

Robotisation, Employment and Industrial Growth Intertwined across Global Value Chains

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Introduction

- Following the recent empirical pioneers in the field (Graetz and Michaels, 2018; Abeliatsky and Prettnner, 2017; and Acemoglu and Restrepo, 2018) and extending the long-run distributed lag framework developed by Autor and Salomons (AS, 2018), this paper provides a comprehensive analysis of the **direct** and **indirect effects** of **industrial robots** on **employment** and **real value-added growth**.
 - The indirect effects capture both **domestic** and **international linkages** along the global value chains (GVCs) which were obtained from inter-country input-output tables.
 - The **expansion of value added** of a given sector could indirectly influence the employment in another sector through **backward** or **forward** linkages.
 - Example: **productivity gains in manufacturing** sectors (higher quality and less expensive products) **transmit to non-robotized services**

Data

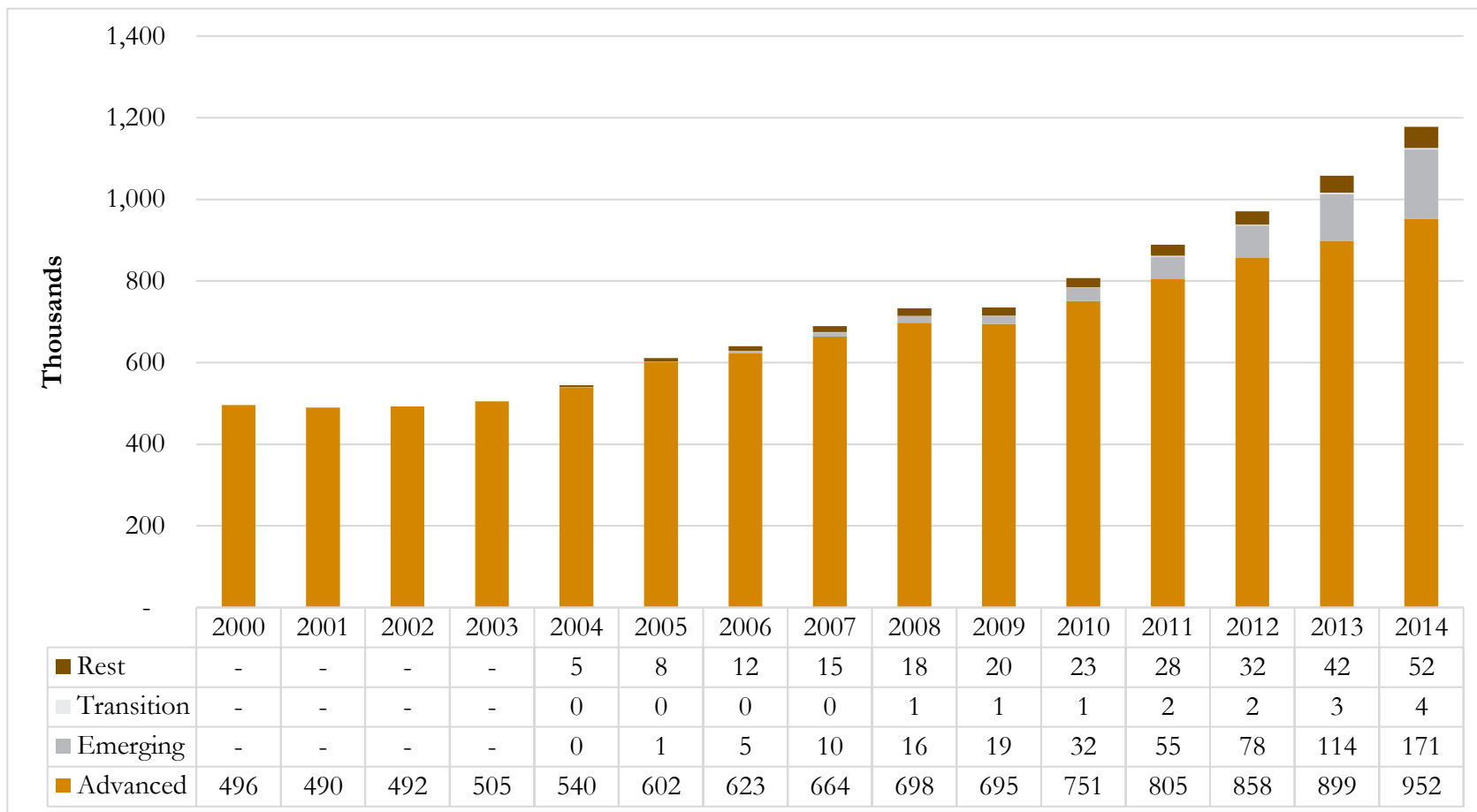
- The econometric model draws on few major data sources:
 1. World Input-Output Database (**WIOD**) (Timmer et al., 2015) including data from accompanying Socio-Economic Accounts (SEA)
 - 44 economies (with the resto of the world), 56 industries over the period 2000-2014.
 2. Stocks of industrial multipurpose robots database collected from the International Federation of Robotics (**IFR**, 2018)
 - Industrial multipurpose robots are defined as: “automatically controlled, reprogrammable multipurpose manipulator programmable in three or more axes”
 3. World Development Indicator (**WDI**) of the World Bank augmented by the Penn World Table (Feenstra et al., 2015)

Countries included in the analysis

| No. | Country | Group | No. | Country | Group |
|-----|----------------|----------|-----|--------------------|------------|
| 1 | Australia | Advanced | 22 | Brazil | Emerging |
| 2 | Austria | Advanced | 23 | China | Emerging |
| 3 | Belgium | Advanced | 24 | Indonesia | Emerging |
| 4 | Canada | Advanced | 25 | India | Emerging |
| 5 | Denmark | Advanced | 26 | Mexico | Emerging |
| 6 | Finland | Advanced | 27 | Turkey | Emerging |
| 7 | France | Advanced | 28 | Bulgaria | Transition |
| 8 | Germany | Advanced | 29 | Romania | Transition |
| 9 | Greece | Advanced | 30 | Russian Federation | Transition |
| 10 | Ireland | Advanced | 31 | Croatia | Rest |
| 11 | Italy | Advanced | 32 | Cyprus | Rest |
| 12 | Japan | Advanced | 33 | Czech Republic | Rest |
| 13 | Luxemburg | Advanced | 34 | Estonia | Rest |
| 14 | Netherlands | Advanced | 35 | Hungary | Rest |
| 15 | Norway | Advanced | 36 | Latvia | Rest |
| 16 | Portugal | Advanced | 37 | Lithuania | Rest |
| 17 | Rep. of Korea | Advanced | 38 | Malta | Rest |
| 18 | Spain | Advanced | 39 | Poland | Rest |
| 19 | Sweden | Advanced | 40 | Slovakia | Rest |
| 20 | United Kingdom | Advanced | 41 | Slovenia | Rest |
| 21 | United States | Advanced | 42 | Switzerland | Rest |
| | | | 43 | Taiwan | Rest |

Data

Stocks of industrial robots by country groups in thousands – 2000-2014



Source: IFR (2018), authors' elaboration.

Data

- **Average annual growth of stocks of industrial robots, 2000-2014**

| Industry Description | World | Advanced | Emerging | Transition | Other |
|----------------------|-------|----------|----------|------------|-------|
| Primary | 7.6% | 6.7% | 61.9% | 34.7% | 42.1% |
| Manufacturing | 6.2% | 4.7% | 67.3% | 52.6% | 24.6% |
| Robotized Services | 2.6% | 1.0% | 68.5% | 32.6% | 10.0% |
| Total | 6.2% | 4.7% | 67.3% | 50.9% | 24.3% |

- Note: Primary includes agriculture, fishing, forestry, and mining. Robotized services include electricity and water supply (DtE), construction (F) and Scientific research and development; Other professional, scientific and technical activities; veterinary activities; Education (MtN&P).
- Source: IFR (2018), WIOD; own calculations.

Data

- Average annual growth of employment in %, 2000-2014

| Industry Description | World | Advanced | Emerging | Transition | Other |
|------------------------|-------|----------|----------|------------|-------|
| Primary | -0.7% | -1.2% | -0.6% | -3.3% | -3.3% |
| Manufacturing | 2.1% | -1.5% | 3.3% | -1.7% | 0.3% |
| Robotized Services | 3.1% | 0.5% | 4.2% | 0.5% | 1.2% |
| Non-robotized Services | 3.0% | 0.9% | 4.5% | 1.8% | 1.5% |
| Total | 1.7% | 0.4% | 2.2% | -0.1% | 0.7% |

- Note: Primary includes agriculture, fishing, forestry, and mining. Robotized services include electricity and water supply (DtE), construction (F) and Scientific research and development; Other professional, scientific and technical activities; veterinary activities; Education (MtN&P).
- Source: WIOD; own calculations.

Data

- **Average annual growth of real value added in %, 2000-2014**

| Industry Description | World | Advanced | Emerging | Transition | Other |
|------------------------|-------|----------|----------|------------|-------|
| Primary | 2.1% | 1.3% | 2.8% | 1.9% | -0.2% |
| Manufacturing | 3.1% | 0.8% | 9.3% | 2.7% | 4.3% |
| Robotized Services | 1.4% | 0.1% | 6.0% | 1.8% | 1.1% |
| Non-robotized Services | 2.3% | 1.5% | 6.1% | 4.0% | 2.4% |
| Total | 2.3% | 1.2% | 6.2% | 3.2% | 2.5% |

- Note: Primary includes agriculture, fishing, forestry, and mining. Robotized services include electricity and water supply (DtE), construction (F) and Scientific research and development; Other professional, scientific and technical activities; veterinary activities; Education (MtN&P).
- Source: WIOD; own calculations.

Methodology

- **Econometric model of TFP (AS, 2018):**

$$\Delta \ln Y_{cit} = \beta_0 + \sum_{k=0}^5 \beta_6^k \Delta \ln TFP_{ci,t-k}^* + \sum_{k=0}^5 \beta_7^k \Delta \ln TFP_{ci,t-k}^{dom-BW} + \sum_{k=0}^5 \beta_8^k \Delta \ln TFP_{ci,t-k}^{dom-FW} + \mu_{ct} + \mu_s + \varepsilon_{ict}$$

$$Y \in \{EMP, HEMP, LSH, VA^{real}, VA^{nominal}\}$$

- $\Delta \ln TFP_{ci,t-k}^*$ is the own-industry other-countries' average TFP growth
- $\Delta \ln TFP_{ci,t-k}^{dom-BW}$ is the TFP growth of domestic suppliers to industry i through domestic backward linkages
- $\Delta \ln TFP_{ci,t-k}^{dom-FW}$ is the TFP growth of domestic buyers to industry i through domestic forward linkages

Methodology

- TFP growth and direct own-industry TFP growth:

$$\Delta \ln TFP_{cit} = \Delta \ln VA_{cit}^{real} - \left(\frac{W_{cit}}{VA_{cit}^{nominal}} * \Delta \ln EMP_{cit} \right) - \left(\left(1 - \frac{W_{cit}}{VA_{cit}^{nominal}} \right) * \Delta \ln K_{cit}^{real} \right)$$

$$\Delta \ln TFP_{cit}^* = \frac{\sum_{f \neq c}^{F^X - 1} \Delta \ln TFP_{fit}}{F^X - 1}, \quad f \in \Phi^X \wedge X \in \{A, A'\}$$

Methodology

- **Extending the AS specification:**
 1. Adding other **WIOD countries** to ensure a cross-country variation that can be econometrically generalised to the world economy.
 2. Adding the change in the stock of **industrial multipurpose robots (R)** at the country-industry level as another proxy for technological change in addition to TFP growth.
 3. Allowing for an **open-economy** setting in the sense that the indirect effects of industrial robots on labour market outcomes and value added also include linkages to industries of foreign countries along the GVC.
 4. Due to the **time dimension** of the WIOD data, the lags are limited to three periods.
 5. **Industry fixed effects** instead of aggregate sector fixed effects are used to control technological change at the global industry level.
 6. Country-time fixed effects are used to control for **business cycles**.

Methodology

- Benchmark econometric model of robots:

$$\begin{aligned}
 \Delta \ln Y_{cit} = & \beta_0 + \sum_{k=0}^3 \beta_1^k \Delta \ln R_{ci,t-k} + \sum_{k=0}^3 \beta_2^k \Delta \ln R_{ci,t-k}^{dom-BW} + \sum_{k=0}^3 \beta_3^k \Delta \ln R_{ci,t-k}^{dom-FW} \\
 & + \sum_{k=0}^3 \beta_4^k \Delta \ln R_{ci,t-k}^{int-BW} + \sum_{k=0}^3 \beta_5^k \Delta \ln R_{ci,t-k}^{int-FW} \\
 & + \sum_{k=0}^3 \beta_6^k \Delta \ln TFP_{ci,t-k}^* + \sum_{k=0}^3 \beta_7^k \Delta \ln TFP_{ci,t-k}^{dom-BW} + \sum_{k=0}^3 \beta_8^k \Delta \ln TFP_{ci,t-k}^{dom-FW} \\
 & + \sum_{k=0}^3 \beta_9^k \Delta \ln TFP_{ci,t-k}^{int-BW} + \sum_{k=0}^3 \beta_{10}^k \Delta \ln TFP_{ci,t-k}^{int-FW} + \mu_{ct} + \mu_i + \varepsilon_{ict}
 \end{aligned}$$

$$Y \in \{EMP, HEMP, LSH, VA^{real}, VA^{nominal}\}$$

Results

■ Benchmark econometric model of robots and TFP – WIOD countries: p1

| Dependent variable: | (1) $\Delta \ln EMP_{cit}$ | (2) $\Delta \ln HEMP_{cit}$ | (3) $\Delta \ln LSH_{cit}$ | (4) $\Delta \ln VA_{cit}^{real}$ | (5) $\Delta \ln VA_{cit}^{nominal}$ |
|---|-------------------------------|--------------------------------|-------------------------------|-------------------------------------|--|
| $\sum_{k=0}^3 \beta_1^k \Delta \ln R_{ci,t-k}$ | .011*** | .01*** | -.001 | .023*** | .009** |
| F-Test of joint significance | (.001) | (.003) | (.67) | (0) | (.031) |
| $\sum_{k=0}^3 \beta_2^k \Delta \ln R_{ci,t-k}^{dom-BW}$ | .024 | .053* | .021 | .007 | .017 |
| F-Test of joint significance | (.239) | (.051) | (.237) | (.801) | (.456) |
| $\sum_{k=0}^3 \beta_3^k \Delta \ln R_{ci,t-k}^{int-BW}$ | .055 | .095** | -.064 | .044 | .19*** |
| F-Test of joint significance | (.157) | (.022) | (.101) | (.478) | (0) |
| $\sum_{k=0}^3 \beta_4^k \Delta \ln R_{ci,t-k}^{dom-FW}$ | -.027* | -.037* | .016 | -.039 | -.054** |
| F-Test of joint significance | (.098) | (.079) | (.326) | (.173) | (.039) |
| $\sum_{k=0}^3 \beta_5^k \Delta \ln R_{ci,t-k}^{int-FW}$ | .037 | .047 | .083*** | .1*** | -.041 |
| F-Test of joint significance | (.219) | (.122) | (.005) | (.006) | (.199) |
| R-sq. | .123 | .139 | .08 | .187 | .261 |
| Obs | 19500 | 19092 | 19500 | 19500 | 19500 |

Note: Sample includes countries as in AS (2018). P-values for the F-test that of joint significance ($\beta^0 + \beta^1 + \beta^2 + \beta^3 = 0$) in parentheses. ***, **, and * indicate statistical significant at the 1%, 5% and 10% level respectively. The size of the coefficients is obtained by summing up the estimated coefficients of the contemporaneous value and the five lagged values. All specifications include country-time fixed effects and sector fixed effects. Estimated with STATA using the *reghdfe* estimation command

Results

■ Benchmark econometric model of robots and TFP – WIOD countries: p2

| Dependent variable: | (1) $\Delta \ln EMP_{cit}$ | (2) $\Delta \ln HEMP_{cit}$ | (3) $\Delta \ln LSH_{cit}$ | (4) $\Delta \ln VA_{cit}^{real}$ | (5) $\Delta \ln VA_{cit}^{nominal}$ |
|--|-------------------------------|--------------------------------|-------------------------------|-------------------------------------|--|
| $\sum_{k=0}^3 \beta_6^k \Delta \ln TFP_{ci,t-k}^*$ | .088 | .072 | -.087 | .141** | .188*** |
| <i>F-Test of joint significance</i> | (.19) | (.314) | (.178) | (.028) | (.002) |
| $\sum_{k=0}^3 \beta_7^k \Delta \ln TFP_{ci,t-k}^{dom-BW}$ | .178 | .334** | -.032 | .47*** | .374** |
| <i>F-Test of joint significance</i> | (.149) | (.016) | (.778) | (.001) | (.014) |
| $\sum_{k=0}^3 \beta_9^k \Delta \ln TFP_{ci,t-k}^{int-BW}$ | .266 | .353* | -.373* | .991*** | .358 |
| <i>F-Test of joint significance</i> | (.179) | (.086) | (.083) | (.001) | (.15) |
| $\sum_{k=0}^3 \beta_8^k \Delta \ln TFP_{ci,t-k}^{dom-FW}$ | .061 | .11 | -.008 | -.152 | .036 |
| <i>F-Test of joint significance</i> | (.405) | (.126) | (.927) | (.156) | -0.685 |
| $\sum_{k=0}^3 \beta_{10}^k \Delta \ln TFP_{ci,t-k}^{int-FW}$ | .252 | .293* | -.4* | .605** | 1.254*** |
| <i>F-Test of joint significance</i> | (.123) | (.078) | (.082) | (.015) | (0) |
| R-sq. | .123 | .139 | .08 | .187 | .261 |
| Obs | 19500 | 19092 | 19500 | 19500 | 19500 |

Note: Sample includes countries as in AS (2018). P-values for the F-test that of joint significance ($\beta^0 + \beta^1 + \beta^2 + \beta^3 = 0$) in parentheses. ***, **, and * indicate statistical significant at the 1%, 5% and 10% level respectively. The size of the coefficients is obtained by summing up the estimated coefficients of the contemporaneous value and the five lagged values. All specifications include country-time fixed effects and sector fixed effects. Estimated with STATA using the *reghdfe* estimation command

Results

- **Potential explanations on positive direct impact of robots on employment growth:**
 1. In the **Ricardo-Viner** model or the **Heckscher-Ohlin** framework (i.e. with capital mobile across industries) **an increase in capital would shift employment** to the capital-intensive industries.
 2. **Old** vintages of **machineries** could also be **replaced** and upgraded by newer machineries (or **robots**) as a form of **process innovation**. Growth of capital and growth of stocks of robots are **not significantly correlated** with each other (including fixed FE).
 3. Smart machines are replacing **unskilled labour**, while complementing skilled labour, which depends on all **substitution elasticities** across production factors.

Results

- **Potential explanations on negative impact of robots in domestic forward linkages on employment growth:**
 1. One reason might be that the new machineries in the downstream industry require less demand for inputs from the upstream industries.
 2. Another reason could be that digitalisation in a downstream industry allows industries to take over some tasks previously undertaken in the upstream industries.

Methodology

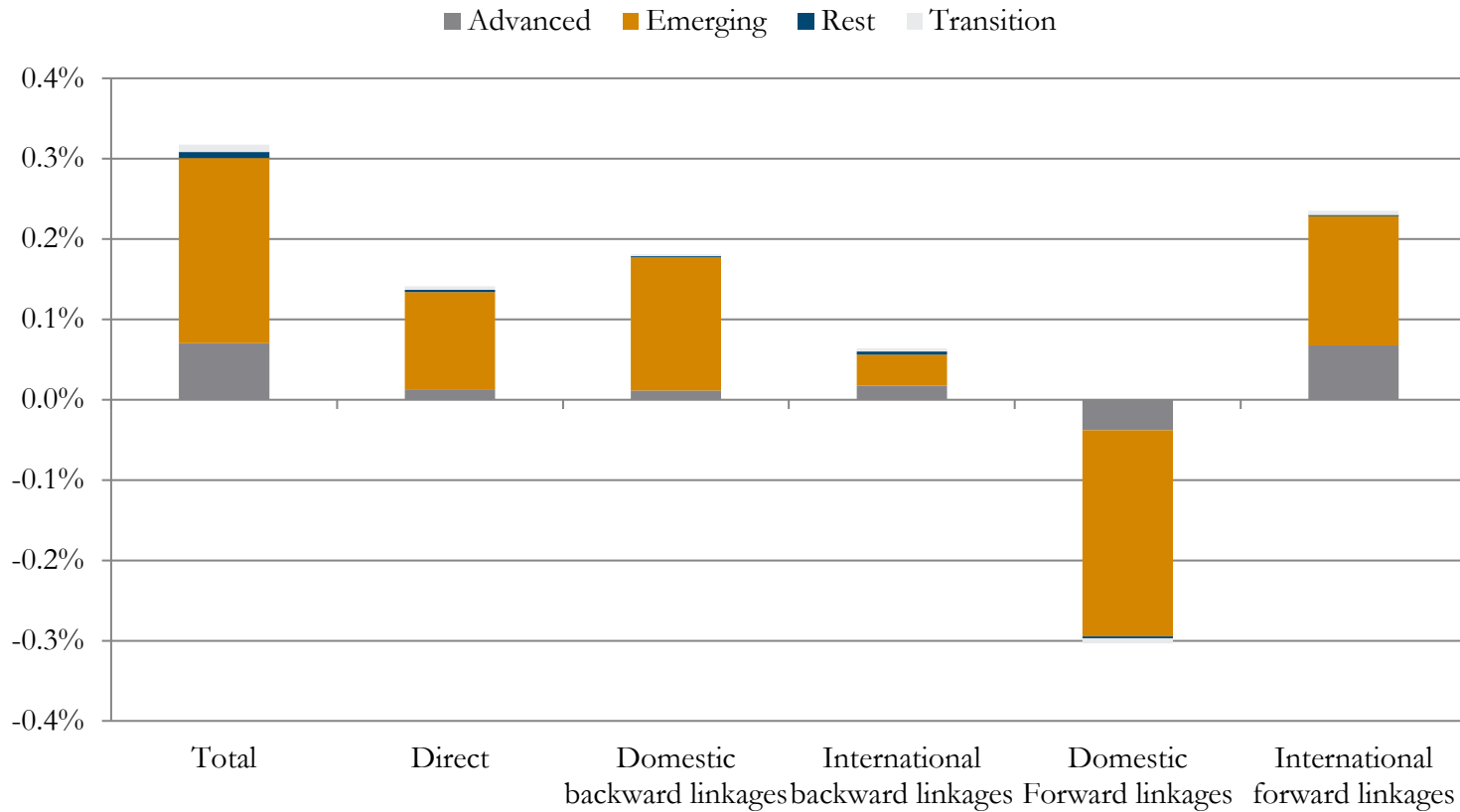
- Quantitative implications based on model predictions – **Destination perspective:**
- Estimation results are now used to retrieve the implied contribution of growth in robot stocks on changes of employment and real value-added

$$\widehat{\Delta \ln Y_t^E} = \sum_{k=0}^3 \hat{\beta}_1^{kY} \sum_c^C \sum_i^I \left[\left(\frac{1}{T} \cdot \sum_t^T \frac{Y_{cit}}{Y_t} \right) \Delta \ln R_{cit}^E \right]$$

$$Y \in \{\text{EMP}, \text{VA}^{\text{real}}\}, \quad E \in \{\text{Direct}, \text{dom} - \text{BW}, \text{int} - \text{BW}, \text{dom} - \text{FW}, \text{int} - \text{FW}\}$$

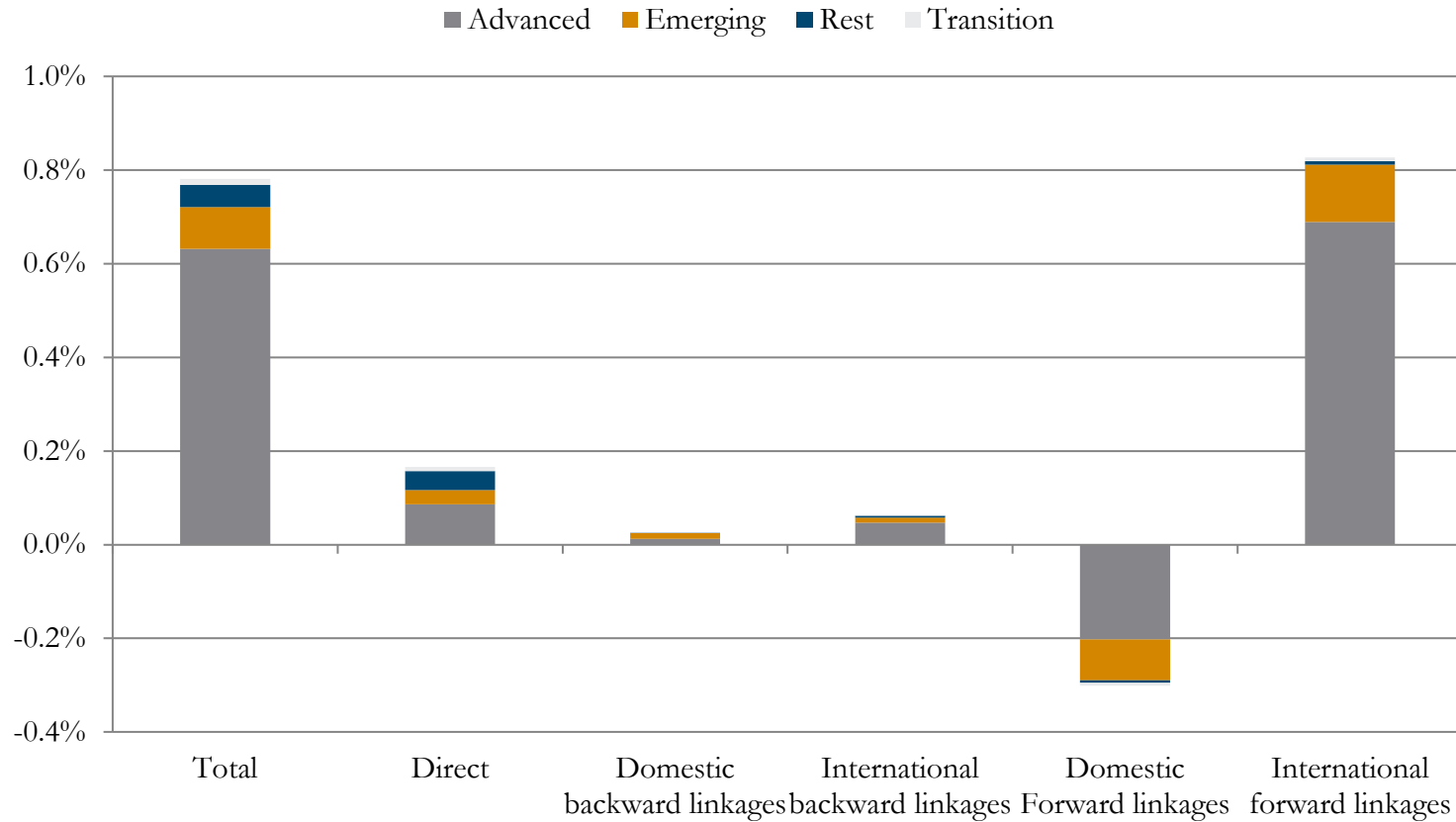
Results

- Predicted effects of the per annum growth of robots on economy-wide employment, **Destination:**



Results

- Predicted effects of the per annum growth of robots on economy-wide real value-added, **Destination:**



Methodology

- Quantitative implications based on model predictions – **Origin perspective**:
- Which country or industry introduced new robots and is therefore originally responsible for the employment (or value added) that has been generated in the destination country or industry?

$$\Delta \ln Y_t^{\widehat{E}^{\text{dom-origin}}} = \sum_{k=0}^3 \hat{\beta}_1^{kY} \sum_c^C \sum_{j \neq i}^J \Delta \ln R_{cjt} \left[\sum_i^I \left(\frac{1}{T} \cdot \sum_t^T \frac{Y_{cit}}{Y_t} \right) \Gamma_{cjt,cit} \right]$$

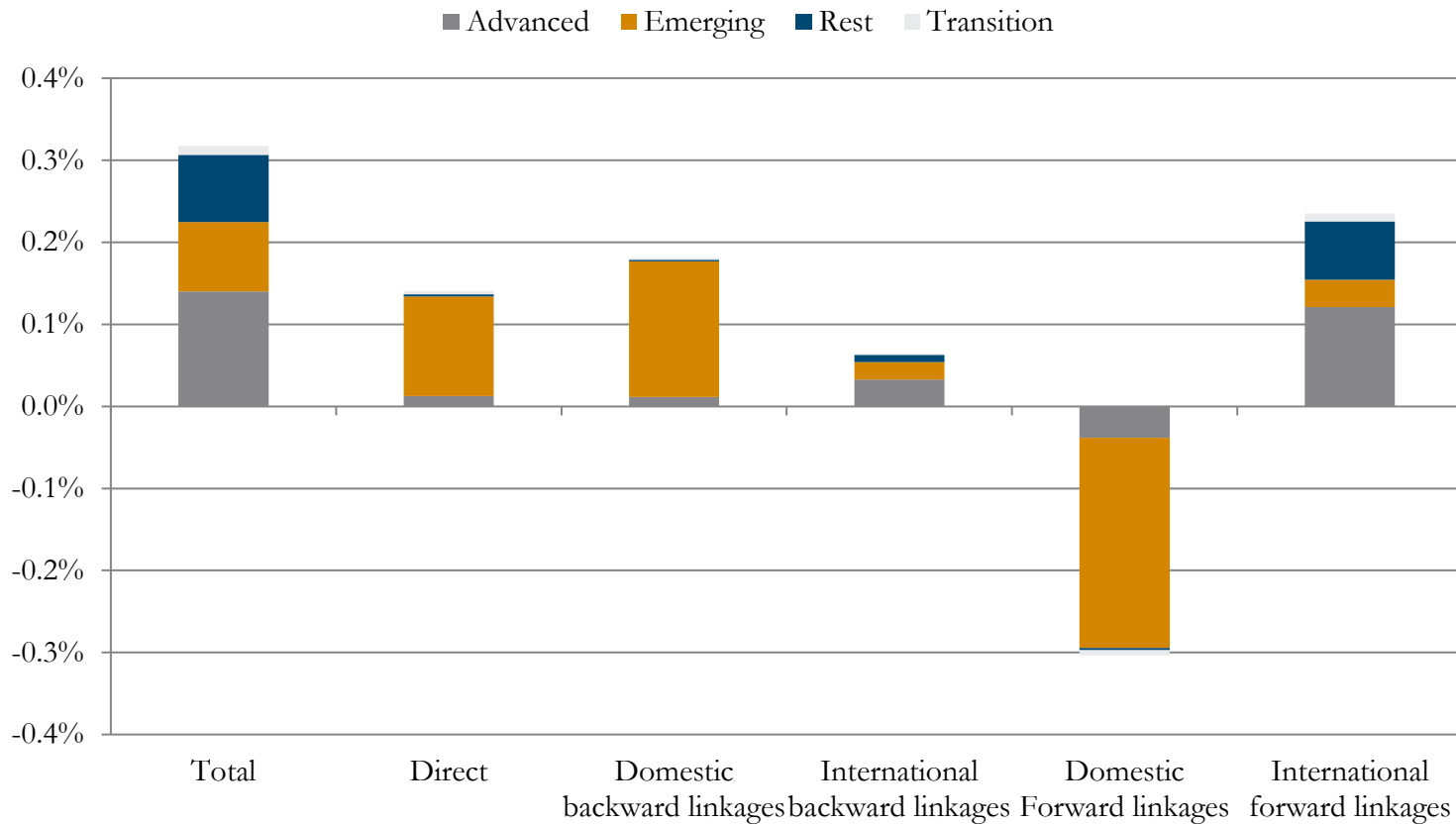
$$Y \in \{EMP, VA^{real}\}, \quad E \in \{dom - BW, dom - FW\}, \quad \Gamma \in \{l, g\}$$

$$\Delta \ln Y_t^{\widehat{E}^{\text{int-origin}}} = \sum_{k=0}^3 \hat{\beta}_1^{kY} \sum_c^C \sum_{j \neq i}^J \Delta \ln R_{cjt} \left[\sum_i^I \left(\frac{1}{T} \cdot \sum_t^T \frac{Y_{cit}}{Y_t} \right) \sum_{f \neq c}^F \Gamma_{fjt,cit} \right]$$

$$Y \in \{EMP, VA^{real}\}, \quad , \quad E \in \{int - BW, int - FW\}, \quad \Gamma \in \{l, g\}$$

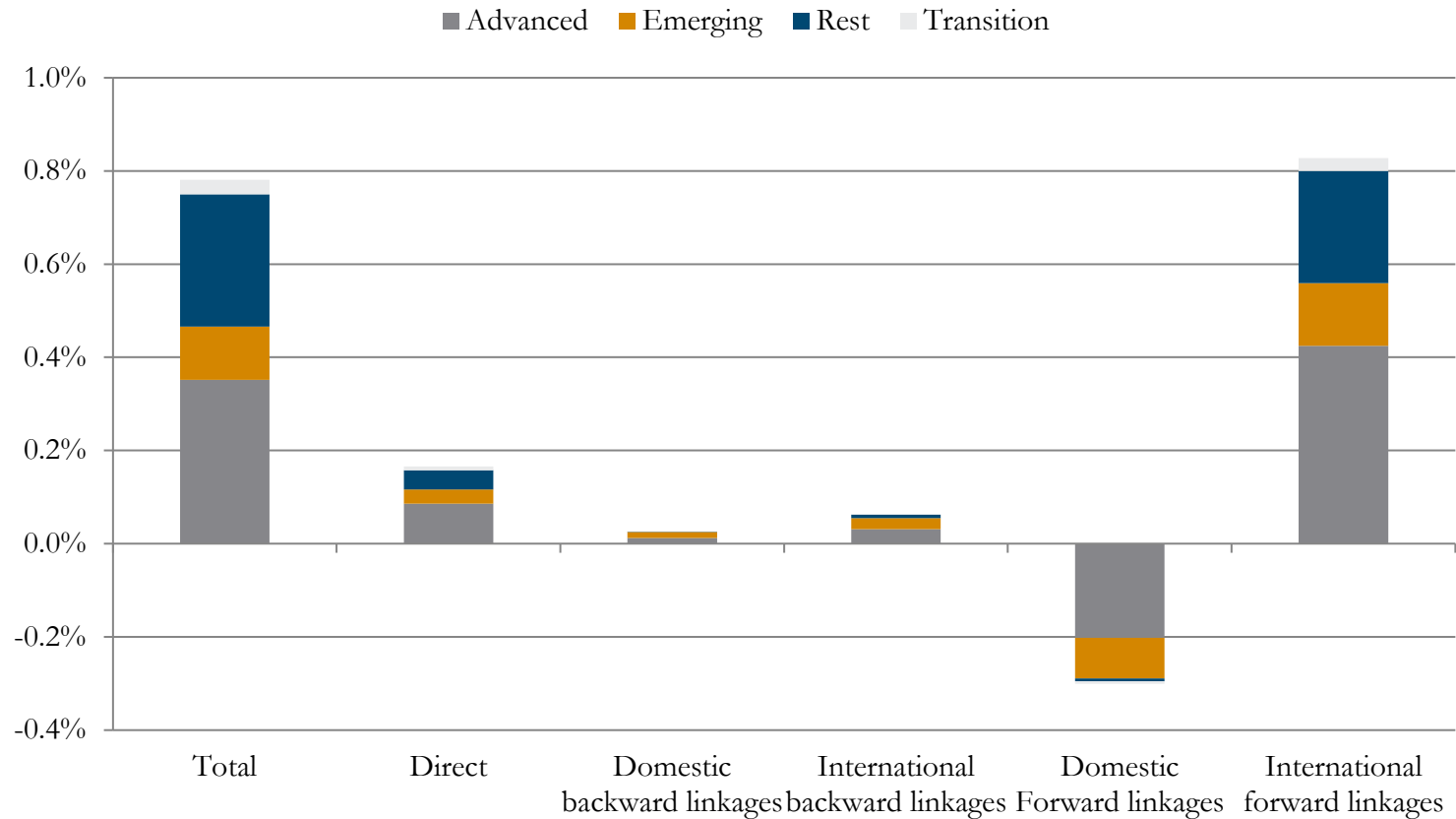
Results

- Predicted effects of the per annum growth of robots in the **origin perspective**, WIOD average
- Employment growth



Results

- Predicted effects of the per annum growth of robots in the **origin perspective, WIOD average**
- Real value-added growth



Concluding Remarks

- This study analysed the **role of robotisation** in the **global economy** by taking the spillover effects of the impacts of TFP growth and robotisation on the **global value chains (GVCs)** into account.
- **Growth in stocks of industrial robots** in an industry **improves the employment growth** and **real value added growth** of the respective industry at a 1% level of significance.
- **Growth in the stocks of industrial robots in suppliers** of an industry that is accumulated along the **domestic supply chains** and the one that is accumulated along the **international backward linkages** improve **employment in hours**, while the **latter** improves also **real value added**.
- However, growth in the stocks of robots in **domestic forward linkages** **reduces employment and value added growth**. Moreover, growth in the stocks of **robots in international forward linkages** **reduces real value added growth**.
- Global robots adoption **contributed** mostly to the **real value added growth in advanced** economies, whereas it contributed mostly to the **employment growth of non-advanced** economies. However, the positive impact of robots adoption **originates mostly from advanced** economies.

**Thank you for
your attention**



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Results

■ Econometric model of TFP (AS, 2018) – selected advanced countries

| Dependent variable: | (1) | (2) | (3) | (4) | (5) |
|--|------------------------|-------------------------|------------------------|------------------------------|---------------------------------|
| | $\Delta \ln EMP_{cit}$ | $\Delta \ln HEMP_{cit}$ | $\Delta \ln LSH_{cit}$ | $\Delta \ln VA_{cit}^{real}$ | $\Delta \ln VA_{cit}^{nominal}$ |
| $\sum_{k=0}^5 \beta_6^k \Delta \ln TFP_{ci,t-k}^*$ | -0.39*** | -0.34*** | .092 | .095 | .022 |
| <i>F-Test of joint significance</i> | (0) | (0) | (.767) | (.434) | (.831) |
| $\sum_{k=0}^5 \beta_7^k \Delta \ln TFP_{ci,t-k}^{dom-BW}$ | .708*** | .869** | .148 | .467 | .627* |
| <i>F-Test of joint significance</i> | (.008) | (.021) | (.65) | (.156) | (.085) |
| $\sum_{k=0}^5 \beta_9^k \Delta \ln TFP_{ci,t-k}^{int-BW}$ | -.903** | -1.176*** | -.126 | .327 | -1.529*** |
| <i>F-Test of joint significance</i> | (.011) | (.003) | (.795) | (.673) | (.001) |
| $\sum_{k=0}^5 \beta_8^k \Delta \ln TFP_{ci,t-k}^{dom-FW}$ | .03 | .017 | -.375* | -.103 | .433* |
| <i>F-Test of joint significance</i> | (.742) | (.89) | (.068) | (.517) | (.097) |
| $\sum_{k=0}^5 \beta_{10}^k \Delta \ln TFP_{ci,t-k}^{int-FW}$ | 1.011*** | 1.093** | -.28 | .615 | 2.251*** |
| <i>F-Test of joint significance</i> | (.002) | (.017) | (.441) | (.207) | (0) |
| Weight | Employment | Hours worked | value added | value added | value added |
| R-sq. | .328 | .35 | .15 | .242 | .289 |
| Obs | 8036 | 8036 | 8036 | 8036 | 8036 |

Note: Sample includes countries as in AS (2018). P-values for the F-test that of joint significance ($\beta^0 + \beta^1 + \beta^2 + \beta^3 + \beta^4 + \beta^5 = 0$) in parentheses. ***, **, and * indicate statistical significant at the 1%, 5% and 10% level respectively. The size of the coefficients is obtained by summing up the estimated coefficients of the contemporaneous value and the five lagged values. All specifications include country-time fixed effects and *sector* fixed effects. Estimated with STATA using the *reghdfe* estimation command

Results

- Extension to econometric model of TFP (AS, 2018) – WIOD countries:

| Dependent variable: | (1) $\Delta \ln EMP_{cit}$ | (2) $\Delta \ln HEMP_{cit}$ | (3) $\Delta \ln LSH_{cit}$ | (4) $\Delta \ln VA_{cit}^{real}$ | (5) $\Delta \ln VA_{cit}^{nominal}$ |
|--|-------------------------------|--------------------------------|-------------------------------|-------------------------------------|--|
| $\sum_{k=0}^3 \beta_6^k \Delta \ln TFP_{ci,t-k}^*$ | .087 | .076 | -.066 | .121* | .163*** |
| <i>F-Test of joint significance</i> | (.167) | (.258) | (.284) | (.057) | (.008) |
| $\sum_{k=0}^3 \beta_7^k \Delta \ln TFP_{ci,t-k}^{dom-BW}$ | .123 | .235** | -.009 | .238* | .209* |
| <i>F-Test of joint significance</i> | (.205) | (.037) | (.923) | (.05) | (.095) |
| $\sum_{k=0}^3 \beta_9^k \Delta \ln TFP_{ci,t-k}^{int-BW}$ | .204 | .293 | -.475** | .909*** | .37* |
| <i>F-Test of joint significance</i> | (.247) | (.114) | (.015) | (.001) | (.095) |
| $\sum_{k=0}^3 \beta_8^k \Delta \ln TFP_{ci,t-k}^{dom-FW}$ | -.009 | -.014 | .004 | -.152* | -.021 |
| <i>F-Test of joint significance</i> | (.876) | (.817) | (.934) | (.072) | (.807) |
| $\sum_{k=0}^3 \beta_{10}^k \Delta \ln TFP_{ci,t-k}^{int-FW}$ | .386*** | .434*** | -.375* | .647*** | 1.277*** |
| <i>F-Test of joint significance</i> | (.005) | (.004) | (.069) | (.006) | (0) |
| R-sq. | 0.118 | 0.135 | 0.079 | 0.174 | 0.246 |
| Obs | 20,609 | 20,191 | 20,609 | 20,609 | 20,609 |

Note: Sample includes countries as in AS (2018). P-values for the F-test that of joint significance ($\beta^0 + \beta^1 + \beta^2 + \beta^3 = 0$) in parentheses. ***, **, and * indicate statistical significant at the 1%, 5% and 10% level respectively. The size of the coefficients is obtained by summing up the estimated coefficients of the contemporaneous value and the five lagged values. All specifications include country-time fixed effects and sector fixed effects. Estimated with STATA using the *reghdfe* estimation command

Data

- Effects by industry, employment per annum growth, origin perspective:

| Industry Description | Total | Direct | Domestic Supplier (BW) | International Supplier (BW) | Domestic Buyer (FW) | International Buyer (FW) |
|------------------------|--------------|--------------|------------------------|-----------------------------|---------------------|--------------------------|
| Primary | 0.08% | 0.06% | 0.01% | 0.01% | -0.02% | 0.02% |
| Manufacturing | 0.21% | 0.05% | 0.15% | 0.05% | -0.26% | 0.21% |
| Robotized Services | 0.04% | 0.03% | 0.02% | 0.00% | -0.03% | 0.01% |
| Non-robotized Services | | | | | | |
| Total | 0.32% | 0.14% | 0.18% | 0.06% | -0.30% | 0.24% |

Results

- Effects by industry, employment per annum growth, destination perspective:

| Industry Description | Total | Direct | Domestic Supplier (BW) | International Supplier (BW) | Domestic Buyer (FW) | International Buyer (FW) |
|------------------------|-------|--------|------------------------|-----------------------------|---------------------|--------------------------|
| Primary | 0.08% | 0.06% | 0.04% | 0.01% | -0.08% | 0.05% |
| Manufacturing | 0.13% | 0.05% | 0.04% | 0.02% | -0.05% | 0.06% |
| Robotized Services | 0.04% | 0.03% | 0.03% | 0.01% | -0.11% | 0.06% |
| Non-robotized Services | 0.07% | 0.00% | 0.06% | 0.02% | -0.07% | 0.06% |
| Total | 0.32% | 0.14% | 0.18% | 0.06% | -0.30% | 0.24% |

Data

- Effects by industry, real value-added per annum growth, origin perspective:

| Industry Description | Total | Direct | Domestic Supplier (BW) | International Supplier (BW) | Domestic Buyer (FW) | International Buyer (FW) |
|------------------------|--------------|--------------|------------------------|-----------------------------|---------------------|--------------------------|
| Primary | 0.02% | 0.01% | 0.00% | 0.00% | -0.02% | 0.03% |
| Manufacturing | 0.74% | 0.14% | 0.02% | 0.06% | -0.25% | 0.77% |
| Robotized Services | 0.02% | 0.02% | 0.00% | 0.00% | -0.03% | 0.02% |
| Non-robotized Services | | | | | | |
| Total | 0.78% | 0.17% | 0.03% | 0.06% | -0.30% | 0.83% |

Results

- Effects by industry, real value added per annum growth, destination perspective:

| Industry Description | Total | Direct | Domestic Supplier (BW) | International Supplier (BW) | Domestic Buyer (FW) | International Buyer (FW) |
|------------------------|-------|--------|------------------------|-----------------------------|---------------------|--------------------------|
| Primary | 0.02% | 0.01% | 0.00% | 0.00% | -0.02% | 0.02% |
| Manufacturing | 0.36% | 0.14% | 0.01% | 0.03% | -0.06% | 0.25% |
| Robotized Services | 0.12% | 0.02% | 0.00% | 0.01% | -0.07% | 0.16% |
| Non-robotized Services | 0.28% | 0.00% | 0.01% | 0.02% | -0.16% | 0.40% |
| Total | 0.78% | 0.17% | 0.03% | 0.06% | -0.30% | 0.83% |