#### DEMOCRATIZING THE ECONOMY OR INTRODUCING ECONOMIC RISK? GIG WORK DURING THE COVID-19 PANDEMIC

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Advancing Research on African American Families in Economically Marginalized Communities

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# What I am sharing with you today

• Motivation of this project.

• Data, variables and method.

## • Results and policy implications.

## Motivation

- The gig economy and its promises
  - Allowing people to become entrepreneurs.
  - Giving people freedom to work on their own time while pursuing their passions.
- $\rightarrow$  Reducing economic inequality.

## Motivation

• The growth of the gig economy has coincided with increased economic precarity in the new economy.

• Unclear how gig work fuels economic insecurity among American families.

## Gig work and Economic Insecurity

- Gig workers lack the social protections associated with standard employment arrangements (Kalleberg 2018: 3; Vallas and Schor 2020).
- Precarity of gig employment came into sharp relief during the pandemic (Ravenelle et al. 2021).
- For some workers, gig work may not be a sign of economic insecurity (Ravenelle et al. 2021; Moulton and Scott 2016).
- People's experiences with gig work vary by their family situations (Ricketts & Boshara 2020).

## Data and Methods

• Sample of 4,756 workers, Socio-Economic Impacts of COVID-19 survey (April 27 to May 12, 2020).

• Qualtrics online panels (Panels of a large number of potential respondents).

### Measuring Self-Employment and Gig Work

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3/5/2024 7

### Measuring Self-Employed

Which best describes your <u>current</u> employment situation?

- Self-employed, full-time for pay (include contract work)
- Self-employed, part-time for pay (include contract work)

### Measuring Gig Work

In the past 3 months, has anyone in your household been paid for any of the following activities? (Yes/No)

- Child or elder care services
- Dog walking, feeding pets, or house sitting
- House cleaning, yard work, or other property maintenance work
- Driving or ride-sharing, such as with Uber or Lyft
- Paid tasks online, such as posting YouTube videos (do not include online surveys)
- Other personal tasks, such as deliveries, grocery shopping, running errands, or helping people move

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3/5/2024 9

Outcome			Non-Gig	
	Sample %	Gig Worker	Worker	p-value
Skipped Bills, Past 3 Months	0.12	0.21	0.08	0.000
Behind on Credit Card, Now	0.08	0.18	0.04	0.000
Unpaid Credit Card Balance, Now	0.30	0.33	0.28	0.006
Lost Job/Income due to COVID-19	0.29	0.40	0.24	0.000
Skipped Housing, Past 3 Months	0.07	0.15	0.15	0.000
Evicted, Past 3 Months	0.03	0.09	0.01	0.000
Skipped Medical Care, Past 3 Months	0.10	0.20	0.06	0.000
Food Insecurity, Past 3 Months	0.25	0.40	0.19	0.000
Weighted Observations	4756	1358	3398	

### Financial Distress Outcomes, by Gig Employment Status (Propensity Score Weighted)

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## Other Key Variables & Controls

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3/5/2024 11

Characteristic	Sample (4,756)	Gig Worker (1,358)	Non-Gig Worker (3,398)	Diff	p-value
Liq. Assets pre-COVID, Mean (SD)	27469.9 (59481.7)	25392.0 (58532.2)	28300.3 (59845.3)	0	0.963
Liq. Assets pre-COVID, Median	4450	4450	6000		
Children, 0 (%)	0.74	0.63	0.78	-0.35	0
Children, 1 (%)	0.13	0.17	0.11	0.18	0
Children, 2 (%)	0.1	0.15	0.08	0.25	0
Children, 3+ (%)	0.03	0.05	0.03	0.11	0

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## Results

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3/5/2024 13

**Figure 1**. Probability of Experiencing Economic Hardship Among Gig and Non-Gig Workers.



14

**Figure 2**. Probability of Experiencing Economic Hardship Among Gig and Non-Gig Workers.



**Figure 3**. Probability of Experiencing Economic Hardship Among Gig and Non-Gig Workers by Liquid Asset Endowment.



Job/Income Loss Due to COVID

#### Unpaid CC Balance, past 3 mo



#### Food Insecurity, past 3mo





## Q1: \$0-\$2,000 Q2: \$2,001-\$8,250 Q3: \$8,251-\$28,900 Q4: >\$28,900

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Figure 4. Probability of Experiencing Economic Hardship Among Gig and Non-Gig Workers by Liquid Asset Endowment.

30%

15%

0%

Q2 Q3 Q4

Liquid Asset Quartile

Q1



#### **Skipped Housing Payment, past 3mo**

Q2 Q3

Q1 Q2 Q3

Q4

Q4

Q1: \$0-\$2,000 Q2: \$2,001-\$8,250 Q3: \$8,251-\$28,900 Q4: >\$28,900



3/5/2024

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17

# **Figure 5**. Probability of Experiencing Economic Hardship Among Gig and Non-Gig Workers by Number of Dependent Children in the Household.



#### Unpaid CC Balance, past 3 mo



#### Job/Income Loss Due to COVID



#### Food Insecurity, past 3mo



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3/5/2024 18

Figure 6. Probability of Experiencing Economic Hardship among Gig and Non-Gig Workers by Number of Dependent Children in the Household.



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19

## Self-Employed and Gig Workers Only

**Figure 7**. Probability of Experiencing Economic Hardship Among Gig and Self-Employed Workers.



**Figure 8**. Probability of Experiencing Economic Hardship Among Gig and Self-Employed Workers.



# Conclusion and Policy Implications

- Economic hardship is starker among gig workers compared with non-gig and other self-employed workers.
- Gig economy is associated with distinct labor mechanisms.
- These findings highlight the protective power of wealth.
- Labor market stability and family economic wellbeing are interconnected issues.
- Gig workers experience starker economic hardships despite that the pool of workers eligible for unemployment benefits was broadened to include gig workers.

## Thank You

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24

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## Q&A Gig work and Entrepreneurship

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3/5/2024 26

		U.S.
	Analytic Populatio	
	Sample	(2019)
Characteristic		
Age (mean)	46.9	48.2
Female	50.1	51.3
Married	51.4	49.0
Has Children<18	25.9	26.5
Bachelor's Degree or Higher	57.7	33.1
Employed (Full-Time or Part-Time)	66.3	60.7
2019 Household Income (mean)	87012.0	92324.0
Race		
White	76.5	72.0
Asian	5.9	5.7
Black	13.3	12.8
Native American/Pacific Islander	2.5	1.1
Other	4.1	8.5
Hispanic	17.4	18.4
Region		
Northeast	20.3	17.1
Midwest	20.7	20.8
South	34.8	38.3
West	23.1	23.9
Observations	4756	

#### Analytic Sample Comparison to U.S. Census Data

Note: U.S. population characteristics are drawn from the 2019 1-year estimates of American Community Survey data.

3/5/2024 27

## Method

- Generalized boosted regression modeling (utilizing machine learning) to account for potential endogeneity.
- We then estimated the relationship between gig employment and household hardship outcomes using linear probability models of the following general form:
- $y_i = \beta_0 + \beta_1 Gig_W orker_i + \beta_2 Liq_Q uartile_i + \beta_3 Num_K ids_i + \gamma_i \pi + \varepsilon_i$

### Conceptual Difference Between Entrepreneurial Self-Employment and Gig Work

• Entrepreneurial self-employed workers have full autonomy over the direction of their work.

• Entrepreneurial self-employed stands to benefit the most from the success of their self-employment activity.

Control Variables			
Gender	Emp., Self-Emp. Full-Time		
Race	Emp., Self-Emp. Part-Time		
Age	Emp., Wage/Salary Full-Time		
Education	Emp., Wage/Salary Part-Time		
Housing, Own w/ Mortgage	Partner Emp., Self-Emp. Full-Time		
Housing, Own In Full	Partner Emp., Self-Emp. Part-Time		
Housing, Rent	Partner Emp., Wage/Salary Full-Time		
Housing, Neither Own/Rent	Partner Emp., Wage/Salary Part-Time		
Health Insurance	Partner Emp., Unemp.		
HH Income in 2019	Partner Emp., Retired/Disabled		
Region, Midwest	Partner Emp., Single		
Region, Northeast	Region, West		
Region, South	Region, Other		

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## **Research Questions**

- To what extent did gig workers experience economic hardship during the COVID-19 pandemic compared with non-gig and other self-employed workers?
- To what extent did the level of household financial endowment prior to the COVID-19 pandemic influence the degree to which gig and non-gig workers experienced economic hardship during the COVID-19 pandemic?
- To what extent did household composition, such as the presence of dependent children in the household, influence the degree to which gig and non-gig workers experienced economic hardship during the COVID-19 pandemic?

## Hypotheses

• Gig workers experience greater economic hardship during the COVID-19 pandemic compared with non-gig and other self-employed workers.

• Household financial endowment prior to the COVID-19 pandemic reduces the likelihood of experiencing economic hardship, whereas the presence of dependent children in the household increases this likelihood.

*Financial Distress Outcomes Attributed to COVID-19, by Gig Employment Status (Propensity Score Weighted)* 

			Type of Gig Work			
			Offline			
				Online	Sales/Misc.	
	No Gig	Any Gig	Services	Tasks	Activities	Multiple
	Work	Work	Only	Only	Only	Gig Jobs
Outcome	(n=3,398)	(n=1,358)	(n=409)	(n=315)	(n=104)	(n=530)
Lost Job/Income due to COVID-19	0.26	0.35***	0.31*	0.3*	0.33	0.43***
Skipped Bills due to COVID-19	0.05	0.14***	0.1**	0.1	0.09	0.22***
Behind on Credit Card due to COVID-19	0.02	0.09***	0.04**	0.05**	0.01	0.19***
Skipped Housing due to COVID-19	0.02	0.08***	0.05	0.03	0.06	0.17***
Food Insecurity due to COVID-19	0.18	0.3***	0.21**	0.26*	0.18	0.47***
Notes: All significance tests conducted in relation to "No Gig Work" category. * $p < 0.05$ ; ** $p < 0.01$ ; *** $p < 0.01$ ; *						
0.001.						

Figure A1. Probability of Experiencing Economic Hardship Among Gig and Non-Gig Workers by Age of Dependent Children in the Household.



Behind on CC, now

Unpaid CC Balance, past 3 mo



Child Age



Food Insecurity, past 3mo



3/5/2024 34 Figure A2. Probability of Experiencing Economic Hardship Among Gig and Non-Gig Workers by Age of Dependent Children in the Household.



3/5/2024 35

## Measuring Self-Employed

Which best describes your <u>current</u> employment situation?

- Self-employed, full-time for pay (include contract work) (1)
- Self-employed, part-time for pay (include contract work) (2)
- Work full-time for pay (3)
- Work part-time for pay (4)
- Temporarily laid off (furloughed) (5)
- Unemployed, looking for work (6)
- Unemployed, not looking for work (7)
- Disabled, not able to work (8)
- Retired (9)
- Homemaker (10)
- Full-time caregiver for family member (11)
- Prefer not to say (12)
### Gig work: We use the Fed's two questions on gig work.

In the past 3 months, has anyone in your household been paid for any of the following activities? (Yes/No)

- Child or elder care services
- Dog walking, feeding pets, or house sitting
- House cleaning, yard work, or other property maintenance work
- Driving or ride-sharing, such as with Uber or Lyft
- Paid tasks online, such as posting YouTube videos (do not include online surveys)
- Other personal tasks, such as deliveries, grocery shopping, running errands, or helping people move

In addition, has anyone in your household been paid for any of the following activities <u>in the past 3</u> <u>months</u>? (Yes/No)

- Selling goods yourself at flea markets or garage sales
- Selling goods at consignment shops or thrift stores
- Selling goods on-line, such as on eBay, Craigslist, or Etsy
- Renting out property, such as your car or your house
- Any other paid activities that you have not already mentioned in the survey (do not include online surveys)

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Characteristic	Sample	<b>Gig Worker</b>	Non-Gig Worker	Differences	p-value
Housing, Own w/ Mortgage (%)	0.40	0.39	0.40	-0.02	0.000
Housing, Own In Full (%)	0.27	0.23	0.29	-0.13	0.000
Housing, Rent (%)	0.28	0.30	0.27	0.07	0.000
Housing, Neither Own/Rent (%)	0.05	0.07	0.04	0.16	0.000
Health Insurance, Yes (%)	0.94	0.93	0.95	-0.11	0.001
	87012.0	84573.6			
HH Income in 2019, Mean (SD)	(70250.7)	(73882.8)	87986.5 (68732.6)	-0.05	0.142
HH Income in 2019, Median	70000.0	65000.0	72000.0		
	27469.9	25392.0			
Liq. Assets pre-COVID, Mean (SD)	(59481.7)	(58532.2)	28300.3 (59845.3)	0.00	0.963
Liq. Assets pre-COVID, Median	4450.0	4450.0	6000.0		
Emp., Self-Emp. Full-Time (%)	0.07	0.12	0.05	0.27	0.000
Emp., Self-Emp. Part-Time (%)	0.03	0.06	0.03	0.19	0.000
Emp., Wage/Salary Full-Time (%)	0.45	0.45	0.46	-0.01	0.000
Emp., Wage/Salary Part-Time (%)	0.10	0.13	0.09	0.15	0.000
Partner Emp., Self-Emp. Full- <u>Time (%)</u>	0.05	0.09	0.01	0.09	0.000
Partner Emp., Self-Emp. Part-Time (%)	0.01	0.02	0.40	0.09	0.000

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	Sample	Gig Worker	Non-Gig		
Characteristic (%)	(4,756)	(1,358)	Worker (3,398)	Differences	p-value
Total (%)	100	28.55	71.45		
Age, Mean (SD)	46.9 (16.8)	39.11 (15.32)	50 (16.4)	-0.65	0.000
Gender, Female	0.50	0.51	0.50	0.03	0.309
Race, Asian (%)	0.05	0.04	0.06	-0.09	0.000
Race, Black (%)	0.12	0.09	0.14	-0.16	0.000
Race, Hispanic (%)	0.17	0.17	0.17	0.00	0.000
Race, White (%)	0.62	0.67	0.60	0.15	0.000
Race, Other (%)	0.03	0.03	0.03	-0.02	0.000
Children, 0 (%)	0.74	0.63	0.78	-0.35	0.000
Children, 1 (%)	0.13	0.17	0.11	0.18	0.000
Children, 2 (%)	0.10	0.15	0.08	0.25	0.000
Children, 3+ (%)	0.03	0.05	0.03	0.11	0.000
Ed, Less than HS (%)	0.01	0.01	0.01	0.01	0.003
Ed, HS Degree (%)	0.28	0.31	0.27	0.09	0.003
Ed, Bachelor's Degree	0.36	0.37	0.36	0.04	0.003
Ed, Graduate Degree	0.22	0.19	0.23	-0.08	0.003

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Characteristic	Sample	Gig	Non-Gig Worker	Differences	p-value
		Worker			
Partner Emp., Wage/Salary Full-Time (%)	0.27	0.27	0.27	-0.01	0.000
Partner Emp., Wage/Salary Part-Time (%)	0.04	0.04	0.05	-0.02	0.000
Partner Emp.,Unemp. (%)	0.09	0.09	0.09	0.00	0.000
Partner Emp., Retired/Disabled (%)	0.13	0.06	0.16	-0.29	0.000
Partner Emp., Single (%)	0.41	0.44	0.40	0.09	0.000
Region, Midwest	0.21	0.20	0.21	-0.02	0.948
Region, Northeast	0.20	0.20	0.21	-0.02	0.948
Region, South	0.35	0.35	0.35	0.01	0.948
Region, West	0.23	0.24	0.23	0.02	0.948
Region, Other	0.01	0.01	0.01	0.01	0.948
Observations	4756	1358	3398		

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#### **Table 4.** Linear Probability Regressions of Household Hardship on Gig Work

Notes: Suppressed demographic controls include age, gender, marital/partner status, and education. Suppressed financial controls include spousal employment, health insurance, homeownership, vehicle ownership, and bank account ownership. Suppressed geographic controls include Census region.

et ome, ID *** 6) 2 1) *** 2)	Skipped Bills, past 3 months 0.094*** (0.011) -0.111**	Behind on CC, Now 0.090*** (0.010)	Unpaid CC Balance, Now 0.043** (0.015)	Skipped Housing, past 3 months 0.069*** (0.010)	Evicted, past 3 months 0.044*** (0.005)	Skipped Medical, past 3 months 0.095***	In: 1 0.
*** 6) 2 1) *** 2)	0.094*** (0.011) -0.111**	0.090*** (0.010)	0.043** (0.015)	0.069*** (0.010)	0.044*** (0.005)	0.095***	0.
*** 6) 2 1) *** 2)	0.094*** (0.011) -0.111**	0.090*** (0.010)	0.043** (0.015)	0.069*** (0.010)	0.044*** (0.005)	0.095***	0.
6) 2 1) *** 2)	(0.011) -0.111**	(0.010)	(0.015)	(0.010)	(0.005)	(0.011)	
2 1) *** 2)	-0.111**	0.100444				(0.011)	(
2 1) *** 2)	-0.111**	0 100 4444					
51) *** 2)	(0.020)	-0.132***	-0.153***	-0.134***	-0.100***	-0.171***	-0.
*** 2)	(0.036)	(0.031)	(0.042)	(0.031)	(0.021)	(0.034)	(
2)	-0.109***	-0.131***	-0.089**	-0.107***	-0.090***	-0.130***	-0.
-1	(0.025)	(0.024)	(0.030)	(0.023)	(0.018)	(0.025)	(
4*	-0.087**	-0.112***	-0.098**	-0.111***	-0.080***	-0.113***	-0.
-1)	(0.032)	(0.028)	(0.035)	(0.027)	(0.019)	(0.030)	(
***	-0.069*	-0.105***	-0.118***	-0.113***	-0.106***	-0.126***	-0.
7)	(0.031)	(0.027)	(0.033)	(0.026)	(0.018)	(0.029)	(
3*	-0.045*	-0.055***	-0.053*	-0.025	-0.027***	-0.043*	-0.
(1)	(0.017)	(0.015)	(0.022)	(0.015)	(0.008)	(0.017)	(
2	-0.113***	-0.108***	-0.189***	-0.082***	-0.045***	-0.091***	-0.
.2)	(0.014)	(0.013)	(0.022)	(0.013)	(0.007)	(0.015)	(
)5	-0.101***	-0.096***	-0.297***	-0.072***	-0.043***	-0.099***	-0.
4)	(0.014)	(0.013)	(0.022)	(0.014)	(0.008)	(0.015)	(
.0	0.055**	0.066***	0.057*	0.068***	0.045***	0.060**	0.
.3)	(0.018)	(0.016)	(0.023)	(0.016)	(0.011)	(0.019)	(
19	0.100***	0.103***	0.071**	0.067***	0.073***	0.051*	0.
.8)	(0.022)	(0.019)	(0.025)	(0.018)	(0.015)	(0.020)	(
4	0.079*	0.071*	0.069	0.052	0.043*	0.003	
	f Managantent	Assi (0.080) ofe	ssor (0f. 938)olog	y, F160:029Atlan	ntic (0.022)	(0.029)	1
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	Model 9	Model 10	Model 11	Model 12	Model 13	Model 14	Model 15	Model
	Lost	Skipped		Unpaid CC	Skipped		Skipped	Food
	Job/Income,	Bills, past 3	Behind on	Balance,	Housing, past	Evicted, past	Medical, past	Insecuri
	COVID	months	CC, Now	Now	3mo	3mo	3mo	past 3m
Gig Employment (Ref = Non-Gig Worker)								
Gig Worker	0.116***	0.131***	0.122***	0.029	0.110***	0.075***	0.147***	0.169*
	(0.026)	(0.024)	(0.020)	(0.026)	(0.021)	(0.012)	(0.023)	(0.027
Liquid Asset Quartile (Ref. = 1st Quartile) 2nd Quartile (\$2,001 -								
\$8,250)	0.055*	-0.038*	-0.040**	-0.048*	-0.007	-0.006	-0.015	-0.066
	(0.022)	(0.016)	(0.012)	(0.023)	(0.013)	(0.007)	(0.015)	(0.02)
3rd Quartile (\$8,251 -								
\$28,900)	0.066**	-0.078***	-0.075***	-0.225***	-0.045***	-0.020**	-0.052***	-0.132*
	(0.023)	(0.014)	(0.011)	(0.021)	(0.012)	(0.007)	(0.013)	(0.02
4th Quartile (\$28,901+)	0.025	-0.057***	-0.069***	-0.295***	-0.035**	-0.018*	-0.046***	-0.108*
	(0.024)	(0.014)	(0.011)	(0.021)	(0.013)	(0.007)	(0.013)	(0.020
Gig Worker-Liquid Asset Interaction Gig Worker*2Q Liquid								
Assets	-0.003	-0.013	-0.030	-0.011	-0.039	-0.044**	-0.057	-0.03
Gig Worker*3Q Liquid	(0.043)	(0.036)	(0.028)	(0.044)	(0.030)	(0.016)	(0.034)	(0.044
Assets	-0.091*	-0.071*	-0.067**	0.075	-0.076**	-0.051***	-0.080**	-0.05
	(0.042)	(0.029)	(0.025)	(0.041)	(0.024)	(0.014)	(0.029)	(0.03
Gig Worker*4Q Liquid								
Assets	-0.062	-0.090***	-0.057*	-0.003	-0.077***	-0.051***	-0.108***	-0.136*
	(0.033)	(0.028)	(0.016)	(0.023)	(0.024)	(0.011)	(0.026)	(0.032
Other Controls								
Demographic	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Financial	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Geographic	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	4756	4756	4756	4756	4756	4756	4756	4756
	0.101	0.156	0.1.1.6	0.014	0.110	0.1.42	0 1 0 5	0.000

**Table 5.** The Interaction between Gig Work, Financial Endowment, and Household Hardship (Linear Probability Models)Notes: Suppressed demographic controls include age, gender, marital/partner status, race/ethnicity, # of children, and education. Suppressed financial controls include householdincome in 2019, respondent employment status, spousal employment, health insurance, homeownership, vehicle ownership, and bank account ownership. Suppressed geographiccontrols include Census region.

Table 6. The Interaction between Gig Work, Children, and Household Hardship (Linear Probability Models)
Notes: Suppressed demographic controls include age, gender, marital/partner status, race/ethnicity, and education. Suppressed financial controls include household income in 2019.
liquid asset quartiles, respondent employment status, spousal employment, health insurance, homeownership, vehicle ownership, and bank account ownership. Suppressed
geographic controls include Census region.

Model 17	Model 18	Model 19	Model 20	Model 21	Model 22	Model 23	Model 24
Lost	Skipped		Unpaid CC	Skipped		Skipped	Food
Job/Income,	Bills, past	Behind on	Balance,	Housing,	Evicted, past	Medical, past	Insecurity,
COVID	3mo	CC, Now	Now	past 3mo	3mo	3mo	past 3mo
0.0==++++	0.055444			0.045444	0.04.64.44	0.0=1.4.4.4	0.000444
0.075***	0.066***	0.060***	0.017	0.046***	0.016***	0.071***	0.098***
(0.019)	(0.013)	(0.010)	(0.018)	(0.010)	(0.004)	(0.012)	(0.017)
0.006	0.022	0.015	0.022	0.025	-0.007	0.007	0.048
(0.027)	(0.017)	(0.014)	(0.023)	(0.014)	(0.005)	(0.017)	(0.025)
-0.004	0.031	0.026	0.010	0.023	0.015	0.003	0.027
(0.033)	(0.024)	(0.019)	(0.030)	(0.021)	(0.017)	(0.022)	(0.029)
-0.027	0.000	0.047	-0.010	0.006	0.001	-0.000	-0.039
(0.047)	(0.032)	(0.032)	(0.045)	(0.026)	(0.015)	(0.031)	(0.042)
0.029	0.070*	0.104**	0.071	0.090**	0.108***	0.109**	0.054
(0.045)	(0.035)	(0.032)	(0.043)	(0.032)	(0.022)	(0.036)	(0.042)
0.027	0.140***	0.155***	0.124**	0.090*	0.117***	0.097**	0.125**
(0.051)	(0.041)	(0.036)	(0.046)	(0.035)	(0.031)	(0.037)	(0.045)
0.026	0.160*	0.051	0.160*	0.093	0.088*	0.009	0.153*
(0.078)	(0.066)	(0.058)	(0.073)	(0.056)	(0.041)	(0.055)	(0.070)
Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
4756	4756	4756	4756	4756	4756	4756	4756
0.123	0.158	0.153	0.216	0.111	0.157	0.137	0.222
	Model 17           Lost           Job/Income,           COVID           0.075***           (0.019)           0.006           (0.027)           -0.004           (0.033)           -0.027           (0.047)           0.029           (0.051)           0.026           (0.078)           Yes           Yes           Yes           Yes           Yes           Yes           0.123	Model 17Model 18LostSkippedJob/Income, COVIDBills, past 3mo $0.075^{***}$ $0.066^{***}$ $(0.019)$ $0.006$ $0.022$ $(0.017)$ $0.006$ $0.022$ $(0.027)$ $(0.027)$ $(0.017)$ $-0.004$ $0.033)$ $(0.024)$ $-0.027$ $0.000$ $(0.047)$ $(0.032)$ $0.029$ $0.070^*$ $(0.045)$ $(0.045)$ $(0.035)$ $0.027$ $0.140^{***}$ $(0.051)$ $(0.041)$ $0.026$ $0.160^*$ $(0.078)$ $(0.066)$ YesYes Yes Yes YesYesYes Yes Yes YesYesYes <td>Model 17Model 18Model 19LostSkippedJob/Income, COVIDBills, pastBehind on CC, Now<math>0.075^{***}</math><math>0.066^{***}</math><math>0.060^{***}</math><math>(0.019)</math><math>(0.013)</math><math>(0.010)</math><math>0.006</math><math>0.022</math><math>0.015</math><math>(0.027)</math><math>(0.017)</math><math>(0.014)</math><math>-0.004</math><math>0.031</math><math>0.026</math><math>(0.033)</math><math>(0.024)</math><math>(0.019)</math><math>-0.027</math><math>0.000</math><math>0.047</math><math>(0.047)</math><math>(0.032)</math><math>(0.032)</math><math>0.029</math><math>0.070^*</math><math>0.104^{**}</math><math>(0.045)</math><math>(0.035)</math><math>(0.032)</math><math>0.027</math><math>0.140^{***}</math><math>0.155^{***}</math><math>(0.051)</math><math>(0.041)</math><math>(0.036)</math><math>0.026</math><math>0.160^*</math><math>0.051</math><math>(0.078)</math><math>(0.066)</math><math>(0.058)</math>YesYesYesYesYesYesYesYesYesYesYesYesYesYesYes0.123<math>0.158</math><math>0.153</math></td> <td>Model 17Model 18Model 19Model 20LostSkippedUnpaid CCJob/Income, COVIDBills, pastBehind on SmoBalance, Now<math>0.075^{***}</math><math>0.066^{***}</math><math>0.060^{***}</math><math>0.017</math><math>(0.019)</math><math>(0.013)</math><math>(0.010)</math><math>(0.018)</math><math>0.006</math><math>0.022</math><math>0.015</math><math>0.022</math><math>(0.027)</math><math>(0.017)</math><math>(0.014)</math><math>(0.023)</math><math>-0.004</math><math>0.031</math><math>0.026</math><math>0.010</math><math>(0.033)</math><math>(0.024)</math><math>(0.019)</math><math>(0.030)</math><math>-0.027</math><math>0.000</math><math>0.047</math><math>-0.010</math><math>(0.047)</math><math>(0.032)</math><math>(0.032)</math><math>(0.045)</math><math>0.029</math><math>0.070^*</math><math>0.104^{**}</math><math>0.071</math><math>(0.045)</math><math>(0.035)</math><math>(0.032)</math><math>(0.043)</math><math>0.027</math><math>0.140^{***}</math><math>0.155^{***}</math><math>0.124^{**}</math><math>(0.051)</math><math>(0.041)</math><math>(0.058)</math><math>(0.073)</math>Yes<td>Model 17Model 18Model 19Model 20Model 21LostSkippedUnpaid CCSkippedJob/Income,Bills, pastBehind onBalance,Housing,COVID3moCC, NowNowpast 3mo0.075***0.066***0.060***0.0170.046***(0.019)(0.013)(0.010)(0.018)(0.010)0.0060.0220.0150.0220.025(0.027)(0.017)(0.014)(0.023)(0.014)-0.0040.0310.0260.0100.023(0.033)(0.024)(0.019)(0.030)(0.021)-0.0270.0000.047-0.0100.006(0.047)(0.032)(0.032)(0.045)(0.026)0.0290.070*0.104**0.0710.090**(0.045)(0.035)(0.032)(0.043)(0.032)0.0270.140***0.155***0.124**0.090*(0.051)(0.041)(0.036)(0.046)(0.035)0.0260.160*0.0510.160*0.093(0.078)(0.066)(0.058)(0.073)(0.056)Yes<!--</td--><td>Model 17Model 18Model 19Model 20Model 21Model 22LostSkippedUnpaid CCSkippedJob/Income,Bills, pastBehind onBalance,Housing,Evicted, past0.075***0.066***0.060***0.0170.046***0.016***(0.019)(0.013)(0.010)(0.018)(0.010)(0.004)0.0060.0220.0150.0220.025-0.007(0.027)(0.017)(0.014)(0.023)(0.014)(0.005)-0.0040.0310.0260.0100.0230.015(0.033)(0.024)(0.019)(0.030)(0.021)(0.017)-0.0270.0000.047-0.0100.0060.001(0.047)(0.032)(0.032)(0.045)(0.026)(0.015)0.0290.070*0.104**0.0710.090**0.108***(0.045)(0.035)(0.032)(0.043)(0.032)(0.022)0.0270.140***0.155***0.124**0.090*0.117***(0.051)(0.041)(0.036)(0.046)(0.035)(0.031)0.0260.160*0.0510.160*0.0930.088*(0.078)(0.066)(0.058)(0.073)(0.056)(0.041)Yes&lt;</td><td><math display="block">\begin{array}{c c c c c c c c c c c c c c c c c c c </math></td></td></td>	Model 17Model 18Model 19LostSkippedJob/Income, COVIDBills, pastBehind on CC, Now $0.075^{***}$ $0.066^{***}$ $0.060^{***}$ $(0.019)$ $(0.013)$ $(0.010)$ $0.006$ $0.022$ $0.015$ $(0.027)$ $(0.017)$ $(0.014)$ $-0.004$ $0.031$ $0.026$ $(0.033)$ $(0.024)$ $(0.019)$ $-0.027$ $0.000$ $0.047$ $(0.047)$ $(0.032)$ $(0.032)$ $0.029$ $0.070^*$ $0.104^{**}$ $(0.045)$ $(0.035)$ $(0.032)$ $0.027$ $0.140^{***}$ $0.155^{***}$ $(0.051)$ $(0.041)$ $(0.036)$ $0.026$ $0.160^*$ $0.051$ $(0.078)$ $(0.066)$ $(0.058)$ YesYesYesYesYesYesYesYesYesYesYesYesYesYesYes0.123 $0.158$ $0.153$	Model 17Model 18Model 19Model 20LostSkippedUnpaid CCJob/Income, COVIDBills, pastBehind on SmoBalance, Now $0.075^{***}$ $0.066^{***}$ $0.060^{***}$ $0.017$ $(0.019)$ $(0.013)$ $(0.010)$ $(0.018)$ $0.006$ $0.022$ $0.015$ $0.022$ $(0.027)$ $(0.017)$ $(0.014)$ $(0.023)$ $-0.004$ $0.031$ $0.026$ $0.010$ $(0.033)$ $(0.024)$ $(0.019)$ $(0.030)$ $-0.027$ $0.000$ $0.047$ $-0.010$ $(0.047)$ $(0.032)$ $(0.032)$ $(0.045)$ $0.029$ $0.070^*$ $0.104^{**}$ $0.071$ $(0.045)$ $(0.035)$ $(0.032)$ $(0.043)$ $0.027$ $0.140^{***}$ $0.155^{***}$ $0.124^{**}$ $(0.051)$ $(0.041)$ $(0.058)$ $(0.073)$ Yes <td>Model 17Model 18Model 19Model 20Model 21LostSkippedUnpaid CCSkippedJob/Income,Bills, pastBehind onBalance,Housing,COVID3moCC, NowNowpast 3mo0.075***0.066***0.060***0.0170.046***(0.019)(0.013)(0.010)(0.018)(0.010)0.0060.0220.0150.0220.025(0.027)(0.017)(0.014)(0.023)(0.014)-0.0040.0310.0260.0100.023(0.033)(0.024)(0.019)(0.030)(0.021)-0.0270.0000.047-0.0100.006(0.047)(0.032)(0.032)(0.045)(0.026)0.0290.070*0.104**0.0710.090**(0.045)(0.035)(0.032)(0.043)(0.032)0.0270.140***0.155***0.124**0.090*(0.051)(0.041)(0.036)(0.046)(0.035)0.0260.160*0.0510.160*0.093(0.078)(0.066)(0.058)(0.073)(0.056)Yes<!--</td--><td>Model 17Model 18Model 19Model 20Model 21Model 22LostSkippedUnpaid CCSkippedJob/Income,Bills, pastBehind onBalance,Housing,Evicted, past0.075***0.066***0.060***0.0170.046***0.016***(0.019)(0.013)(0.010)(0.018)(0.010)(0.004)0.0060.0220.0150.0220.025-0.007(0.027)(0.017)(0.014)(0.023)(0.014)(0.005)-0.0040.0310.0260.0100.0230.015(0.033)(0.024)(0.019)(0.030)(0.021)(0.017)-0.0270.0000.047-0.0100.0060.001(0.047)(0.032)(0.032)(0.045)(0.026)(0.015)0.0290.070*0.104**0.0710.090**0.108***(0.045)(0.035)(0.032)(0.043)(0.032)(0.022)0.0270.140***0.155***0.124**0.090*0.117***(0.051)(0.041)(0.036)(0.046)(0.035)(0.031)0.0260.160*0.0510.160*0.0930.088*(0.078)(0.066)(0.058)(0.073)(0.056)(0.041)Yes&lt;</td><td><math display="block">\begin{array}{c c c c c c c c c c c c c c c c c c c </math></td></td>	Model 17Model 18Model 19Model 20Model 21LostSkippedUnpaid CCSkippedJob/Income,Bills, pastBehind onBalance,Housing,COVID3moCC, NowNowpast 3mo0.075***0.066***0.060***0.0170.046***(0.019)(0.013)(0.010)(0.018)(0.010)0.0060.0220.0150.0220.025(0.027)(0.017)(0.014)(0.023)(0.014)-0.0040.0310.0260.0100.023(0.033)(0.024)(0.019)(0.030)(0.021)-0.0270.0000.047-0.0100.006(0.047)(0.032)(0.032)(0.045)(0.026)0.0290.070*0.104**0.0710.090**(0.045)(0.035)(0.032)(0.043)(0.032)0.0270.140***0.155***0.124**0.090*(0.051)(0.041)(0.036)(0.046)(0.035)0.0260.160*0.0510.160*0.093(0.078)(0.066)(0.058)(0.073)(0.056)Yes </td <td>Model 17Model 18Model 19Model 20Model 21Model 22LostSkippedUnpaid CCSkippedJob/Income,Bills, pastBehind onBalance,Housing,Evicted, past0.075***0.066***0.060***0.0170.046***0.016***(0.019)(0.013)(0.010)(0.018)(0.010)(0.004)0.0060.0220.0150.0220.025-0.007(0.027)(0.017)(0.014)(0.023)(0.014)(0.005)-0.0040.0310.0260.0100.0230.015(0.033)(0.024)(0.019)(0.030)(0.021)(0.017)-0.0270.0000.047-0.0100.0060.001(0.047)(0.032)(0.032)(0.045)(0.026)(0.015)0.0290.070*0.104**0.0710.090**0.108***(0.045)(0.035)(0.032)(0.043)(0.032)(0.022)0.0270.140***0.155***0.124**0.090*0.117***(0.051)(0.041)(0.036)(0.046)(0.035)(0.031)0.0260.160*0.0510.160*0.0930.088*(0.078)(0.066)(0.058)(0.073)(0.056)(0.041)Yes&lt;</td> <td><math display="block">\begin{array}{c c c c c c c c c c c c c c c c c c c </math></td>	Model 17Model 18Model 19Model 20Model 21Model 22LostSkippedUnpaid CCSkippedJob/Income,Bills, pastBehind onBalance,Housing,Evicted, past0.075***0.066***0.060***0.0170.046***0.016***(0.019)(0.013)(0.010)(0.018)(0.010)(0.004)0.0060.0220.0150.0220.025-0.007(0.027)(0.017)(0.014)(0.023)(0.014)(0.005)-0.0040.0310.0260.0100.0230.015(0.033)(0.024)(0.019)(0.030)(0.021)(0.017)-0.0270.0000.047-0.0100.0060.001(0.047)(0.032)(0.032)(0.045)(0.026)(0.015)0.0290.070*0.104**0.0710.090**0.108***(0.045)(0.035)(0.032)(0.043)(0.032)(0.022)0.0270.140***0.155***0.124**0.090*0.117***(0.051)(0.041)(0.036)(0.046)(0.035)(0.031)0.0260.160*0.0510.160*0.0930.088*(0.078)(0.066)(0.058)(0.073)(0.056)(0.041)Yes<	$\begin{array}{c c c c c c c c c c c c c c c c c c c $

\* p < 0.05; \*\* p < 0.01; \*\*\* p < 0.001.

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	Model 25	Model 26	Model 27	Model 28	Model 29	Model 30	Model 31	Model 32
					Skipped		Skipped	Food
	Lost	Skipped		Unpaid CC	Housing,	Evicted,	Medical,	Insecurity,
	Job/Income,	Bills, past	Behind on	Balance,	past 3	past 3	past 3	past 3
	COVID	3months	CC, Now	Now	months	months	months	months
Gig Employment (Ref = Non-Gig Worker)								
Gig Worker	0.104*	0.147***	0.223***	0.142***	0.132***	0.132***	0.176***	0.187***
	(0.042)	(0.031)	(0.027)	(0.036)	(0.028)	(0.021)	(0.031)	(0.035)
Liquid Asset Quartile (Ref. = 1st Quartile, \$0 - \$2,000)								
2nd Quartile (\$2,001 - \$8,250)	0.012	-0.037	-0.027	-0.046	-0.037	-0.059	-0.068	-0.076
	(0.058)	(0.047)	(0.041)	(0.053)	(0.043)	(0.033)	(0.046)	(0.052)
3rd Quartile (\$8,251 - \$28,900)	-0.001	-0.159***	-0.187***	-0.196***	-0.157***	-0.151***	-0.152***	-0.164**
	(0.060)	(0.041)	(0.034)	(0.054)	(0.036)	(0.026)	(0.043)	(0.052)
4th Quartile (\$28,901+)	-0.004	-0.126**	-0.152***	-0.386***	-0.129**	-0.117***	-0.187***	-0.196***
	(0.062)	(0.047)	(0.037)	(0.052)	(0.041)	(0.034)	(0.045)	(0.051)
Number of Children (Ref. $= 0$	( , , , ,		(*****)	( , , , ,	()	()	( )	(, , , ,
Children)								
1 Child	0.088	0.129**	0.148***	0.143**	0.112**	0.093**	0.157**	0.185***
	(0.056)	(0.045)	(0.043)	(0.050)	(0.040)	(0.033)	(0.049)	(0.048)
2 Children	0.126	0.202***	0.288***	0.208***	0.174**	0.166***	0.188**	0.222***
	(0.068)	(0.061)	(0.054)	(0.058)	(0.054)	(0.050)	(0.059)	(0.057)
3+ Children	-0.119	0.100	0.026	0.033	0.106	0.043	0.052	0.128
	(0.098)	(0.074)	(0.063)	(0.079)	(0.078)	(0.050)	(0.078)	(0.075)
Other Controls								
Demographic	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Geographic	Yes	Yes	Yes	Yes	Yes	Yes	Yes	Yes
Observations	676	676	676	676	676	676	676	676
R-squared	0.101	0.226	0.356	0.293	0.257	0.328	0.257	0.334
* = < 0.05, $* * = < 0.01$ , $* * * = < 0.001$	N ( ( ( ( ( ( ( ( ( ( ( ( ( ( ( ( ( ( (		000000000000000000000000000000000000000	2001/09/2011/09/2012	1/2012/01/12/02/02/02/02	111111111111111111111111111111111111111	100000000000000000000000000000000000000	7755007777777777777777777

**Table 7.** Linear Probability Regression of Household Hardship on Gig Work and Among Self-Employed HouseholdsNotes: Suppressed demographic controls include age, gender, marital/partner status, and education. Suppressed financial controls include spousal employment, health insurance,<br/>homeownership, vehicle ownership, and bank account ownership. Suppressed geographic controls include Census region.

\* p < 0.05; \*\* p < 0.01; \*\*\* p < 0.001.

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## Q&A Race and Entrepreneurial Return to Education

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3/5/2024 45

# **Table 2:** White-Black Mean Differences(Entrepreneurs)

	Mean(white)	Mean(Black)	Diff.	Std. Error	Obs.
Business profit	22,798.92	8,889.89	13,909.03***	3,401.60	9,258
Positive profit	0.56	0.37	0.19***	0.02	9,258
Profit \$10k+	0.42	0.26	0.16***	0.02	9,258
Household student loan debt	7,538.58	12,158.94	-4,620.36***	1,064.55	9,258
Household Net worth	1,522,849.56	599,739.88	923,109.68***	280,015.93	9,258
High school degree or less	0.36	0.40	-0.04**	0.02	9,258
Some college/associate degree	0.27	0.33	-0.06***	0.02	9,258
College degree	0.22	0.15	0.07***	0.02	9,258
Graduate degree	0.15	0.12	0.03**	0.01	9,258

# **Figure 1**: Probability of Making a Positive Profit by Race and Education Levels.







Predicted Probability of Business Ownership by Household Value of Net Worth for Black and White.



3/5/2024 50

University

# Q&A Income Inequality and Entrepreneurship

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#### Inequality Weakens Entrepreneurship More at the Top Income Strata





#### Inequality Increases Necessity-motivated Entrepreneurship More Lower-Income Strata.



### Inequality Increases Necessity-motivated Entrepreneurship More Among Lower-Education Strata.







Pairwise correlations		1	2	3
Entrepreneurial entry	1	N/A	N/A	
Mativated by look of botton antion for concreting income	2			
Motivated by lack of better option for generating income	L	IN/A	IN/A	
Income (Upper 33%tile)	3	0.0540***	-0.175***	
Educational attainment (postsecondary Graduate Education)	4	0.024**	-0.189***	0.283***

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**Figure 1:** Predicted Probability of Entrepreneurial Entry for three Income Categories, by Societal level Income Inequality (95% CI).



<sup>3/5/2024 62</sup> 

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### Figure 2: Predicted Probability of Entrepreneurial Entry for the three Education Categories, by



**Figure 4:** Predicted Probability of Necessity-motivated Entrepreneurship for the three Education Categories, by Societal level Income Inequality (95% CI).



**Table 3:** Mixed-effects Regression Estimates of the Moderating Effect of Inequality on the Effect of Education and Income on the Log-odds of Entrepreneurial Entry.

		Income	Education
		X Gini	X Gini
	(1)	(2)	(3)
	*		
Constant	-4.043*	1.976	1.954
	(1.979)	(2.401)	(2.414)
Middle 33 <sup>rd</sup> income percentile	0.124	0.108	0.123
	(0.0241)	(0.122)	(0.0241)
Upper 33 <sup>rd</sup> income percentile	0.308***	0.626***	0.305***
	(0.0246)	(0.118)	(0.0247)
Secondary education	0.123***	0.126***	$0.618^{***}$
	(0.0242)	(0.0242)	(0.127)
Postsecondary and graduate education	0.281***	$0.288^{***}$	0.945***
	(0.0254)	(0.0255)	(0.124)
Gini Coefficient		-0.0612***	-0.0589**
		(0.0130)	(0.0130)
Cross-level interactions			
Middle 33 <sup>rd</sup> income percentile X Gini Coefficient		0.000299	
-		(0.00285)	
Upper 33 <sup>rd</sup> income percentile X Gini Coefficient		-0.00790**	
		(0.00279)	
Secondary education X Gini Coefficient			-0.0115**
			(0.00291
Postsecondary and graduate education X Gini Coefficient			-0.0158**
			(0.00290
Individual controls	Yes	Yes	Yes
Country level controls	Yes	Yes	Yes
	(1.236)	(1.310)	(1.322)
Between-country intercept s. d.	1.058	1.211	1.236
	(0.272)	(0.297)	(0.302)
Observations	142,840	142,840	142,840
Number countries	50	50	50
Log likelihood	-43,867.3	-43.847.5	-43,838.
			,

Daniel Auguste | MLK Visiting Assistant Professor, MIT Sloan School of Management Assistant Professor of Sociology, Florida Atlantic University p < 0.05, p < 0.01, p < 0.001

<sup>66</sup> 

			Income X Gini	Education X Gini
		(4)	(5)	(6)
	Constant	$5.741^{**}$	$5.016^{*}$	$5.368^{*}$
	Middle 33 <sup>rd</sup> income percentile	-0.362***	0.0791	-0.361***
	Upper 33 <sup>rd</sup> income percentile	(0.0493) -0.881 <sup>***</sup>	(0.272) -0.476	(0.0493) -0.876 <sup>***</sup>
		(0.0527)	(0.278)	(0.0529)
	Secondary education	-0.293*** (0.0491)	-0.287*** (0.0492)	-0.461 (0.277)
	Postsecondary and graduate education	-0.727***	-0.722***	-0.0914
	Gini Coefficient	(0.0562)	(0.0563) 0.0115 (0.0141)	(0.282) 0.00621 (0.0141)
	<i>Cross-level interactions</i> Middle 33 <sup>rd</sup> income percentile X Gini Coefficient		-0.0102	(0.01.1)
	Upper 33 <sup>rd</sup> income percentile X Gini Coefficient		(0.00621) -0.00945	
	Secondary education X Gini Coefficient		(0.00640)	0.00394
	Postsecondary and graduate education X Gini Coefficient			-0.0152* (0.00650)
	Individual controls	Yes	Yes	Yes
	Country level controls	Yes	Yes	Yes
	Between-country intercept s. d.	0.359*** (0.101)	0.350 <sup>***</sup> (0.0988)	0.360*** (0.102)
	Observations	15,760	15,760	15,760
	Number countries	50	50	50
	Log likelihood Chi-squared	-8,170.6 808.5	-8,169.0 812.7	-8,165.4 814.1
Daniel Auguste   MLK Vis. University	iting Assistant Professor, MIT Sloan School and Antagement Assistant $p < 0.05$ , $p < 0.01$ , $p < 0.01$	fessor of Sociol .001	ogy, Florida Atla	ntic

**Table 4:** Mixed-effects Regression Estimates of the Moderating Effect of Inequality on the Effect of Education and Income on the Log-odds of Necessity Entrepreneurship.

3/5/2024	67
5/5/2021	0

University

Effect of Income Inequality on Entrepreneurial Entry by Government Spending in Education (in Log-odds)



University



Effect of Poverty on Entrepreneurial Entry by Government Spending in Education (in Log-odds)



Predicted probabilities of business ownership for men and women, by country-level belief that men have more right to jobs.




Predicted probabilities of Necessity motivated entrepreneurship by Country-level Economic Inequality.





University

