17 and what the groupings are in terms of who is in them and what the circumstance 18 Then we'll show you some maps 19 is. 20 detailing the geography of poverty in 21 Philadelphia. And then Octavia will 22 spend some time speaking with you about 23 the perspectives that have been offered to Pew through their survey work and 2.4 25 focus groups on what the situation is to

Page 24 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. live as somebody who is below the poverty line. 4 So with that, the poverty 5 measure that we are all used to thinking 6 about is one that was created in the 1960s by now sort of folk hero in this 7 space, Molly Orshansky, and other than 8 9 basically the Consumer Price Index, that has not changed over the course of time. 10 11 It has the benefit of being 12 very consistently measured, and so you can sort of benchmark yourself from time 13 14 to time, but it's got a lot of problems with it. 15 There have been innovations in 16 17 the measurement of poverty, one of which I'll talk about in a moment called the 18 19 Supplemental Poverty Measure, and that I think is in some ways going to be more 20 21 instructive to you as you think about the kinds of intervention strategies and 22 23 policies and programs and legislation that will help move the situation for 2.4 25 people who are poor here in the City of

Page 25 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Philadelphia. Nevertheless, as I said, 3 the official poverty measure is the one that people generally will look to move 4 5 with these things. That's an awful graphic for 6 7 people sitting anywhere other than close by, but we offer it mostly just to give 8 9 people a sense of what the income thresholds are for people who are poor. 10 11 Basically when you're thinking about a four-person family, depending upon the 12 constellation of that family, you're 13 14 looking at an annual income that's in and around \$24,000, \$25,000. So these are 15 16 very low incomes predicated on the notion 17 that back in the 1960s people spent about three times what they spent on food for 18 the rest of the things that they needed, 19 20 and that became the poverty threshold. 21 They are, by no means, I think a way of really understanding the totality of the 22 circumstance, nor do they take into 23 account a variety of other things, which 2.4 25 the Supplemental Poverty Measures do.

Page 26 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Like, for example, the fact that people 3 who are of modest means, depending upon the program that we're talking about, 4 5 they'll have access to other things that sort of add to their economic viability. 6 7 On the other hand, there are also things that take away from it, things like, as 8 9 Councilman Domb mentioned, taxes take away from that. And so the Supplemental 10 11 Poverty Measure is a way of essentially 12 trying to sort of equalize this situation for people. 13 14 It's also the case that the 15 poverty measure is invariant across the 16 country. And so while we've always felt 17 that Philadelphia is a pretty affordable 18 place to be, the fact of the matter is, we are above the average in terms of the 19 20 basic costs of living. We're not as bad 21 as San Jose, for example, but we are substantially less affordable than a 22 23 place like even San Antonio or Memphis or some of the other places that we like to 2.4 25 think ourselves in comparison to or, by

Page 27 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. the way, even Chicago. 3 This chart is also showing you that there are great differences in the 4 5 extent to which housing costs vary across 6 the country. So, for example, the 7 National Low Income Housing Coalition does a terrific job with their annual Out 8 9 of Reach study that looks at the extent to which if you are earning a minimum 10 wage how many hours you would need to 11 12 work to be able to afford a typical 13 two-bedroom apartment. And in the City 14 of Philadelphia, somebody who is working 15 at minimum wage would have to work about 16 127 hours to have the typical two-bedroom apartment without being substantially 17 18 cost burdened. And in some measure, it's our cost of housing, but another measure, 19 20 it's also our minimum wage, which is still stuck at that seven and a quarter 21 level, where if you look at some of the 22 23 other places like New York state or California, some of those other 2.4 25 communities have raised their minimum

Page 28 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. wages. And so the number what minimum 3 wage is and thus the number of hours you need to work, all else being equal, will 4 5 be more advantageous in those places. 6 So what do the drivers of 7 poverty look like? The fact of the matter is, education is a huge driver. 8 9 This particular chart is really designed as a way of trying to help understand 10 11 what it means to have different levels of 12 education, and critically I think is the fact that without a high school degree, 13 14 so for heads of household who are 25 15 years old who have not completed high 16 school and don't have a degree, roughly 17 50 percent of those households live at or 18 near poverty. The simple -- not the simple, but the fact of completing high 19 20 school and getting that degree drops that 21 percentage down pretty significantly to about 35 percent if you add in at or near 22 23 poverty, but 28 percent below poverty. So getting that high school degree 2.4 25 reduces quite dramatically the likelihood

Page 29 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. that somebody would be in poverty, as does increasing levels as you get up 3 4 through some college and a college 5 degree. 6 So clearly education is a huge 7 driver. Helping people get to the point where their educations are above that 8 9 high school level in and alone is going to make, I think, a pretty significant 10 difference in the poverty rates here in 11 12 Philadelphia. This is unemployment. 13 14 obviously unemployment is a huge driver. The black line at the -- well, in recent 15 16 years in the top is Philadelphia's unemployment rate. The other lines on 17 18 that chart are the other large cities, 19 the nine other largest cities in the country. And what you'll notice is in 20 the beginning of the period that we've 21 charted here in that 2009 or '10 period, 22 we were sort of in the middle of the 23 pack, and we've all gotten better in 2.4 25 terms of what the unemployment rate looks

Page 30 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. like. Nevertheless, at this point now 3 Philadelphia is leading in a bad way the unemployment situation of those other 4 5 nine large American cities. So we're 6 much better off than we were at the peak 7 of unemployment, but we're less well off vis-a-vis the other large American 8 9 cities. Our labor force, which is the 10 next two lines, and the number of people 11 12 who are employed in the City of Philadelphia has in fact grown, but at a 13 14 relatively slow pace. Nevertheless, if 15 you look back to 2009, if you were to 16 think about just the labor force, there are about 40 or so thousand more people 17 18 in the labor force and a good almost 80,000 people who are employed. So those 19 20 are real numbers, and when you think about the drivers, as Council President 21 Clarke said, moving 100,000 people out of 22 23 poverty, the one way to do that is to get people meaningfully attached into the 2.4 25 labor force.

Page 31 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. This is a very difficult chart to understand and to read and certainly 3 at a distance even more difficult, so my 4 5 apologies for that. What we're trying to 6 get at here for those who can see it up close is what jobs pay in different 7 communities over time. So the first set 8 9 of bars on the left are Philadelphia, then another set being Chicago, then 10 11 Memphis and Milwaukee. Why Memphis and Milwaukee? Because those are two other 12 cities that have poverty rates that are 13 14 in and around what Philadelphia's are, 15 although they are smaller cities than 16 Philadelphia. And what you'll notice is that 17 18 the percentage of jobs paying the lowest 19 wages -- so these are jobs, not people 20 with jobs. These are actual jobs in the 21 City of Philadelphia. Compared to Chicago at least, we have a larger 22 23 percentage of our jobs paying the lowest wages and a higher percentage of jobs 2.4 25 persistently paying the higher wages. So

Page 32 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. work, just physical jobs located in the 3 City of Philadelphia, pay less than they pay, and certainly in Chicago, and we're 4 5 better off vis-a-vis this measure than 6 Memphis and Milwaukee. Now, if you think about the earnings that people have, and these are 8 9 residents of Philadelphia, regardless of where they work, if they work in the City 10 11 or if they work in the suburbs, we have 12 higher percentages of people earning, of our residents, earning less money whether 13 14 or not they're working here in 15 Philadelphia or not. And so you have 16 this sort of mismatch between the jobs 17 that are here and the people that are 18 getting them. 19 The next chart shows you --20 well, Philadelphia, we did the 21 estimations. Philadelphia would have to have 20,000 fewer people in the lowest 22 23 wage category and 17,000 fewer people in that middle wage category just to have a 2.4 25 profile that looks like Chicago in 2017.

Page 33 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. That's almost 38,000 people, roughly 3 speaking, that would be moved sort of 4 above that line if we're able to sort of 5 move those wages up a bit. 6 So this next chart is showing 7 you if we were to compare ourselves to those same set of cities, but notably I 8 9 think Chicago on this particular chart, again, is that we have a lower percentage 10 11 of our jobs paying \$3,300 a month, which 12 is just \$40,000 a year, but that's the best way you can break these data. 13 14 although that has been rising, we remain 15 consistently below that comparator city 16 of Chicago. 17 COUNCIL PRESIDENT CLARKE: Ira, 18 excuse me. 19 I'm sorry to interrupt, Madam Chair. 20 21 One quick question. With respect to the incomes rising, did you 22 23 dig down to determine if that was because of people moving into the City with 2.4 25 higher incomes or people who currently

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1	10/10/19 - POVERTY REDUCTION - RES. 190239	
2	live here incomes are rising? Because	
3	that's kind of important in terms of what	
4	we're tying to do.	
5	DR. GOLDSTEIN: Yes. And the	
6	next slides well, we do have that	
7	data, yes.	
8	COUNCIL PRESIDENT CLARKE: All	
9	right.	
10	DR. GOLDSTEIN: We can tell you	
11	that. We can tell you whether or not	
12	they've moved in.	
13	COUNCIL PRESIDENT CLARKE:	
14	Sorry to break up your flow.	
15	DR. GOLDSTEIN: No, no, no.	
16	I'll get it back.	
17	So this is the percentage of	
18	folks who have let's see where we are.	
19	I'm sorry. The percentage of higher	
20	paying jobs to non-residents. So this	
21	pattern has actually been pretty	
22	consistent over time. In other words,	
23	people who live outside of the City of	
24	Philadelphia and work in Philly generally	
25	make more than people who live inside of	

Page 35 1 10/10/19 - POVERTY REDUCTION - RES. 190239 Philly and work inside of Philly, and 2. 3 that difference is not trivial, and you'll see that chart is showing that 4 5 those differences is maintained. Even if 6 you compare us to Chicago, which has a 7 similar pattern, we are worse off than them in that regard. 8 9 Just a little bit more stats and then you'll get to see some 10 11 geography. These are just the raw 12 numbers, and no need to dwell on these. I think that people understand at this 13 14 point that we had in 2016 close to 15 400,000 people who were living at or 16 below the poverty line and that we were fortunate that it has dropped down by 17 18 about 23,000, 25,000 people, but it's still an awful lot of people in the City 19 20 of Philadelphia who are suffering from 21 the economic deprivation that attaches to 22 such low wages. 23 These are the poverty rates of 2.4 the ten largest cities. Philadelphia is 25 the top. And, again, although we are

Page 36 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. down off that peak, we're not down all 3 that much off that peak, and we are in some ways going in a different direction 4 5 than some of the other large cities where 6 they're dropping at a rate that is a good 7 bit faster than our drop-off. And one estimation that we did 8 9 in comparing us again to the other large cities that have a more typical current 10 11 poverty rate of about 18 percent, we would have to have that hundred thousand 12 people fewer in poverty just to get to 13 14 that more typical rate of poverty. 15 Comparing Philadelphia to the 16 other poorest cities, what you'll find is that Philadelphia amongst the poorest 17 18 cities as opposed to amongst the largest cities, but amongst the poorest cities, 19 20 we are still in the upper reaches of 21 that. So, you know, higher rates of the 22 poorest cities in the country are Milwaukee, Memphis, Cleveland, and 23 Detroit. And a little better off than us 2.4 25 are places like Fresno, New Orleans,

Page 37 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Tucson, Miami, and Atlanta. But these are very different communities, and if 3 you think about what's happening to 4 5 Detroit and Cleveland and Milwaukee, 6 these are not places that I think that as 7 a lifelong Philadelphian, that economically that I'd like to compare 8 9 myself to, frankly. The groups met, the different 10 11 subgroups met, and identified several 12 different categories of people that they would like us over the course of the next 13 14 month and a half to drill into a little 15 bit, and those are children who are 16 living in households that are below the 17 poverty line. So people under 18. are people who are 18 to 64 but in the 18 labor force, and then 18 to 64 and out of 19 20 the labor force, and then people who are 65 and older. 2.1 22 Why those breakouts? Those 23 breakouts are really critical because the kinds of policies, programs, legislation, 2.4 25 intervention strategies and the like are

Page 38 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. very different when you're talking about 3 somebody who is over the age of 65 than 4 for children, and the flows of funds, as 5 Councilman Domb mentioned, the EITC and 6 the variety of other things, those attach differently to different populations. 7 And so the ability to sort of be able to 8 9 drill down on those groupings is going to be critical, and we will continue to 10 11 drill down on those as we go. 12 One of the -- I'm skipping over some charts in the interest of time and 13 14 the fact that they really aren't visible 15 from this distance anyhow. So what I'd 16 like to just show here is the number of 17 people in those groupings and, for 18 example, using the 2006 American Community Survey data. What you find is 19 20 that there's about 123,000 people who are 21 under the age of 18 who are in poverty. You've got about 94,000 who are 18 to 64 22 but in the labor force, about 143,000 who 23 are 18 to 64 and out of the labor force, 2.4 25 and then about 31,000 who are over the

Page 39 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. age of 65. 3 So that gives you a sense of 4 the magnitude of the populations that you 5 are thinking about and the different 6 poverty rates for those different groups. 7 So, for example, those people who are 18 to 64 and out of the labor force, far and 8 9 away the highest poverty rates. And as you'll see in some later 10 slides, one of the critical things is 11 12 undoubtedly going to be figuring out how to get them to reconnect into that labor 13 14 force, because that does make a huge difference in what the conditions are 15 16 economically that people confront. This next chart -- or this 17 figure is showing you the different 18 19 income levels of the groups. So you're 20 going to see, for example, the far left, 21 under 18 and in poverty and then under 18 but not in poverty. The little lines 22 23 inside those boxes that you can see if you stare closely, that is the middle 50 2.4 25 percent of the population in that group.

Page 40 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. The little green dot is the median. 3 that's the typical income for people in that group. And then the large box shows 4 5 you about the middle 80 percent. So 80 6 percent of the people in that group fall 7 within that box. And you'll see there's actually some overlap in those particular 8 9 boxes. And so when you think about what the conditions are, they're actually 10 11 quite variant. For example, if you look at in 12 poverty but not in the labor force versus 13 14 not in poverty and not in the labor 15 force, there's actually, like some of 16 those other groupings, there are some 17 different overlappings, but you'll also notice those typical incomes vary a good 18 19 bit. 20 Let me just go back to one 21 slide and point out, without giving any detail, that we will, before this is all 22 said and done, be able to give you the 23 numbers of people who are self-reporting 2.4 25 with a variety of different disability

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2	conditions, like people's ability to	
3	cognitive disabilities or ambulatory	
4	disabilities or ability to live	
5	independently. These are all	
6	self-identified conditions in the	
7	American Community in the census, in	
8	the American Community Survey, and there	
9	are tens of thousands of people in the	
10	City of Philadelphia in each of these	
11	groupings that identify with each of	
12	those disabilities. The complexity here	
13	is that not all of them necessarily, not	
14	all of those disability conditions	
15	preclude one from participating in the	
16	labor force. And it's going to be a	
17	little difficult to disentangle all of	
18	that, but it's important to know that	
19	those populations do exist and the size	
20	of them.	
21	So with that, I'm going to turn	
22	it over to my colleague Emily, who is	
23	going to speak to you a little bit about	
24	the geography of poverty in Philadelphia.	
25	MS. DOWDALL: Good afternoon.	

Page 42 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. So I'm going to talk a little bit about 3 the geography of poverty and then take a 4 look at how some of the geographic 5 patterns differ across the different 6 drivers of poverty, as well as some of the factors that influence and affect the 7 lives of those living in poverty and then 8 9 also the policy and programming intervention possibilities for those 10 11 folks living in poverty. 12 So, first, this is a map of people living in poverty. Those darker 13 14 areas are areas where there are more 15 households in poverty. This map will 16 look pretty familiar, I think, to many 17 people in this room. Poverty is somewhat concentrated in the City, but there are 18 pockets of poverty in every Council 19 20 district. And so I think it's very clear 2.1 that this is an issue that impacts all communities across the City. 22 23 Now, looking at one of those major drivers of poverty, these are 2.4 25 concentrations of adults who do not have

Page 43 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. a high school diploma. The very darkest 3 areas there are areas where at least 50 4 percent of the population does not have a 5 high school diploma. 6 Again, going back to some of the numbers that Ira went through, the 7 high school diploma is something that is 8 9 the difference between you being able to earn a wage that is going to keep you in 10 11 poverty or earn a wage that will allow 12 you to rise above poverty. And I would just point out that those sort of dark 13 14 yellow areas where 25 percent of the 15 population has no high school degree are 16 areas where there's also roughly 25 17 percent of the population or more also living in poverty, and that's no surprise 18 19 given how important and essential that 20 high school diploma is. 21 Now, this is areas where the population has some college but did not 22 23 earn a degree. And so they may have the burden of the debt taken out to attend 2.4 25 college, but none of the income benefit

Page 44 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. of having the actual degree. This can contribute to general 3 financial instability, and this is also 4 5 an area for possible intervention where 6 there can be efforts to connect people back with education to finish the degree. 7 This is the share of residents 8 9 who are earning \$40,000 a year or less. That's about the median income for the 10 11 City of Philadelphia. So these are areas where incomes are below median for the 12 City. And, again, this looks very 13 14 similar to that map of poverty. This is, 15 again, for workers, no matter where they 16 work, but who live in these areas. Now, this is moving down to 17 workers earning \$15,000 a year or less. 18 So these are places where people are 19 20 working, but not earning a living wage 21 through that work and where they're 22 concentrated in the City. 23 Now, this is a map of the share of tax returns that had an EITC claim 2.4 25 with them. We would expect this to look

Page 45 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. a lot like that map of low-income 3 earners. And combining this map, comparing this with that last map could 4 5 also help point you to places where 6 people might be under-enrolled and not 7 taking advantage of that program. This is a map of households 8 9 with single parents and children. know that the poverty rate is 10 11 substantially higher for children than it 12 is for the general population, and many of those children live in single-parent 13 14 households. 15 This is the population that 16 lacks health insurance. It looks a little bit different from that last map 17 18 in part because children have access to more programs than, of course, single 19 adults do, but lacking health insurance 20 21 can open you up to the insecurity of being thrown off by one emergency medical 22 bill, and it can also interfere with 23 attachment to the workforce if you're not 2.4 25 addressing underlying health problems

Page 46 10/10/19 - POVERTY REDUCTION - RES. 190239 1 because of that lack of insurance. Now we're going to look at some maps related to housing and poverty. 4 5 these are areas -- the darker areas are 6 where homeowners are spending more than 30 percent of their income on housing. That's how cost burden is defined. 8 9 When we then take a look at renters, things look a lot worse. 10 11 are a lot of -- more than 50 percent of 12 renters across the City are cost burdened. And I think the notable thing 13 14 about this map is that it really 15 resembles a lot more the map of low 16 incomes than it would resemble a map of 17 high prices. So a lot of the cost burden 18 is associated with very low incomes, and these are households who just can't 19 20 afford really anything that's on the 21 market given their income. 22 MS. GLADSTEIN: Could you give us the numbers of households at some 23 later point, not just the percent, in 2.4 25 both those categories?

Page 47 1 10/10/19 - POVERTY REDUCTION - RES. 190239 MS. DOWDALL: Sure. MS. GLADSTEIN: Thank you. 3 4 MS. DOWDALL: And now here are 5 subsidized renter households. And so for 6 that subgroup of low-income households who are able to access subsidized housing, this does alleviate that burden 8 9 and can make for better financial stability, but we know that the demand 10 11 and the need for subsidized housing far 12 outstrips the availability of those 13 units. 14 This is a map of foreclosures 15 in Philadelphia as a share of sales. 16 can see that in many cases they are 17 concentrated in high-poverty areas and areas just outside those high-poverty 18 areas. Foreclosures, of course, 19 destabilize households but also the 20 communities around them. We know there's 2.1 22 a lot of research on how they can affect 23 housing prices and investment in the surrounding neighborhood. 2.4 25 These are eviction filings as a

Page 48 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. share of renters in neighborhoods. Again, we see some concentrated pockets 3 that also tend to be in high-poverty 4 5 areas. Evictions, of course, are also a 6 very destabilizing event. And now I'm just going to walk quickly through our market value 8 9 analysis. So this is an analytic tool that Reinvestment Fund developed. 10 11 sorts all of the block groups in the City 12 of Philadelphia into one of ten categories here from the strongest "A" 13 markets to the weakest "I" markets and 14 15 everything in between. This analysis 16 allows us to think about how to 17 strategically use resources and match the right policy tools and interventions to 18 different types of real estate markets. 19 20 The first one was done back in 2001, and this, I believe, is our seventh one that 21 we just completed. 22 23 So these are the indicators, which, of course, the print is much too 2.4 25 small for anyone to see, but there are

Page 49 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. about 12 different data indicators that go into making these market categories. 3 So while sale price is important and 4 5 you'll see if you look from "A" to "I" 6 that the sale price goes from higher to lower, it's not the only thing that makes 7 a real estate market. 8 9 So, for example, when you look at the foreclosure column, you'll see 10 11 that the highest foreclosures are in 12 those yellow middle markets. And I think for our purposes today, it's also really 13 14 important to look at the subsidized 15 renters and where they are and the 16 ability of renters living in subsidized 17 housing to access these more stable, 18 stronger markets. And you see that in fact in those "D" markets, you do see a 19 20 relatively high share of subsidized 21 housing units. Now comparing the 2018 MVA back 22 23 to the one in 2015, in some ways it's been seen as good news that you see 2.4 25 growth of the stronger market categories.

Page 50 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. We saw a lot more purple and blue. We saw a lot of orange turning to yellow up 3 in the Lower Northeast and in the 4 5 Northwest, but we also see that the 6 lowest -- the weakest markets have grown 7 smaller, and what this means is that there's also fewer affordable markets. 8 9 So while investment in the City is a good thing, it means that there are fewer 10 11 areas that are accessible to our lowest income families. 12 And another way of looking at 13 14 this is this next series of maps that I'm 15 going to show you, which is something we 16 think about as our geography of 17 opportunity. So what types of real estate markets can people of different 18 19 income levels access in the City. So here we take the median 20 2.1 income for a white household in the City of Philadelphia and see what markets 22 23 they're able to access as determined by the median sale price on those markets. 2.4 25 And so you can see that this family can

		Page 51
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2	access areas that are many of the yellow	
3	middle markets as well as the orange and	
4	the salmon color weaker markets.	
5	Now when we look at the median	
6	income for African American households,	
7	you see that the geography becomes more	
8	constrained, because that's a lower	
9	median income. When we look at the	
10	median income for Hispanic households,	
11	it's even more constrained.	
12	We also looked at markets	
13	compared to the poverty line income for a	
14	family of four in Philadelphia, and you	
15	see that households earning poverty	
16	income are only going to be able to	
17	access really these very weak markets.	
18	So with that, I'm going to turn	
19	it over to Octavia.	
20	COUNCILWOMAN SANCHEZ: Can you	
21	hold it a second. Just for the purposes	
22	of flow, I didn't know if any of the	
23	Co-Chairs had some immediate questions	
24	around the maps, and then I'll have	
25	Octavia, if it's okay with you.	

Page 52 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. MS. MATLOCK-TURNER: Yes. curious when we sort of look at the data 3 around poverty, especially around 4 5 household income, how do we count or do we have the ability to count unrelated 6 7 adults who are sharing households? does that data get incorporated? 8 And I 9 think of that just to wonder whether or not there could be some incentives as 10 11 we're looking for transitional support 12 and driving down expenses to that 30 13 percent number. 14 But I'm just curious, do we 15 have any sense at all in how we collect 16 the data how unrelated adults maybe with children are ultimately cohabitating? 17 18 DR. GOLDSTEIN: I think the answer to that is, we don't know that 19 right this second, and I think that the 20 21 data that we're going to have accessible to us to do that is going to be somewhat 22 23 less than perfect, but I think we can get it to be indicative, particularly if 2.4 25 we're trying to figure out in general

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2	what that means in terms of contribution	
3	to the household income.	
4	I think we will be kind of	
5	hard-pressed to be able to say that there	
6	are X number of households very	
7	specifically that are in that condition,	
8	but I think we can get at it as the	
9	subcommittees opine on the different	
10	intervention strategies.	
11	MS. MATLOCK-TURNER: Thank you.	
12	COUNCILWOMAN SANCHEZ: Any	
13	other questions from the Committee?	
14	(No response.)	
15	COUNCILWOMAN SANCHEZ: Octavia.	
16	MS. HOWELL: Well, thank you.	
17	As Ira mentioned, I am Octavia Howell, a	
18	researcher with Pew Charitable Trusts	
19	here in Philadelphia, and we have done a	
20	lot of research about poverty in the past	
21	several years. As a part of that	
22	research, we commissioned a poll of 1,600	
23	Philadelphia residents of all income	
24	levels and asked them a number of	
25	questions about their lives. These poll	

Page 54 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. results provide much of what I'm going to 3 talk about here today. 4 Our survey gave us two ways to 5 identify people living in poverty. One 6 was based on matching the size of their households and incomes to the federal poverty threshold and the other was their 8 9 answer to a simple question: Do you consider yourself to be poor? And the 10 two methods produced very different 11 12 results. We found that 45 percent of 13 14 those whose income and family size met 15 the federal definition of poverty said 16 that they were not poor. We have a quote 17 here from Lucy, which is one of my 18 favorite quotes from our focus group's work, where she said, "We have to work 19 20 for things, but I don't think anybody here is really, really poor. We have 21 water. We have food." 22 23 Through our polling, we learned that those experiencing poverty in 2.4 25 Philadelphia share many of the same

Page 55 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. concerns as other residents in the City, 3 including crime, education, and jobs, but they face unique obstacles. 4 5 As Emily outlined, one of those 6 obstacles is the ability to find and maintain stable housing. And we learned in our poll that 39 percent of 8 9 respondents to Pew's 2016 poll said they grew up poor and that they had moved a 10 lot when they were young, compared to 11 12 only 14 percent of those who had not 13 grown up poor. And more than a quarter 14 of them said they had been evicted or 15 homeless during childhood, compared with 16 only 6 percent of those who did not grow 17 up poor. 18 Data from the American Housing 19 Survey sheds additional light on the 20 problem of housing instability for 21 Philadelphia's poor residents. Results 22 from this census survey revealed that more than four out of five households in 23 Philadelphia, poor households in 2.4 25 Philadelphia, live in private market

Page 56 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. housing with no rent subsidies, and among 3 them, 94 percent were cost burdened, 4 spending more than 30 percent of their 5 incomes on housing. Eighty percent were 6 spending more than half of their incomes 7 on rent, mortgage, and utility payments. Households with incomes below 8 9 the poverty level also have greater exposure to inadequate housing, with 17 10 11 percent reporting such conditions, 12 compared to 8 percent of those with incomes from above the poverty line. 13 14 Education was another issue 15 that was very important to Philadelphia 16 residents, including those with incomes below the poverty line. And like all 17 18 residents, poor Philadelphians are concerned that public schools are doing a 19 20 poor job -- poor or fair job, excuse me, 21 educating students. Sixty-nine percent of the poor felt this way, along with 75 22 23 percent of those with higher incomes. Yet, 70 percent of poor parents 2.4 25 responding to our poll said that their

Page 57 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. children -- they had children that 3 attended District-run public schools, 4 compared with 46 percent of non-poor 5 residents. 6 Data from the Philadelphia School District revealed that 2 percent 7 of all poor students in Philadelphia 8 9 public schools attend schools with high achievement ratings, while 75 percent 10 went to schools that received the lowest 11 achievement rating of intervene, compared 12 with 59 percent of those who are not 13 14 poor. 15 Only 61 percent of poor 16 students who entered District-run schools in 2012 went on to graduate in 2016, 17 18 compared to 70 percent of the non-poor. Nationally, 84 percent of public high 19 20 school students graduated in four years. 2.1 The impact of these metrics is significant. You'll recall that Ira and 22 23 Emily both pointed out that poverty is higher among those who do not have a high 2.4 25 school degree. According to the 2016

Page 58 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. American Community Survey, the poverty 3 rate for Philadelphians without a high school diploma is 10 percentage points 4 5 higher than the citywide rate. 6 By far, Philadelphians ranked crime as the most important issue facing 7 the City, including 52 percent of poor 8 9 residents and 41 percent of those of higher income responded to the poll. 10 11 those living in areas with high poverty 12 have greater exposure to the City's crime. Police Department data reveal 13 14 that there was an average of 40 crimes 15 per thousand residents in the City's 16 poorest neighborhoods, compared to 14 in 17 areas where poverty rates were below 20 percent. This contributes to a sense of 18 insecurity for many poor residents. 19 In our poll, 48 percent of 20 2.1 those with incomes below poverty said they felt safe outside in their 22 23 neighborhoods at night, compared to 64 percent of those with higher incomes. 2.4 25 We sought to understand the

Page 59 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. impact that poverty had on the physical 3 and emotional well-being of Philadelphia 4 residents. Surveys by the 5 Philadelphia -- excuse me; by the 6 Philadelphia Health Management 7 Corporation showed us that Philadelphians in poverty had higher levels of stress 8 and had higher levels of chronic illness 9 than other residents. Research from the 10 11 Virginia Commonwealth University also 12 revealed that those living in the poorest parts of the City have life expectancies 13 14 as much as 20 years lower than 15 Philadelphia residents -- in wealthier 16 parts of the City. Excuse me. 17 From our own poll, we learned 18 that Philadelphians who said they grew up 19 poor were twice as likely to report 20 having experienced neglect and abuse 21 while they were growing up. Thirty-eight 22 percent who said they had grown up poor 23 also said they had been cared for during their childhood by someone with mental 2.4 25 illness, depression, or drug problems.

Page 60 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Of those who were not poor as children, 3 16 percent had caregivers with those 4 issues. 5 In our focus group, we talked 6 to residents about their experiences with work, and as might be expected, many shared that they often performed one or 8 9 more jobs under the table, taking advantage of their talents and skills to 10 11 earn money, and this type of work 12 typically does not show up in workforce participation figures. 13 14 Among those who do sustain 15 formal employment, the American Community 16 Survey data revealed that only 21 percent 17 were employed full time all year at work. 18 An additional 20 percent worked part 19 time. And, finally, I want to talk 20 21 about mobility from poverty as we revealed in our survey. Numerous studies 22 23 have found that many of the life experiences faced by the poor, 2.4 25 particularly in their childhood and young

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2	adult years, helped to perpetuate the	
3	condition of poverty, but Philadelphians	
4	reported that they experienced both	
5	upward and downward mobility.	
6	More than half of those who	
7	were poor at the time of the Pew survey	
8	reported having been poor during	
9	childhood, and others saw themselves as	
10	examples of upward mobility. More than	
11	half of those who said that they grew up	
12	poor were not in poverty at the time of	
13	our survey.	
14	COUNCILWOMAN SANCHEZ: Thank	
15	you, Octavia.	
16	Any questions from our panel?	
17	Sharmain Matlock Turner.	
18	MS. MATLOCK-TURNER: Thank you	
19	very much, all of you, for the data. It	
20	certainly is daunting to try to overlay	
21	all of this information on top of the	
22	challenge that we face to figure out what	
23	are the right kind of strategies and/or	
24	interventions that we really need to put	
25	in place. And so I was curious on one of	

Page 62 1 10/10/19 - POVERTY REDUCTION - RES. 190239 the slides, Ira -- and, again, I may have 2. 3 missed this, so forgive me if I did, but I know we talked about Philadelphia's --4 5 or Pennsylvania's minimum wage being at 6 7.25 as being one of the lowest when we 7 did some comparisons to other cities and what their statement on wages are, but 8 9 I'm curious, do you also have a sense of how many people in Philadelphia are 10 11 actually working at 7.25 an hour? 12 even though that is the minimum wage, if we sort of look at the working poor, are 13 14 half of them -- is that their wage or 15 not? 16 DR. GOLDSTEIN: So I couldn't 17 give you the actual number off the top of 18 my head, but if you think back to those charts that had a lot of bars going 19 20 across divided into three categories, 21 that lowest category, the one -- that 22 chart there. The portion of the chart 23 that's in red, those are people who -those are jobs that are paying under 2.4 25 \$1,250 a month. That's the lowest income

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2	cut that we could make. And if you sort	
3	of think about seven and a quarter times	
4	40 times four, that gets you very close	
5	to that total. So we will be able to get	
6	you reasonably close to that answer, but	
7	not precisely there. But we'll get you	
8	close.	
9	MS. MATLOCK-TURNER: Okay.	
10	Thank you. Because I know that's	
11	certainly one of the areas that we spend	
12	a lot of time talking about.	
13	DR. GOLDSTEIN: And this is	
14	really important. It's an important	
15	distinction too, because we often times	
16	look at household income, but apropos to	
17	your other question, like there are	
18	people who contribute, but these are the	
19	actual jobs what they pay. So it will	
20	get us closer to your answer.	
21	MS. MATLOCK-TURNER: Thank you.	
22	COUNCILWOMAN SANCHEZ:	
23	Councilman Domb.	
24	COUNCILMAN DOMB: Thank you.	
25	Great presentations by all of	

Page 64 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. you, by the way. Thank you very much. Great information. I just have a few 3 4 questions I wanted to ask. 5 You have a chart in here that 6 talked about Philadelphia's unemployment and it talked about how we have not 7 really -- I think it's Page 10. Why do 8 9 you think that is occurring, in your 10 opinion? DR. GOLDSTEIN: Well, I think 11 12 that there's probably a few reasons that First of all, our economy over the 13 14 last 40, 50 years, whether or not you 15 look at housing, whether or not you look 16 at employment, you'll notice even in just 17 this relatively shorter period of time we 18 are way less volatile than other cities. 19 Some people will draw the 20 analogy between a stock and a bond. 21 Philadelphia is more like a bond. doesn't fluctuate as much as some other 22 23 cities. So I think that is a piece of 2.4 it. 25 I think it's also the case that

Page 65 1 10/10/19 - POVERTY REDUCTION - RES. 190239 our economy in its comeback has come back with a different set of jobs that are 3 going to be a little -- the implications 4 5 for the incomes are less. 6 So I don't know that I can give 7 you a fully definitive answer, other than some of the other cities have more 8 9 diverse or less diverse economies if you think about the regions of the country, 10 11 like San Jose. You know, that is sort of 12 the tech corridor of the country for sure, and so it's going to be influenced 13 14 by a variety of other much stronger 15 driving forces. Ours are going to be 16 driven by service, health, and education. COUNCILMAN DOMB: Would you say 17 18 that maybe because we were such a large manufacturing base 30, 40, 50 years ago, 19 20 that the economy changed so dramatically 2.1 that we did not keep up with the training of our workforce and what we're teaching 22 23 in schools to compete with today and 2.4 tomorrow's economy? 25 DR. GOLDSTEIN: I think that

Page 66 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. that's a fair extrapolation from that 3 change for sure, and I think what we have to figure out to a degree is whether or 4 5 not we've done a better or worse job than some other cities. If it's a factor, we 6 7 need to do a better job. So the comparative piece there I think is less 8 9 important than the absolute of are we doing a good job doing that. 10 11 COUNCILMAN DOMB: One other 12 last question I have. I think it was Page 24. You talked about people not in 13 14 the labor force. Are people not in the 15 labor force because they've given up 16 looking for a job, or what would be the reason not to be in the labor force? 17 18 DR. GOLDSTEIN: Typically not 19 in the labor force means, for whatever 20 reason, I've given up work. It's not that I'm in school. It's not that I'm 21 temporarily laid off. It's that I've 22 23 given up looking for work. So, you know, I think we can, 2.4 25 over the course of the next month and a

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2	half, try and determine what the largest	
3	drivers of that are, whether or not it's	
4	that they tried for a given period of	
5	time and just gave up or whatever else is	
6	the situation that attaches. That will	
7	require us to look at other data than	
8	typical census data, but we can figure	
9	some of that stuff out, and that might be	
10	helpful in terms of a lever to pull	
11	people in, since that labor force	
12	participation is so important to getting	
13	the incomes above poverty.	
14	COUNCILMAN DOMB: Because	
15	that's your biggest number in here, isn't	
16	it, the not in the labor force?	
17	DR. GOLDSTEIN: It's a big	
18	number, yeah.	
19	COUNCILMAN DOMB: Okay. Thank	
20	you.	
21	DR. GOLDSTEIN: And it's not	
22	older people, because we've broken it out	
23	separately.	
24	COUNCILMAN DOMB: Right. Thank	
25	you very much. Thank you.	

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2	COUNCILWOMAN SANCHEZ: For	
3	clarity purposes, would a lot of those	
4	folks fall in the category of permanently	
5	disabled?	
6	DR. GOLDSTEIN: It could be	
7	that some are permanently disabled, yes.	
8	I don't know the exact composition,	
9	though, but, yes, that could be it.	
10	COUNCILWOMAN SANCHEZ: For the	
11	purposes of the mapping around	
12	particularly like North Central where you	
13	have Temple, what variation do you allow	
14	for that student population that moves	
15	around as it relates to that concentrated	
16	pocket?	
17	DR. GOLDSTEIN: So that is an	
18	issue for sure in the official poverty	
19	data. The best estimates that I've seen	
20	is that Philadelphia's student population	
21	accounts for a couple percentage points	
22	possibly overall in the poverty rate. In	
23	any given area it could be a lot more,	
24	though. And so in those few census	
25	tracts in and around Temple where	

Page 69 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. students have found housing off campus, 3 then, yes, particularly if they're going 4 through the exercise of -- assuming for 5 purposes of the census saying that's 6 their residence, that could make the 7 poverty rate in those places look higher. But there are communities that are entire 8 9 communities where the Census Bureau has estimated 10, 12 percent of the poverty 10 11 rate is a function purely of a student place like Ann Arbor, for example, where 12 it's all students. 13 14 COUNCILWOMAN SANCHEZ: Of these 15 data -- I know one of the things that we 16 want to be mindful of in creating these 17 categories is like how do we measure the 18 impact of what we want to do moving 19 forward. What other variations should we 20 be paying attention to with the limitations of the data, right? The data 21 is the data, but what other things should 22 23 we keep in mind, whether it's geography based or category based? 2.4 25 DR. GOLDSTEIN: Well, I think a

Page 70 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. couple --3 MS. GLADSTEIN: Could I add in 4 margin of error in terms of these 5 one-year reports. Talk about that a 6 little bit. DR. GOLDSTEIN: Sure. And you'll notice on a lot of these things, 8 9 we've tried not to use one year's worth of data. We've tried to sort of fold 10 11 together two or even three years, because 12 we think that there's a lot of volatility in the data. 13 14 Particularly the supplemental 15 poverty measures, those are based on data 16 from the current population survey, and 17 those are samples of, I think, closer to 18 500 or 600 for the entire City. So we 19 felt it was important to roll together a 20 few years. 2.1 Some of the other things that I 22 think might be useful to pay attention to 23 here are people who are sort of close to those categories and will be migrating in 2.4 25 and out, and so that you're trying to

Page 71 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. sort of think about a problem that you're 3 tackling now and that's the "now" 4 snapshot, but there's a whole other 5 cohort of people moving through and a whole cohort of people moving out. 6 So I think a little bit of that mobility will be important for us to take 8 9 account of, because, again, that's going to influence your ability to sort of hit 10 11 your target number. 12 I also think it probably makes sense to pay attention to the people who 13 14 are sort of near the margin. So that's 15 why one of the charts we showed you 125 16 percent of poverty. 17 For other things that we've 18 been involved in, when you look at, for example, school lunch participation, 19 20 subsidized school lunch participation, 21 that percentage of people who are sort of near poverty, get the reduced price 22 23 lunches, they're a very volatile, economically volatile, group, because one 2.4 25 day they're above and then something

Page 72 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. happens and all the sudden they're below 3 poverty. 4 So I think in some ways we 5 should probably think about broadening 6 our lens on some of these things to look 7 at some of those populations just above poverty, because I also think that some 8 9 of those families are going to be in a situation where a really constructive 10 hand could sort of get them above and 11 12 more permanently out and not be in that position of volatility. 13 14 COUNCILWOMAN SANCHEZ: 15 MS. GLADSTEIN: I think the 16 Council President might have asked this 17 question, but I just want to stress it. 18 One of my concerns always is that while we want to encourage support, activate 19 20 mobility out of poverty, knowing who is 21 moving in also and what their needs may be and are there like finite ways to 22 23 actually measure that is one question I 2.4 have. 25 My second is different,

Page 73 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. unrelated. So why don't you respond to 3 that. 4 DR. GOLDSTEIN: The first one 5 is difficult. I mean, I think that what we should probably be doing is paying 6 7 attention to a variety of different pieces of data. Like, for example, 8 9 people are born every day in Philadelphia and people pass away every day in 10 11 Philadelphia, and there is reasonable 12 data that the Health Department has access to that can help us think through 13 14 that process of who is being born to 15 whom, where in the City, making sure if 16 we think about connection to benefits and things like that, that we're very focused 17 on making sure that people do connect at 18 the earliest moment or when people pass, 19 20 that property passes in a reasonable way. 21 So I think that that's probably 22 another piece of data that will help us 23 design closer to in time and immediately responsive kinds of things, whether or 2.4 25 not we're talking about benefits,

Page 74 1 10/10/19 - POVERTY REDUCTION - RES. 190239 frankly, or housing or workforce, because 2. 3 those all attach. 4 We also at our organization pay 5 a lot of attention to childcare, and one 6 of the things that has drawn us to that is this notion that firstly high-quality 7 childcare, the evidence shows, prepares 8 9 people, prepares children to get into kindergarten and perform in school, which 10 11 is going to hopefully address that 12 disparity in terms of high school graduation. It's a long slav to get to 13 14 it, but we think that that's a piece of it. 15 16 So sort of thinking through 17 what are the investment strategies that 18 we're going to get at essentially early that will have a long term but more 19 20 permanent return I think is probably 21 important for us. 22 MS. GLADSTEIN: My second is more of a comment, which is as we do our 23 work, I think it's important for us to 2.4 25 both think about individuals and think

Page 75 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. about them as households as well, because 3 a lot of the data is based upon 4 households, and understanding if the 5 individuals in that household, for 6 example, are under or over 65, have a 7 disability that prevents them from working or not, are young and should not 8 9 be expected to be in the workforce, or hopefully should be able to attach to the 10 11 workforce. 12 So it's just kind of something for us all to be conscious about, because 13 14 I think we kind of flip back and forth between the household and the 15 16 individuals, without ever really knowing 17 who was in the household. And if we can 18 make an investment in a potential wage 19 earner in the household or help an 20 individual, a child in the household get 21 health insurance or early childhood education, like what are those 22 23 differences. 2.4 DR. GOLDSTEIN: We will do 25 that.

Page 76 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. COUNCILWOMAN SANCHEZ: going to recognize Sharmain and then 3 4 Councilman Domb. 5 MS. MATLOCK-TURNER: In sort of 6 thinking about the people who have 7 multiple barriers to success, I mean, I think of people who are homeless, people 8 9 who have drug and alcohol problems, maybe people who have some kind of criminal 10 11 record. How do you account in the data 12 for people -- are they in the disabled group? Are they just a part of the 13 14 poverty number? How do you account for 15 folks that we know have multiple barriers 16 and need very specific interventions just 17 for health and safety and an opportunity 18 to be able to make it from day to day? 19 DR. GOLDSTEIN: I think from a 20 data perspective, honestly, it's a tough 2.1 I think that there probably are significant undercounts of some 22 populations in the official data and then 23 reasonable counts of other of the 2.4 25 groupings that you had mentioned, the

Page 77 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. other -- for example, I suspect that a 3 lot of the homeless population is probably not getting picked up, 4 5 notwithstanding the best efforts of 6 outreach around the American Community 7 Survey and the like. My quess is that there's probably a reasonably significant 8 9 undercount. On the other hand, people who 10 11 are -- you had mentioned folks who have criminal records and the like. 12 I think that to the extent that any population is 13 14 being counted or not, I don't know that 15 there's any particular reason to suspect, 16 and I haven't seen a lot of data to 17 suggest, that there's a differential undercount there. So I think that that 18 19 population is probably as well as any other counted in the census and then 20 2.1 accounted for in the poverty rate. I do think that probably some 22 23 of the drug and alcohol, particularly if it's so severe that people are not 2.4 25 accessible to those who are collecting

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2	data or the homeless population of	
3	Philadelphia, my guess is that there's	
4	not a trivial undercount there.	
5	Now, what I would say is that	
6	it's probably not different than other	
7	cities, but, again, I don't think the	
8	comparison to other cities is as	
9	important as we got a job here, right?	
10	And whether or not we have more or less	
11	than Chicago or whether or not the	
12	differential undercount is the same or	
13	different, less important, but the fact	
14	that it is a population that's going to	
15	need attention.	
16	COUNCILWOMAN SANCHEZ: I want	
17	to acknowledge that Councilwoman Blondell	
18	Reynolds Brown has joined us. Thank you,	
19	Councilwoman.	
20	Councilman Domb.	
21	COUNCILMAN DOMB: I have	
22	another question on the data. I guess	
23	from what I can tell, and these maps are	
24	very good, clearly the poverty is focused	
25	in Council District 7, Council District 5	

Page 79 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. and 3 and 8, with 5 and 7 having the 3 most, according to the chart, poverty. And then when you go to the educational 4 5 attainment, I don't know why, but District Council 7 seems to have the 6 7 lowest educational attainment of any of the -- this is great information, I 8 9 think. And then when you go to the share of residents that are less than 15,000 a 10 11 year, I would have thought it was District Council 7, but it's District 12 Council 5. Why do you think that is? 13 14 DR. GOLDSTEIN: I don't have an 15 answer for you on that one, but I think that was an anomaly. You do see it a 16 17 little bit -- you see that population pop up in District 5. That's where you're 18 seeing that hotspot. 19 20 But Emily may --2.1 MS. DOWDALL: And it could also 22 be that in that area which has so many 23 people without high school diplomas, there are going to be a higher share of 2.4 25 people who aren't in the workforce at

Page 80 1 10/10/19 - POVERTY REDUCTION - RES. 190239 all, so they're not showing up as earning 2. 3 less than \$15,000 a year, because that's only earners captured in that figure. 4 5 COUNCILMAN DOMB: So if we were thinking of ideas, just to throw this out 6 there, would it make sense for us to 7 focus on those four districts and 8 9 specifically 5 and 7 and figure out like can we use the libraries to get people 10 11 high school diplomas, get them the 12 courses, can we get Community College to have courses offered in those areas, like 13 14 could we focus on those districts where 15 we know there's issues? 16 DR. GOLDSTEIN: I think you're 17 going to hear from Benefits Data Trust 18 and United Way, for example, on some of this notion of targeting based on where 19 20 the need is and getting the service right 21 there, and I'm sure they'll have something very specific to say about 22 23 that. But I think on its face, it's a reasonable thing to think about. 2.4 25 COUNCILWOMAN SANCHEZ: I just

Page 81 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. want to add the factor of race, because 3 the difference between the 5th and 7th is 4 that I have a large Hispanic population, 5 language minority population. That also 6 counts for some of that variation. So if 7 you would take it by race in those two districts, you would see, it would pop up 8 9 for you. COUNCILMAN DOMB: 10 Okay. Thank 11 you. Thank you very much. 12 COUNCILWOMAN SANCHEZ: Thank 13 you. 14 I'm going to recognize Mel 15 Wells. 16 MR. WELLS: I have to excuse 17 myself, so I did not want to be rude as 18 we're speaking about our family members 19 that are in poverty, but one thing I really heard -- maybe you didn't say the 20 21 word -- was generational poverty and expectations, and it sounds like a lot of 22 23 individuals are stuck in poverty because that's their expectation that that's how 2.4 25 they're supposed to be, especially when

Page 82 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. you ask the question of are you poor and 3 the person I believe in North Philadelphia said that, oh, I'm not poor, 4 5 this is just how it is and so forth. 6 So I believe one of my jobs here and also the Co-Chairs is also to 7 change the expectations that are in the 8 9 community. Once we do that, then people can utilize the resources more. 10 11 So thank you for your time. 12 think you did a great presentation. 13 Thank you. 14 DR. GOLDSTEIN: Can I say one 15 thing about that? And we talked a little 16 bit about it in the meetings leading up 17 to this, is this notion of -- I think it's an increasingly popular notion that 18 people are now thinking that, for 19 20 example, there is more to our well-being 21 than how much money we have in our pocket or the GDP or GNP at the City or state or 22 23 county or country, and it's this notion of social well-being, and it's sort of 2.4 25 multidimensional and it gets to things

Page 83 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. like how safe do I feel in my community, 3 how connected do I feel to my neighbors, how is my health, what's the environment 4 5 within the community within which I live. And I think that in some ways, what that 6 shows is that economic well-being. 7 If you were to map that in 8 9 Philadelphia, and we have those dimensions, what you'll see is there are 10 11 places where sort of the adverse side of 12 those things piles up, but there are other places where the economics are 13 14 challenging, but there are strengths on some of those other dimensions. And I 15 16 think that although the Committee is very 17 much focused on sort of dropping that rate and getting 100,000 people to earn 18 at a level that gets them above the 19 20 poverty line, I think it's not -- it would be wise also to think about what 21 some of these other dimensions are and 22 23 help to think about what are the other conditions of people who are living 2.4 25 notwithstanding what their economic means

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2	are.	
3	MR. WELLS: Thank you so much.	
4	COUNCILWOMAN SANCHEZ:	
5	Sharmain.	
6	MS. MATLOCK-TURNER: Yes.	
7	Certainly one of the areas that everyone	
8	is concerned about whenever you talk	
9	about you want to change the poverty rate	
10	is, are we talking about a Philadelphia	
11	way, which is helping those who are	
12	really in need move up, or are we talking	
13	about a Seattle way where we move	
14	everyone out of Seattle because they can	
15	no longer afford to live there.	
16	When you look at the data, what	
17	do you look at to make sure that we are	
18	not looking at a displacement strategy	
19	for change, that we're really looking at	
20	support for those who are here to be able	
21	to move up?	
22	DR. GOLDSTEIN: Yeah. I think	
23	that before December, we'll have a much	
24	better handle on it, but those data that	
25	we use to create some of those challenge	

Page 85 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. slides, wherever they are, about the wage 3 levels -- I'm trying to get you into this 4 right place. This really sort of talks 5 about where people live and where they 6 work and talks about the income levels 7 that they have and a variety of other circumstances. It's a very unique 8 9 dataset and one that's not often used when you sort of think about these 10 11 poverty-related issues. I think we will be able to use that a little bit to 12 figure out like are we doing that. 13 14 Now, as much as I mentioned 15 that Philadelphia's cost of living and 16 our housing is higher, it's not anywhere near what a Seattle is, and so I don't 17 know that we have -- or a Portland or a 18 variety of other places. So I think that 19 20 there's probably some of that sort of 21 displacement of poor people into our sort of inner rings, but we will be able to 22 23 test that a bit and figure out whether or not we are achieving a decline in poverty 2.4 25 as a result of really helping people get

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 2.
         out of it or helping people out of the
         City. And, again, we would hope it would
 3
         be the former, but we should be able to
 4
 5
         get a handle on that.
 6
                   MS. MATLOCK-TURNER:
                                         Thank you.
                   COUNCILWOMAN SANCHEZ:
                                           Thank
 8
         you.
 9
                   I want to open it up real
         quickly to any of the Committee members,
10
11
         the general Committee members that are
12
         here, if you have a question.
         Sergeant-At-Arms has a microphone.
13
14
         of the general Committee members, if you
15
         have a question for our panel, the
16
         Sergeant-At-Arms has a microphone.
17
                    (No response.)
18
                   COUNCILWOMAN SANCHEZ:
                                           Okav.
         So one of the things to Sharmain's point,
19
20
         because poverty causes the level of
21
         mobility that it does, what should we be
22
         looking at as we create strategies or
23
         tactics and how would you make the
         modification in the data for the purposes
2.4
25
         of evaluating our work moving forward?
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Page 87 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. DR. GOLDSTEIN: That's a very good and difficult question. I guess we 3 4 could look at some of the surrounding 5 areas and see if they are essentially 6 feeling a change that's sort of 7 commensurate with ours or in an opposite direction than ours. So that's one 8 9 thing. I think it's also going to be 10 11 important for us to sort of think through 12 some of the complexities, I think the philosophical complexities, of let's say 13 14 that the City of Philadelphia is 15 successfully able to help a set of 16 families out of poverty and it becomes 17 their choice to go to Elkins Park right outside the City because it's an 18 opportunity for reasonably priced housing 19 and decent public education and things 20 2.1 that were the irritants that people identified in the survey. I mean, I 22 23 think that that's something that we ought to think about, like that's not 2.4 25 necessarily a bad thing, but I think that

Page 88 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. we need to figure out, we need to be able to quantify when that's happening versus 3 4 a sort of force-out. 5 COUNCILWOMAN SANCHEZ: And the 6 reason I say that, because so much of 7 this is about housing insecurity, which causes the mobility. So I want to make 8 9 sure, again, that we're being mindful of that as we look at our numbers and the 10 11 impact that we want to have. 12 DR. GOLDSTEIN: And I think that we dropped a lot of housing data in 13 14 here without, I feel in some ways, 15 without the tightest connection, because 16 in some ways, influencing somebody's 17 housing situation doesn't directly affect 18 how much money they have in their pocket, which doesn't directly affect whether or 19 20 not they count as poor or not. Nevertheless, it does influence the 21 situation and the condition on which 22 23 people are, A, living and, B, trying to get out of economic poverty. 2.4 25 So I think that we're going to

Page 89 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. need, before all this is said and done, really to sort of dig in on some of the 3 housing-related things in a more 4 5 constructive way so that we can get at 6 the issue that you're talking about. COUNCILWOMAN SANCHEZ: I iust think it's important for our framing. I 8 9 know that we're being rather aggressive and bold in what we're trying to do, and 10 11 just remind people some of the nuances to this and we're going to try to measure 12 this for one year, five years, and ten 13 14 years since Council President has put out 15 those numbers. So I just -- it's not 16 going to be a perfect science, but that 17 we are mindful of all of the things. mean, School District is here. They can 18 tell you that in January kids go to 19 different schools, right? That happens 20 21 often, right? The principals talk about it. So we just want to be a little bit 22 more mindful. 23 Eva -- oh, Councilwoman 2.4 25 Blondell Reynolds.

Page 90 1 10/10/19 - POVERTY REDUCTION - RES. 190239 COUNCILWOMAN BROWN: Yes. afternoon. Good afternoon to the 3 4 panelists. 5 Three things. One, I want to 6 commend my colleague Councilwoman Sanchez and President Clarke for stepping back and taking an airplane view at this 8 9 challenge that we continue to face as a city and, two, say thank you to the 10 11 partners who believe that this exercise, 12 which is more than that, is necessary if we want to be serious and strategic about 13 14 kicking this ball down the road. 15 And so with that, two things. 16 One, when I came here in 2000, one of my 17 disturbing revelations was that systems 18 were not talking to each other, and so that made remedies and the implementation 19 of remedies exceedingly difficult if DHS 20 and the School District and OESS was 2.1 22 dealing with the same family and neither 23 one of those systems knew that there were other systems involved in that child's 2.4 25 life. I think there's been great

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2	improvement.	
3	The data that you have here is	
4	timely as we look to next year when we	
5	have to wrap our arms and heads around	
6	the census. And so where and what role	
7	do you believe this data can have so that	
8	we participate in the census in a way	
9	that drives the dollars we believe we	
10	deserve as a city since what we get is	
11	purely based on those who sign up for the	
12	census?	
13	DR. GOLDSTEIN: That's a	
14	fascinating and I think super practical	
15	question. I mean, to my mind, what we do	
16	know is that if you look at data about	
17	census participation, first of all,	
18	generally, the rates have been coming	
19	down over the decades. They're not what	
20	they used to be.	
21	COUNCILWOMAN BROWN: That's	
22	alarming.	
23	DR. GOLDSTEIN: And we know	
24	also that not every population	
25	participates at the same rate, and we	

Page 92 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. know that Philadelphia is a place where 3 its various populations, although not as 4 bad as it was decades ago, we're still a 5 reasonably segregated city by race and ethnicity and income. And so what that 6 means to me is if the issue is making 7 sure that everybody gets counted, it's 8 9 going to those places where you're going to have the highest concentrations of 10 11 people who are most likely not to be 12 counted and give them the comfort and tools to be able to be counted. 13 14 Some of the innovations in the 15 census for 2020 involve things like 16 filing electronically, like using the web 17 to fill out your census form, but we also know that that's going to help some 18 populations more than others giving 19 20 people access. 2.1 COUNCILWOMAN BROWN: 22 Absolutely. 23 DR. GOLDSTEIN: So I think embedded in your question I think is the 2.4 25 very smart, practical thing, which is

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2	find those places where you have the	
3	greatest concentrations of people who	
4	have the greatest likelihood of not	
5	participating and give them the comfort	
6	and tools to participate.	
7	COUNCILWOMAN BROWN: And the	
8	one group that comes to mind immediately	
9	are those who are most vulnerable, the	
10	homeless. How and what are we going to	
11	do as a city that we capture every single	
12	individual who is in that circumstance?	
13	And so I'm of the thinking that	
14	anticipating the work required around the	
15	census count can be a part of this	
16	conversation and discussion as well.	
17	Thank you, Madam Chairwoman.	
18	COUNCILWOMAN SANCHEZ: Thank	
19	you.	
20	I think Eva wants to respond a	
21	little bit to that.	
22	MS. GLADSTEIN: I did, and then	
23	ask a further question about housing. I	
24	think it's a great question,	
25	Councilwoman. I know Census 2020, which	

Page 94 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Council helped support through funding, is looking very closely specifically at 3 vulnerable populations that we had to 4 5 target for outreach and, for example, 6 working in partnership with the Office of Homeless Services, among other organizations, to make sure there is a 8 9 good count, but this data can help feed that effort and make it even stronger. 10 11 COUNCILWOMAN BROWN: T would submit that there's a role for the School 12 District of Philadelphia and other sister 13 14 stakeholders in that process as well. 15 if we don't capture them in one step, do 16 we get them in another. 17 MS. GLADSTEIN: Absolutely. 18 The other point I was going to 19 make as we talked about housing stability and the reason I asked for the numbers 20 and you had percents of overburdened 21 homeowners and percents of overburdened 22 23 renters, and I know of the eviction work you're doing at The Reinvestment Fund. 2.4 25 think particularly for the Housing

		Page 95
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2	Subcommittee starting to look at the real	
3	numbers, because we know that evictions	
4	drive people further into poverty and	
5	there's a cascading effect. So looking	
6	at the housing solutions, just getting	
7	some of the more detail, particularly	
8	from your eviction work, I think will be	
9	very helpful.	
10	DR. GOLDSTEIN: Absolutely.	
11	If I could add just one other	
12	thing about this census. We focus on the	
13	resources, but if you think about the	
14	constitutional purposes, representation,	
15	and we've lost representation over the	
16	decades. If you go back 50, 60, 80	
17	years, we've lost representation. So	
18	it's not only the immediacy of the money,	
19	but it's all the other things that having	
20	a larger representative body has for us	
21	in Washington.	
22	COUNCILWOMAN SANCHEZ:	
23	Councilman Domb.	
24	COUNCILMAN DOMB: I have two	
25	more questions I wanted to ask. On Page	

Page 96 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. 13, you're comparing Philadelphia, 3 Chicago, Memphis, and Milwaukee, and I quess my question is, knowing that in the 4 5 last several years millennials and baby boomers wanted to live urban, because 6 7 people are coming into cities like Philadelphia and Chicago, would that have 8 9 skewed those numbers to show that growth? DR. GOLDSTEIN: Yeah, I think 10 11 that it could, and I also think that 12 there's a piece of it which is, we are more vital economies than -- well, than 13 14 certainly Milwaukee at this point. So 15 our population has begun to grow. 16 started to grow in 2006, I think was --17 2005-06 was about the trough. And so 18 that sort of growth, I think, fuels a bit of a stronger economy. I don't have 19 20 Chicago's population numbers in my head, but I think that the vitality of the 21 22 economy helps understand that. And, yes, we do draw more millennials than some of 23 the other communities. We're noted for 2.4 25 having some of the highest rates of that

Page 97 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. in the country. 3 COUNCILMAN DOMB: And the other 4 chart that I don't understand why but 5 maybe you do, Page 15, why the higher 6 paying jobs to non-residents is higher 7 than Chicago's and increasing. DR. GOLDSTEIN: That was 8 9 puzzling, right? And honestly, I think that puzzled me as well and Emily and our 10 team. We tried to understand it better. 11 I don't know if -- I mean, it's not a 12 minimum wage issue. That's not what's 13 14 driving it, and it's not the fact that 15 there are imperfections in the data in 16 terms of what gets reported, and it's not 17 a function of the fact that they're sort of economic centers of their regions. 18 19 So we don't fully understand 20 that yet, but I think we will, because if 21 that ends up becoming a policy lever for us, it's worth understanding why that 22 23 pattern exists. But it puzzled us as 2.4 well. 25 COUNCILMAN DOMB: Okay. Thank

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2	you. Thank you.	
3	Thank you, Madam Chairwoman.	
4	COUNCILWOMAN SANCHEZ: Thank	
5	you.	
6	There being no other questions,	
7	I'm going to thank our first panelists.	
8	For the purposes of logistics,	
9	we had allotted some time for minimal	
10	public testimony at this time. The	
11	people who had signed up unfortunately	
12	had to go. I didn't know if there was	
13	anybody because we advertised it that	
14	way, I wanted to make sure. The two	
15	people who had signed up had to go at	
16	4:30. Is there anyone here before I call	
17	our next panel?	
18	(No response.)	
19	COUNCILWOMAN SANCHEZ: Okay.	
20	Because we had asked folks to sign up at	
21	the table.	
22	So with that, thank you so	
23	much. Again, the Committee has access to	
24	The Reinvestment Fund through this	
25	process, and that's why we wanted to get	

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 1
         some clarity on this data and the numbers
 3
         and how we were going to measure or
         attempt to outline goals and impact that
 4
 5
         could be measurable based on the
 6
         categories. Not perfect science, but at
         least the ability for folks to see the
 7
         way we wanted to break up individual
 8
 9
         versus household.
                   So thank you very much to Ira
10
11
         and to Reinvestment Fund team and the Pew
         Charitable Trusts.
12
                   So I'm going to allow for a
13
14
         five-minute break, because I know our
15
         stenographer, and then I'm going to ask
16
         our next panelists, because we're ahead
17
         of time, which is not a bad thing to be.
18
         So I'm going to ask United Way, Benefits
         Data Trust, and Sister Mary Scullion to
19
20
         be ready, and we will reconvene at 5
         o'clock.
2.1
22
                   Thank you.
23
                    (Short recess.)
2.4
                    (Witnesses approached witness
25
         table.)
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Page 100 10/10/19 - POVERTY REDUCTION - RES. 190239 1 COUNCILWOMAN SANCHEZ: going to reconvene. Sharmain learned 3 that I know how we are in Council, 4 5 there's no such thing as five minutes. 6 MS. MATLOCK-TURNER: She is so 7 right, as usual. COUNCILWOMAN SANCHEZ: 8 T want. 9 to thank Mitch Little and Mecca Robinson for joining us. They're also members 10 11 at-large of this Committee who have 12 joined us. And if we can proceed with our 13 14 next panel, Bill Golderer and Pauline 15 Abernathy, and of course everyone knows 16 Sister Mary Scullion. This is our next 17 and final panel and then, as promised, we will allow for public testimony at 6:30. 18 19 Please note that we were 20 intentional in doing this hearing in the 21 afternoon to allow people, school, work, and when I close today and give some of 22 23 the logistics around the subcommittees, all of us have been charged to make these 2.4 25 public hearings available and accessible

Page 101 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. to all. So thank you very much. 3 So with that, Bill, you want to 4 start. 5 MR. GOLDERER: All right. 6 Councilwoman, thank you. And I couldn't 7 say enough of how honored I am to serve the mission, the newly refined mission, 8 9 of the United Way of Greater Philadelphia and Southern New Jersey, which is to 10 11 meaningfully and measurably reduce 12 intergenerational poverty in our region. And something that I think is not lost on 13 14 anybody here, but we will be talking a 15 lot about data today, which is obviously 16 very, very important, but I come to this conversation having served for 15 years 17 as the founder of an organization called 18 Broad Street Ministry. When you see 19 20 9,000 unduplicated neighbors walking through your front door on an annual 21 basis looking to achieve their dreams of 22 23 security and well-being and prosperity, we're talking about data and statistics, 2.4 25 but really what we're talking about is

Page 102 1 10/10/19 - POVERTY REDUCTION - RES. 190239 people, and they're our neighbors. And 2. that's the fierce urgency of now that we 3 need to be pursuing this with. So that's 4 5 the first comment. The second comment before I 6 launch into this is, I'm sensing that we 7 are entering a very important moment. 8 have the highest millennial retention 9 rate of any major city in the United 10 11 States. We have a renaissance that's 12 underway in Philadelphia that isn't touching everyone, and we'll be talking 13 14 about that later, but more and more 15 neighbors who are -- who have capacity, 16 who have skills, who have a desire to lean in want to see this needle move. 17 18 They don't want to carry around this 19 badge of shame that we are the poorest 20 large city. And I feel like if we could 21 energize and serve as a catalyst and galvanize and capitalize a real strategy, 22 23 I think people will come and people will invest and people will share what they 2.4 25 have so that our neighbors can experience

Page 103 10/10/19 - POVERTY REDUCTION - RES. 190239 1 their slice of the American dream. So with that said, I'm incredibly grateful to our friends at 4 5 Benefits Data Trust as well as my friend 6 of many years, Mary Scullion, as well as Ira and his team at The Reinvestment Fund for framing the conversation in data, 8 9 because when we talk about the interventions that are actually going to 10 11 help people and move people forward, if 12 we're talking about changing the narrative around poverty, we're not 13 14 talking about people who are doing 15 anything other than striving, and we need 16 to know to a better degree how hard people are working to overcome their 17 challenges. And the interventions that 18 we select, the things we capitalize, 19 20 require that we be rooted in data and 21 analyzing the interventions that we know work, but better capitalizing them in a 22 23 strategic fashion. So when the United Way has been 2.4 25 leaning on our friends -- and some of

Page 104 1 10/10/19 - POVERTY REDUCTION - RES. 190239 this is going to be duplicative, the data 2. 3 that you're going to be hearing, but 4 that's by intention. I want us all to be 5 mindful that all data is not the same, that this idea of the disaggregation of 6 the data that is in front of us helps us 7 understand not only disparities among 8 9 race and gender but also in terms of neighborhood and the proximity of -- it's 10 11 going to be very important for us to be 12 proximate to people suffering. challenges in one neighborhood are very 13 14 different from the challenges of another 15 neighborhood. And so that's why the 16 disaggregation of this data requires the 17 best science that we can get our hands around. 18 19 And so I want to be very clear, 20 the swimlane of the United Way and the data has not been to do the work 21 ourselves, but to invite partnership with 22 some of the best in the business who do 23 this, frankly, for profit. And I wonder 2.4 25 if there's ways -- we were wondering

Page 105 1 10/10/19 - POVERTY REDUCTION - RES. 190239 whether there were ways they could do it 2. 3 for good. 4 So some of this is a little bit 5 of breaking news, but we have for over a 6 year now have been leaning on the 7 analysis of Mark Zandi, who is the Chief Economist at Moody's, who has in a pro 8 9 bono fashion put five senior analysts alongside the Monitor Institute, which is 10 11 part of the Deloitte practice, which is a 12 global practice that is not rooted in Philadelphia. We were part of a 13 14 competition to see if they would lean 15 into our challenges to help us better understand this data before we start 16 17 making assumptions, and the reality is both of these interventions that are 18 19 coming in terms of research and analysis and as I've spoken with my colleagues 20 21 here, we are only halfway there. We are nowhere near answers, but what we're 22 learning is surprising, a little bit 23 troubling, but there are also a lot of 2.4 25 opportunities, and these represent two

Page 106 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. entities from the private sector that are leaning into our challenges as though it 3 were as important as meeting their bottom 4 5 line obligations. 6 So moving ahead, things we've 7 already heard today, and I'll move quickly here. We know that while 8 9 imperfect and dated, what we're talking about with poverty is a measure of 10 11 income. And so what I'll be talking 12 about here is, we are making prudent and important investments in a holistic set 13 14 of interventions already with people who 15 are experiencing poverty, but I think 16 what we might be considering today is if 17 we're talking about moving people higher 18 and higher in their income, the interventions that we are zeroing in on 19 have to do with what are the barriers 20 21 that need to be removed for people to achieve higher income, what are the 22 23 benefits that could be pulled down for people to be lifted up, and also what are 2.4 25 the new opportunities that need to be

Page 107 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. created and how do we capitalize those opportunities. 3 4 And here's some stark things 5 that are important for Council. I'm sure 6 you know this, but I don't know if you've 7 ever seen this graphic. This is the same data used by Ira Goldstein, but it's the 8 9 American -- it's sourced from the American Community Survey, which if you 10 walk around Center City, people really do 11 12 wonder, well, how could we possibly still have this high incidence of poverty when 13 14 there's all these gleaming towers and 15 beautiful restaurants, and the reality 16 is, this points to that during the 17 largest sustained economic recovery in 18 the history of our nation, and Philadelphia has benefited from that, if 19 20 you are in the top four quintiles of 21 income, meaning if you are in the top 80 percent, Philadelphia is as good as any 22 23 place in the country to live. However, the only city in the 2.4 25 United States where during the same

Page 108 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. period of economic recovery, if you are 3 in the bottom 20 percent, every other market that we looked at, people, our 4 5 neighbors, moved forward in other cities, 6 our neighbors moved backward. It's a 7 statistical anomaly that we really need to get curious about. 8 9 We've heard a lot about labor force participation. I am not an expert 10 11 on this, but this also shows that we are 12 lagging behind our peer cities in terms of attachment to the workforce. And I 13 14 feel like when we talk about measure of 15 income, what we're talking about here is 16 income supports and getting more people 17 to living wage employment. Those are really the focus, with an important 18 attachment to criminal justice, housing, 19 20 and health. These are things that we 2.1 know. This I just mentioned. 22 23 been well documented by Council, and I really appreciate your advocacy and work 2.4 25 on this. This just points out that some

Page 109 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. information as it relates to the barriers to employment for people who are coming 3 out of incarceration are steep, and we 4 5 need more than a strategy. We need a 6 legislative agenda. We need a policy 7 agenda. But really we need a private sector comprehensive agenda about how we 8 9 create opportunity for people who are coming out of prison for a second chance, 10 11 and that's at the core of some of this, 12 but just to see where we rank in terms of other peer cities in terms of the 13 14 barriers of incarceration. We heard a lot about the heat 15 16 map before, and I really want to emphasize this, because in addition to 17 18 pointing out the other principals that I've laid out before, another principal 19 20 is, this is a city of neighborhoods. 21 all know that, and one size will not fit all, and the interventions that are 22 outlined here are not intended to be a 23 blanket or some kind of ivory tower, but 2.4 25 they need local application with

Page 110 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. incredibly talented community leaders who 3 have been asked for years and years to do 4 more with less. And I'm going to be 5 suggesting something really strange, 6 which is that we give them more to do but with greater resource, because that would 7 be really -- make this idea of lifting 8 9 100,000 people out of poverty a real possibility. 10 11 So how do we respond? 12 appreciated Council President Clarke's idea of the moonshot. My parlance has 13 14 been the Marshall Plan, because the 15 reality is after the Second World War, 16 after Europe was broken, there were two 17 motivations for rebuilding Europe. One, 18 it's the right thing to do. It's the morally right thing to do, and it was in 19 our economic self-interest to do so. 20 21 it was -- the three principles that were 22 underneath the Marshall Plan was, one, an 23 aspirational vision. This problem is not 2.4 bigger than who we are. We are capable 25 of doing this and we are going to do it

Page 111 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. with resolve. Second, you don't just walk 3 over to Europe with your shovel in your 4 5 garage and start digging away. You need 6 an actual plan. You need to actually 7 think through what goes first, second, and third, and what are we going to need 8 9 to do if we want to be successful. And then, third, this is the 10 11 controversial part. This is going to 12 cost more than anybody wants it to cost. We are going to have to invest our way 13 14 out of this. It's just that simple. 15 Council is going to be part of that, I 16 hope, but I also believe that the private 17 sector is waiting. The philanthropists 18 of the community are waiting. Normal neighbors who feel fortunate want to get 19 20 engaged, but we need to first set the 21 table for something that is measurable 22 and meaningful in terms of an engagement 23 here. So here are a few things. 2.4 25 promised you a couple of ideas. And I'm

Page 112 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. trying to watch my time. The first has 3 to do with, unsurprising, and we've heard about this, the imperative for job 4 5 growth. But when we talk about job 6 growth, many people think about, well, 7 who are people who are currently in poverty and where is their best 8 9 employment opportunity? According to the data we have, 52 percent of employment in 10 11 our community comes from small business 12 enterprise and entrepreneurship, and of that 52 percent, 78 percent of people in 13 14 poverty attach to small business. So it 15 seems logical that one of the 16 interventions we need to meaningfully and measurably engage in is small business 17 18 development as well as entrepreneurship. 19 The data I want to make sure I 20 get correct. I think I have it, is that 21 when we talk about owned enterprise, we have a very robust small business 22 23 ecosystem, and yet in a majority minority city, African American and Latinos 2.4 25 represent more than 50 percent of our

Page 113 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. population and yet they are only 3 3 percent of business owners. statistical anomaly when you benchmark 4 5 against other cities. So United Way is proposing, along with a host of other investors, the 7 idea that like Detroit has done before us 8 9 and Oakland and the Bronx, a robust ecosystem from idea to open be created 10 11 for entrepreneurs of color. And that 12 would not be led by me, but it would be led by the immediate past CEO of the 13 14 African American Chamber of Commerce, 15 Mike Banks, who has left that post to 16 join United Way to drive and inspire that 17 effort. Second, barriers to entry. 18 These are surgical interventions that are 19 20 customized for the problem we're trying 21 to solve. When we're talking about removing barriers to labor force 22 23 participation, we're talking about the majority we heard earlier about 2.4 25 home-based childcare, which is huge in

Page 114 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. our region and underinvested in in terms of ensuring that those are quality and 3 safe, not just for the nurturing of the 4 5 child, which is a long-term labor 6 participation strategy, but for people 7 and parents who are trying to get to work and keep a job. If they do not have 8 9 stable childcare, their chances of retaining that job is very low. 10 11 Working with a comprehensive 12 plan hopefully with the Chamber of Commerce to meaningfully engage HR 13 14 functions to pivot away from a talent 15 strategy that is really currently a 16 majority place of risk mitigation instead 17 of courageous approaches to 18 non-traditional candidates who are really looking at opportunity, who are capable 19 and qualified, but right now are ruled 20 21 out and not able to participate in a meaningful way, and then diversion, 22 23 expundement, and preventative healthcare, as well as stable housing. 2.4 These are 25 things we know about it.

Page 115 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. The last has to do with -- I 3 know Councilman Domb might give me a standing O for this one, but the reality 4 5 of the matter is, he quoted 450 million. 6 That's correct. I have 130 million just in Philadelphia County alone, that if we 7 were better able to capitalize 8 9 high-performing, proven, and tested organizations like Benefits Data Trust, 10 11 Campaign for Working Families and look at this holistically, there are so many 12 approaches to Philadelphia's poverty 13 14 problem that are noble and also -- but 15 they're a bit boutique in their 16 application. We need to be talking 17 scale, we need to be talking long term, 18 and we need to be putting out milestones 19 and markers in order to move this stubborn needle. 20 2.1 So what is it going to require? Massive -- and I must say that three 22 The time for small ideas is over. 23 times. We need big ideas with big commitment, 2.4 25 multi-year, citywide campaign to reduce

Page 116 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. poverty. This will not sit at United 3 Way. We are putting all of our resources 4 behind an idea where it could be possibly 5 powered in part by United Way, but I 6 think everyone needs to see themselves 7 inside these solutions. That's, I think, important. 8 9 Second, unprecedented public and private investment to grow the 10 11 capacity of key organizations that have a 12 proven track record that they could, if we challenge them to scale, they could do 13 14 it. They could meet the demands of more 15 capital and more urgency around moving 16 people over the poverty line and even further. 17 18 The reality, though, this is, 19 as we all know -- some of you know I'm an ordained minister, and a minor prophet in 20 21 our times was Hank Williams, Jr., who once said romance without finance don't 22 23 stand a chance. And you cannot move this needle unless we find more capital, 2.4 25 courage capital, committed capital,

Page 117 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. capital that is not risk averse. 3 And measurement and accountability. I think we all know part 4 5 of what contributes to the Philadelphia 6 shrug is that there's bold proclamations, bold ideas that are trotted out there, and then we have a big event and then we 8 9 leave and go do the next thing. I really think measurement and accountability will 10 11 reduce the cynicism of people who want to see this needle move, but they want to 12 know we're serious. 13 14 The thing that I would point to 15 is, there are other major markets that 16 are -- we have an opportunity to learn 17 from them. I think a lot of people in 18 Philadelphia know we're special, and I know that. Born here, raised here, I'm 19 from here. But I am curious about how 20 other markets have moved the 2.1 constellation of the private sector, 22 23 traditional philanthropy, high-performing community organizations, and public 2.4 25 policy marching in concert forward toward

Page 118 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. a common goal. This has happened in other markets to great effect. 3 So the mandate that I bring to 4 5 you today is this idea that there needs 6 to be an articulation, and you're getting 7 there, measurable goals, incremental goals that need to be drilled down that 8 9 would move our region forward. interventions that we can put against 10 11 those goals -- and you'll hear -- I've shared some. You'll hear others in a 12 minute. The stakeholder alignment is 13 14 crucial, and that's above my pay grade, 15 but at some level, we're going to have to 16 overcome the MET if we're ever going to 17 get there. I'm sure you know the MET, 18 money, eqo, and turf. That's our biggest challenge here. And new transformational 19 investment off the sidelines. 20 2.1 Considerations for Council, and I'll leave it here. I appreciate the 22 23 leadership you've shown in bringing this conversation to the forefront and putting 2.4 25 some numbers to it, and I think that the

Page 119 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. platform you have and the leverage that 3 you produce can create this fierce sense of urgency, and bringing proven leaders 4 5 like Eva along with Sharmain around to 6 articulate a common agenda I think is 7 crucial. Capitalization, there are many 8 9 models of public, private models where we agree that there needs to be an 10 11 allocation and a strategy for deployment 12 of capital in a way that makes sense, and we need to agree upon measurement, what 13 14 is it that we think is worth measuring, 15 because what gets measured gets done. 16 Thank you. 17 COUNCILWOMAN SANCHEZ: Before 18 we move over to Pauline, I had a couple of questions. I didn't want to lose the 19 flow around this. 20 I want to drill down a little 2.1 22 bit around some of the business, the 23 commerce piece that you've talked about. We've had competing goals departmentally, 2.4 25 right? And I think as government,

Page 120 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. sometimes the bureaucracy, people are spinning, I always say, the different 3 portfolios that people have. On the one 4 5 hand, in every index that we use about 6 what is a choice neighborhood, right, or a neighborhood that has all those amenities, transportation, walkability, 8 9 all those things, we talk about mixed-income, commercial, residential. 10 11 And so part of our zoning strategy and 12 our development strategy has been about pushing for this commercial development, 13 14 right, that is empty. 15 I represent ten El stops. 16 MR. GOLDERER: Yes. 17 COUNCILWOMAN SANCHEZ: Right? 18 Ouite a challenge. And this number around the lack of capital for 19 20 entrepreneurship among communities of 2.1 color. 22 Can you give me one or two 23 things that in that world -- and we talked about investment, but what are two 2.4 25 or three things that quickly rise to you

Page 121 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. about what we should be doing around I know we have the Power Up program with Community College, and we've 4 5 done a lot of incremental things, but if 6 you could do one or two big-ticket items 7 other than just investing, what do you think? 8 9 MR. GOLDERER: One is more atmospheric and then the other one is 10 11 concrete. The atmospheric one is, I've 12 been trying to quietly bring people on field trips to look at how other markets 13 14 have done this, and what you hear from this market is, oh, we're already doing 15 16 that. And so I'm going to give you a 17 very concrete example. In Detroit, since the 18 19 Entrepreneurs of Color fund has been 20 launched, which is about five years ago, 21 18,000 living wage jobs have been created and sustained, with the average credit 22 23 rating of the applicant for those funds is 420. 2.4 25 COUNCILWOMAN SANCHEZ: WOW.

Page 122 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. How much money are we talking about? 3 MR. GOLDERER: So the total 4 value of the fund is \$30 million. 5 it's not -- again, romance without finance don't stand a chance. Remember 6 7 that? And the second part is, lending 8 9 institutions do not lend capital to lose money. So how could philanthropy and 10 others who have gone on an 11 12 entrepreneurial journey and had success -- and there are many of those in 13 14 our region -- how could we convince them 15 to create an opportunity for there to be 16 courage capital that's deployed in three 17 ways. More robust investment in our 18 technical assistance. We have great providers here, but they're doing it on a 19 20 shoestring. 2.1 Second, backstop some of these loans. Like if you want to see different 22 23 results, you need to give some of the people who are lending money some stomach 2.4 25 comfort. In their desire to do the right

Page 123 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. thing, you want to make sure that they're encouraged to do the right thing. 3 4 And the third part is really 5 doing a little bit more with our 6 entrepreneurs or would-be entrepreneurs 7 to deliberate -- you can have a great business plan, but do you have the 8 9 passion test, do you have the ability to see it through. Because that started a 10 bunch of things, and it gets hard and 11 12 there are dark days, and we need to really do more diligence on -- everyone 13 14 has ideas, but who is going to execute. 15 So those would be the three 16 things. 17 COUNCILWOMAN SANCHEZ: So one 18 of the things we're hearing from some of the small businesses is, they're getting 19 20 priced out, particularly if you don't own 21 your facility. In the models that you looked at or that you're beginning to 22 23 look at, did any of them -- and I know when Eva was overseeing the American 2.4 25 Street Empowerment Zone, one of the

Page 124 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. toolboxes was one-year subsidy, rental 3 subsidy. Did you see any other programs similar that you think Philadelphia 4 5 should be considering? I think the 6 Department of Commerce identified 100-and-something commercial corridors of 7 some sort in the City. Did you --8 9 MR. GOLDERER: There's an incredible hub concept. I know I'm 10 11 talking a lot about Detroit, but I'm 12 enamored of what they've done, and this was partly from people acting differently 13 14 and behaving differently. 15 So just to give you an example, 16 for 150 years the Kellogg Family 17 Foundation only funded kids, safe, unobjectionable, uncontroversial kids. 18 And then one day they woke up and read 19 20 the paper and it said Detroit is 21 bankrupt. And so then they thought, well, what should we do differently, and 22 23 one of the things they backed, in addition to this Entrepreneurs of Color 2.4 25 idea, was basically a comprehensive land

Page 125 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. banking system that matched people who were looking to enter a commercial 3 corridor and also people who were aging 4 5 out, who were retiring and looking to 6 move for whatever reason, and gave first 7 shot to people who were younger who were trying to establish themselves at that. 8 9 And so that was a deep, deep partnership between their Chamber or the equivalent, 10 11 their City Council, and also 12 philanthropy. And we don't have those kinds of things here, but I don't think 13 14 it's impossible. 15 COUNCILWOMAN SANCHEZ: So \$30 16 million, five years, 18,000 living wage 17 jobs. 18 MR. GOLDERER: It's very 19 different. And we need to continue, I 20 must say. My friends here, from Pat 21 Clancy, I'm not suggesting that we move 22 away from workforce development, or Sheila, don't stab me, if you're still 23 here. I believe in that, but I feel like 2.4 25 this is about additive, what are we going

Page 126 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. to do on top of. It's not about rearranging the chairs. It's about 3 bringing new chairs. 4 5 COUNCILWOMAN SANCHEZ: 6 We've made it clear this is not about either/or. It's an "and" and an "and" 7 and an "and." So I appreciate that. 8 9 Sharmain. MS. MATLOCK-TURNER: 10 Yes. 11 Bill, thank you very much for giving us, 12 I think, something that we can sort of really think about from a macro process, 13 14 but also this sort of idea and some bold and different ideas and how also we learn 15 16 from others. 17 One of the things that I liked 18 about the City Council report was with the suggestions that they made, they 19 20 actually did look at other cities and say this has been tried here, it's been tried 21 22 here. But I want to jump to the 23 measurement side, because when we talk about trying to make sure that we're 2.4 25 moving 100,000 people out of poverty --

Page 127 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. and I love your finance and romance and we need the money. Money, money, money. 3 How can you with what you're doing really 4 5 help the broader community, the City 6 government, the private sector, 7 individuals, when we come back as a group and say in order to move 100,000 people, 8 9 we need X millions and millions of dollars over a five-year, 25-year span? 10 11 How do we build that case and put the 12 kind of structure in place that people can see the incremental change, but know 13 14 that it's the total investment that 15 ultimately is going to drive that change 16 as opposed to these are some really great 17 ideas and you guys can do it with what 18 you already have? 19 MR. GOLDERER: Thank you, 20 Sharmain, for the question. There's two 21 components to that. The first is, I think a very significant -- and I have a 22 23 donor in mind for this -- a significant investment needs to be made in knowledge 2.4 25 management, a dashboard that everybody

Page 128 1 10/10/19 - POVERTY REDUCTION - RES. 190239 can understand, that is about are we making progress whenever we launch, 3 whenever we move out to market and say 4 5 this is what we're trying to do and by 6 when, this is the knowledge management. And then affixed to that answer is a price tag. 8 9 So I believe that what we should be talking about in terms of these 10 11 ideas that are generated, there needs to 12 be an intervention map or a solution map where you see if this happens, if you are 13 14 able to assemble five years, 25 million, 15 and you are interested in this 16 neighborhood because of this 17 hyper-customized intervention, the price 18 tag for that is this. And then somebody would say -- I mean, I don't know if you 19 do this, but I've had zero success in 20 21 Philadelphia restaurants negotiating the price of an entree. I'd like that, but I 22 I'd like it for \$12. No. 23 It costs what it costs and we have to be firm on that. 2.4 25 You don't have to do that, but

Page 129 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. I'll do it, because for years -- and Eva 3 knows this -- that's really -- people who are delivering for our communities don't 4 5 feel like they can stand up and say 6 actually you're trying to make me do a \$30 job with \$4, and because I love the people, I'm going to do it, but the 8 9 reality is, it's high time there be some real conversation about are we 10 11 appropriately capitalizing interventions that we know work, they are proven in 12 this market and elsewhere. But we are 13 14 not -- but also to be fair, if you're 15 going to ask an investor to be 16 sacrificial in their investment, you owe 17 a higher rate of return in terms of proving it. Measurement, knowledge 18 information, and accountability. 19 20 MS. MATLOCK-TURNER: And, Bill, 21 just let me follow up on that, because it seems like to me married to that 22 measurement is the idea about how is the 23 City going to look better for, 2.4 25 quote/unquote, the investors.

Page 130 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. measuring certainly the effect of the 3 change that we're looking at, but are you also going to have a metric and be able 4 5 to communicate that metric effectively that it's costing us less to do this, 6 7 that, or the other because we've made this intervention or taxes are different 8 9 because of this intervention. What's your sense about how you sort of marry 10 11 the benefit as well as the change? MR. GOLDERER: So the benefit 12 of the engagement with Moody's and 13 14 Deloitte is -- I mean, I have 15 limitations, as you well know. I'm more 16 of a moralist. Okay? But they're economists, and they're benchmarking us 17 18 against other cities, and the playbook that we're playing out of is, I was 19 amazed at how powerfully people who I 20 know hate each other convened around the 2.1 prospect of Amazon, right? Because they 22 23 could imagine what that would do for the economy. What you're suggesting with 2.4 25 100,000 people moving out of poverty, it

Page 131 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. certainly -- there is no tolerable amount 3 of poverty in our city, but what we're 4 looking for is progress, and what I've 5 asked the analysts that are working on this project with us is to forecast two 6 7 scenarios. One, the aspirational scenario. How would our lives be 8 9 different, how would the lives of the individuals be different, how would all 10 the systems be different, how would our 11 taxation structure be different. 12 If we achieved this, what would that look like. 13 14 And then also how is the Philadelphia renaissance going to look if we don't. 15 16 What's the cost of that, not just to 17 individual lives but the viability of the 18 City where we are -- I don't think 19 anybody here was celebrating a move of 20 25.7 to 24.5, but really we are still a 21 huge delta away from the top ten cities in terms of poverty. We are still 87.5 22 23 thousand people behind. We're the New York Knicks. We're behind, and we need 2.4 25 to get serious.

Page 132 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. COUNCILWOMAN SANCHEZ: MS. GLADSTEIN: I so hope you 3 are successful. We all so hope you are 4 5 successful. 6 I think there are some truisms, if you will, that we all think about our 7 city and our city as we think about this 8 9 issue. One is that we have sort of a low corporate participation rate, because we 10 11 don't have headquarters located here. Second is, we don't have as 12 robust a philanthropic and foundation 13 14 community as other cities. And for 15 years, we were not on the growth curve, 16 although we are now. 17 So if you could talk a little 18 bit about like as you've been doing this work, what makes you think some of those 19 20 first two factors might change, and then 21 a little bit more about the specific role that United Way would play in doing this. 22 23 Because from my perception, you're looking at significantly altering the 2.4 25 role of United Way, which I think is --

Page 133 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. ultimately I hope is a great thing, but as you know, those organizations that 3 have been doing \$30 of work for \$4 are 4 5 going to view that as a real thread and 6 have a concern. COUNCILWOMAN SANCHEZ: Explain your MET matrix to us. 8 9 MR. GOLDERER: Sure. I'm going 10 to go in reverse order. I give the Board a lot of credit of United Way. 11 In 2021, 12 United Way will celebrate 100 years in this community. And I'm stewarding a 13 14 mission that will live well beyond me and 15 was going on well before me, and I give 16 the Board credit to recognize that the disparity in our city is not something 17 that they were willing to tolerate. And 18 so the question that I had to ask them 19 20 was, okay, you've adopted a new purpose, 21 but you're built the way you've always 22 been built. So are you prepared to 23 transform yourself into something -- if 2.4 you're trying to address this urgent 25 need, then we have to change, and it's

Page 134 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. not incremental. And some of you, if you 3 follow this stuff, the change over the last 18 months -- and Sharmain is a Board 4 5 member -- it has not been incremental. 6 So we are in service -- for me, 7 the mission is always first. institution comes second. 8 9 Second, I don't think there's any -- I've been doing, when I can't 10 11 sleep, just a brief study of -- this is a 12 trope in our -- and I don't disagree with you, but I would say that there are two 13 14 things that are about behavior change, 15 but I don't think there's scarcity here. 16 My closet, my quiet hobby has been 17 studying the valuation of companies in 18 this region over the last ten years, and philanthropy is flat or down and 19 valuation is up. 20 2.1 United Way is invited into 400 companies throughout the region. We're 22 23 really -- that's really one of our own -that is a value to be able to enter those 2.4 companies at the invitation of the 25

Page 135 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. corporation, and we have an invitation to 3 ask if we were to assemble something real, would you invest something big, and 4 5 I feel like there's a lot of folks over 6 the last 10 to 12 years whose personal 7 income has grown seismically, and to ask them to be involved in something small 8 9 would not be the right thing, but to invite them into something big, that 10 seems to be the strategy that we're 11 12 trying to go after. Now, the reality is, everybody 13 14 is inviting everybody to care about 15 everything, and the reality is, you can 16 do anything, but you can't do everything. So the reality is, there is going to be 17 some struggle here, because that's just 18 real, but if we're really serious about 19 20 putting the neighbors first, then we need 21 to arrive on that as the principal, and we're going to try and capitalize 22 23 strategies that move people forward, and that could be -- I mean, we all know 2.4 25 Philadelphia -- social change in

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2	Philadelphia is a contact sport. I don't	
3	need to tell you that.	
4	COUNCILWOMAN SANCHEZ: Any	
5	other questions of Bill?	
6	(No response.)	
7	COUNCILWOMAN SANCHEZ: Thank	
8	you, Bill. I'm going to ask you to stay	
9	at the podium. I just wanted to make	
10	sure we kept on the flow.	
11	Pauline.	
12	COUNCILWOMAN BROWN: Madam	
13	Chair?	
14	COUNCILWOMAN SANCHEZ:	
15	Councilwoman Blondell Reynolds Brown.	
16	COUNCILWOMAN BROWN: Could I	
17	please hear an elevator snapshot of your	
18	vision of how the work of your small	
19	business entrepreneurial expert in back	
20	of you will help and be a part of this	
21	equation that you're talking about?	
22	MR. GOLDERER: Can you say that	
23	one more time?	
24	COUNCILWOMAN BROWN: You've	
25	mentioned that you brought on the former	

Page 137 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Executive Director, Mr. Banks, of the 3 African American Chamber. So in a nutshell, can you simply share with us 4 your vision on how his work is going to 5 6 fold in and be a part of this equation? MR. GOLDERER: I think one of the things that we cannot discount -- and 8 9 I don't want to brag on Mike, but Mike represents this rare combination -- well, 10 11 it's not rare, unfortunately, but of lived experience, of understanding the 12 anatomy of business, of how -- of where 13 14 and what the gaps actually are and what 15 they look like. And so when we're going 16 to try and bridge the gap between 17 venture, courage, capital to lift something like this off, you need someone 18 who can really drill down on where the 19 20 exact barriers to this job growth and upscaling and moving people forward are 21 and what new capital could yield us in 22 23 terms of new results. So you need -- I mean, part of this is credibility, like 2.4 25 is this a credible thing you're telling

Page 138 10/10/19 - POVERTY REDUCTION - RES. 190239 1 And I think we have reason to be suspicious, because there's been a lot of 3 big ideas, but we're trying to lead with 4 5 the lived experience of overcoming 6 challenges and demonstrating what capital could do. 7 COUNCILWOMAN BROWN: Well, he 8 9 certainly has his own testimony that's a wonderful story, so there should be no 10 11 question about credibility, and it is a 12 novel, innovative approach that Councilman Domb talks about all the time. 13 14 So to know that now we have an entity in 15 place that's going to put legs on the 16 idea is encouraging. 17 Thank you. Thank you, Madam Chairwoman. 18 19 COUNCILWOMAN SANCHEZ: Let me add the boldness around the investment. 20 21 I went to a pension training by minority pension investors several years ago for 22 the summer. Yeah, that's what we do in 23 the summer. And one of the things that I 2.4 25 appreciated was that the minority

Page 139 1 10/10/19 - POVERTY REDUCTION - RES. 190239 investors were really trying to encourage 2. 3 pension funds to do some of the social 4 impact investment, and some of the 5 examples that they gave were when in fact 6 they were flexible in their criteria, when they created product lines for particular constituencies, including 8 9 undocumenteds, their returns were incredibly high and their defaults were 10 11 almost zero. And you sit through that 12 and you say, how can we as government be even more creative around our pension. 13 14 Because that is, to your point, you're 15 saying I have access to 400 corporations, 16 right? Imagine if you went and said 10 17 percent of your pension is going to be 18 invested in social impact within your 19 footprint, right, where you make money, 20 your customer base. I mean, what does 21 that translate to? But, you know, where people have been bold in that kind of 22 23 investment, the return is there, right, let alone the human investment and the 2.4 25 capital.

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2	We got to get away from that	
3	Philly shrug and really do that.	
4	Councilman Domb.	
5	COUNCILMAN DOMB: Thank you,	
б	and thanks for your presentation. It was	
7	very good. I just had a couple questions	
8	and comments I wanted to make.	
9	We talk about entrepreneurs,	
10	and I know you touched on this, that 78.8	
11	percent of the entrepreneurs in	
12	Philadelphia are white and 10 percent are	
13	Asian and 2.4 African American, which is	
14	a disgrace to the City with a population	
15	of 45 percent.	
16	This Council back in February,	
17	we put forth a resolution, and I'm going	
18	to ask United Way if you can, with all	
19	the companies involved, if they could get	
20	behind this and get our legislators in	
21	Harrisburg to move it forward. I'll tell	
22	you what the legislation is basically,	
23	but I want you think in this context	
24	before I mention it. Twenty years ago if	
25	you could name the top five companies in	

Page 141 1 10/10/19 - POVERTY REDUCTION - RES. 190239 the United States, they probably would 2. have been maybe Exxon, maybe AT&T, maybe 3 General Motors, maybe Ford. And today 4 5 who are the top five companies? Probably 6 Amazon, probably Google, probably Apple. I don't know what else is out there, but 7 that's the concept, the tech. 8 9 So the resolution put forth by every member of this Council was to ask 10 11 the State of Pennsylvania -- after I met with Governor Wolf, he said do this, 12 which we did. We asked for four things. 13 14 One, teach financial literacy pre-K to 15 12th grade. So four-year-olds up to 12th 16 grade. Teach in our schools coding and 17 technology pre-K to 12th grade. could tell you right now most 18 four-year-olds are probably better in 19 20 tech than I am. They're really good. 21 And then number three, teach entrepreneurship, teach it in the 22 23 schools, 7th, 8th, 9th, 10th, 11th, and 12th grade. This was all in the 2.4 25 resolution. And then the last piece was

Page 142 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. a page out of the Cristo Rey High School. 3 Allow students the option of working one day a week in 9th, 10th, 11th, and 12th 4 5 grade, get credit for the one day, and go 6 to school for the four days. But those four different job experiences would be 7 invaluable to them. 8 9 I mean, if we were trying to change poverty in the big picture -- and 10 11 so we sent this up to Harrisburg. Every 12 member here supported it. We haven't gotten anywhere yet. If we could enlist 13 14 United Way or your companies to call the 15 Governor, call the legislators and say, 16 hey, we need to change -- this is not, by 17 the way, this probably should be changed across the country, not just 18 Philadelphia, but we need to change what 19 20 we're doing so we can prepare our 21 students for not just today's economy but 22 tomorrow's economy. This is going to be 23 a very different economy tomorrow than it 2.4 is today. So whatever United Way could 25 do to help.

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2	And this entrepreneurship	
3	thing, teaching it in school would be a	
4	great thing, because I think you'd	
5	increase, as we saw from earlier, the	
6	high school graduation rates. I know	
7	myself, if I'm interested in a topic, I'm	
8	going to pay attention, and if I'm not	
9	interested, you lost me. So I think if	
10	we can engage the students, it would be a	
11	really positive thing.	
12	Anyway, thank you for your	
13	presentation. It was very good. Thank	
14	you.	
15	MR. GOLDERER: Thank you,	
16	Councilman.	
17	COUNCILWOMAN SANCHEZ: Sorry,	
18	Pauline.	
19	MS. ABERNATHY: Thank you,	
20	Council President Clarke, Councilmembers	
21	Quinones-Sanchez, Reynolds Brown, and	
22	Domb, Special Committee Co-Chairs, and	
23	everyone else who is here today to share	
24	their thoughts and insights about how to	
25	reduce poverty in Philadelphia.	

Page 144 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. My name is Pauline Abernathy. I'm Chief Strategy Officer at Benefits 4 Data Trust. BDT is a national non-profit 5 organization headquartered here in 6 Philadelphia with a mission to help people live healthier, more independent lives by creating smarter ways to access 8 9 essential benefits and services. year BDT helps tens of thousands of 10 11 people make ends meet using a mix of data, technology, targeted outreach, and 12 policy change. 13 14 Since 2005, BDT has screened more than one million households 15 16 nationwide, securing over \$7 billion in cumulative benefits and services. 17 Philadelphia alone, we have helped more 18 than 110,000 Philadelphians secure more 19 than 330 million in benefits to help pay 20 for food, healthcare, housing, childcare, 21 and other basic needs. 22 23 We are proud of the fact that many of our employees exemplify how 2.4 25 benefits can help people through rough

Page 145 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. times and on a path to a better life. I'll share the story of one staff member 3 who I'll call Maria. 2011 was a dark 4 5 year for her. As Maria puts it, she had 6 three children, a pack of hotdogs in the 7 freezer, and no money in her bank account, despite having worked since age 8 9 She literally didn't know how she was going to feed her kids. She tried 10 11 applying for SNAP and found the 12 experience utterly demoralizing and, in fact, demeaning. I'm thrilled to say 13 14 that today Maria is a Benefits Outreach 15 Specialist helping others in a dignified 16 way on their own journeys, and her 17 children are in college. 18 Last year, Mayor Kenney came to 19 BDT's office across the street from here 20 to mark the 10th anniversary of 21 BenePhilly, a citywide initiative to combat poverty and increase economic 22 23 security. BenePhilly started as a partnership with the City and state to 2.4 25 help low-income seniors apply over the

Page 146 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. phone for benefits to help them pay for 3 groceries, housing, and healthcare. 2014, with the support from the Mayor's 4 5 Office of Community Empowerment and 6 Opportunity, BenePhilly expanded to include in-person assistance to people of all ages through a network of community 8 9 partners using our software. Today, low-income people of all 10 ages can walk into BenePhilly Centers run 11 12 by organizations they know and trust and be screened for up to 19 benefits and 13 14 receive help applying, while being treated with dignity. I want to take a 15 16 moment to commend our outstanding 17 partners, including Catholic Social 18 Services, Esperanza, Impact Services Corporation, Penn Asian Senior Services, 19 20 Philadelphia FIGHT, Project HOME, SEAMAAC, UESF, and United Communities 21 Southeast Philadelphia. 22 23 Through BDT and our partners, last year alone 5,200 Philadelphia 2.4 25 households received an average of \$5,100

Page 147 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. in critical benefits, which translates 3 into \$26 million in benefits for the City 4 as a whole. 5 The return on investment is 6 compelling. A recently published 7 randomized experiment conducted by economists at MIT's Poverty Lab with 8 9 Benefits Data Trust found our targeted outreach tripled SNAP enrollment and 10 11 generated \$20 in benefits for every \$1 12 invested. While there is no single 13 14 solution to poverty, benefits provide a 15 bridge to economic mobility for low-income families and adults. 16 It is difficult to focus on getting a job if 17 18 you don't know how you're going to feed 19 your children or where they're going to 20 sleep tonight. So it's not surprising 21 that rigorous research shows that benefits improve health, employment, 22 earnings, and educational achievement. 23 Yet, despite the work of 2.4 25 BenePhilly, thousands of struggling

Page 148 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Philadelphians are still not receiving the benefits for which they are eligible. 3 As Councilman Domb said at the beginning 4 5 of this hearing, we estimate at BDT that 6 Philadelphians are collectively leaving 7 \$450 million in untapped federal and state benefits on the table every year. 8 9 To put this number in perspective, \$450 million is more than the City spends 10 11 annually on public and behavioral health, human services, parks and recreation, and 12 SEPTA combined. 13 14 Oh, that's hard to read also. What this shows is even --15 16 there are over 60,000 Philadelphians who 17 are eligible to participate in SNAP who are not participating, which is nearly 18 enough to fill the entire Eagles stadium. 19 20 That's how many are eligible and not 21 receiving it. And this estimate of 450 22 23 million takes into account only five programs for which we could make reliable 2.4 25 estimates, EITC, SNAP, Medicaid, PACE and

Page 149 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. PACENET prescriptions assistance, and the 3 Commonwealth's property tax and rent rebate program. Adding in other critical 4 5 benefits such as WIC, Medicare Savings programs and the Child Tax Credit would 6 increase the number further. 7 There are several reasons why 8 9 eligible people don't apply for benefits. Lack of information is one of them. 10 11 simply don't know that assistance is 12 available or that they qualify, which makes sense if you think of what Pew's 13 14 poll showed, that half of people living 15 in poverty don't think of themselves as 16 poor, so in that case, you might not 17 think you're eligible for assistance. 18 Others know they're eligible, but need help applying, and the process 19 can be complicated and each benefit can 20 21 require a separate application. And as anyone in this room who has ever 22 struggled to fill out their own taxes 23 knows, the language on government forms 2.4 25 can be confusing. It is doubly so for

Page 150 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. individuals with low literacy or for whom English is not their first language. 3 4 While these and other barriers 5 are real, they are entirely surmountable 6 and addressing them should be part of any 7 comprehensive strategy to reduce poverty and increase economic mobility. 8 9 Narrowing the Gap suggested, Philadelphia can dramatically improve benefits access 10 11 by leveraging existing access points, technology, and data sources. 12 people cannot be automatically enrolled 13 14 in a benefit, we should alert them to 15 their eligibility and make it easy to 16 apply, either in person, on the phone, or 17 online. 18 A few years ago there was basically only one way to do banking. 19 You went into a bank and saw a teller. 20 21 Today, with the proliferation of mobile phones and apps, the financial world has 22 23 changed. Banks quickly changed and realized that there was no 2.4 25 one-size-fits-all way of serving their

Page 151 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. customers. They came up with new ways of doing business using technology. 3 4 Imagine if technology serving 5 families in economic distress were up to 6 par with that technology serving financial services. Imagine if our neighbors in need could get help without 8 9 having to ask for it. This Committee and the City of Philadelphia can do that. 10 11 offer two main recommendations for doing 12 so as part of a broader strategy. First, leverage existing access 13 14 points across Philadelphia. One of the 15 biggest opportunities to achieve impact 16 at scale is by leveraging existing infrastructure, institutions, and 17 18 community organizations. For instance, our city's libraries and schools create 19 20 opportunities for people to be screened 21 and apply for benefits at places they already know and visit. Similarly, the 22 network of CareerLinks and training 23 programs operated by Philadelphia Works 2.4 25 offer opportunities to screen and assist

Page 152 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. jobseekers in applying for benefits. 3 City's RISE programs are another example 4 of an access point that can be leveraged 5 to assist returning citizens who were 6 previously incarcerated. Technology-driven tools have made it possible for people to begin the 8 9 process of applying for benefits from just about anywhere. For example, with 10 support from the United Way, we worked 11 12 with the Campaign for Working Families during the most recent tax season to 13 14 pilot a new tool called Benefits Launch. 15 People waiting for free tax assistance in the VITA centers could use Benefits 16 17 Launch to see if they were likely eligible for up to 19 benefits. The 18 screening takes ten minutes or less, can 19 be completed on a mobile phone or a 20 tablet. It identifies the benefits 21 22 people are likely eligible for and provides options to apply in person, over 23 the phone, or online. BDT can then 2.4 25 follow up with text messages to help

Page 153 1 10/10/19 - POVERTY REDUCTION - RES. 190239 people follow through on the path they 2. 3 select. Because Benefits Launch is adaptable, it could even be used in a 4 5 canvassing effort in high-need 6 neighborhoods. Second, leverage existing data for targeted outreach and streamlined 8 9 enrollment. BDT has been able to scale its work in six states using a model that 10 11 combines data-matching technology and targeted outreach. For example, by 12 cross-matching Medicaid and SNAP 13 14 enrollment lists, one can identify people 15 who are unenrolled and likely eligible 16 for SNAP by virtue of the fact that they 17 receive Medicaid. We then conduct targeted outreach to the identified 18 people, in partnership with trusted state 19 or local entities, to let people know 20 21 they are likely eligible and offer 22 assistance applying. 23 In New York City, with support from the Robin Hood Foundation, we have 2.4 25 worked with the City Department of Social

Page 154 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. Services to conduct proactive outreach to those on Medicaid but not SNAP and use 4 verified information from their Medicaid 5 applications to greatly streamline the SNAP enrollment process. 6 Many Philadelphia agencies have data that can be used to conduct targeted 8 9 outreach to people who are likely eligible for help paying for essentials 10 11 like groceries, healthcare, and housing. 12 For example, the Philadelphia Water Department and Gas Works offer discounts 13 14 to low-income families who may qualify for other federal, state, or local 15 16 assistance. Similarly, some Medicaid 17 managed care organizations that operate 18 in Philadelphia are interested in strategies focused on the social 19 determinants of health, including 20 2.1 promoting good nutrition for their members, which has been shown to reduce 22 23 healthcare costs and improve health. Data matching can identify their members 2.4 enrolled in Medicaid but not in SNAP and 25

Page 155 10/10/19 - POVERTY REDUCTION - RES. 190239 1 provide the basis for targeted outreach. 2. 3 The Health Department may have similar 4 opportunities to increase SNAP enrollment 5 among people who use the City health 6 centers. Finally, all families in public housing have limited incomes, and it may 8 9 be possible to work with the Philadelphia Housing Authority on a data match and 10 11 outreach campaign. 12 By leveraging existing access points, technology, and data sources, the 13 14 City can literally secure tens of millions of dollars of federal and state 15 16 benefits for Philadelphians, benefits that have been proven to improve student 17 18 test scores, graduation rates, earnings, employment, and health, create jobs and 19 20 stimulate the economy. We know how to do 21 it. We simply need the will to do it. When combined with other proven and 22 innovative policies and services 23 encompassing housing, jobs, and 2.4 25 education, more Philadelphians will be

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2	able to look forward to a brighter	
3	future.	
4	Thank you.	
5	COUNCILWOMAN SANCHEZ: Thank	
6	you, Pauline, and I know that you're	
7	going to be working very closely with Eva	
8	and the Safety Net.	
9	What, if any, challenges did	
10	you face in other cities around the data	
11	matching?	
12	MS. ABERNATHY: It is always a	
13	challenge, but we've now done it in	
14	multiple states and with multiple	
15	agencies, and what we find is much more	
16	is permitted than agencies often realize.	
17	And so by taking the time to explain what	
18	is possible and working through, we've	
19	been able to surmount them.	
20	COUNCILWOMAN SANCHEZ: Were	
21	there any legal obstacles? Sometimes	
22	Health Department, HIPAA. Did you	
23	confront any challenges that way around	
24	data sharing?	
25	MS. ABERNATHY: It absolutely	

Page 157 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. is challenging, but it's doable. working now with health plans, so we have 3 to go through a very rigorous process in 4 5 order to do data sharing. In some cases where an agency doesn't want to share 6 data, we've been able to have -- help 7 them do the data match themselves and 8 9 then have them do the outreach that sends people to us for assistance. So even 10 11 where a data match isn't possible or they 12 don't want to share the data externally, we're able to help them do it themselves 13 14 and still be able to do targeted 15 outreach. 16 COUNCILWOMAN SANCHEZ: What are 17 the targeted outreach strategies that you 18 find most effective that you know we're not doing in Philadelphia currently? 19 There's a lot of great efforts, right, 20 21 but what are we not doing in Philadelphia 22 right now? 23 MS. ABERNATHY: I think we're doing the right things. We're just not 2.4 25 doing them at the proper scale. With

Page 158 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. more funding, we could be reaching out to many more people. We have the list 3 already. There are obviously a whole 4 5 host of additional lists, but we just 6 aren't funded adequately. With more partners, we can see when we map who are enrolled through BenePhilly. You can see 8 9 they're clustered where we have a BenePhilly partner. So having partners, 10 11 having sufficient people to answer the 12 phones to more aggressively promote the hotline, those are all things that 13 14 would -- I mean, again, we know how to do 15 this. We are just underresourced and our 16 partners are underresourced. 17 COUNCILWOMAN SANCHEZ: So one 18 of the benefits of having this process, I'm not the no person. I'm like tell me 19 20 how, right, and if it's a law, then I'm going to write it. Too many times in the 21 22 bureaucracy we plan for what we have, not 23 for what we need. And so part of this action plan has to take us out of that 2.4 25 box that we put ourselves in.

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 2.
                   So we appreciate the work, and
         we'll be asking very specifically how do
 3
         we make the investment -- you said a 20
 4
 5
         to 1 return on our investment -- so that
         we could do something aggressively. I'll
 6
         talk to Councilman Domb. He's always
 7
         coming up with ideas. He walks in the
 8
 9
         office, he's like what about if we, what
         about if we.
10
11
                   This is one of those issues
12
         that is a non-negotiable coming out of
         this process, and I know that Eva is very
13
14
         committed to it, and it's very
15
         frustrating that as a government, we're
16
         not doing a better job. We can't ask
17
         private sector to do what we don't do,
         right? And so I really, really thank you
18
         for all the work that's been done and
19
20
         with our partners, but this is one of
21
         those things that we're going to -- so
22
         let's plan for what we need, not for what
23
         we have.
2.4
                   Thank you.
25
                   Sharmain.
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Page 160 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. MS. MATLOCK-TURNER: 3 MS. GLADSTEIN: You go first. 4 MS. MATLOCK-TURNER: I just 5 wanted to say thank you very much for 6 this, because I was really happy to see 7 that you're continuing to move not just for this to be a retail model, which is 8 9 somebody has to show up and ask for help, when we know just from the work that we 10 11 saw with Ira, there's a lot of data out 12 there that tells us what block people live on, what neighborhood they're in. 13 14 The census is certainly going to give 15 us -- Councilwoman Blondell Reynolds 16 Brown talks about the census. We're going to have a lot of new data. 17 So it 18 seems like to me getting to a more wholesale plan as opposed to us having to 19 20 invest so much on the retail side just to 21 go find people -- is there something in 22 the structure that you're creating here 23 that can push us even a little bit faster to identifying and then have a department 2.4 25 say, okay, I'm signing you up and it's an

Page 161 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. opt out as opposed to an opt in strategy? 3 MS. ABERNATHY: So great 4 question. We actually worked with New 5 York City Mayor's Office in response to a request from their City Council, quite 6 frankly, legislation they passed to mandate a study to look at to what extent 8 9 there could be automatic enrollment and whether there could be a single 10 11 application, all those different ways of streamlining it. And so the report was 12 presented, and I've shared that with the 13 14 Safety Net Subcommittee, and I think it 15 would be worthwhile to do something 16 similar in Philadelphia. There won't be 17 quite as many opportunities, because in New York City they actually administer 18 19 SNAP in a way that we don't here, but certainly all of the City benefits, the 20 City-funded benefits, there's no reason 21 why they couldn't be aligned to the 22 23 eligibility for SNAP, for instance, or Medicaid and then we can make it almost 2.4 25 automatic. You could have pop-ups

Page 162 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. that pop up as soon as you click off. 3 Particularly in partnership 4 with the state, one of the things we 5 worked with the state to do is that when 6 you're applying for SNAP, you can get a 7 pop-up if you're eligible to say would you like to be enrolled in Medicaid, and 8 9 you don't even have to fill out a full Medicaid application. You're just 10 11 consent. And so we submitted over 60,000 12 applications for that kind of a consent to be enrolled in Medicaid when Medicaid 13 14 was first expanded, and now it's just been institutionalized. And as much as 15 16 possible, we want all of the data 17 matching and everything to get institutionalized, so where we can be 18 19 helpful and do it for agencies or teach them how to do it, we're happy to do so. 20 MS. MATLOCK-TURNER: Just one 2.1 follow-up on the EITC. I've had these 22 23 conversations with Councilman Domb. curious if you have looked at the Earned 2.4 25 Income Tax Credit at the federal level to

Page 163 1 10/10/19 - POVERTY REDUCTION - RES. 190239 see whether or not either Congress or the IRS could automatically, based on the 3 data that they have, formulate at least 4 5 some percentage of what people would be 6 eligible for for EITC and just give it to 7 them either on a monthly basis or at the end of the year. Has there been any 8 9 conversation like that in your space, and why is something like that maybe crazy 10 11 and won't work? MS. ABERNATHY: 12 There was actually a pilot done of that, and others 13 14 here may recall what some of the 15 challenges were, but I'd be happy to dig 16 that up and share that with you to 17 understand, because it does seem like 18 common sense, but I remember they ran into some challenges. 19 20 MS. GLADSTEIN: Particularly if 21 it could be done on a monthly basis 22 rather than an annual payment, wouldn't we all love it. 23 I want to thank BDT for all it. 2.4 25 has done and is doing and share one thing

Page 164 1 10/10/19 - POVERTY REDUCTION - RES. 190239 and then ask Pauline a question. The one thing I wanted to share is, with this new 3 kind of self-administered application, 4 5 we're looking at using that in our health centers, in our community schools, and in 6 our prisons for family members who are 7 waiting for visits. So that's also 8 9 increased the opportunity that there are in addition to the seven BenePhilly 10 11 centers based in community organizations. 12 CEO spends a very large percentage of its grant on BenePhilly and 13 14 I think believes that -- Mitch had to 15 leave. I think he would testify that 16 it's well spent and it's a very good 17 investment. 18 The question I wanted to ask is, it's always helpful to know when we 19 20 are an outlier compared to other cities. So in terms of our utilization of the 2.1 five big benefits that you charted out 22 23 for us, could you share with us whether or not -- I had always heard like some of 2.4 25 our numbers were pretty good, not that

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2	they were at 100 percent utilization, but	
3	is 85 percent good or actually like is	
4	the norm 97 percent, so 85 percent really	
5	isn't good?	
6	MS. ABERNATHY: It's certainly	
7	higher than many states, but we also have	
8	much higher poverty. So one would have	
9	to look at the find cities with	
10	similar poverty levels and compare their	
11	participation rates to ours to really	
12	assess it. But what we clearly know is,	
13	there are over 60,000 people who are	
14	eligible and not participating even if we	
15	are doing well as a result of these	
16	partnerships that have taken hold.	
17	MS. GLADSTEIN: Maybe that's	
18	something we could look at in our	
19	subcommittee, comparing ourselves to some	
20	other high-poverty areas and their	
21	utilization rates. Maybe we could learn	
22	something from them as well.	
23	COUNCILWOMAN SANCHEZ: Thank	
24	you.	
25	Councilman Domb.	

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2	COUNCILMAN DOMB: Thank you.	
3	Thank you for your	
4	presentation. By the way, I've taken a	
5	tour of BDT and you guys are doing a	
6	great job really. Very impressive.	
7	MS. ABERNATHY: Thank you.	
8	COUNCILMAN DOMB: Council	
9	President and I had a question. I just	
10	wanted to clarify. Is the \$450 million	
11	Philadelphia residents or is that	
12	Philadelphia and the suburbs?	
13	MS. ABERNATHY: Philadelphia	
14	itself. Just Philadelphia.	
15	COUNCILMAN DOMB: Just	
16	Philadelphia, okay.	
17	MS. ABERNATHY: Yeah.	
18	COUNCILMAN DOMB: And then the	
19	second question I have I have three	
20	how much is the City contributing to BDT	
21	on an annual basis?	
22	MS. ABERNATHY: \$1.3 million.	
23	COUNCILMAN DOMB: 1.3. And	
24	you're saying, Councilwoman Sanchez	
25	confirmed, that it's a 20 to 1 return?	
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Page 167 1 10/10/19 - POVERTY REDUCTION - RES. 190239 MS. ABERNATHY: Well, the total of BenePhilly's investment, including the private sector from our foundation 4 5 partners and the state, for that matter, 6 is 2.2 million, and so the 2.2 million generated 26 million. So for that, it 7 was 12 to 1. The research I cited was 8 9 for SNAP outreach in particular rather than this model that -- this more 10 11 comprehensive model. 12 COUNCILMAN DOMB: But even if it's 10 to 1 or 20 to 1, what we're 13 14 saying here is that of the \$450 million 15 that I assume that money comes out of all 16 of our federal withholding tax, basically that's where it's coming from, and when 17 it doesn't come back to the residents in 18 the City, to a degree it's a transfer of 19 wealth out of this region. You could 20 2.1 look at it that way. And so it would 22 behoove us to increase the funding --23 MS. ABERNATHY: Absolutely. COUNCILMAN DOMB: 2.4 25 dramatically to make sure this money

Page 168 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. comes back to the people in the City. 3 The other question I had was, 4 do you think there's any benefit if we 5 could figure out a way -- we send out 6 576,000 real estate tax bills every year 7 and we send out water and sewer bills, I think, monthly. Is there any way to 8 9 include these benefits in that mailing to raise awareness and who to call to 10 11 enroll? 12 MS. ABERNATHY: Yes. It would certainly help as long as we've got 13 14 enough resources to handle the phone 15 calls and provide the assistance. What 16 we found is, the more it can be personalized, the more effect it has. 17 18 often just generic information, here's a program. Particularly, again, if you 19 20 don't think you're poor, you don't think 21 you're eligible. 22 Well, let me COUNCILMAN DOMB: 23 ask you a question. Would it make sense for us to focus on properties that are 2.4 25 AVI valued under a certain level?

Page 169 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. MS. ABERNATHY: Absolutely. What again that MIT study showed is that 3 when we did targeted outreach that said 4 5 from a trusted source, in this case it 6 was the state, to someone saying we think you are eligible for SNAP -- these were seniors on Medicaid -- we think you 8 9 personally are eligible, there was an 81 percent increase in people who never even 10 11 called us. They just went and did it themselves based on the information 12 alone, which would suggest if you have 13 14 targeted information that is personalized 15 from a trusted source, for many people 16 that's enough. For other people, they 17 need the assistance. So we found that 18 our offering them assistance tripled it, not just an 80 percent increase, but a 19 20 tripling. And for others still, they're 21 going to want to have it from someone 22 they trust in person or they're not 23 connected already to a benefit. you're not connected to a benefit, then 2.4 25 you need a different way.

Page 170 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. But absolutely. I think the more we could target the information so 3 that it was more personalized, we'll see 4 an increase. And then if we can send 5 6 them to places where they can be served 7 in the way they want, whether it's online, in person, or on the phone, even 8 9 better, we'll see something more like that tripling. 10 11 COUNCILMAN DOMB: So I don't 12 know if you're allowed to discuss this, but I'll ask the question anyway. In the 13 14 contract that we have with your company, is that a contract -- how does that work? 15 16 Is that a contract for X amount of hours 17 or is it X amount of people? How is that 18 determined? 19 MS. ABERNATHY: It's for a 20 certain amount of money and in exchange, 21 we and the BenePhilly partners commit to doing a certain number of applications. 22 23 COUNCILMAN DOMB: Applications, okay. So it's per application basically. 2.4 25 So the more we --

Page 171 1 10/10/19 - POVERTY REDUCTION - RES. 190239 MS. ABERNATHY: Well, we don't 2. 3 get paid more if we do more applications. 4 It's a fixed amount. 5 COUNCILMAN DOMB: Okay. Ι 6 think I understand. MS. GLADSTEIN: There's a certain payment that goes to each of the 8 9 outreach sites, and for that payment, which pays for a little bit more than one 10 11 staff person maybe or maybe just a staff 12 person, they commit to doing X number of applications, and then beyond that, there 13 14 are costs to BDT, which is still 15 operating the call center and doing all 16 of the back-office support and actually 17 getting the applications in, like the 18 follow-up documentation and the training 19 for all those sites. 20 COUNCILMAN DOMB: Got it. Would you be equipped if -- we noticed 21 that Council District 5 and 7 have by far 22 23 the highest level of poverty. Would you be equipped to be able to have people on 2.4 25 laptops or iPads or whatever to go into

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2	the field and actually take applications	
3	in the field, whether they're knocking on	
4	doors, as Councilwoman Sanchez and I	
5	talked about earlier, or in the District	
6	Council offices where they could have a	
7	Tuesday evening from 6:00 to 9:00 or	
8	something?	
9	MS. ABERNATHY: Yeah. In fact,	
10	the mobile unit at CEO does go out and do	
11	exactly that using our software.	
12	SISTER MARY: We do it too.	
13	COUNCILMAN DOMB: Okay. That's	
14	great. Thank you very much.	
15	MS. ABERNATHY: Again, we have	
16	the tools. We know how to do it. It is	
17	really just a resource question.	
18	COUNCILMAN DOMB: So you need	
19	the money, is what you're saying.	
20	MS. ABERNATHY: Yes.	
21	COUNCILMAN DOMB: Thank you.	
22	Thank you.	
23	COUNCILWOMAN SANCHEZ:	
24	Councilman, we know who the people are	
25	and we know where they live. We just	

Page 173 1 10/10/19 - POVERTY REDUCTION - RES. 190239 need to connect the dots. 2. Thank you. 3 Thank you, Sister Mary, for 4 5 your patience. 6 SISTER MARY: Okay. I'm really 7 excited to be here today. I want to thank Council President Clarke, who has 8 9 been such a phenomenal leader in our community and in our city to end poverty, 10 11 and we appreciate all his support and 12 leadership over many, many years. And to Councilwoman Maria Quinones-Sanchez, we 13 14 are really excited again to be here with 15 you and to see your hard work and 16 determination in Kensington, and we hope 17 soon to be partnering with you there as well. And, of course, to Eva Gladstein, 18 who I don't want to say how many years 19 20 we've worked together, and the same with Sharmain, but let's just say it's a 21 little bit longer than even Council 22 23 President Clarke. Started very, very 2.4 young. 25 I'm really excited by the

Page 174 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. report that this City Council issued on 3 the poverty and the gap. I read it. I 4 think there's some really interesting and 5 exciting ideas there, and I'm hoping that 6 you can prioritize and, you know, move forward with those ideas. They were 7 great. I especially loved -- and we see 8 9 it every day. I think the idea of really connecting Philadelphians who live in the 10 City of Philadelphia who are poor and 11 12 unemployed with jobs in Philadelphia is the most strategic one that we're really 13 14 excited about and we really want to work 15 on, and we'll get in more detail, not 16 today but at a future time to do that. 17 So anyway, I was asked to give 18 a big idea, so I'm going to do it. And that big idea is, I believe here in the 19 20 City of Philadelphia we can end chronic 21 street homelessness, and the reason I believe that we can do it is because of 22 23 the ten largest cities in the United States, we have the highest poverty rate. 2.4 25 Of the ten largest cities in the United

Page 175 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. States, we have the lowest number of 3 people living on the streets. That is an 4 incredible statistic, and I invite 5 anybody here that thinks, "really?" to go to LA; San Francisco; Seattle, 6 Washington; Portland, Oregon; New York 7 City; or Washington, DC or any other 8 9 place you want to go to and you'll come home and Philadelphia will look a lot 10 11 better than when you went there. 12 really want to acknowledge the presence of Liz Hersh, who leads our homeless work 13 14 here in Philadelphia. Her energy and 15 determination and really creativity in bringing new ideas and efforts to this 16 has really made an impact as well. 17 18 So I am going to talk a little 19 bit -- I'm not going to go a big long 20 presentation, because it can be boiled down into smaller and counterintuitive 2.1 things. So that's what I'm going to do. 22 23 First thing is, single most important thing to ending homelessness in 2.4 25 our city is affordable housing. Single

Page 176 1 10/10/19 - POVERTY REDUCTION - RES. 190239 most important thing to ending 2. 3 homelessness in the long run is a quality education for every single child. 4 5 Homelessness is symptomatic of some our deeper societal problems. It's only 6 7 going to be solved by sustained work, thoughtful, intelligent, continued, 8 9 sustained efforts over long periods of There's no quick fixes. You know, 10 11 just putting it out there right away. 12 Also, the three things that I believe the City needs to invest in is 13 14 affordable housing for everyone that 15 needs it. Okay? So, of course, my 16 particular focus is for the very, very 17 poorest members of our community, but I also believe we need to invest in 18 affordable housing for people who are 19 20 working and low income, and the way the 21 prices are rising in our city, we need to be able to do something to support the 22 working families in our communities as 23 well. Housing, affordable housing. 2.4 25 Second, employment. And I just

Page 177 1 10/10/19 - POVERTY REDUCTION - RES. 190239 said it. Project HOME, the good news is, 2. 3 last year we got 183 people jobs, and some of those men and women have lived on 4 5 our streets for 10 and 20 years and are 6 now working. Ending homelessness -- this is the first counterintuitive thing. 8 Endina 9 homelessness saves lives, but it also 10 saves money. 11 Our prisons are our largest 12 mental hospitals. Our prisons are our largest hopefully recovering communities 13 14 as well, but about 70-some percent of all 15 the people incarcerated have histories of 16 addiction. Let's be preventative. Let's work with men and women and families with 17 18 special needs to get the same thing that we all need, a safe place to live, 19 20 employment opportunities, and a quality 21 education. And at Project HOME, any day of the week that you feel maybe this 22 23 can't work, you just come and we'll show you people that are working all 2.4 25 throughout the City of Philadelphia that

Page 178 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. have been homeless, that have been homeless for many, many years, and now 3 are paying taxes, are paying rent, with 4 5 no subsidy. They may have needed a 6 subsidy at one time, but now no subsidy, 7 are voting, and are leaders in our community. 8 9 I have so much hope that we can do this because of years of experience 10 11 and thousands of lives that are out there that have overcome homelessness with a 12 hand up, not a hand out. Okay? 13 14 So second thing is that we need 15 affordable housing, and in the City of 16 Philadelphia, we have about 5,000 units 17 of supportive housing. To end chronic street homelessness in our city, we need 18 another 2,500 units of housing dedicated 19 20 to people with special needs. 21 People that are on the streets 22 is the prophetic presence saying, hey, 23 there's something radically wrong here. It's a smaller percentage of people that 2.4 25 are homeless, but it's the most visible.

Page 179 1 10/10/19 - POVERTY REDUCTION - RES. 190239 The people who live on the streets mainly 2. 3 have special needs, mental health or 4 addiction or both or physical 5 disabilities. Two thousand five hundred 6 units would give us the capacity in the City to address the needs as people become homeless who are new that year and 8 9 people who leave homelessness, which again is the best news. People leave 10 11 homelessness, leave subsidies when they 12 get employment. So to build that capacity is what we need to do to end 13 14 chronic street homelessness. 15 two-thirds of the way there. Many cities 16 in the United States would die to be as far along as we are, and we just have to 17 18 finish the job. It's not that hard. 19 Subsidies and permanent supportive housing, safe havens and 20 21 recovery housing, and there's -- it's the 22 power of we. There's so many 23 collaborators, ODAAT, Housing First, Pathways to Housing, Mental Health 2.4 25 Partnership, you know, Self, so many

Page 180 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. organizations, but it's the power of we. 3 So that's one thing. 4 The second thing I just want to 5 talk about is another -- this is 6 counterintuitive, but as I said, saving lives and saving money. At Project HOME 7 we do a lot of development of housing 8 9 that's affordable. According to Econsult, in all the zip codes that 10 11 Project HOME works with -- now, this 12 study is going to come out soon. This is a draft. This is a preview. You're the 13 14 first to hear this. Okay? 15 But of all the zip codes that 16 Project HOME works with, if you look at 17 the property values and how they've risen 18 in those zip codes -- and they've all 19 risen. That's one number. But if you 20 look at the properties that are located a 21 quarter of a mile from Project HOME sites, the property values have risen 22 23 greater than what they have in the general zip code. 2.4 25 Now, what Econsult estimates --

Page 181 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. and I'm just saying this because it is 3 counterintuitive and you're going to just 4 think, really? But I'm going to tell you 5 anyway. 6 In total, the presence of 7 Project HOME sites are estimated to add around \$1.4 billion in values to the 8 9 housing prices of the properties around Project HOME. That's \$1.4 billion in 10 11 homeowners' pockets and maybe even in 12 Domb real estate's pockets, because we have some in Center City. So that's 13 14 okay. I'm teasing. I'm just teasing. 15 But it's really true. It's also true. 16 Okay? 17 But if you apply the City's 18 mileage rate of 1. -- it's almost 2 19 percent to this additional \$1.4 billion in housing values, the results is an 20 21 estimated \$20 million in property tax revenue to the City of Philadelphia and 22 23 to the School District of Philadelphia from 2010 to 2019, and that's in Project 2.4 25 HOME's area of development.

Page 182 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. So, again, that is counterintuitive, but the reason I'm 3 bringing that up, these numbers would 4 5 also be true for the Philadelphia Housing Authority, for other developers, CDCs, 6 7 Pennsylvania CDCs, the great CDCs in Kensington. By doing affordable housing 8 9 in neighborhoods, you're adding value to the existing homeowners. So that is a 10 11 real amazing thing. 12 The second thing that I just wanted to say is, I also think that as 13 14 part of the strategy to ending poverty 15 and ending homelessness, like the Earned 16 Income Housing Tax Credit -- and I think 17 Councilman Domb was the one who 18 recommended it. While the City is getting a surplus, so now we're talking 19 20 about when the City's fiscal year ends 21 and you look at the surplus that the City has, in those years where there's a 22 23 surplus, I think we should give a rebate to the Earned Income -- like an Earned 2.4 25 Income Tax Credit here in the City of

Page 183 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Philadelphia to anybody making \$30,000 or 3 \$40,000 or less. You know, I know you 4 said 24,000. I'd take it up a little 5 bit, you know, because families need it, 6 and we have a surplus. So I think we should also give money to people, you know, not just 8 9 programs as much as I value Project HOME and all the other organizations, but 10 11 figure out a way to give money to people. 12 I think that would go a long way too. And I also think there should be 13 14 incentives for philanthropists in the 15 United Way to raise money when we have a 16 surplus by doing challenge grants in 17 whatever strategy you come up with to 18 ending poverty to incentivize people to do it by leading by example and using 19 some of the additional surplus from the 20 General Fund to incentivize others to 21 22 join in this very, very, very important 23 effort. I also wanted to talk about 2.4 25 BenePhilly a little bit. We love

Page 184 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. BenePhilly. It's such a great program. 3 And I think in our big idea of ending 4 chronic street homelessness, which we 5 definitely think is doable and definitely 6 think can be done, when we do BenePhilly, we think we should add registering to 7 vote, right? Like in any program that we 8 9 do, we should register poor and homeless people to vote so that their voice will 10 11 be heard as we move forward with building 12 an agenda. I also believe in our effort to 13 14 end and prevent chronic street 15 homelessness, one of the things that I 16 think is critically important is that we 17 do it with the people that are experiencing homelessness through focus 18 groups, through other ways of engaging 19 20 them as leaders and being part of 21 envisioning how this can be done, because they know the best way out of 22 23 homelessness and poverty, because they're the ones that are directly experiencing 2.4 25 it. So I think that that has to be

Page 185 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. critically important as well. 3 I also think for the BenePhilly 4 thing, to include tax services and links 5 to financial and homeownership counseling 6 as well as to career counseling through 7 BenePhilly. Again, Pauline is absolutely right, people don't realize the 8 9 opportunities that are out there for them, and the more that we can connect 10 11 and talk to people who are actually 12 experiencing poverty and homelessness and get them educated and involved in being 13 14 part of the solution, we're going to have 15 a stronger solution and a stronger city. 16 Lastly, I just really want to say in Philadelphia, I think the most 17 18 amazing thing is that we're a city with so many leaders that are passionate about 19 20 this effort and organizations like SEPTA, organizations like the Franklin 21 Institute, the Free Library. So another 22 23 great example of really connecting with 2.4 people that are on the streets that maybe 25 on some given day you may pass and think

Page 186 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. this is a hopeless situation, with the 3 Restroom Attendant program, now we hired 4 19 people who were living on the streets 5 to be restroom attendants in MSB and in 6 the libraries. It's a win-win situation 7 for everybody. The bathrooms are cleaner, you know, safer. People are 8 9 working. They're the kinds of innovative partnerships that Sharmain and Eva and 10 11 the Councilwoman and the Council President have come up with, along with 12 others in this room, that I just think we 13 14 have to keep expanding. 15 I have so much hope, and I 16 think this is a great opportunity. 17 so excited about your leadership. We're willing to work shoulder to shoulder with 18 you in any way we can. I think we are 19 20 poised in this city to make a big leap 21 forward by working together and thinking it through and being accountable and all 22 23 the things everybody else has said. just didn't want to repeat, repeat. 2.4 25 I just also want to thank everybody in

Page 187 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. this room for believing with us that none of us are home until all of us are home 3 and that we can never be complacent as 4 5 long as one of our sisters and brothers 6 are on the street, and our city will be 7 so much stronger and so much vibrant, that all those corporations in LA and San 8 9 Francisco will want to move here because we have nobody living on our streets, and 10 11 that is a quality of life that is 12 cherished, I think, by many people. I think by doing this, it could be a way 13 14 of enhancing the overall economic 15 potential and growth of our community. 16 Thank you. 17 (Applause.) 18 COUNCILWOMAN SANCHEZ: Thank you, Sister Mary, for doing God's work, 19 20 and thank you so much for your 21 willingness to really help us in Kensington in addressing a very 22 23 complicated issue, and we really look forward to you being there boots on the 2.4 25 ground at this particular time as

Page 188 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. everybody knows with everything that's 3 going on in Kensington. 4 Real quickly, you talked about 5 the 183 clients who got jobs. I know you 6 have your own store that you run and some 7 of the other things. One or two things -- and I know Sharmain is leading 8 9 the education/jobs piece. What is the one or two things we could be doing at a 10 better scale to help people transition 11 from homelessness to a job in terms of 12 13 the skills gaps? 14 SISTER MARY: So the most 15 important thing that we learned if we 16 could do one thing is, incentivize people 17 with a little bit of money to come 18 through what we call certificate programs. So people have to be trained 19 20 to do the library. They have to be 21 trained to do the peer -- CPS, certified peer specialist training, A Plus 22 certification, SafeServ, SafeServ 23 Supervision, customer service. And we're 2.4 25 working with an organization from New

Page 189 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. York, Perskolis (ph), on more technology 3 skills, and by the training and working 4 right -- and it includes internships and 5 apprenticeships, so you have to work at the places. And sometimes it takes two 6 or three times for it to stick, but the 7 issue is, it eventually does stick, and 8 9 that's what you have to do to continue the opportunities for employment in 10 11 various different ways that makes it 12 possible for people to succeed. companies like even Amazon, UPS, the Gap, 13 14 other bigger companies will come and do 15 job fairs, and then if we can continue to 16 work the training piece and the 17 technology piece, Microsoft Office and 18 all that, they really get the jobs. might start part time, but eventually get 19 20 full-time jobs. Just because you said at 21 SEPTA where we've been pretty successful 22 there, one of the people who worked there 23 and now is vested just bought his own home. You know, it's wonderful. 2.4 It's so 25 much -- it's so exciting to see someone

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2	who once lived on the streets, ten years	
3	later again, it's not like there's	
4	no instant things, but there has to be a	
5	pathway forward. There has to be	
6	opportunities. It can't be shelter or	
7	revolving door that leads nowhere, only	
8	back to the street. There has to be	
9	interventions where people can work,	
10	where people can live, live a dignified	
11	life and have a sense of community and	
12	purpose. And, you know, we're just so,	
13	so, so encouraged.	
14	COUNCILWOMAN SANCHEZ: Thank	
15	you.	
16	Any questions?	
17	MS. MATLOCK-TURNER: Yes. I	
18	just want to say thank you for everything	
19	that you've done. One of the first	
20	meetings that I attended when I joined	
21	the Coalition just a few years ago was	
22	with you around the whole effort about	
23	ending homelessness, a Blueprint to End	
24	Homelessness.	
25	SISTER MARY: We can do it. We	

Page 191 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. can do it. 3 MS. MATLOCK-TURNER: And so that's my question. We're sort of 4 5 sitting here now saying we think we have 6 a blueprint to end poverty. What about 7 that movement that gets us to the place where we are the best city on really 8 9 creating space and place and dignity for those that are homeless? What can you 10 11 share with us as we start this movement? SISTER MARY: So I think it 12 consists of two -- a couple things. 13 14 first thing is, I think under Liz's 15 leadership with a lot of support from Eva 16 and David Jones with the Department of 17 Behavioral Health, there is a plan and 18 they know what needs to be done, which is there's no one way. So it's housing 19 20 first. It's recovery housing, but not --21 certified or dignified recovery housing, because there's so many issues with that. 22 23 So recovery housing. Some safe havens for people that are mentally ill and a 2.4 25 lot of permanent housing with both with

Page 192 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. subsidies and through development. 3 that's the housing picture. 4 Every day -- and this has been 5 another really great process with CCD; 6 that's the Center City District -- the 7 Philadelphia Police Department, Project HOME, Reading Terminal, and the Fashion 8 9 District. During the day when people are most visible, we work as a team and 10 11 engage people during the day, and with 12 dedicated placements for people to go, meaning that if someone says, I want to 13 14 come in, we can get them in, and then we 15 follow them, so there's employment 16 opportunities and actually a pathway out of homelessness. We'd like to bring that 17 18 more to scale, and I know that not -- I'm using the Ambassadors of Hope as one 19 20 example, but that's not the only example. 21 Bethesda, One Day At A Time has been phenomenal, you know, with the encampment 22 23 beds and -- they're a strong partner in this whole effort. 2.4 25 So, again, it's just

Page 193 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. understanding who the people are that are out there and then having the appropriate 3 placements for them where they can move 4 5 forward. Sometimes it's a shelter. Actually, most of the people that are on 6 7 the street come and go. We don't know where they came from; we don't know where 8 9 they went. They don't need that heavy intervention, but for the chronically 10 11 homeless who are seriously mentally ill 12 and seriously long-term addicted, they need an intervention with support based 13 14 on their -- like no one can do it for 15 you, but when they're ready to make those 16 changes, you have to have the opportunity 17 and the support there where they can continue on the journey home. And now 18 they're working and contributing, they're 19 taxpayers, they're voters. 20 21 MS. MATLOCK-TURNER: So it. 22 sounds like to me we're going to need the 23 three of you to continue to tell the story, because I think an important part 2.4 25 of what happened around the homeless area

Page 194 10/10/19 - POVERTY REDUCTION - RES. 190239 1 2. was having that kind of civic and 3 business leadership, along with the great work of government, to really continue to 4 5 work together to say we're not going to 6 let small disagreements get in the way of 7 really staying focused on the big picture and working that through a lot of 8 9 different political administrations and different businesses coming in. 10 11 seems like to me sort of keeping the 12 community involved, engaged, sharing information around success, and then 13 14 challenging us around the challenges to 15 say there's more that we could do, but 16 really saying that we can solve this 17 problem. It won't get done tomorrow, but 18 we absolutely can solve this problem. 19 SISTER MARY: Correct. And I 20 think, Sharmain, I think with Council 21 President Clarke's leadership, along with Maria Quinones-Sanchez and Allan Domb and 22 23 all the other Councilpeople, to really put some resources and challenges out 2.4 25 there that need to be matched. The City

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         can't do it alone, but to give some
         leadership and incentives for others to
 3
         get involved, I think we could make a lot
 4
 5
         more progress, and I think to do it in
 6
         years when there's a surplus. You have a
 7
         surplus this year.
                   I mean, I know the City has a
 8
 9
         lot of needs and I'm like the common good
         kind of person, but a little bit -- you
10
         know, we need a nice chunk for poverty,
11
12
         because, again, it has a return on its
         investment if you look at the Econsult
13
14
         report that we're looking at right now.
15
                   MS. MATLOCK-TURNER:
                                         Thank you.
16
                   COUNCILWOMAN SANCHEZ:
                                           Thank
17
         you.
18
                   Any questions from any other
         folks from the panel?
19
20
                    (No response.)
21
                   COUNCILWOMAN SANCHEZ:
                                           Just,
22
         again, thank you to our panel.
23
         really, really appreciate it, and this is
         going to be an ongoing conversation.
2.4
25
         know some of you are participating in
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Page 196 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. some of the subcommittee activities, and definitely all of you are helping inform 3 the action steps for what we're going to 4 5 produce. I want to go quickly through 7 some of the logistics as we move forward. As I mentioned at the beginning of this 8 9 hearing, we have a Housing Subcommittee chaired by Councilman Domb and I, and 10 11 we're going to be focusing on some of the 12 scenes that we heard here, the preservation of affordability, how to 13 14 protect homeowners, and ensuring fair 15 housing for all, with particular eye in 16 making sure that we figure out ways to 17 incentivize access for housing, 18 particularly for returning citizens and others who that becomes even a bigger of 19 20 a challenge. The Jobs and Education 2.1 Subcommittee, which Sharmain and Mel are 22 23 chairing, will focus on strategies that close the skills gap and really looking 2.4 25 at creating those family-sustaining jobs

Page 197 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. and protecting workers. And, of course, Eva is chairing 3 the Social Safety Net Subcommittee, which 4 5 will focus on some of the things we 6 talked today about, strategizing to maximize benefits for enrollment, sustaining family income, and promoting 8 9 financial fairness. I want to really thank everyone 10 who came out today to kick off this big 11 12 ideas kind of discussion, the framing, allowing us to frame poverty and the 13 14 working poor, with keeping in mind that 15 the glass is half full, that there are a 16 lot of great partners and there's an 17 excitement. And I really do believe that the business sector is waiting for us to 18 challenge them with some very specific 19 20 action items. I think everyone is there. I think as we look at the 2.1 22 growth in the City and you hear, whether it's the Chamber's Inclusive Growth 23 2.4 Strategy, the neighborhood strategy, 25 absolutely with our partners at

Page 198 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. Pennsylvania Association of CDCs and others, really looking at commercial 3 4 corridors and this livable, walkable 5 communities, I think we're going to get a 6 lot of great energy. And, again, really 7 want to think and plan for what we need, not what we have, and let us challenge 8 9 ourselves to figure out how we close 10 those gaps. 11 All the chairs, all of the 12 coordination is happening through Chris Goy in the President's office. We will 13 14 be listing the series of hearings. We 15 have tasked every subcommittee to have a 16 public hearing. We've encouraged folks 17 to go out so that this is not the only 18 space where we're having conversation. 19 We will have one more of these sessions, and in the interim, all of the Committee 20 21 members that are on our one-page fact 22 sheet, which is a lot of names, are going 23 to be producing some tangible action 2.4 steps. 25 This is not a production of

Page 199 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. another report, but an action plan that we hope to be able to turn over to the 3 Administration and charge them with 4 5 significant investment for next year, 6 impacting our five-year and our ten-year plan, and that's why the framing, the data, and how we're going to measure and 8 9 monitor our progress was important, that we shape that today. With the support of 10 11 Ira and his team, we look forward to 12 doing that. If there is no one else here to 13 14 testify, no one else signed up at the 15 table, then I will recess early, which 16 never happens. I want to invite any of 17 my Co-Chairs, if they want to have any closing words, and we will reach out to 18 folks, as I said, for the remainder of 19 20 the hearing, the subcommittee work, and 21 then our closing public hearing before issuing the report. 22 23 Eva. 2.4 MS. GLADSTEIN: I just wanted 25 to thank everybody who is here for their

Page 200 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. work, for their hopefulness, and encourage them to do two things. One is 3 tell the success stories, and the other 4 5 is encourage other people to participate in this process as we move forward. 6 7 really important. 8 Again, we hope to get out into 9 the neighborhoods through our three next hearings in November, but just to, again, 10 11 bring your energy out to the community and back into the other activities. And, 12 13 again, thank you, everybody. 14 COUNCILWOMAN SANCHEZ: 15 Sharmain. 16 MS. MATLOCK-TURNER: Yes. Т 17 just want to say even though the three of us are sitting up here and we look really 18 good and smart, I really want to thank 19 20 the team of people who are working with We talked about staff from Council 21 President Clarke's office, but Marcus 22 23 Kellam from my staff is doing 100 other things that I have him doing. He's doing 2.4 25 a terrific job in supporting this work

Page 201 1 10/10/19 - POVERTY REDUCTION - RES. 190239 2. and is working on getting other things I know Eva has the same thing 3 scheduled. with people on her team, also Mel Wells 4 5 as well. 6 So this really is, I think, an act of love. We're all committed to 7 really driving this forward. And so 8 9 everybody is volunteering, so please give us the best of your time and energy so 10 11 that we can come up with an action plan 12 that is going to get the rest of the City excited about the fact that we can end 13 14 this. It's not just about intervention 15 anymore. It's about really getting to 16 the place that we're the city where 17 people want to come because we figured 18 out how to solve this problem. 19 So stay with us. Follow us. 20 Share with others that we're not giving 21 in and that we're not giving up. 22 Thank you. 23 COUNCILWOMAN SANCHEZ: Thank 2.4 you. 25 Councilman Domb, do you have

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2	any closing words for us?	
3	COUNCILMAN DOMB: No.	
4	COUNCILWOMAN SANCHEZ: Thank	
5	you, everyone, and I look forward to this	
6	short journey and a lot of action.	
7	Thank you.	
8	(Special Committee on Poverty	
9	Reduction and Prevention concluded at	
10	6:41 p.m.)	
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                      CERTIFICATE
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              I HEREBY CERTIFY that the
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     proceedings, evidence and objections are
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     contained fully and accurately in the
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     stenographic notes taken by me upon the
     foregoing matter, and that this is a true and
 7
     correct transcript of same.
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