



# The labour market impact of COVID-19 lockdowns: Evidence from Ghana

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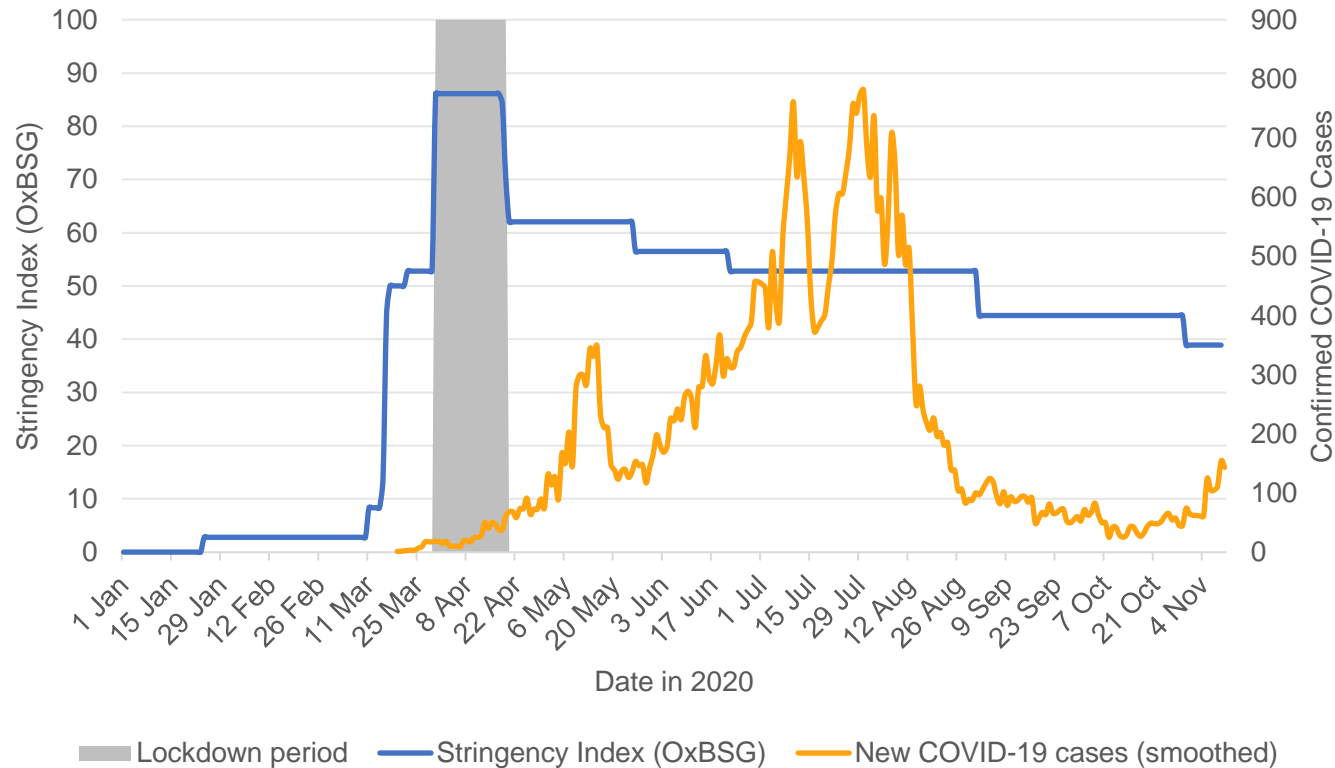
# Motivation

The COVID-19 pandemic has posed important risks not only for people's health but also economic wellbeing. Workplace closures and restrictions on movements have affected the livelihoods of workers in Ghana.

- How have the COVID-19 pandemic and related government response measures affected workers in Ghana?
- To what extent have workers been affected differently, depending on...
  - ... the type of government response measures in place?
  - ... the type of activities that they engage in?

# Background: Partial lockdown in Ghana

## COVID-19 cases and government response stringency index



Note: Based on Hale et al. (2020) and Roser et al. (2020).

The first two cases of COVID-19 were reported in Ghana on 12 March 2020.

As a first response, on 15<sup>th</sup> March, all public gatherings were banned, and all schools and universities were closed. On 23<sup>rd</sup> March, all borders were closed.

On 30<sup>th</sup> March, restrictions on movement of persons were implemented in the **Greater Accra** and **Greater Kumasi Metropolitan Areas** and **contiguous districts**, identified by the Ghana Health Service as **'hotspots.'** This was lifted after **three weeks** on 20<sup>th</sup> April, while the other measures remained in force.

# Data

- Ghana: Socioeconomic Panel Survey – COVID-19 Rapid Survey
  - Joint effort between UNU-WIDER and the Institute of Statistical, Social and Economic Research (ISSER), at the University of Ghana, Legon.
  - Three previous waves of panel data (2009/10, 2013/14, 2018/19).
  - Conducted phone surveys between 19 August and 17 September 2020 with ca. 670 respondents who were working in last panel wave.
  - Retrospective questions regarding the situation in February and April 2020.
- Identify the effect of strict lockdown policies by comparing the labour market outcomes in areas with different policy responses.

# Methodology

Difference-in-differences (DID) design: Comparison between changes in these labour market outcomes among respondents located in **lockdown districts**, considered **‘treated’**, and respondents located in **no-lockdown districts**, considered **‘control’**.

Analyze changes in labour market outcomes between three points in time:

- 1) **February 2020**, base period before COVID-19 pandemic had reached Ghana
- 2) **April 2020**, when parts of Ghana were under lockdown  $\rightarrow POST_{1t}$
- 3) **Aug/Sept 2020**, when the most stringent policy measures had been relaxed  $\rightarrow POST_{2t}$

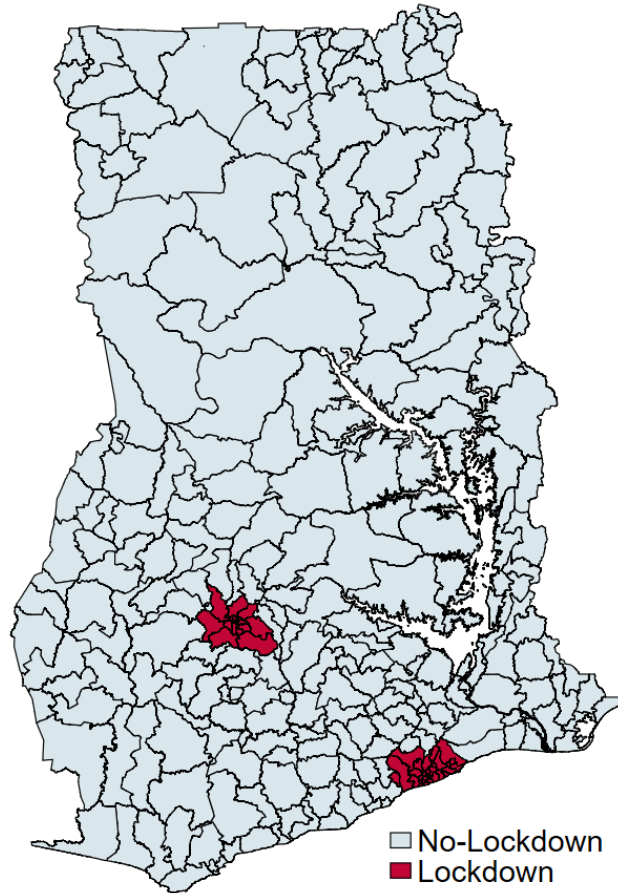
We write the DID regression model as:

$$Y_{idt} = \beta_0 + \beta_1 LOCKDOWN_d + \beta_2 (LOCKDOWN_d \times POST_{1t}) + \beta_3 (LOCKDOWN_d \times POST_{2t}) + \beta_4 X_i + \theta_t + \varepsilon_{idt}$$

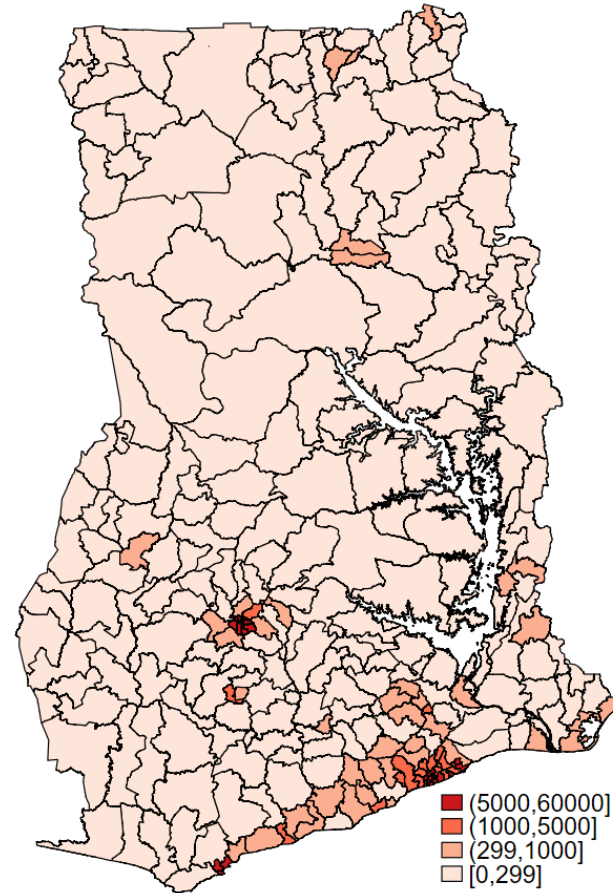
We also estimate a second model that controls for worker-fixed effects,  $\mu_i$ :

$$Y_{idt} = \delta_0 + \delta_1 (LOCKDOWN_d \times POST_{1t}) + \delta_2 (LOCKDOWN_d \times POST_{2t}) + \mu_i + \theta_t + \varepsilon_{idt}$$

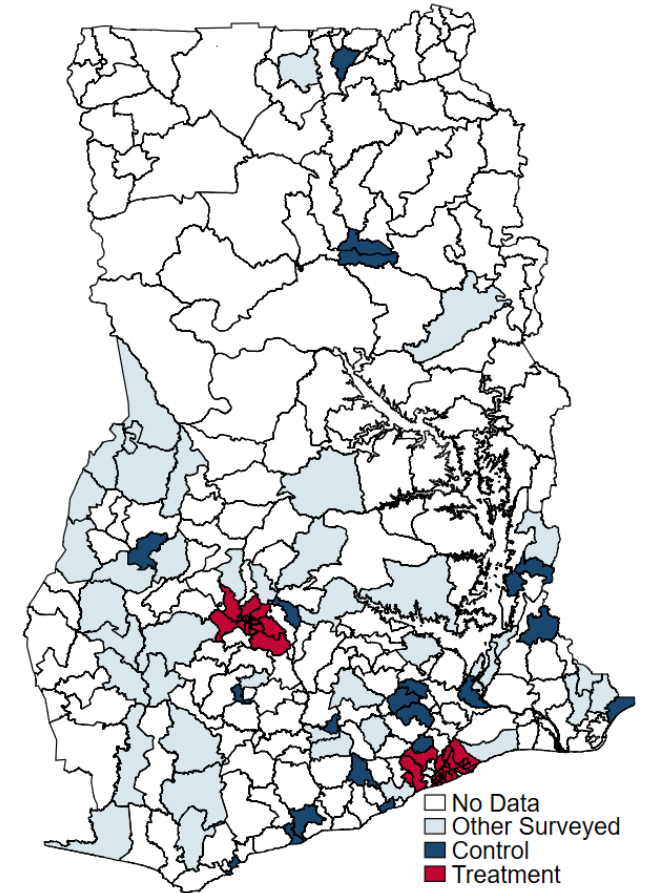
# Study areas: Lockdown versus no-lockdown



**Greater Accra and Greater Kumasi Metropolitan Areas and contiguous districts were under lockdown.**



**Population density across districts (population per km<sup>2</sup>)**

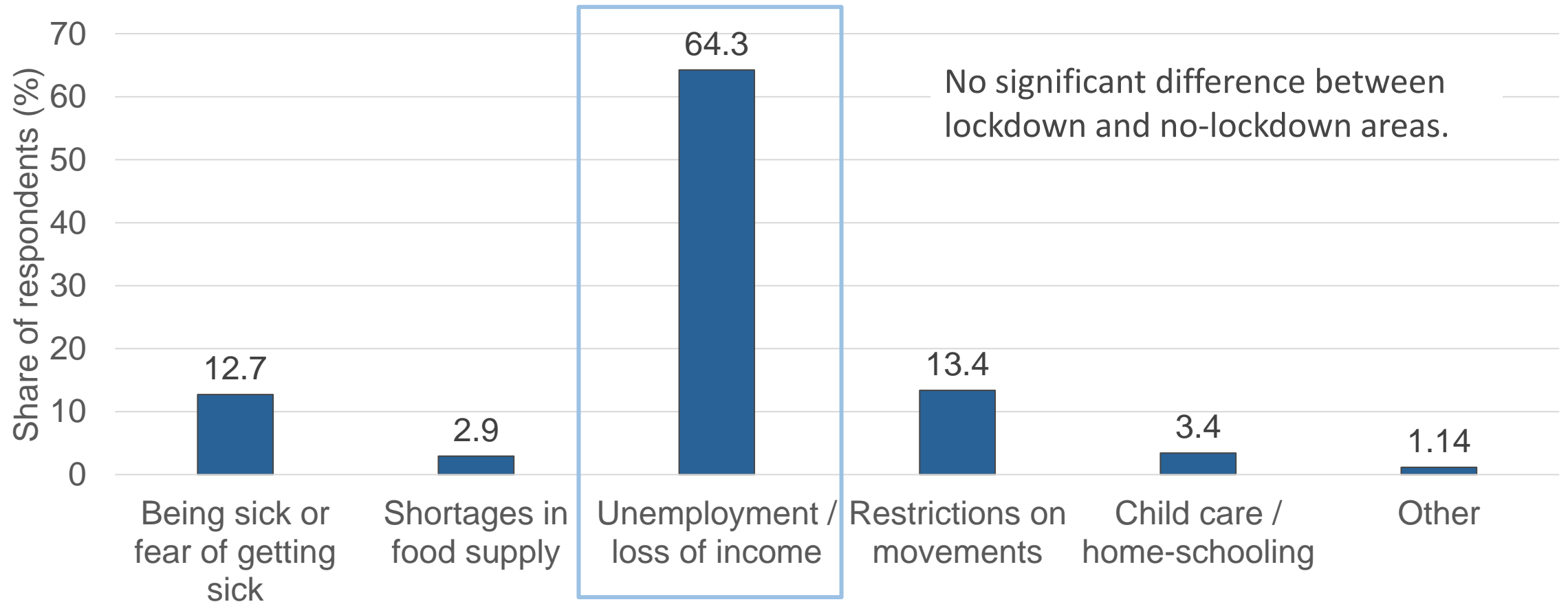


**Control districts limited to those with population density above 300/km<sup>2</sup>.**

# T-tests: Lockdown versus no-lockdown

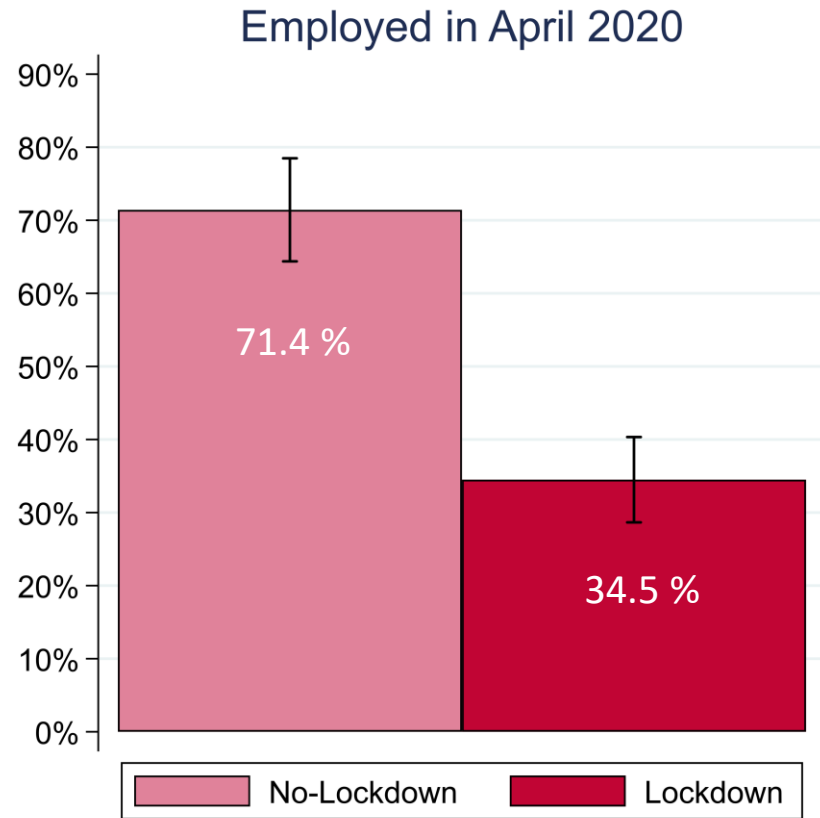
Covariates	(1) Lockdown	(2) No-Lockdown	(3) No-Lockdown size cut-off	(1)-(3) Difference	P-value Ha: diff != 0
Female	0.453 (0.030)	0.504 (0.026)	0.491 (0.038)	-0.038 (0.048)	0.4282
Head of household	0.833 (0.022)	0.813 (0.020)	0.838 (0.028)	-0.005 (0.036)	0.8938
Household size	2.609 (0.094)	3.434 (0.105)	3.428 (0.158)	-0.819*** (0.172)	0.0000
Moved since last interview	0.087 (0.017)	0.122 (0.017)	0.139 (0.026)	-0.052* (0.030)	0.0844
Working in Feb 2020	0.935 (0.015)	0.919 (0.014)	0.925 (0.020)	0.010 (0.025)	0.6872
Formal work (Feb 2020)	0.244 (0.027)	0.279 (0.024)	0.267 (0.035)	-0.023 (0.044)	0.6010
Wage work (Feb 2020)	0.422 (0.031)	0.341 (0.025)	0.385 (0.038)	0.037 (0.049)	0.4500

# Which aspect of the COVID-19 pandemic has the greatest impact on you personally?





# Strong fall in employment in lockdown areas



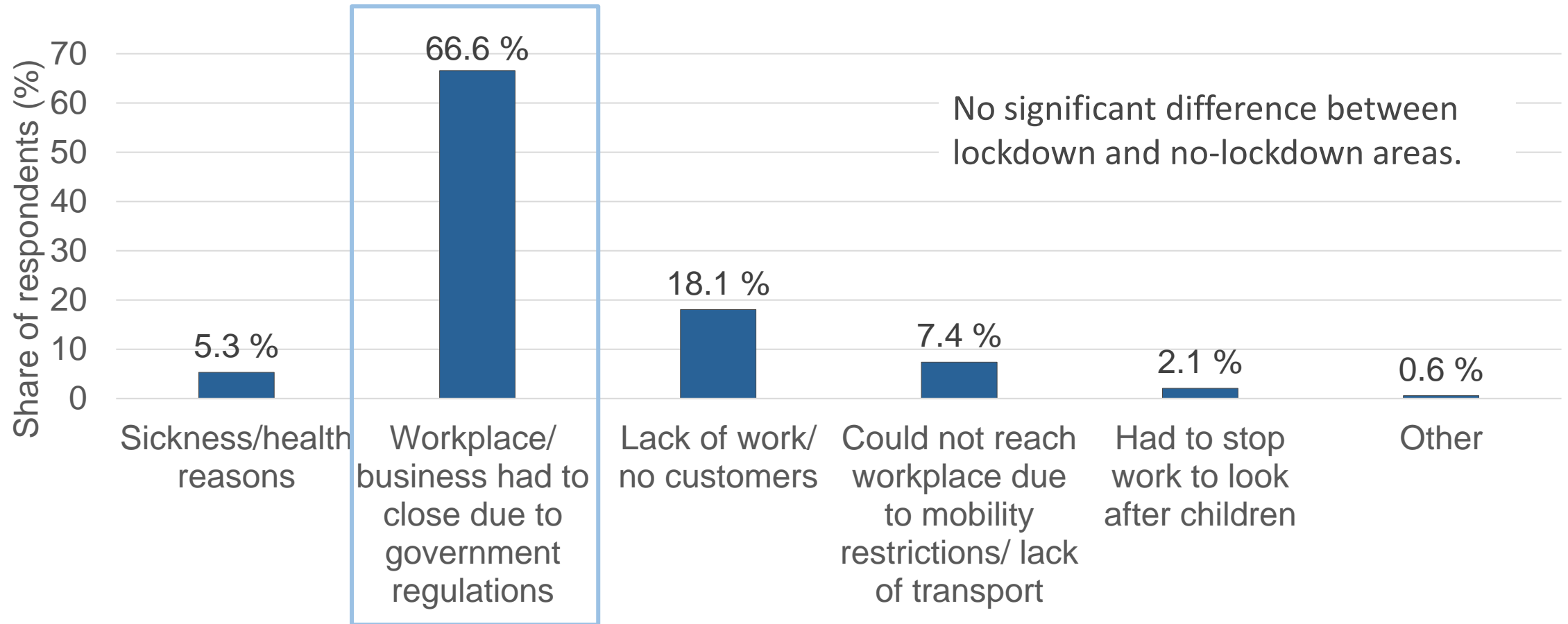
Note: Sample limited to respondents who had been working February 2020. No-Lockdown districts limited to those with population density above 300/km<sup>2</sup>.

- Sizeable and statistically significant treatment effect on employment probability in April 2020.
- Workers in lockdown areas were more than twice as likely to stop working in April.

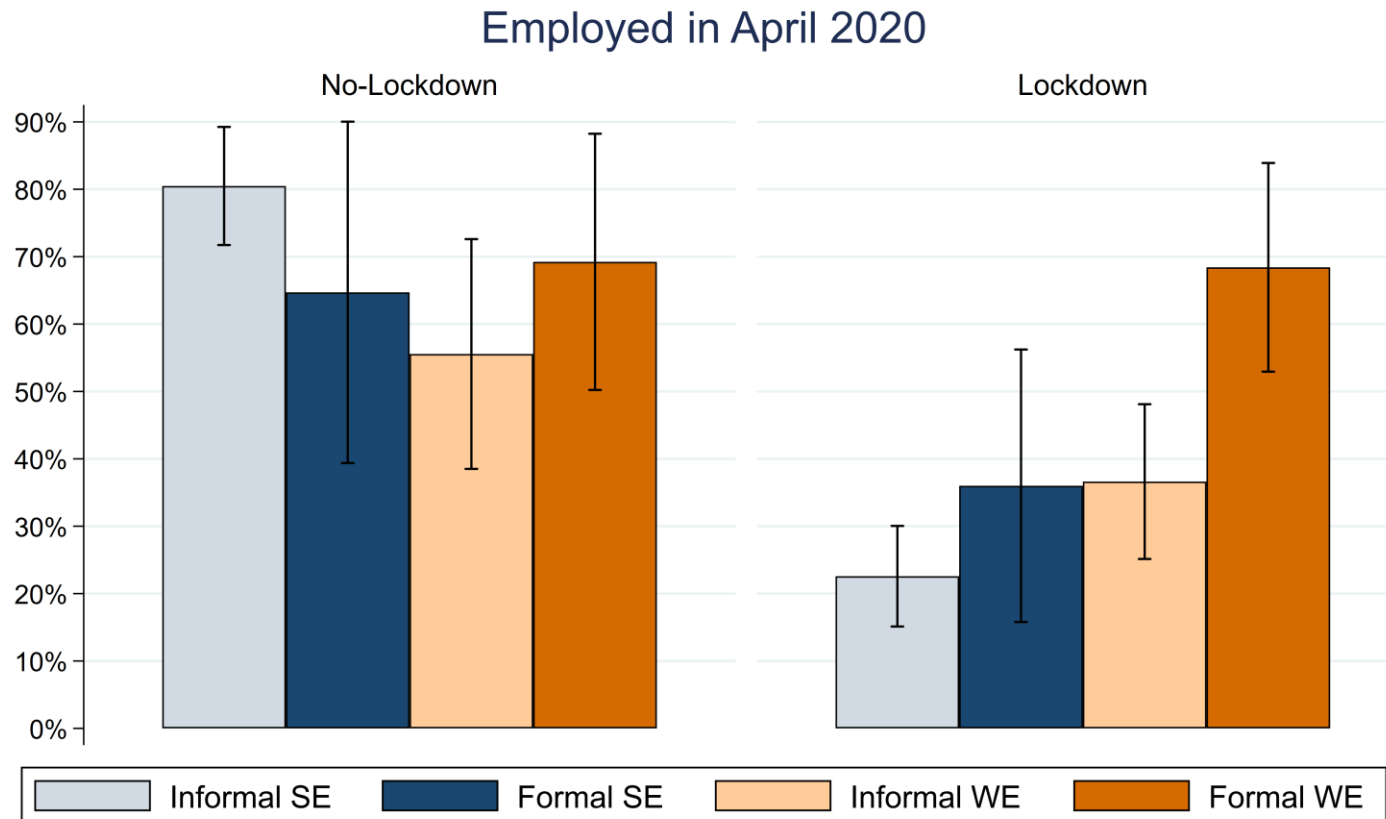
Dependent variable: Working in period $t$ (=1 if YES)	(1) Full sample	(2) District size cut-off	(3) District size cut-off with covariates	(4) District size cut-off with worker FE
Post-period (base Feb 2020)				
April 2020	-0.253*** (0.034)	-0.212*** (0.046)	-0.216*** (0.046)	-0.283** (0.047)
Lockdown	0.015 (0.029)	0.010 (0.033)	0.017 (0.036)	
Lockdown × April 2020	-0.332*** (0.049)	-0.372*** (0.058)	-0.370*** (0.058)	-0.351*** (0.051)
Observations	1936	1318	1318	1318

Note: Linear probability model; s.e. in parentheses; \* $p < 0.05$ , \*\* $p < 0.01$ , \*\*\* $p < 0.001$ .

# What was the main reason why you had to stop working?

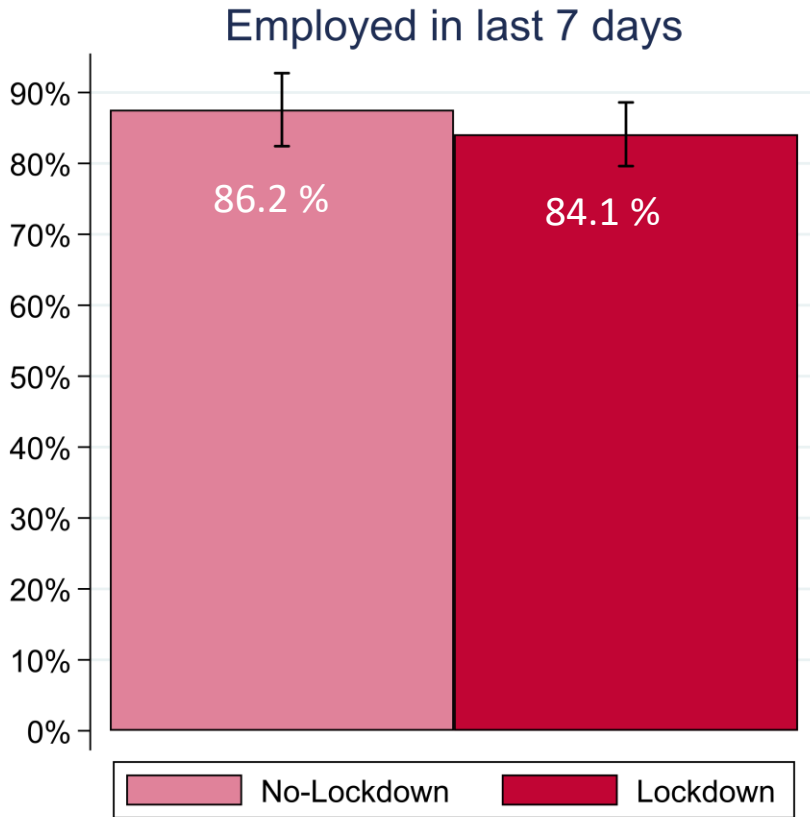


# Informal self-employed workers were most affected in lockdown areas



- In lockdown areas, 77.4% of all informally self-employed workers had stopped work in April 2020, compared to 28.3% in no-lockdown areas.
- We find no statistically significant differences between workers in upper-tier vs. lower-tier informality.

# Strong employment recovery up to Aug/Sept



Note: No-Lockdown districts limited to those with population density above 300/km<sup>2</sup>.

- No statistically significant difference in the probability to be working in Aug/Sept between lockdown and no-lockdown areas

Dependent variable: Working in period $t$ (=1 if YES)	(1) Full sample	(2) District size cut-off	(3) District size cut-off with covariates	(4) District size cut-off with worker FE
Post-period (base Feb 2020)				
Last 7 days	-0.125*** (0.019)	-0.116*** (0.026)	-0.116*** (0.026)	-0.116*** (0.026)
Lockdown	0.015 (0.029)	0.010 (0.033)	0.017 (0.036)	
Lockdown × Last 7 days	-0.013 (0.024)	-0.022 (0.030)	-0.022 (0.030)	-0.022 (0.030)
Observations	1936	1318	1318	1318

Note: Linear probability model; s.e. in parentheses; \* $p < 0.05$ , \*\* $p < 0.01$ , \*\*\* $p < 0.001$ .

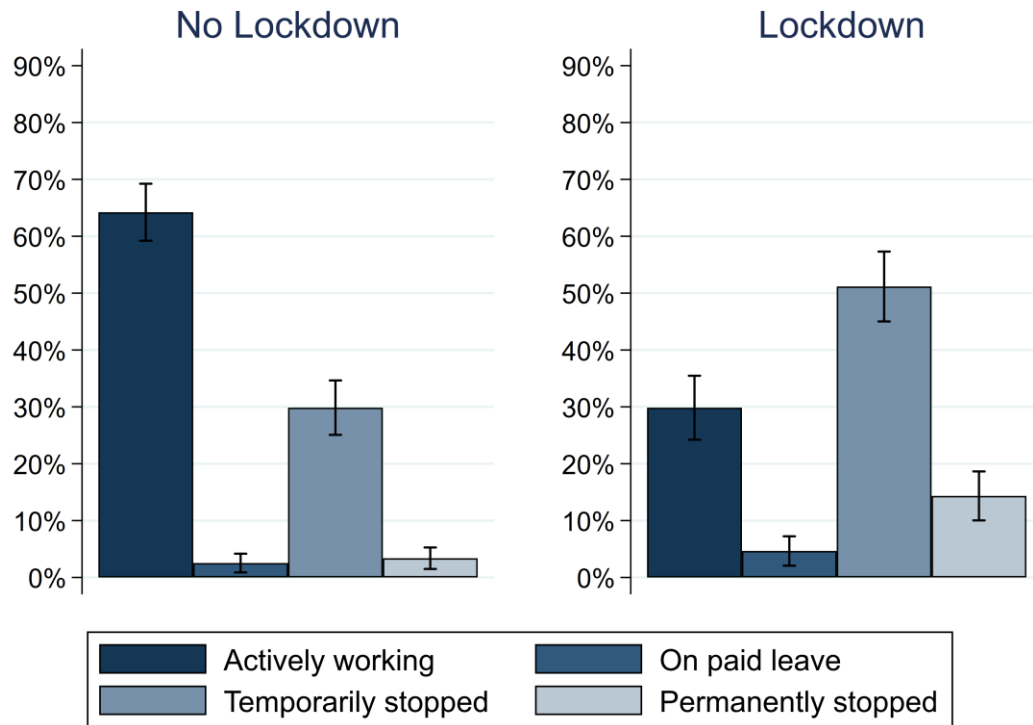
- Across the sample, **18 % of all men and 29 % of all women** who had stopped work in April, were still not working in Aug/Sept.

# Breaks in employment were mainly temporary

83.3 % of those who stopped work, considered this break to be temporary.

Chances to resume work were higher in lockdown than in non-lockdown areas.

Employment status in April 2020



Note: Use full sample to have sufficient observations in small groups (paid leave/permanently stopped) in no-lockdown areas.

	(1)	(2)
Employed in last 7 days	Full sample	Full sample
Lockdown	-0.007 (0.024)	-0.018 (0.026)
Status in April 2020 (base: Actively working)		
On paid leave	0.070*** (0.016)	0.012 (0.028)
Temporarily stopped	-0.184*** (0.042)	-0.185*** (0.044)
Permanently stopped	-0.430** (0.172)	-0.455*** (0.167)
Lockdown#On paid leave	0.007 (0.024)	0.051* (0.031)
Lockdown#Temporarily stopped	0.149** (0.060)	0.167*** (0.060)
Lockdown#Permanently stopped	-0.033 (0.196)	0.015 (0.191)
Observations	612	612
Covariates	No	Yes

Note: Linear probability model; s.e. in parentheses; \*p<0.05, \*\*p<0.01, \*\*\*p<0.001.

# ...But employment, working hours and earnings remain below pre-COVID levels

	(1) Working in period $t$ (=1 if YES)	(2) Log weekly earnings (constant 2018 prices) in period $t$	(3) Weekly hours worked in period $t$
Post-period (base Feb 2020)			
April 2020	-0.283** (0.047)	-0.658*** (0.107)	-14.226*** (4.581)
Last 7 days	-0.116*** (0.026)	-0.459*** (0.067)	-6.955*** (1.811)
Lockdown × April 2020	-0.351*** (0.051)	0.346*** (0.111)	8.902* (5.147)
Lockdown × Last 7 days	-0.022 (0.030)	0.135 (0.084)	0.631 (2.059)
Observations	1318	710	761
Panel effects	FE	FE	FE

Note: s.e. in parentheses; \* $p < 0.05$ , \*\* $p < 0.01$ , \*\*\* $p < 0.001$ .

Respondents working in August/September, on average still saw a...

... 36.8% **decline in weekly earnings** compared to base of 271 GH¢ in Feb 2020.

... 14.3% **decline in working hours** compared to a base of 49 hrs/week in Feb 2020.

<b>Dependent variable: Log weekly earnings in period <math>t</math> (constant 2018 prices)</b>	<b>(1) Full sample with covariates</b>	<b>(2) Full sample with worker FE</b>
Post-Period (base Feb 2020)		
Last 7 days	-0.187** (0.087)	-0.222*** (0.055)
Self-employed in Feb 2020	0.271*** (0.099)	
Formal work in Feb 2020	0.187* (0.101)	
Female	-0.230** (0.101)	
Last 7 days × Self-employed in Feb 2020	-0.222** (0.106)	-0.170** (0.076)
Last 7 days × Formal in Feb 2020	0.099 (0.119)	0.067 (0.089)
Last 7 days × Female	-0.144 (0.115)	-0.205** (0.087)
Observations	876	876
District fixed effects	YES	NO
Panel effects	RE	FE

Indicative evidence that the earnings of

- self-employed workers
- (informal workers)
- women

remain most heavily affected.

# Robustness checks

To ensure the robustness of our findings, we estimate several variants of our preferred model specification:

- ✓ We test whether our data support the assumption of **common pre-treatment trends** in labour market outcomes of treated and control groups, underlying the DID identification.
- ✓ To test for potential bias due to **self-selection**, we examine whether our results are robust to the exclusion of workers who have moved since the 2018/19 panel round.
- ✓ To ensure that our results are not driven exclusively by the **two major metropolitan districts**, which together account for 65.6 per cent of the treated observations, we exclude the Accra Metropolitan and Kumasi Metropolitan districts from the estimation.
- ✓ We estimate a specification with random treatment assignment of districts.



# Summary and conclusions

We provide **causal evidence** of the impact of **stringent lockdown policies** on labour market outcomes at the extensive and intensive margin, using Ghana as a case study.

- Take advantage of a **specific policy setting**: Strict stay-at-home orders were issued and enforced in two spatially delimited area (major metropolitan centres), while in the rest of the country less stringent regulations were in place.

Using a **difference-in-differences** design we find that:

- The lockdown measures implemented in response to the COVID-19 pandemic heavily affected economic activity in the affected districts during April 2020.
- The shock was felt the most by **workers in vulnerable forms of employment**.

# Summary and conclusions ctd.

- Workers in **informal self-employed**, who need to earn a living on a day to day basis, **were most often forced to stop their activities during the lockdown**, while those in formal wage employment were most likely to continue their work.
- Overall, there has been a **strong recovery in employment** up to Aug/Sept 2020. However, **employment levels remain below pre-COVID levels**, and the recovery has been slower for women than men.
- In addition, there is a **persistent negative effect on working hours and earnings**, affecting particularly the earnings of **self-employed workers and women**.
- The COVID-19 pandemic and related government response measures tend to have **accentuated existing vulnerabilities** in the Ghanaian labor market.



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