The Impact of Minimum Wage Shock on Robot Adaption in Turkey

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Introduction

- In recent years, the utilization of industrial robots in the manufacturing industry has experienced a notable surge.
- According to the International Federation of Robotics (IFR), the operational stock of robots has tripled since 2010.
- This transformative shift in manufacturing has prompted scholars to investigate the impact of robotics on employment and wages. While many studies suggest that the adoption of robots significantly reduces employment and earnings, especially in developed (US, France) countries there are also studies demonstrating no negative employment effects and, in some cases, positive effects of robotization in developing (Indonesia, Turkey) and certain developed countries (Finland, Italy, EU regions).
- Various mechanisms, such as productivity effects, diminishing returns, and product innovation, have been proposed to explain this puzzle.

Introduction

- However, in the literature, determinants of robot adoption, particularly cost-related variables, are not extensively studied due to two main challenges. The first challenge is the measurement issue of robots.
- Another issue in robot studies is the endogeneity of possible determinants with robot adoption.
- This study aims to investigate how a minimum wage shock alters the robot adoption decisions of Turkish enterprises. In 2016, Turkey implemented a sudden 33.5% increase in the minimum wage. This shock provides an opportunity to isolate exogenous shocks to labor costs and test the price mechanism.

Introduction

- Moreover, we can observe how labor market shocks affect the robotization process in a developing country at the early stages of automation.
- Our empirical strategy relies on differences-in-differences to examine how firms are affected by the 2016 minimum wage increase when considering the purchase of robots.

Literature

- Empirical evidence regarding the effect of labor market shocks on robot adoption is limited.
- As for causal evidence, Deng et al (2021) and Fan et al.(2021) use minimum wage variation as a quasi-natural experiment for Germany and China, respectively, and observe that the minimum wage shock is more likely to drive firms toward robot adoption.

Deng et al.

- Deng et al. use IAB establishment panel survey and merge it with employer-employee data.
- They utilized the natural experiment that Germany introduced for the first time its statutory minimum wage. But share of minimum wage employment in Germany is 11% approx.

Fan et al.

- Fan et al. utilized the variation of regional minimum wage.
- Like our study, they use custom data.
- Firms with high TFP are more likely to adopt a robot.
- SOEs do not insignificantly import robot.

Data

- We utilize two administrative datasets provided by Enterprise Information System (EIS) between 2009 and 2019.
 - Employer-employee dataset by Social security institute
 - Trade dataset by Ministry of Custom and Trade

Robot importers in Turkey

Number of firms

| Sector | Zero employment | Micro | Small | Medium | Large | Total |
|---------------|-----------------|-----------|---------|--------|-------|-----------|
| Manufacturing | 199,135 | 211,073 | 46,725 | 10,120 | 2006 | 465,587 |
| All | 1,749,101 | 1,156,491 | 161,035 | 26,705 | 5,290 | 3,085,177 |

• Number of robot importers in 2010-2019

| Sector | Zero employment | Micro | Small | Medium | Large | Total |
|---------------|-----------------|-------|-------|--------|-------|-------|
| Manufacturing | 29 | 62 | 146 | 221 | 258 | 716 |
| All | 60 | 156 | 227 | 257 | 274 | 966 |

Employment composition of robot importers

- Young-aged workers consist of 56% of total robot importer firms, which is higher compared to all manufacturing firms (50%)..
- Most of the employment (85%) is male.
- 4.8% of workers employed in robot importers earn minimum wage.

Number of minimum wage earners and total employment by size

| Size | Min. wage earners | Total | Share (%) |
|--------|-------------------|-----------|-----------|
| Micro | 424,420 | 483,199 | 88 |
| Small | 511,807 | 845,971 | 60 |
| Medium | 329,759 | 976,239 | 34 |
| Large | 147,902 | 1,337,369 | 11 |

Geographic distribution of robot importers

Number of firms by district (2010-2019)



Specification

• We employ difference-in-differences with continuous framework. Specifically, we estimate the following equation:

$$y_{it} = \alpha + X_{it}^{'} \Phi + \beta$$
minimum wage share $_{it} + D_i + D_t + D_{kt} + D_{pt} + \varepsilon_{it}$

where the outcome variable y_{it} takes two forms. Firstly, the extensive margin is a dummy variable equal to 1 if firm i adopts robotization in its production at time t. As the second form of the outcome variable, the intensive margin is calculated as the log of cumulative monetary value and quantity (weight) of robot imports adopted by firm i.

- Finally, β denotes the effect of the ratio of the minimum wage employment share of a firm in 2015 on the likelihood of being a robot adapter. We track minimum wage employees using their daily wages.
- Additionally, we interact the variable of interest with firm size to test the hypothesis that large firms are more likely to implement robotization than their smaller counterpart.

Extensive margin and intensive margin

| Dependent Variables: | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) |
|------------------------------|---------------------|-------------------------|----------------------------|----------------|-------------------------|----------------------------|
| Model: | (1) | (2) | (3) | (4) | (5) | (6) |
| Variables | | | | | | |
| minimum wage shareit | 0.0009 | 0.0081 | 0.0060 | 0.0008 | 0.0069 | 0.0052 |
| | (8000.0) | (0.0091) | (0.0062) | (8000.0) | (0.0091) | (0.0062) |
| minimum wage shareit | | | | 0.0045*** | 0.0457** | 0.0308*** |
| × medium | | | | (0.0011) | (0.0114) | (0.0080) |
| minimum wage shareit | | | | 0.0270*** | 0.3251*** | 0.2278*** |
| × large | | | | (0.0066) | (0.0740) | (0.0528) |
| Fixed-effects | | | | | | |
| Firm | Yes | Yes | Yes | Yes | Yes | Yes |
| Year | Yes | Yes | Yes | Yes | Yes | Yes |
| NACE × year | Yes | Yes | Yes | Yes | Yes | Yes |
| Province × year | Yes | Yes | Yes | Yes | Yes | Yes |
| Min. wage share | Yes | Yes | Yes | Yes | Yes | Yes |
| quantile $FE \times t.trend$ | | | | | | |
| Fit statistics | | | | | | |
| Observations | 440,735 | 440,735 | 440,735 | 440,735 | 440,735 | 440,735 |
| R ² | 0.76755 | 0.76768 | 0.78697 | 0.76762 | 0.78301 | 0.78703 |
| Within R ² | 0.00620 | 0.00685 | 0.00770 | 0.00649 | 0.00780 | 0.00799 |
| Clustered (Firm) stand | ard arrors in naran | theses | | | | |

Clustered (Firm) standard-errors in parentheses

Signif. Codes: ***: 0.01, **: 0.05, *: 0.1

Extensive margin and intensive margin min. wage share \times blue collar employment intensity

| Dependent Variables: | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) |
|--|----------------|-------------------------|----------------------------|----------------|-------------------------|----------------------------|
| Model: | (1) | (2) | (3) | (4) | (5) | (6) |
| Variables | | | | | | |
| minimum wage share it | 0.0001 | 0.0001 | -0.0008 | 0.0003 | 0.0024 | 0.0018 |
| | (0.0009) | (0.0006) | (0.0070) | (0.0009) | (0.0102) | (0.0069) |
| minimum wage share _{it} | | | | 0.0029** | 0.0287** | 0.0201** |
| × medium | | | | (0.0013) | (0.0135) | (0.0094) |
| minimum wage share _{/t} | | | | 0.0123 | 0.1437 | 0.1019 |
| × large | | | | (0.0083) | (0.0984) | (0.0702) |
| minimum wage share _{it} | 0.0004 | 0.0049 | 0.0038 | -0.0005 | -0.0050 | -0.0027 |
| × blue collar share of min. wage | (0.0007) | (0.0073) | (0.0052) | (0.0006) | (0.0067) | (0.0048) |
| minimum wage share _{/t} | | | | 0.0038 | 0.0404 | 0.0254 |
| imes blue collar share of min. wage $	imes$ medium | | | | (0.0030) | (0.0327) | (0.0234) |
| minimum wage share _{/t} | | | | 0.0259* | 0.3208** | 0.2213° |
| × blue collar share of min. wage × large | | | | (0.0140) | (0.1612) | (0.1142) |
| Fixed-effects | | | | | | |
| Firm | Yes | Yes | Yes | Yes | Yes | Yes |
| Year | Yes | Yes | Yes | Yes | Yes | Yes |
| NACE × year | Yes | Yes | Yes | Yes | Yes | Yes |
| Province × year | Yes | Yes | Yes | Yes | Yes | Yes |
| Min. wage share | Yes | Yes | Yes | Yes | Yes | Yes |
| quantile FE × t.trend | | | | | | |
| Fit statistics | | | | | | |
| Observations | 392,718 | 392,718 | 392,718 | 392,718 | 392,718 | 392,718 |
| R ² | 0.76787 | 0.78434 | 0.78956 | 0.76797 | 0.78444 | 0.78966 |
| Within R ² | 0.01333 | 0.01687 | 0.01733 | 0.01377 | 0.01734 | 0.01779 |

Clustered (Firm) standard-errors in parentheses

Signif. Codes: ***: 0.01, **: 0.05, *: 0.1

Effect of minimum wage shock on robot adaption interacted by R & D expenditure

| Dependent Variables: | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) |
|----------------------------------|----------------|-------------------------|----------------------------|-----------------------|-------------------------|----------------------------|
| R&D definition | Log(R&D+1) | Log(R&D+1) | Log(R&D+1) | Log(R&D per capita+1) | Log(R&D per capita+1) | Log(R&D per capita+1) |
| Model: | (1) | (2) | (3) | (4) | (5) | (6) |
| Variables | | | | | | |
| minimum wage shareit | 0.0006 | 0.0046 | 0.0033 | 0.0006 | 0.0048 | 0.0035 |
| | (0.0011) | (0.0122) | (0.0085) | (0.0011) | (0.0122) | (0.0085) |
| minimum wage share _{it} | 0.0011** | 0.0129** | 0.0091** | 0.0010 | 0.0121° | 0.0085* |
| × Log(R&D expenditure+1) | (0.0005) | (0.0053) | (0.0037) | (0.0006) | (0.0062) | (0.0043) |
| Fixed-effects | | | | | | |
| Firm | Yes | Yes | Yes | Yes | Yes | Yes |
| Year | Yes | Yes | Yes | Yes | Yes | Yes |
| NACE × year | Yes | Yes | Yes | Yes | Yes | Yes |
| Province × year | Yes | Yes | Yes | Yes | Yes | Yes |
| Min. wage share | Yes | Yes | Yes | Yes | Yes | Yes |
| quantile FE × t.trend | | | | | | |
| Fit statistics | | | | | | |
| Observations | 397,173 | 397,173 | 397,173 | 397,173 | 397,173 | 397,173 |
| R ² | 0.77968 | 0.79546 | 0.80150 | 0.77967 | 0.79545 | 0.80149 |
| Within R ² | 0.00476 | 0.00572 | 0.00578 | 0.00471 | 0.00567 | 0.00573 |

Clustered (Firm) standard-errors in parentheses

Signif. Codes: ***: 0.01, **: 0.05, *: 0.1

Extensive margin and intensive margin min. wage share \times HHI

| | | | . (2 |
|----------------------------------|----------------|-------------------------|----------------------------|
| Dependent Variables: | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) |
| Model: | (1) | (2) | (3) |
| Variables | | | |
| minimum wage shareit | 0.0045*** | 0.0522*** | 0.0362*** |
| | (0.0010) | (0.0118) | (0.0081) |
| minimum wage share _{it} | -0.1195*** | -1.4700*** | -1.0120*** |
| × HHI | (0.0248) | (0.2909) | (0.2032) |
| Fixed-effects | | | |
| Firm | Yes | Yes | Yes |
| Year | Yes | Yes | Yes |
| $NACE \times year$ | Yes | Yes | Yes |
| Province \times year | Yes | Yes | Yes |
| Min. wage share | Yes | Yes | Yes |
| quantile $FE \times t.trend$ | | | |
| Fit statistics | | | |
| Observations | 455,490 | 455,490 | 455,490 |
| R^2 | 0.76423 | 0.77912 | 0.78291 |
| Within R ² | 0.00792 | 0.00847 | 0.00861 |

Clustered (Firm) standard-errors in parentheses Signif. Codes: ***: 0.01, **: 0.05, *: 0.1



Extensive margin and intensive margin interacted by (export + import) / total sales

| Dependent Variables: | Robot importer | Log(Cum. robot value+1) | Log(Cum. robot quantity+1) |
|----------------------------------|----------------|-------------------------|----------------------------|
| Model: | (1) | (2) | (3) |
| Variables | | | |
| min wage share _{it} | 0.0004 | 0.0029 | 0.0016 |
| | (8000.0) | (0.0091) | (0.0063) |
| min wage share _{it} | 0.0073*** | 0.0752*** | 0.0499*** |
| \times (export + import/sales) | (0.0020) | (0.0207) | (0.0141) |
| Fixed-effects | | | |
| Firm | Yes | Yes | Yes |
| Year | Yes | Yes | Yes |
| $NACE \times year$ | Yes | Yes | Yes |
| Province \times year | Yes | Yes | Yes |
| Min. wage share | Yes | Yes | Yes |
| quantile FE × t.trend | | | |
| Fit statistics | | | |
| Observations | 444,327 | 444,327 | 444,327 |
| R ² | 0.76455 | 0.77749 | 0.78026 |
| Within R ² | 0.00624 | 0.00748 | 0.00758 |

Clustered (Firm) standard-errors in parentheses Signif. Codes: ***: 0.01, **: 0.05, *: 0.1



Conclusion

- Our preliminary results reveal that minimum wage shock insignificantly effect robot adaption.
- On the other hand, we document that firm size matters. Medium and large firms are more likely to import robort when they face a huge labor cost shock.
- Task composition of firms is also an important issue to comprehend the robotization decision of firms.
- Firms engaging in R&D are associated with robotization.

Thank you for you listening!

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