Employment Effects and Welfare Consequences of Short-Time Work Programs

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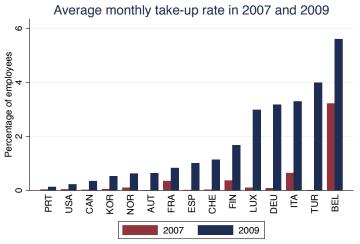
London School of Economics

XIII IZA Conference: Labor Market Policy Evaluation IZA, Bonn - October, 5-6 2017

Motivation

Short-time work

- Subsidy for hour reductions to firms experiencing temporary shocks
- Twofold objective:
 - Insure workers against income fluctuations due to variations in employment at intensive margin
 - Prevent potentially excessive layoffs in response to shocks and the resulting increase in unemployment
- Policy tool widely used in OECD countries during the crisis
 - \triangleright 3-6% of the workforce and 0.1-0.3% of GDP in 2009
 - ▶ Expenditure on UI was 0.7% and 1.1% of GDP in 2007 and 2009
- Little knowledge about the effect of STW on firms and workers
 - Limited evidence due to lack of good-quality data and credible identifying variation
 - ▶ Lack of conceptual framework to evaluate welfare effects



Source: Hijzen and Venn (2010), OECD data

Research question

STW programs are public schemes intended to preserve jobs at firms experiencing temporary shocks

- Is there an effect of STW programs on employment?
- What are the long-term consequences of STW on individuals and firms?

In a labor market with frictions, shocks to firms may induce inefficient layoffs (Hall and Lazear, 1984) and layoffs may respond to the subsidy

- Are layoffs inefficiently high? Is reducing layoffs welfare enhancing?
- What is the optimal design of STW as a tool to prevent excessive layoffs in a frictional labor market?



Literature

- Following earlier cross-country empirical analyses (Abraham and Houseman, 1993), renewed interest in STW at the onset of the crisis:
 - Cross-country studies find evidence of a positive effect of STW on employment and a negative one on average hours worked (Hijzen and Venn, 2010; Boeri and Bruecker, 2011; Cahuc and Carcillo, 2011; Hijzen and Martin, 2013)
 - Analysis at firm-level remains scarce and inconclusive due to limited data availability and credible exogenous variation (Boeri and Bruecker, 2011; Brenke et al., 2013; Calavrezo et al., 2009)
- Early theoretical literature shows that STW reduces layoffs compared to UI, but generates distortions at the intensive margin leading to underemployment (Burdett and Wright, 1989)
- Recent theoretical work shows that, by preventing increases in unemployment during a recession, STW decreases allocative efficiency due to a reduction in the vacancy filling rate (Cooper et al., 2017)

This project

- Unique administrative data on STW at the individual and firm level from Italian Social Security records
- Matched with balance-sheet data at firm level
- Credible sources of quasi-experimental variation in Italian policy rules to identify causal effects of STW on:
 - Employment at the intensive and extensive margins
 - Long-term individual and firm outcomes
- Twofold empirical strategy
 - Baseline triple difference to identify short-run effects
 - Triple difference à la Cellini et al. (2010) to identify dynamic effects
- Use empirical evidence to analyze the welfare consequences of STW programs as policy tools to limit potentially excessive layoffs



Outline

- Institutional background and data
- 2 The effects of STW during the Great Recession
- Opposition of the state of t
- Selection and heterogeneity
- Discussion



Institutional features of the Italian STW program

- Subsidy for hour reductions either partial or full-time available to workers in the private sector and administered by the Social Security
- Replaces about 80% of foregone earnings due to hours not worked
- Firms intending to use the program must file an application to the Social Security or the Ministry of Labor
 - Provide justification of economic need
 - Develop a recovery plan
- Weak conditionality requirements
 - No compulsory training
 - No prohibition of dismissal
 - No job-search requirement for employee
- Cost to employer is a fixed percentage of subsidized hours (with exemptions)

Three pillars of Cassa Integrazione Guadagni

| | Target | Duration | Avg spell | Industry | Size |
|------|---|-----------|------------|--|-----------------------------|
| CIGO | Transitory, exogenous shocks | 13 weeks | 2 weeks | Manufact Construct Transport | |
| CIGS | Crisis Restructuring Insolvency | 2 years | 3.5 months | Manufact Construct Retail Transport | > 15 > 15 > 50 any |
| CIGD | Extend coverage or duration of CIGO/CIGS | By decree | 3.5 months | Sectors and regions set by decree | |

Administrative data from Italian Social Security Archives

- Universe of matched employer-employee data for the private sector
- Monthly data 2005-2015 and annual data 1983-2015
- Information on workers and firms
 - Demographics
 - Working histories
 - Social insurance and social assistance program participation
 - Firm characteristics (size, sector)
- Information on CIG eligibility, applications, authorizations, duration and payment for the years 2005-2015
- Matching with firm-level balance-sheet data (approx. 50%)

Identification of STW effects

- We use the Great Recession as our main source of shock
- Identification exploits variations in eligibility for CIG Straordinaria based on:
 - **Size**: having employed more than 15 employees in full-time equivalent in six months prior to application
 - Industry: mostly industries in manufacturing and construction
- Compare the difference in outcomes around the 15-threshold in eligible and non-eligible industries, before and after the start of the Great Recession
 - Industry variation in eligibility for STW allows controlling for the confounding effect of employment protection legislation

McCrary Test

Empirical strategy

Reduced-form graphical representation

 Plot the difference in outcomes between firms above and below the 15-threshold, and in eligible and non-eligible industries relative to 2007

Structural-form regression-based estimation

- Instrument the probability of receiving CIGS using the interaction between a 15-threshold dummy, industry and calendar year
- Structural form

$$y_{ist} = \beta_0 + \beta_1 T_{ist} + u_{ist}$$



Empirical strategy (cont.)

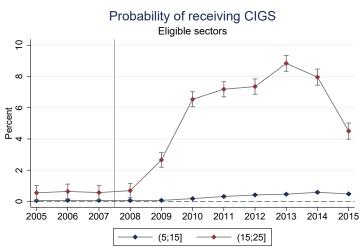
First stage

$$\begin{split} T_{ist} = & \gamma_0 + \sum_{j \neq 2007} \gamma_1^j \mathbbm{1}[s \in elig] * \mathbbm{1}[size > 15] * \mathbbm{1}[j = t] \\ & + \sum_{j \neq 2007} \sum_k \gamma_2^{jk} \mathbbm{1}[k = s] * \mathbbm{1}[j = t] \\ & + \sum_{j \neq 2007} \gamma_3^j \mathbbm{1}[size > 15] * \mathbbm{1}[j = t] \\ & + \sum_k \gamma_4^k \mathbbm{1}[k = s] * \mathbbm{1}[size > 15] \\ & + \sum_k \gamma_5^k \mathbbm{1}[k = s] + \sum_{j \neq 2007} \gamma_6^j \mathbbm{1}[j = t] \\ & + \gamma_7 \mathbbm{1}[size > 15] + v_{ist} \end{split}$$

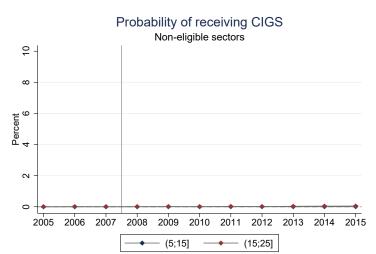
where i is firm, s industry at 5-digit level and t calendar year

Empirical implementation

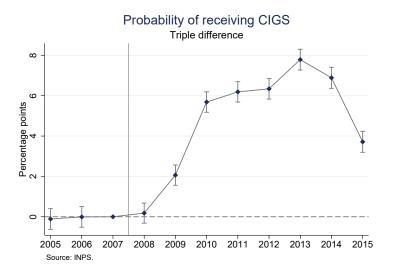
- Annual data from 2000 to 2015
- Panel of all ever-active firms and of their workers (in eligible job positions)
- Eligibility status is defined dynamically based on maximum 6-month average firm size and industry in each year
 - Select firms with 6-month full-time equivalent firm size \in (5; 25]
- Definition of a STW event
 - Any month in which a CIG episode is reported, which is also authorized according to the authorization data
 - When aggregating at the annual level, an event is defined as having at least one CIG episode during the year



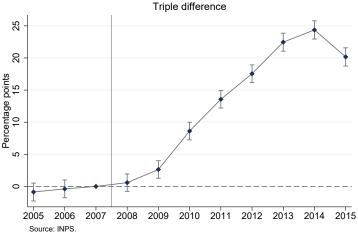
Source: INPS.



Source: INPS.



Probability of ever receiving CIGS in the past 5 years

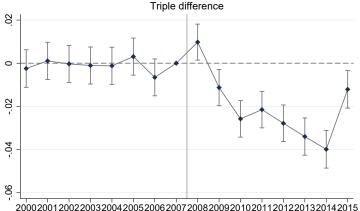






Intensive-margin response

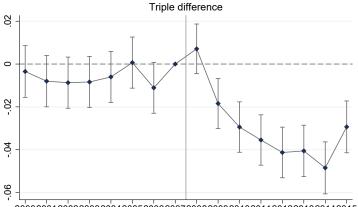
Log annual weeks worked per employee



2000200120022003200420052006200720082009201020112012201320142015

Note: IV estimate -0.486 (s.e. 0.033). Source: INPS.

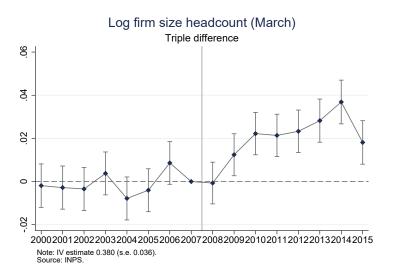
Log annual wage bill per employee



2000200120022003200420052006200720082009201020112012201320142015

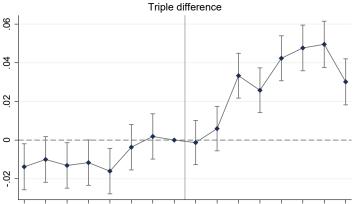
Note: IV estimate -0.585 (s.e. 0.044). Source: INPS.

Extensive-margin response



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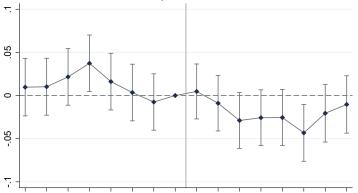
Log open-ended contracts (March)



2000200120022003200420052006200720082009201020112012201320142015

Note: IV estimate 0.616 (s.e. 0.043). Source: INPS.

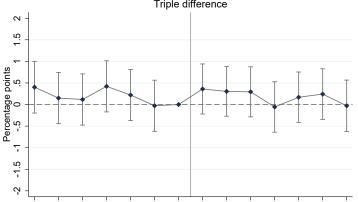
Log fixed-term contracts (March) Triple difference



2000200120022003200420052006200720082009201020112012201320142015

Note: IV estimate -0.403 (s.e. 0.117). Source: INPS.

Probability of firm survival in t+1 Triple difference



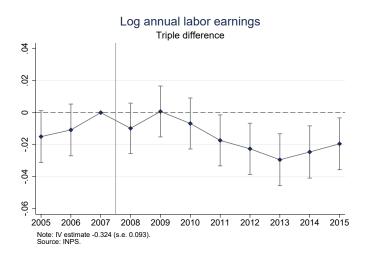
2000 2001 2002 2003 2004 2005 2006 2007 2008 2009 2010 2011 2012 2013

Note: IV estimate -0.017 (s.e. 0.022). Source: INPS.

Balance-sheet outcomes

| | Net revenue | Profit | Labor product | Liquidity | Invest. intang. | Invest. tang. |
|------|-----------------------|---------------------|------------------|------------------------|-----------------------|-----------------------|
| CIGS | 690.458 (2873.104) | 52.638 (602.295) | 1.644 (1.710) | 701.563** (337.087) | -210.891 (525.144) | -227.865 (481.722) |
| Obs. | 10950 | 10950 | 10950 | 10950 | 10950 | 10950 |

Total labor earnings for hours worked



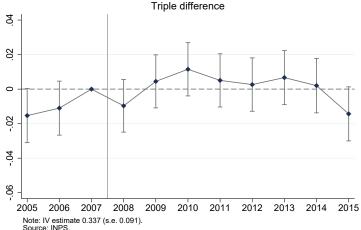


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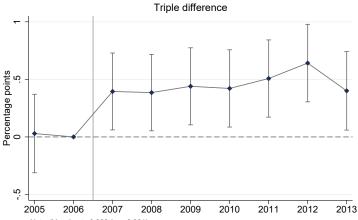


Insurance role of STW compensation

Log annual labor earnings including STW



Probability of employment in t+1



Note: IV estimate 0.062 (s.e. 0.021). Source: INPS.





Dynamic effects

- The previous analysis captures the contemporaneous effects of STW on outcomes
- We are also interested in long-term effects:
 - On firms
 - On workers, to capture longer term insurance value of STW
- A challenge in the identification of long-term effects is the possibility of dynamic treatment
- In order to isolate the effect of treatment in t-k on outcomes in t controlling for treatment between t-k and t, we adopt an empirical strategy similar to Cellini et al. (2010)
- Preliminary evidence suggests that effects dissipate quickly over time



Treatment effects with dynamic treatment

We can write $y_{i,t}$ as a function of the full history of treatment:

$$y_{i,t} = \sum_{t=0}^{T} T_{i,t-k} \beta_k$$

Effect of treatment on the treated (TOT): The effect of treatment in t - k on outcome in t, absent any additional treatment between t - k and t

$$\beta_k^{TOT} = \frac{\partial y_{i,t}}{\partial T_{i,t-k}} = \beta_k$$

Intention-to-treat effect (ITT): The effect of treatment in t-k on outcome in t, allowing for the possibility of additional treatment between t-k and t

$$\beta_k^{ITT} = \frac{\mathrm{d}y_{i,t}}{\mathrm{d}T_{i,t-k}} = \frac{\partial y_{i,t}}{\partial T_{i,t-k}} + \sum_{j=1}^k \frac{\partial y_{i,t}}{\partial T_{i,t-k+j}} * \frac{\mathrm{d}T_{i,t-k+j}}{\mathrm{d}T_{i,t-k}}$$
$$= \beta_k^{TOT} + \sum_{j=1}^k \beta_{k-j}^{TOT} * \pi_j$$

Identification of dynamic treatment effects

1. Estimate $\hat{\pi}_k$ instrumenting $T_{i,t}$ using the interaction between size, industry and calendar year

$$T_{i,t+k} = \pi_0 + \pi_k T_{i,t} + \varepsilon_{i,t+k}$$

2. Estimate $\hat{\beta}_k^{ITT}$ instrumenting $T_{i,t}$ using the interaction between size, industry and calendar year

$$y_{i,t+k} = \beta_0 + \beta_k^{ITT} T_{i,t} + \eta_{i,t+k}$$

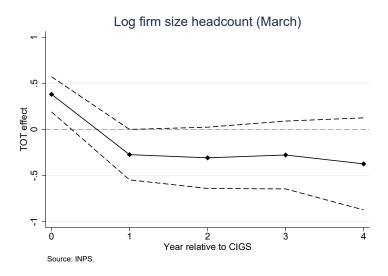
3. Retrieve β_k^{TOT} recursively and estimate standard errors via bootstrapping

$$\beta_0^{TOT} = \beta_0^{ITT}$$

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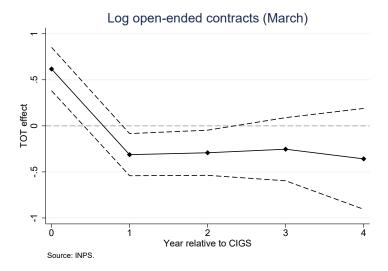
$$\beta_k^{TOT} = \beta_k^{ITT} - \sum_{h=1}^{k-1} \pi_h \beta_{k-h}^{TOT}$$

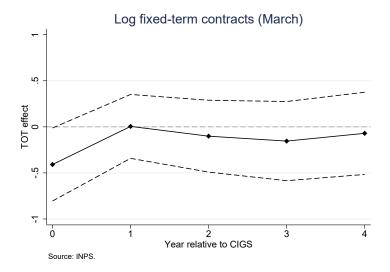


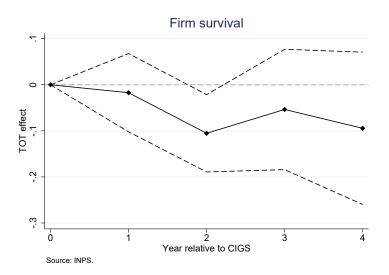




ITT effect





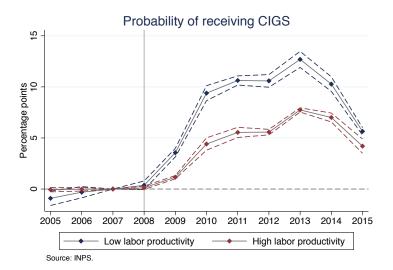


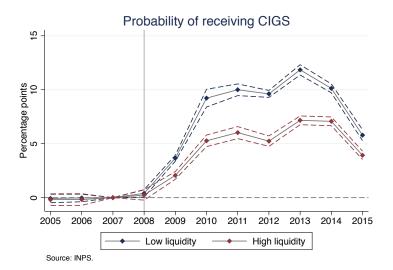
Selection and heterogeneity

- With heterogenous firms and workers, it is key to understand which firms and workers are taking up STW (selection) and how treatment effects vary across different firms (heterogeneity)
- Productivity: STW subsidizes low-productivity matches and may prevent the efficient reallocation of workers in the labor market, as suggested by the long term effects of STW
- Liquidity/financial constraints: STW may provide liquidity to financially constrained firms and prevent excess layoff sensitivity to productivity shocks (Schoefer, 2016)

Selection into STW

- Explore heterogeneity in take-up across different firm characteristics
- Use balance sheet data to construct measures of heterogeneity:
 - Liquidity (proportion of liquid assets out of total assets)
 - Labor productivity (value added per employee)
- For each of these dimensions, rank firms by their pre-crisis level of liquidity/credit constraints/labor productivity, conditional on 2-digit industry
- Compare take-up for firms in the top and bottom quartiles of their industry-specific ranking





Summary of empirical results

- Exploit richness of administrative data on STW and exogenous variation in Italian STW regulations to assess the impact of STW on firms and workers
- Document large increase in STW take-up by eligible firms during the Great Recession
- Find sharp reduction of employment at the intensive margin and positive effects on the extensive margin, but entirely to the benefit of open-ended contracts
- No significant effect on firm survival
- Matching with firm-level balance-sheet data allows exploring selection and heterogeneous effects by firm-level characteristics and types of shocks
- Some initial evidence of heterogeneity in take-up by degree of liquidity and labor productivity



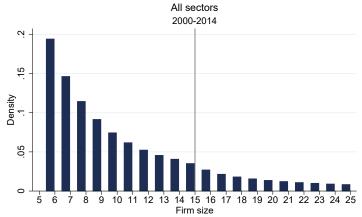
Theoretical framework for welfare analysis

- Develop a general-equilibrium framework to analyze the welfare effects of STW as a policy tool to prevent layoffs in the presence of frictions and a dual labor market
- Characterize the optimal design of the subsidy
- Use empirical estimates to assess the optimality of the current system and inform policy choices

Additional slides



Distribution of firms by firm size



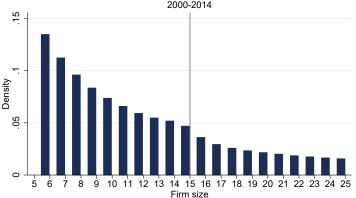
Note: McCrary test -0.009 (s.e. 0.005). Source: INPS.





Distribution of firms by firm size





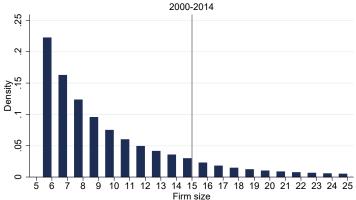
Note: McCrary test -0.024 (s.e. 0.008). Source: INPS.





Distribution of firms by firm size

Non-eligible sectors

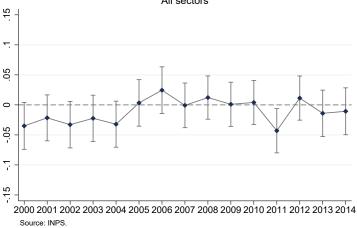


Note: McCrary test 0.003 (s.e. 0.007). Source: INPS.





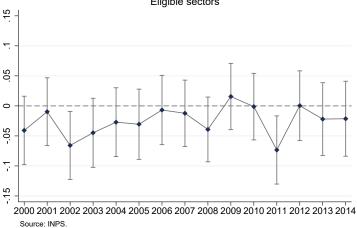








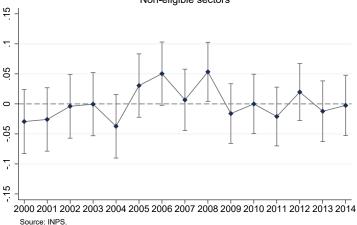
McCrary Test Eligible sectors







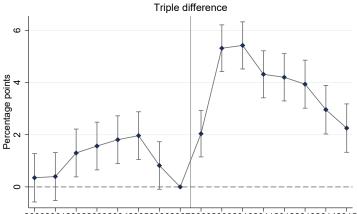








Probability of receiving any CIG

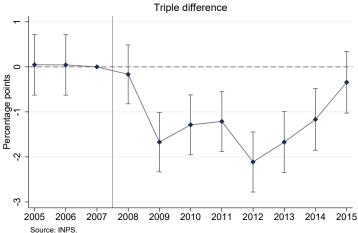


2000200120022003200420052006200720082009201020112012201320142015 Source: INPS.





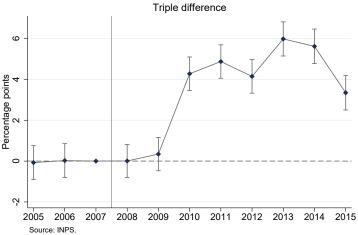
Probability of receiving CIGD







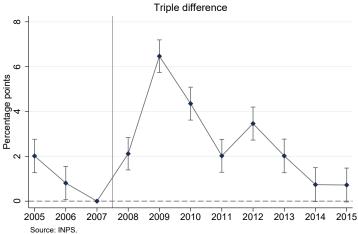
Probability of receiving CIGS or CIGD







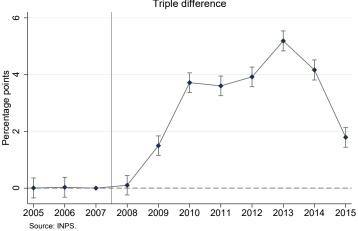
Probability of receiving CIGO







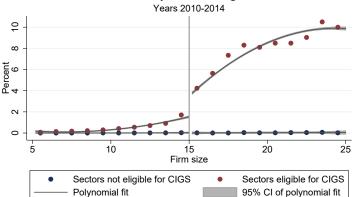
Probability of receiving CIGS Triple difference







Probability of receiving CIGS



Note: Discontinuity in eligible sectors 1.970 (s.e. 0.077).

Discontinuity in non-eligible sectors -0.020 (s.e. 0.063).

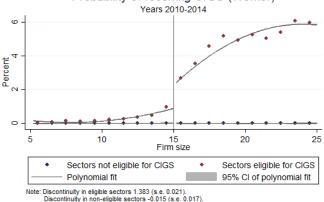
Double-discontinuity estimate 1.990 (s.e. 0.100).

Source: INPS.









Discontinuity in non-eligible sectors -0.015 (s.e. 0.017). Double-discontinuity estimate 1.398 (s.e. 0.027). Source: INPS.





IV Difference-in-Discontinuity Design - Firm-level outcomes

| | Log wage bill per empl | Log total weeks per empl | Log employment |
|------|---------------------------|-----------------------------|-------------------|
| CIGS | -1.403*** (0.437) | -1.291*** (0.302) | -0.330 (0.273) |
| Obs. | 1424248 | 1423439 | 1424786 |



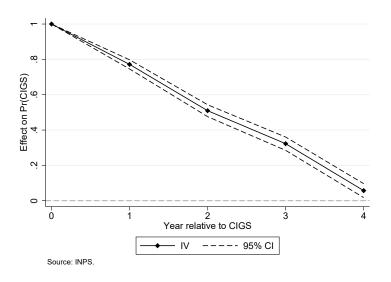


IV Difference-in-Discontinuity Design - Worker-level outcomes

| | Log weeks worked | Log earnings for hours worked | Log earnings including STW |
|------|----------------------|----------------------------------|----------------------------|
| CIGS | -0.672*** (0.203) | -0.464 (0.307) | 0.158 (0.305) |
| Obs. | 14573660 | 14723369 | 14723369 |











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