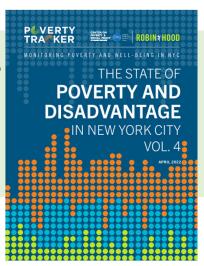




The State of Poverty and Disadvantage in New York City

Latest findings from the Poverty Tracker

Sophie Collyer Center on Poverty and Social Policy Annual Conference May 26, 2022







Launched in 2012 to capture a more complete picture of poverty and disadvantage in New York City and to track long-term trends and dynamics.

Surveying a representative sample of more than 3,000 New Yorkers every three months for up to four years.





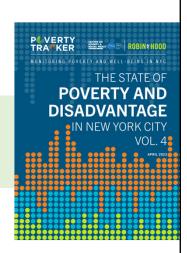






Measure more than just poverty. Annual measure rates of income poverty, material hardship, health problems. Also regularly collect data on factors related to disadvantage, such mental health, life satisfaction, assets and debts, employment, among others.

Poverty and disadvantage in New York City are widespread *and* citywide rates mask stark disparities.



In 2020, the city's poverty rate was well above the national average.



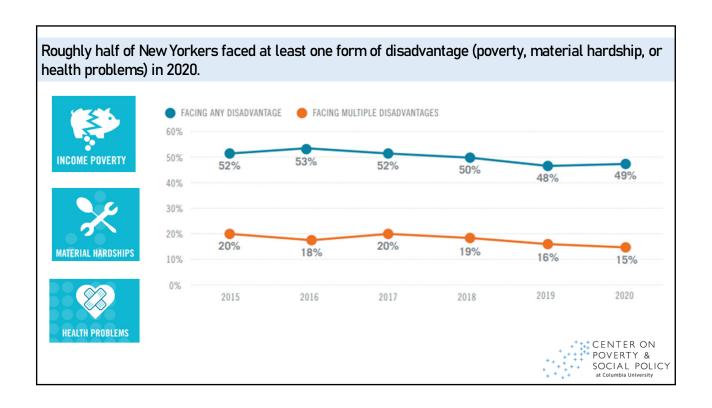
16% of adults

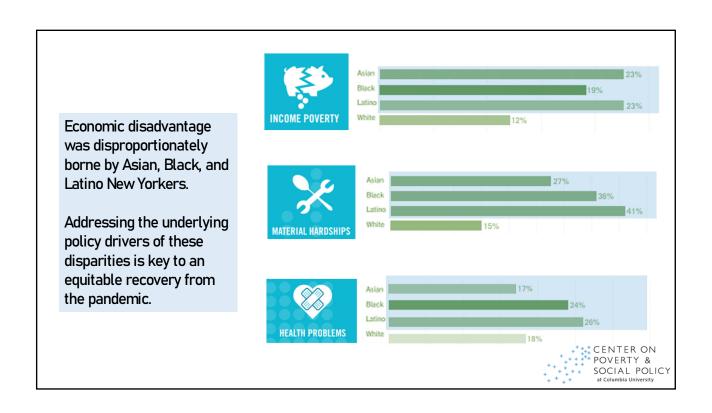
18% of children

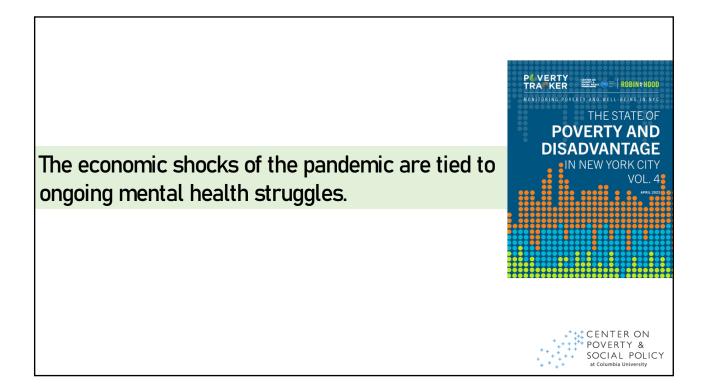


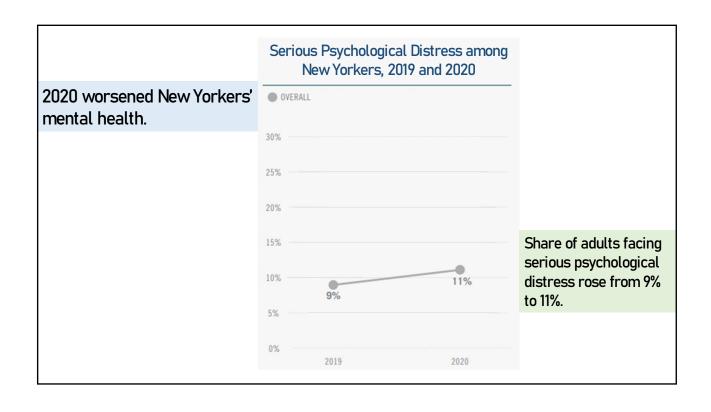
9% of adults

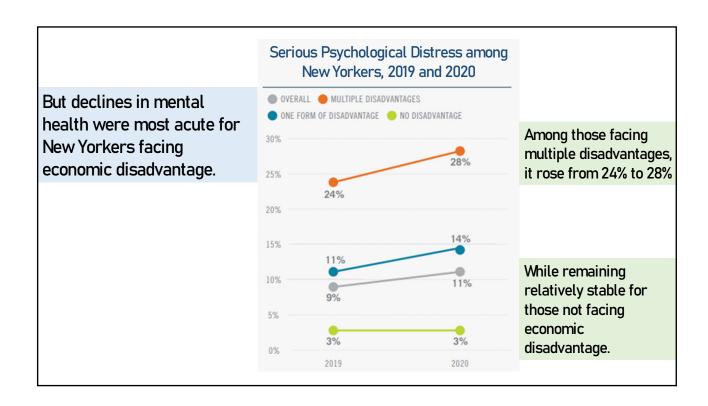
10% of children









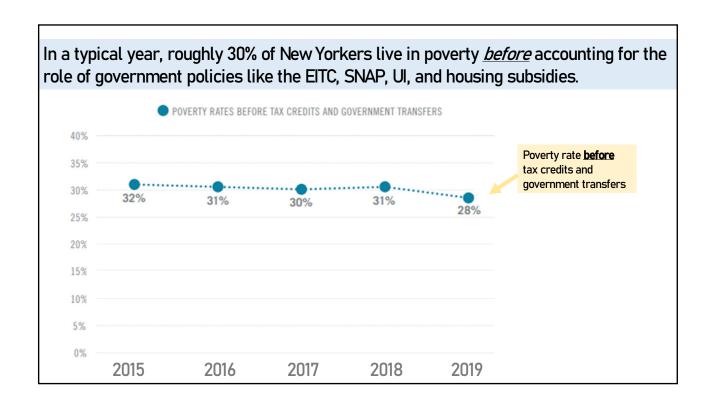


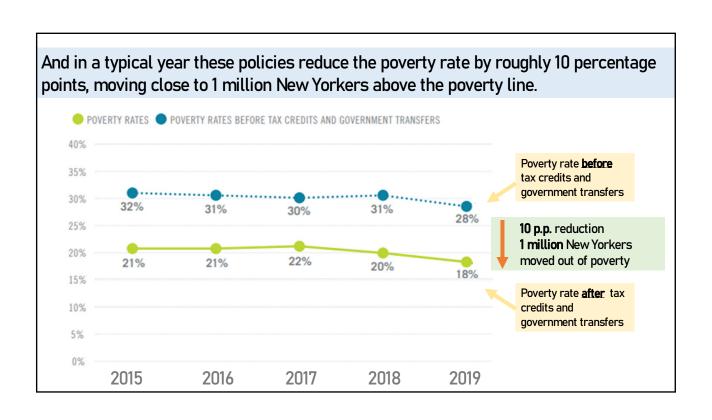
Policy reforms can have a substantial effect on the economic hardships and disadvantages that New Yorkers face today.

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IN NEW YORK CITY VOL. 42

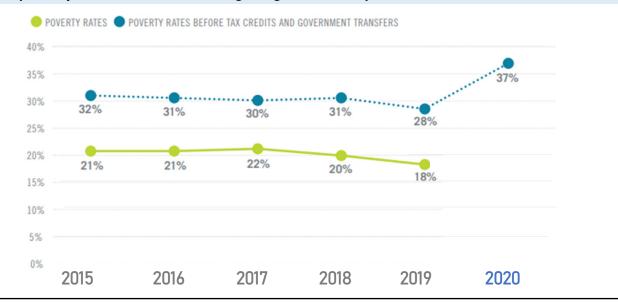
VOL. 42

SOCIAL POLICY & SOCIAL POLICY ALL CHARLES AND POVENTY & SOCIAL POLICY ALL CHARLES AND POLICY AND POLICY ALL CHARLES AND POLICY ALL CHARLES AND POLICY AND POLI





In 2020, as millions of New Yorkers lost work or income, we saw a sharp increase in the poverty rate before accounting for government policies

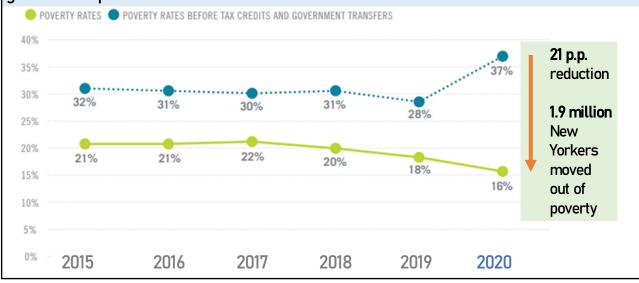


2020 also saw <u>temporary</u> reforms made all levels of government to strengthen the safety net and stabilize incomes in uncertain times.

- Temporary expansion to Unemployment Insurance benefits
- Economic Impact Payments or "stimulus checks"
- Expanded SNAP benefits
- Pandemic EBT
- Eviction moratoria



Due to these policy expansions, the poverty rate in New York City actually fell between 2019 and 2020 and 1.9 million New Yorkers were kept above the poverty line by government policies.



Altogether ...

- highlight the high rates of disadvantage in New York City
- · economic disadvantage and mental health, and
- · the effects of policy reforms
- Single metric vs. multi-dimensional measures
- Multitude of social forces affecting well-being and economic security
- The potential of policy reforms in addressing these multiple and varying challenges



Macire Aribot, Abraham Arriaga, Alexander Auyeung, Chantal Bannerman,
Lily Bushman-Copp, Tingyi Cao, Chloé Cargill, Daniel Castillo, Eunho Cha, Sophie Collyer,
Isaiah Colmenero, Lolita Colon, Genesis De Los Santos, David Faes, Ava Farrell, Katherine Friedman,
Jill Gandhi, Qin Gao, Irwin Garfinkel, Janira Gayle, Elizabeth Gonzalez, Mikayla Greeley, Sonia Huq,
Yajun Jia, Xuan Jia, Barbara Lantz, Young Seo Lee, Kevin G. Li, Danli Lin, Xiaofang Liu,
Matthew Maury, Ronald Mincy, Angie Moran, Kathryn Neckerman, Juan Rincon, Schuyler Ross,
Daniel Salgado, Ao Shen, Julien Teitler, Serdil Tinda, Luis Gasca Trivino, Jane Waldfogel,
Kahlen Washington, Nicholos Wilkinson, Christopher Wimer, Ho Yan Wong, Xinyu Xia, Christopher Yera

Thank you!

This report would not be possible without the support and partnership of Robin Hood. We are especially grateful to all members of the Robin Hood team who provided feedback and support in the production of this report. We also would like to thank all of the Poverty Tracker interviewers at Columbia University, past and present — we would not be able to say much without their excellent work.



Spotlight on Asian New Yorkers: Experiences of Poverty, Disadvantage, and Discrimination

Xiaofang Liu, Survey Coordinator, Poverty Tracker Studies



Background - about Asian Americans

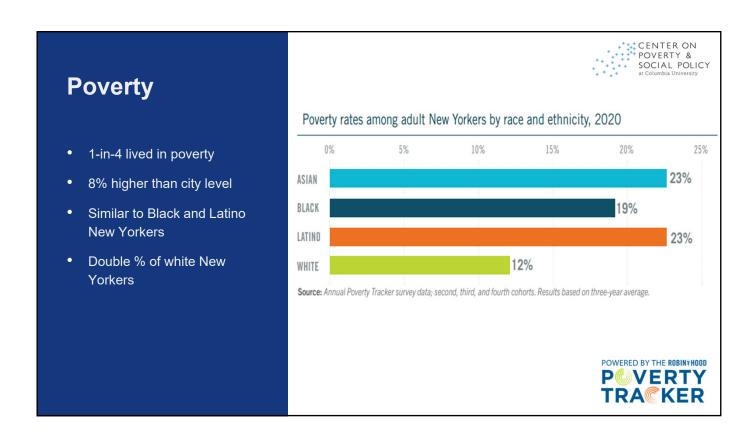
- Fastest growing racial and ethnic group in NYC and the U.S.
- Among the most understudied racial and ethnic groups
- Underrepresented in many data sources, including those that collect information on poverty and economic disadvantage
- "Model minority" myth
- Incredibly diverse

It's important to have accurate and timely data on poverty and disadvantage among Asian New Yorkers.



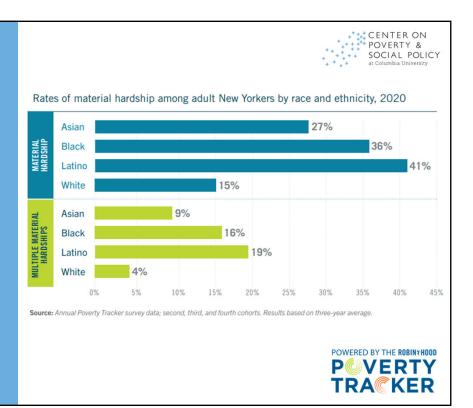
Poverty Tracker's Oversample of Asian New Yorkers

- In 2020, PT began to survey in Mandarin and recruited an over-sample of Asian New Yorkers
- Increased the representativeness of Asian New Yorkers
- The only source of longitudinal information on poverty (SPM) and other forms of disadvantages among Asian New Yorkers
- Language limitation: Asian New Yorkers being interviewed in English, Mandarin, or Spanish



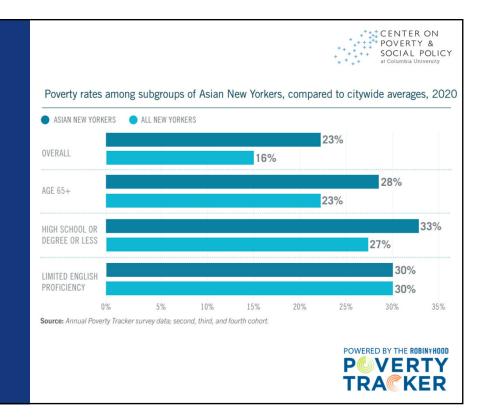
Material Hardship

- More than ¼ faced material hardship
- In-line with city level
- Double % of White New Yorkers
- Lower than what might be expected >> additional research is needed



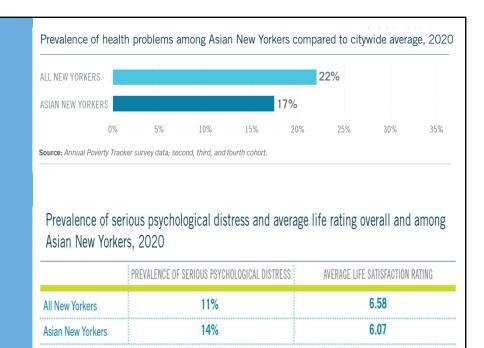
Vulnerable Subgroups

- Aged 65 or older
- with a high school degree or less
- with limited English proficiency



Health & Well-being

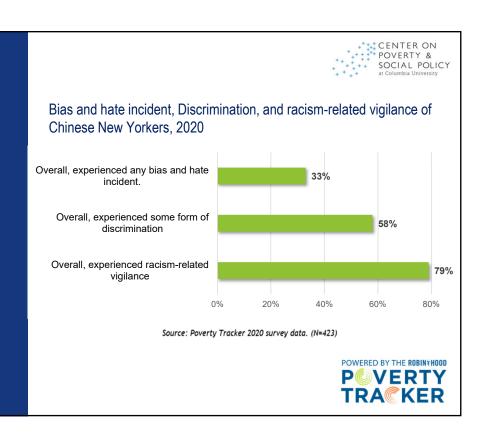
- Less health problems
- Higher serious psychological distress
- Lowest life rating



Source: Annual Poverty Tracker survey data; second, third, and fourth cohort.



- 1-in-3 encountered bias and hate incident
- Over half experienced discrimination
- Majority endured racismrelated vigilance





Implications

- Greater attention to the experiences of Asian New Yorkers facing economic disadvantage
- "Double pandemic" is likely to have serious negative impacts
- Policies and culturally/linguistically supportive programs serving these vulnerable populations are essential
- Asian Americans need to be sufficiently represented in all data sources used to make policy-related decisions

Thank you

Questions about this research? Contact:

Xiaofang Liu, Study Coordinator, Poverty Tracker Study xl2761@columbia.edu

Qin Gao, Professor of Social Policy and Social Work qin.gao@columbia.edu

Acknowledgments

- We thank Robin Hood for funding the Poverty Tracker, which provided data for this study.
- We are grateful to the feedback and support from members of the Poverty Tracker research team, particularly Tingyi Cao, Sophie Collyer, Irwin Garfinkel, Suchen Huang, Sonia Huq, Xuan Jia, Yajun Jia, Dani Lin, Lin Mao, Matthew Maury, Kathryn Neckerman, Schuyler Ross, Ao Shen, Julien Teitler, Jane Waldfogel, Christopher Wimer, and Xinyu Xia.

Links to relevant reports:

- The State of Poverty and Disadvantage in New York Citv
- Double Pandemic:
 Discrimination Experiences of
 New Yorkers of Chinese
 Descent During COVID-19

Web: povertycenter.columbia.edu Email: cpsp@columbia.edu Twitter: @cpsppoverty

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When Money and Mental Health Problems Pile Up: The reciprocal relationship between income and psychological distress

Oscar Jiménez-Solomon, MPH

Pre-Doctoral Poverty Research Fellow, Center on Poverty and Social Policy, Columbia University

Research Scientist, New York State Psychiatric Institute, Columbia University Medical Center

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+ SOCIAL POLICY
+ at Columbia University

Our Methods

Data: New York City Longitudinal Study on Wellbeing (Poverty Tracker)

- 2015-2019 cohort: 5-wave with yearly measure for key outcomes
- 18-64 year old New Yorkers (N=3,103)

Measures:

- 1. Individual earnings in past 12 months (percentiles)
- 2. Psychological distress over the last 30 days: Kessler-6 for distress (e.g., felt nervous, hopeless, depressed)

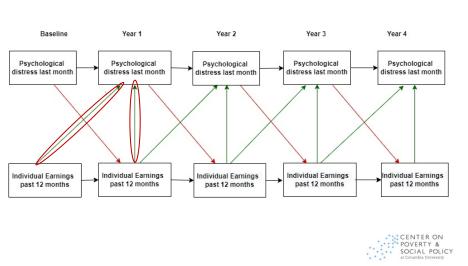
Control variables: gender, age, race/ethnicity, education level, immigrant status, partnership status, number of children

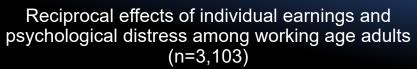
Analytical approach: cross-lagged panel model with fixed effects

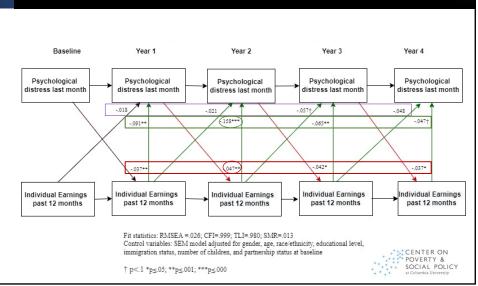
- · all relationships in the same model
- · controlling for unobserved differences between individuals

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at Columbia University

Effect of individual earnings → psychological distress in last month







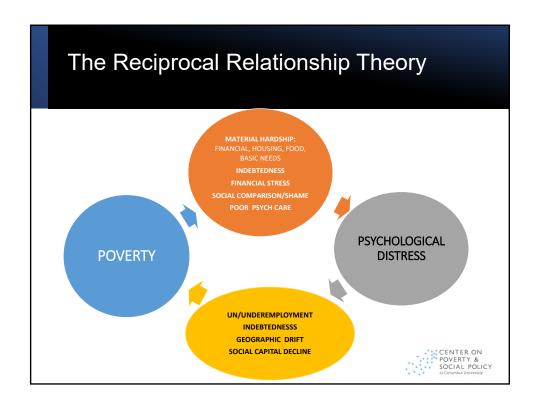
Conclusions

Money & mental health problems pile up!

Individual earnings←→ psychological distress:

- Individual earnings → psychological distress
- Psychological distress → individual earnings





Conclusions

Money & mental health problems pile up!

$\textbf{Individual earnings} \boldsymbol{\leftarrow} \boldsymbol{\rightarrow} \textbf{psychological distress:}$

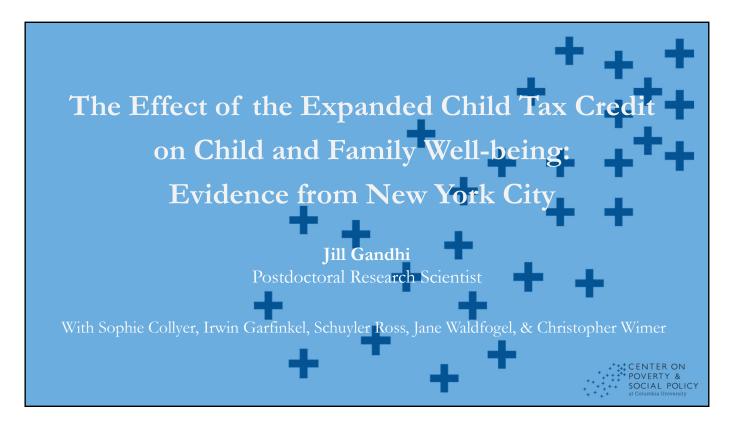
- Individual earnings → psychological distress
 Psychological distress → individual earnings
- Recent earnings (last 12 months): stronger impact on psychological distress than income further back in time (13 -24 months prior)
- Family income (last 12 months): stronger impact than income 13-24 months prior
- Need to integrate economic policies (e.g., cash, unemployment supports) and mental health services
- Future research:
 - · Replicate with other Poverty Tracker cohorts
 - Other measures of poverty (e.g., material hardship)
 - · Different/shorter time lags
 - Longer observation periods to examine cumulative effects



THANK YOU FOR JOINING US







Policy Context: The Expanded Child Tax Credit under the American Rescue Plan

Three key changes:



Expanded eligibility to include those with moderate, low, or no earnings previously left out



Increased annual benefit levels to a maximum of \$3,000 per child aged 6 to 17 and \$3,600 per child under age 6



Delivered payments on a monthly, rather than annual, basis between July and December 2021–reaching more than *61 million children* in over 36 million households



The potential effects of the expanded CTC

- The monthly CTC payments moved millions of children out of poverty in the months that payments were made (Parolin et al., 2021).
- The monthly payments also had the potential to improve the well-being of families and children in other domains:
 - material hardship
 - the need to use emergency food assistance
 - parents' mental health
 - ability to make child-related investments
 - parents' employment

Data and Methods



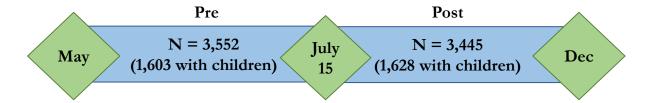
Poverty Tracker

Longitudinal study with surveys every three months

Early Childhood Poverty Tracker (ECPT)

• A parallel longitudinal study of families with young children (3-7 years old during expanded CTC payments)

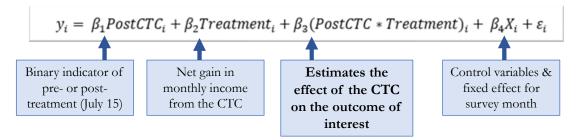
Both provide critical information on poverty, material hardships, psychological distress, and other factors related to well-being.



Data and Methods: Estimating effects

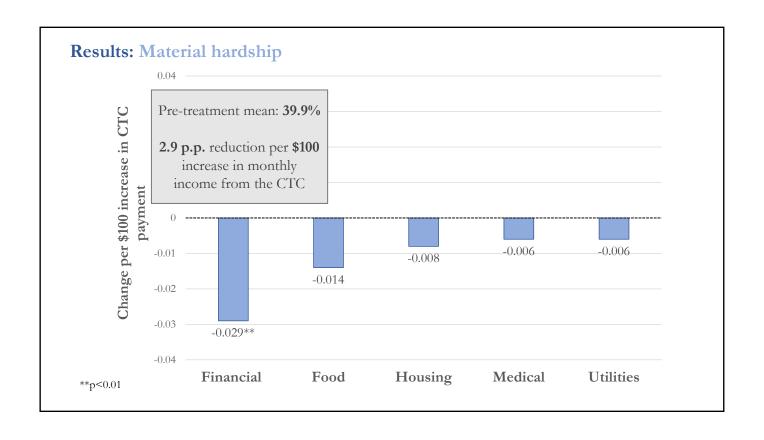
With each \$100 increase in monthly income from the CTC, how did outcomes change for families with children?

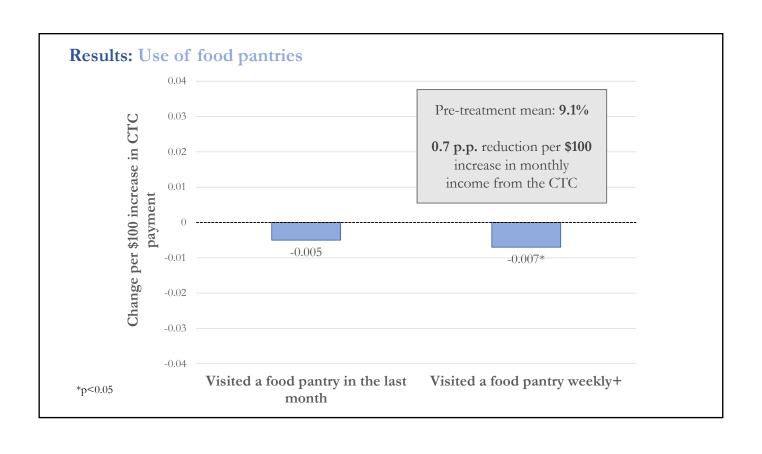
• Employed a **difference-in-differences** framework (Parolin et al., 2021).

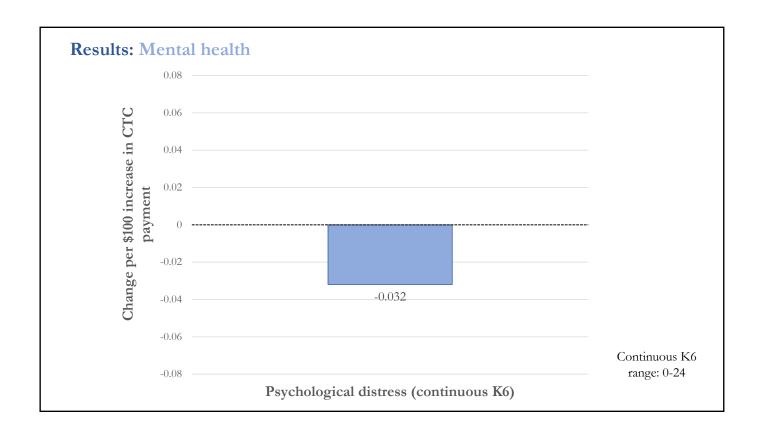


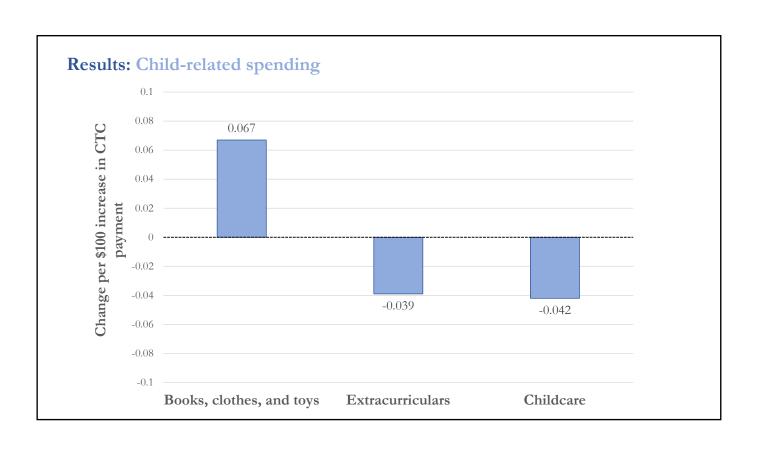
• Presenting intent-to-treat results of our outcomes, using continuous treatment for the full sample of New Yorkers.

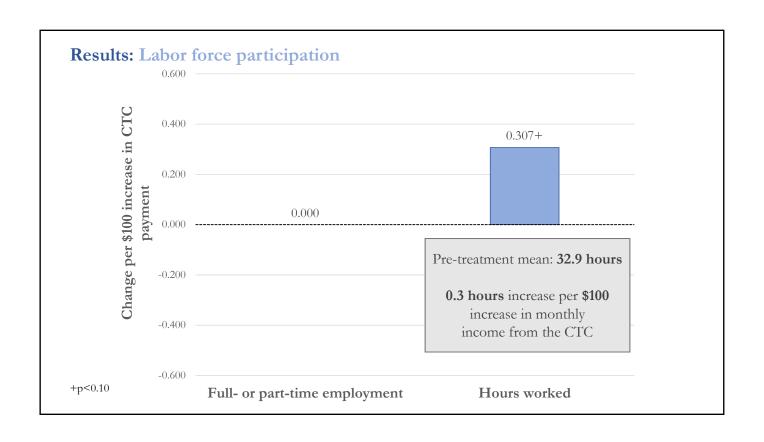


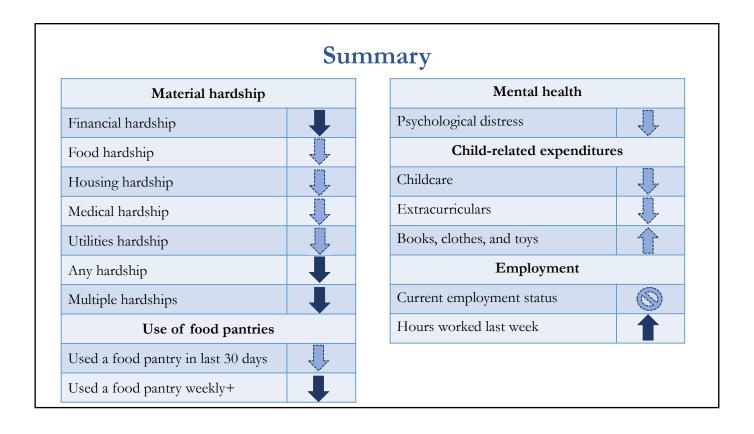


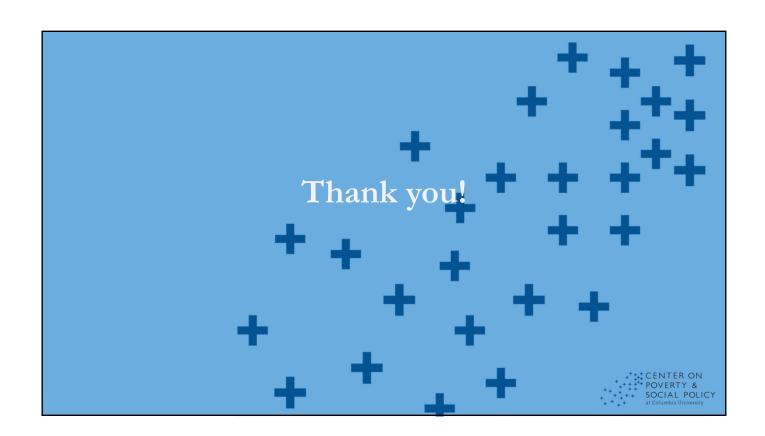












Research Roundup:

What do we know about the expanded Child Tax Credit?

Megan A. Curran Policy Director, Center on Poverty and Social Policy

What do we know about the expanded Child Tax Credit?



POVERTY & SOCIAL POLICY REPORT

Research Roundup of the Expanded Child Tax Credit: The First 6 Months

- onthly payments may be reducing financial stress and other hardships





Impact of the Expanded Child Tax Credit

ACCESS

Reached over 61 million children, but outreach needed to newly-eligible children in low income families

INCOME

Buffered family incomes amidst continuing pandemic & uncertain economy

POVERTY

Significantly reduced child poverty (by approx. 30%) during the six months it was in place

SPENDING

Families spent it on basic household needs – most common item: food

HUNGER

Reduced food insecurity, particularly among families with lower incomes

STRESS

Reduced financial stress and material hardship.

EMPLOYMENT

No evidence of reductions in parental work

EQUITY

Children of color stand to benefit the most



By September 2021, checking account balances of families with low incomes were 70% higher (approx \$1,000) than pre-pandemic

(JPMorgan Chase)



The most common way families spent the Child Tax Credit was on food. This was true for families with incomes up to \$150,000 & for families with incomes less than \$35,000

(Social Policy Institute & Center on Budget & Policy Priorities)

Findings



The initial Child Tax Credit payments reduced food hardship by 25%

(Center on Poverty and Social Policy)



Child Tax Credit Effects on Child Poverty

The fifth Child Tax Credit payment kept

3.8 million children

from poverty in November.

iiiii

The CTC's anti-poverty effects have grown by 800,000 children

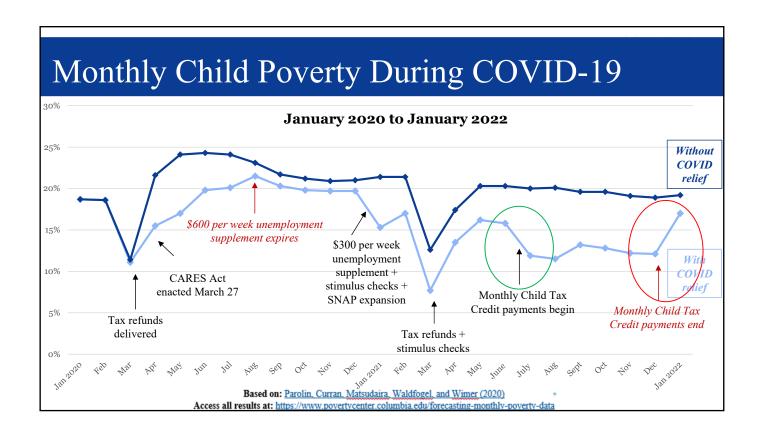
since the first payment, which kept 3 million children from poverty in July.

Due to the Child Tax Credit, monthly child poverty rates have dropped.

> 15.9% JUNE 2021

> 12.2% NOV. 2021

*3.7 million children kept out of poverty in December 2021 with a December child poverty rate of 12.1%







Effects of the Expanded Child Tax Credit on Employment Outcomes: An Update

Center on Poverty and Social Policy Annual Conference 2022

Benjamin Glasner, Postdoctoral Research Scientist May 26, 2022

Written with Elizabeth Ananat, Christal Hamilton, and Zach Parolin

Research Question: How did expanded CTC payments affect parents' labor supply?

Employment ↓

- Unconditional transfer
- · Removal of phase in
- Cut in relative wage and an increase in non-labor income
- Simulations report reductions in parental employment

No Effect or Employment ↑

- Simulations based on 1980s 2000s
- Lower willingness to leave work
- Volatile nature of low-wage work
- · Canadian child allowances null
- Parents increased work (5%) and decreased work (5%)

Note: CTC expansion was temporary and this is a short-run analysis



Approach

• Data:

- Current Population Survey (Jan 2021 to Feb 2022) Employment and Labor Force Participation
- Household Pulse Survey (Jan 2021 to Feb 2022) Employment

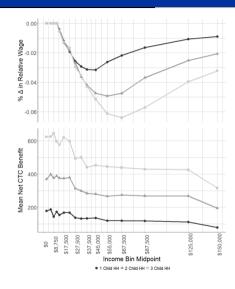
Treatments:

- · Dichotomous: Children vs. no children
- Continuous: Predicted net change in CTC benefit (tests income effect)
- · Continuous: % Change in return to work (tests substitution effect)

Design:

- · We use a two-way fixed effect difference-in-differences approach
- Condition on age, sex, and education status of the household head
- Include robustness checks using alternative treatment timing, event studies, and CENTER ON group-dosage response designs

Treatments



- Two Children, \$8,750:
 - · Children vs. no children 1
 - % Change in return to work ~0%
 - Net change in monthly CTC benefit ~\$380
- Two Children, \$67,500:
 - Children vs. no children 1
 - % Change in return to work ~-5%
 - Net change in monthly CTC benefit ~\$275
- Two Children, \$125,000:
 - Children vs. no children 1
 - % Change in return to work ~-3%
 - Net change in monthly CTC benefit ~\$270



Results

Table 1: Difference-in-Differences Estimates of the Effect of the Expanded CTC on Employment and LFP

	Binary Treatment		Continuous Treatment		Continuous Treatment	
	1 = Household with Child(ren)		\$100s of Net Monthly Benefit		1% Change in Relative Wage	
CPS (N=822,933)	1:Employed	2:Active in Labor Force	3:Employed	4:Active in Labor Force	5:Employed	6:Active in Labor Force
Treatment	0.043***	0.037***	0.000	0.000	-1.032***	-0.834***
	(0.003)	(0.003)	(0.006)	(0.005)	(0.053)	(0.056)
Treatment X Post	-0.002	-0.000	0.001	0.001	0.04	-0.022
	(0.004)	(0.003)	(0.001)	(0.001)	(0.064)	(0.057)
Pulse (N=818,009)	1:Employed (Intent-to-Treat)	2:Employed (Treatment-on- Treated)	3:Employed (Intent-to- Treat)	4: Employed (Treatment-on- Treated)	5:Employed (Intent-to- Treat)	6: Employed (Treatment-on- Treated)
Treatment	0.006	0.005	-0.009	-0.009	-0.958***	-0.987***
	(0.004)	(0.005)	(0.005)	(0.005)	(0.089)	(0.072)
Treatment X Post	0.004	0.007	0.002	0.003	-0.073	-0.006
	(0.006)	(0.011)	(0.001)	(0.002)	(0.091)	(0.009)

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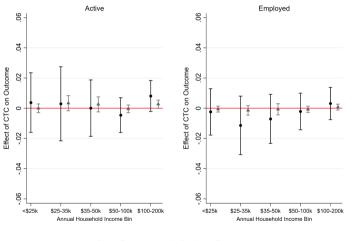


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	(0.003)	(0.003)	(0.006)	(0.005)	(0.053)	(0.056)
Treatment X Post	-0.002	-0.000	0.001	0.001	0.04	-0.022
	(0.004)	(0.003)	(0.001)	(0.001)	(0.064)	(0.057)
Pulse (N=818,009)	1:Employed (Intent-to-Treat)	2:Employed (Treatment-on- Treated)	3:Employed (Intent-to- Treat)	4: Employed (Treatment-on- Treated)	5:Employed (Intent-to- Treat)	6: Employed (Treatment-on- Treated)
Treatment	0.006	0.005	-0.009	-0.009	-0.958***	-0.987***
	(0.004)	(0.005)	(0.005)	(0.005)	(0.089)	(0.072)
Treatment X Post	0.004	0.007	0.002	0.003	-0.073	-0.006
	(0.006)	(0.011)	(0.001)	(0.002)	(0.091)	(0.009)

Heterogeneity by Income Bin





Conclusion

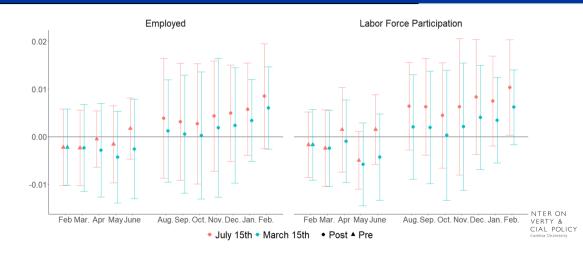
- Our employment analyses do <u>not</u> support the claim that the CTC expansion resulted in reduced employment or labor force participation
- · Our findings are robust:
 - · Across three measures of the CTC expansion
 - · Across both the CPS and Pulse
 - · Using both an Intent-to-Treat and Treatment-on-Treated design
 - We find no indication of a violation in parallel trends or lagged effects on employment/labor force participation
 - · When testing for Group-dosage response variation



Appendix



Event Study on the Effect of the CTC Expansion using both the March 15th and July 15th Treatment Definitions



Group-dosage response

 Table 7: Difference-in-differences Estimates of the Effect of the CTC Expansion on Employment

 Outcomes Using the Callaway and Sant' Anna (2020) Methodology and Defining Treatment Group

 by the Additional Monthly CTC Payment Received (CPS, January 2021 through December 2021)

		Effect Estimate		
Treatment-Group: Monthly Additional CTC Payments	Treated Units	1: Employed	2: Active in Labor Force	
All Treated Households	208,572	0.004	0.006	
\$100	6,701	(0.004) -0.009	(0.004) -0.006	
\$100	0,701	(0.013)	(0.012)	
\$125	17,000	-0.011	-0.009	
****	17,000	(0.009)	(0.009)	
\$150	10,787	0.007	0.009	
		(0.014)	(0.013)	
\$175	39,616	0.0003	0.004	
		(0.007)	(0.007)	
\$200	5,355	0.006	0.011	
		(0.019)	(0.019)	
\$225	7,123	0.022	0.035	
		(0.019)	(0.017)	
\$250	17,930	-0.010	0.0001	
		(0.009)	(0.009)	
\$325	11,419	-0.008	-0.003	
		(0.013)	(0.011)	
\$350	24,882	-0.005	-0.001	
		(0.009)	(0.008)	
\$375	11,787	-0.003	0.001	
		(0.011)	(0.01)	
\$525	17,082	0.007	0.012	
		(0.010)	(0.010)	
\$675	6,182	0.024	0.025	
		(0.016)	(0.015)	



Group-dosage response

Table 8: Difference: in-Differences Estimates of the Effect of the CTC Expansion on Employment Outcomes Using the Callaway and Sant' Anna (2020) Methodology and Defining Treatment Group by The Number and Age of Children (CPS, January 2021 through December 2021)

			Effect Estimate		
Treatment- Group: Children ages: $0 \le x < 6$	Treatment- Group: Children ages: $6 \le x < 18$	Treated Units	1: Employed	2: Active in Labor Force	
All Treated	Households	208,572	0.004 (0.004)	0.006 (0.004)	
0	1	46,206	0.002 (0.005)	0.003 (0.005)	
0	2	37,703	0.001 (0.006)	0.002 (0.005)	
0	3	13,158	-0.0003 (0.006)	0.003 (0.006)	
0	4	3,523	-0.0001 (0.007)	0.0001 (0.007)	
1	0	31,145	-0.003 (0.006)	0.001 (0.005)	
1	1	21,477	0.006 (0.008)	0.006 (0.007)	
1	2	11,045	-0.001 (0.006)	0.0001 (0.006)	
1	3	3,990	-0.0001	0.002	
			(0.006)	(0.007)	
2	0	15,782	-0.003	-0.0007	
			(0.007)	(0.006)	
2	1	6,402	-0.001	0.0004	
			(0.007)	(0.007)	
2	2	2,649	0.001	0.002	
			(0.007)	(0.007)	



Group-dosage response

Table 9: Difference-in-Differences Estimates of the Effect of the CTC Expansion on Employment Outcomes Using the Callaway and Sant'Anna (2020) Methodology and Defining Treatment Group by the Change in the Relative Wage (CPS, January 2021 through December 2021)

		Effect Estimate			
Treatment-Group: Change in the Relative Wage	Treated Units	1: Employed	2: Active in Labor Force		
All Treated Households	208,572	0.004 (0.004)	0.006 (0.004)		
-1% Δ in Relative Wage	18,263	-0.0001 (0.007)	0.003 (0.006)		
-2% ∆ in Relative Wage	27,455	(0.001)	0.004 (0.006)		
-3% ∆ in Relative Wage	41,780	-0.001 (0.005)	0.002 (0.005)		
-4% ∆ in Relative Wage	16,830	0.003 (0.007)	0.006 (0.006)		
-5% ∆ in Relative Wage	22,265	-0.002 (0.006)	0.002 (0.006)		
-6% ∆ in Relative Wage	16,686	0.003 (0.007)	0.004 (0.007)		
-7% ∆ in Relative Wage	8,735	0.0003 (0.006)	0.003 (0.007)		
-8% ∆ in Relative Wage	3,869	0.002 (0.007)	0.004 (0.007)		
≤-9% ∆ in Relative Wage	2,174	-0.0001 (0.007)	0.002 (0.007)		



Poverty Around the Time of a Birth and the Role of Social Policies

with Jane Waldfogel, Chris Wimer, and Laurel Sariscsany

Christal Hamilton

Postdoctoral Research Scientist
May 26, 2022

Introduction



- The arrival of a newborn can have significant financial implications for mothers and families.
- First-time mothers may be less likely to be employed in positions with the flexibility needed for child rearing.
- Mothers from minority backgrounds are more likely to be low-income and to be employed in positions without maternity leave.

Research Objective

To estimate poverty status in the six months before and after the arrival of a newborn for all mothers and by birth parity and race, and assess the extent to which current social supports mitigate economic losses surrounding a birth.

Data and Methods



Data

- 2014 and 2018 panels of the Survey of Income and Program Participation (SIPP)
 - Detailed monthly information on income, program participation, demographic characteristics, household composition, and fertility

Sample

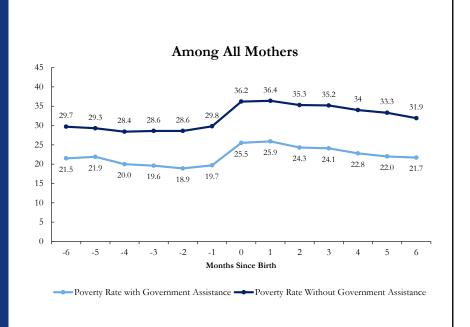
- Women with a birth during SIPP panel years
 - 966 mothers

Outcome Measure

Poverty Status (Supplemental Poverty Measure)

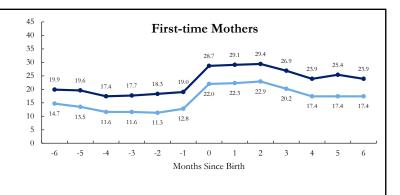
Findings

- Poverty rates among mothers significantly increase after birth and rates with government support is consistently lower.
- Increase in poverty rate is lower after childbirth with government support.

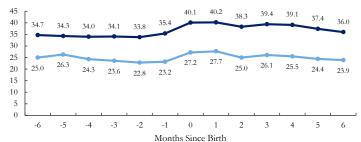


Findings

- First time mothers have consistently lower poverty rates than mothers with previous children, but sharper increase in poverty after birth
- Sharp increase in poverty immediately after childbirth even with government support

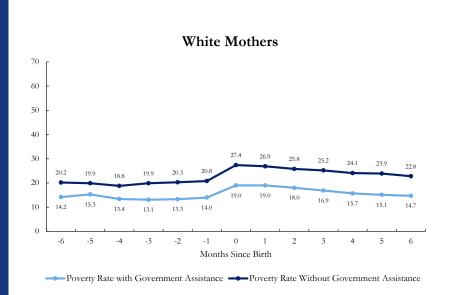


Mothers with Previous Children



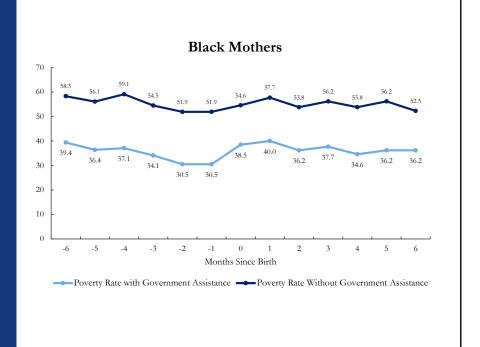
Findings

- Similar trend in poverty rates as with all mothers
- Increase in poverty rate is lower after childbirth with government support.



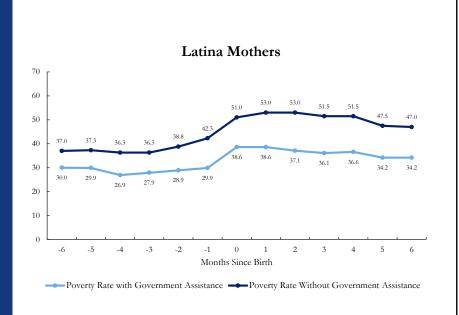
Findings

- Black mothers had the highest poverty rate of all racial groups.
- Government support important both before and after birth
- Substantial increase in poverty immediately after childbirth even with government support



Findings

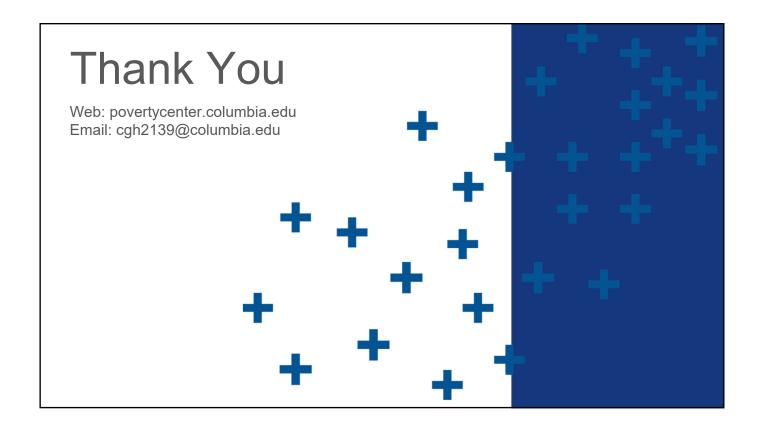
- Higher poverty rate than White mothers, but lower than Black mothers
- Similar trend in poverty rates as with all mothers.
- Government support particularly important after childbirth



Discussion and Conclusion



- Government support helps reduce poverty among mothers with newborns, but additional help still needed.
- Additional Supports:
 - Family Leave Policy
 - Child Tax Credit
 - Birth Grants
- Future research to explore the impact these polices can have on the poverty rate of women around the time of birth.



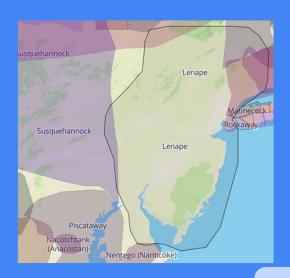
American Indian & Alaska Native Homelands

Amber Lewis, MSW
Racial Justice Research Assistant, Center on Poverty & Social Policy



Lenapehoking

Land Acknowledgement



The Lenape Center, 2022

Poverty & Infrastructural Inequities

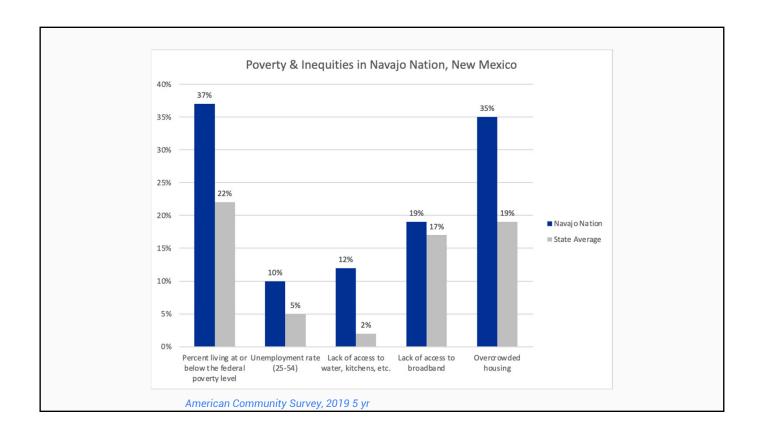
- Poverty
 - o 28% of all AIAN live at or below the Federal Poverty Level (FPL). (2019 ACS 5 yr)
 - of AIAN residing on and 24% of AIAN residing off Homeland Territory (HT) live at or below the FPL. (2019 ACS 5 yr)
- Infrastructural inequities.
 - 6% of AIAN residing on and 1% of AIAN residing off HT lack access to basic infrastructure, including piped water, plumbing, and kitchens. (2019 ACS 5 yr)
 - 0 10% of AlAN residing on HT living at or below the FPL lack access to piped water, plumbing, or kitchens. (2019 ACS 5
 - 4% of AIAN residing on HT living above the FPL lack access to piped water, plumbing, or kitchens. (2019 ACS 5 yr)

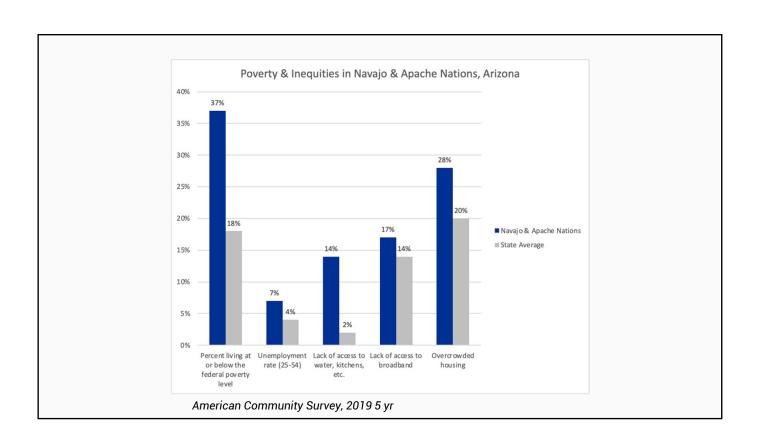
AIAN are systematically undercounted and underrepresented in key reports, including national poverty measures.

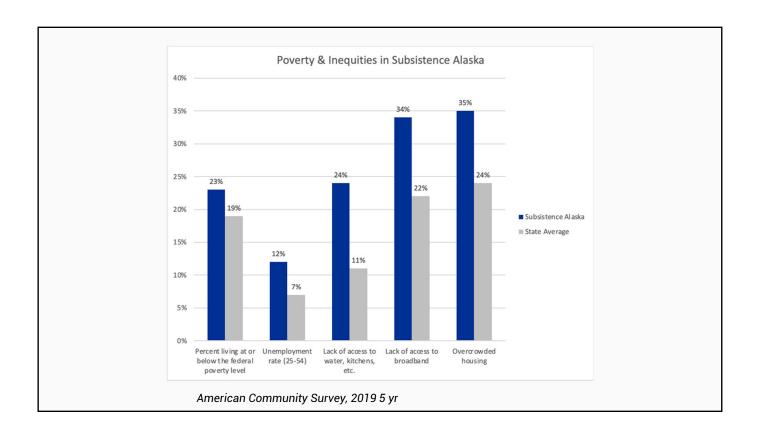
Case Study in Context: Homeland Territory

- Identified top 3 Public Use Microdata Areas (PUMA) with highest concentration (majority) of AIAN. These PUMA comprised of 100% HT.
 - Navajo Nation, NM 89% AIAN
 - Navajo and Apache Nations, AZ 73% AIAN
 - Subsistence Alaska, AK 72% AIAN

Research question: What is the significance of place in relation to poverty for AIAN?







Policy Implications

• Indigenous Rights

- Right to water and to regulate water resources. (Congressional Research Service 2022; Crepelle 2019; United Nations Declaration of the Rights of Indigenous Peoples (UNDRIP) 2007; City of Albuquerque v. Browner 1993)
- O Right to self determination. (Crepelle 2021, 2019; UNDRIP 2007)
- Right to free, prior, and informed consent prior to approval of any project affecting AIAN lands, territories, and resources. (UNDRIP 2007)

Addressing Poverty

- o Permanent expansion of Child Tax Credit. (Arizona Center for Economic Progress 2021)
- o Comprehensive data collection. (Stamatopoulou 2021)
- Increased employment opportunities.

Thank you!

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