

# **The contribution of robots to productivity and GDP growth in advanced economies over 1960-2022**

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# 1. Introduction: motivation and context

## ➤ Abundant literature concerning robot impact on...

### ○ Employment

- Among others: Brynjolfsson & McAfee (2014); Autor (2015); Acemoglu & Restrepo (2020); Acemoglu *et al.* (2020); Aghion *et al.* (2020); Aghion *et al.* (2022); Acemoglu *et al.* (2023), ...
- For Kapetanious & Pissarides (2024) or Shahin *et al.* (2024), the substitution between robots and jobs depends on the country's innovation capabilities, and on openness
- No consensual results...

### ○ Labor share

- Among others: Koch & Manuylov (2023) or Shahin *et al.* (2024), both based on analyses on Spanish data, and Shimizu & Momoda (2023), through a theoretical approach, ...
- Consensual results: robot diffusion => labor share decline

# 1. Introduction: motivation and context

- **Few papers concerning robot impact on productivity**
  - **Graetz & Mitchels (2015, 2018)**
    - Industry level database of 17 countries from 1993 to 2007
    - Increased robots per worker => +0.36pp to annual productivity growth
  - **Acemoglu *et al.* (2020)**
    - French firm level database, manufacturing sector, 2010, 2015
    - Positive impact on productivity
  - **Cette, Devillard & Spiezia (2021, 2022)**
    - Country level analysis
    - 30 advanced countries from 1960 to 2019
    - Use elasticities from G&M (2015, 2018)
  - **Bekhtiar *et al.* (2024)**
    - Same type of data as G&M
    - G&M suffer from a positive because of industry-level heterogeneity
    - They halve the impact
  - **Almeida and Sequeira (2024)**
    - Same type of data as G&M
    - Elasticity of productivity to robotization has at least halved from 2008

# 1. Introduction: motivation and context

## ➤ Our paper...

- Proposes **new evaluation** of robot impact on productivity growth
- **Country level analysis, 29 countries, 1960 to 2022**
- Standard **growth accounting approach**
  - Value added elasticity to robots: share of robots in total input remuneration
  - For this, need to estimate robot user cost
- **Two methodologies** to estimate robot user cost
  - **Methodology 1:** Use the Jorgenson (1963) relation  
Price of robots proxied by the US price index of “*information processing equipment*” from the BEA
  - **Methodology 2:** derived from elasticity of labor productivity to robots estimated by G&M (2015)

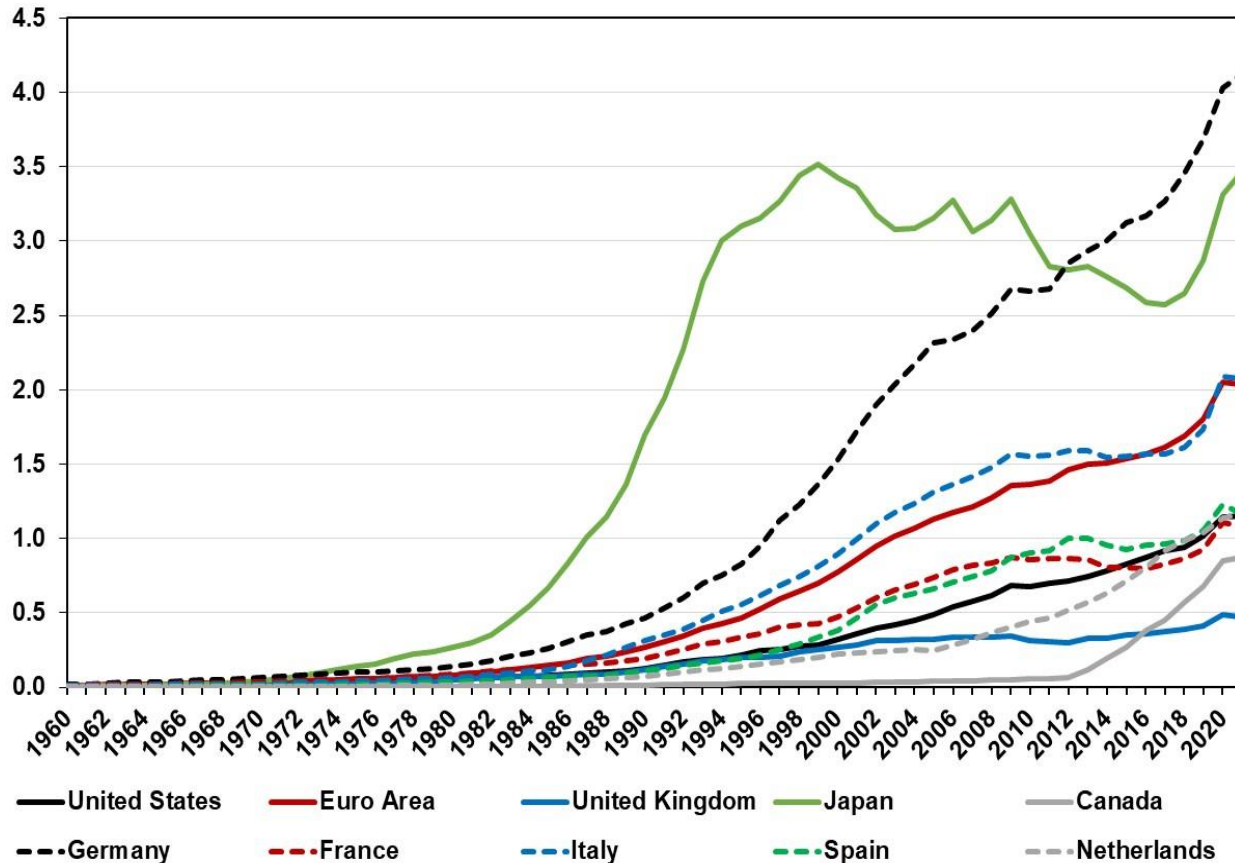
## 2. Data

- **An update of the database built by Cette, Devillard & Spiezia (2021 & 2022)**
  - **Country level, from 1950 to 2022, 29 advanced countries, Euro Area** reconstituted
  - **GDP, employment, hours:** Bergeaud *et al.* (2016), OECD databases, Conference Board TED, & several other databases
  - **Interest rates:** Jordà *et al.* (2019), OECD databases
  - **Robots:**
    - **Industrial robots only**  
ISO definition (ISO 8373:2012): an “*automatically controlled, reprogrammable multipurpose manipulator programmable in three or more axes*”
    - **Number of robots:** IFR completed by us through estimates on ICTs

### 3. Robot diffusion

#### ➤ Chart 1. Robot diffusion, 1960-2021

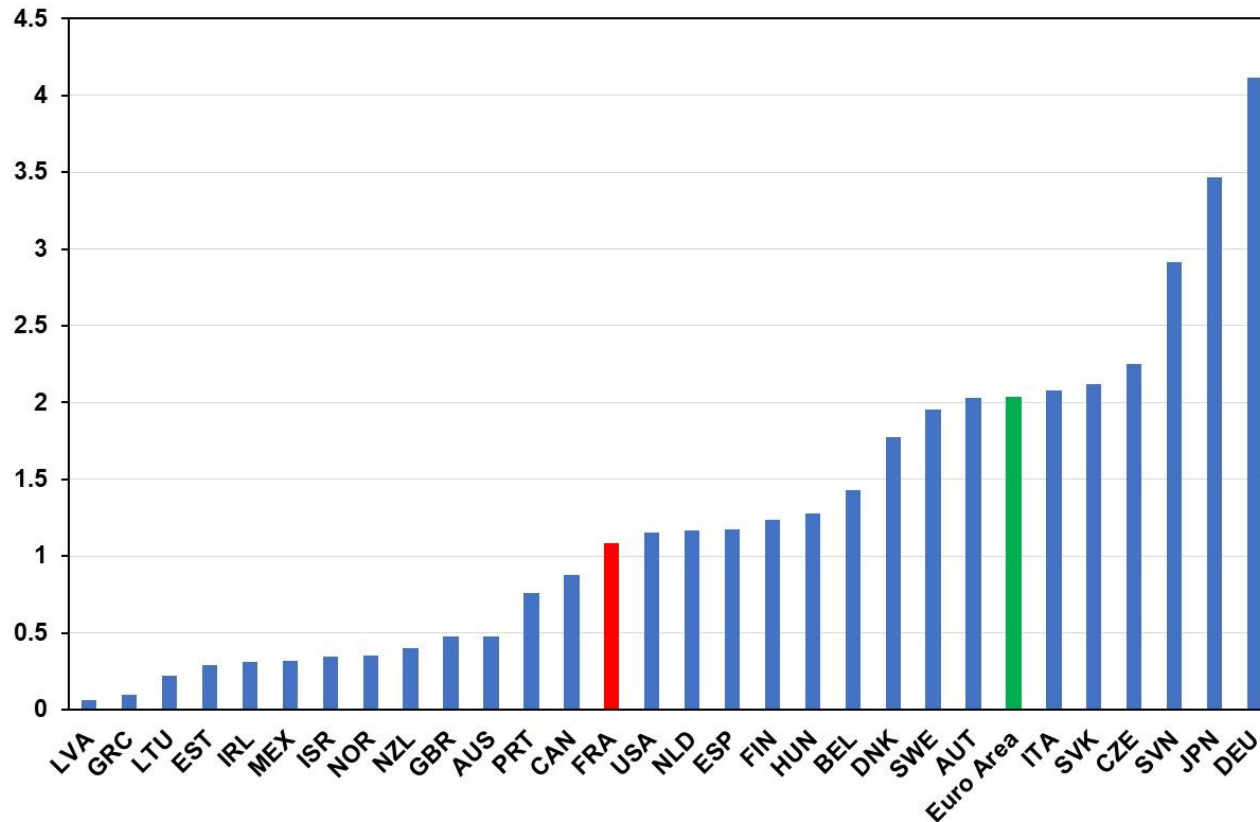
Number of robots per million hours worked in selected countries



- Continuously increasing diffusion in all countries, except Japan from the early 2000s to the mid 2010s
- Explained by the robot price decrease: G&M (2018) estimate that the price of robots in six major developed economies (France, Germany, Italy, Sweden, the United Kingdom and the United States) in 1990-2005 fell by about 50% in nominal terms and 80% when adjusted for quality

### 3. Robot diffusion

➤ Chart 2. Robot diffusion, 2021  
Number of robots per million hours worked



- Contrasted level of robot diffusion
- In 2021, highest diffusion: Germany, Japan
- The observed patterns of diffusion also reflect country-specific specializations: robots tend to be concentrated in few manufacturing sectors



## 4. The two methodologies

### ➤ Growth accounting approach (Solow approach)

### ➤ Capital deepening contribution channel

- $RCG_t = \alpha r_t \cdot (\Delta kr_{t-1} - \Delta n_t - \Delta h_t)$

$kr$ : log of robot capital,  $n$ : log of employment;  $h$ : log of hours per worker

- Törnquist index on  $\alpha r_t = (CR_t \cdot KR_{t-1}) / (PQ_t \cdot Q_t)$

$CR$ : user cost of robot capital,  $KR$  robot capital, constant price quality adjusted;  $PQ$ :  $GDP$  price;  $Q$ :  $GDP$  in constant price

## 4. The two methodologies

### ➤ Methodology 1

- $CR_t = PR_t \cdot (i_t - \Delta pr_t + \delta R)$  following Jorgenson (1963)  
 $PR$ : price of the robots, quality adjusted;  $pr$ : in log;  $i$ : interest rate;  
 $\delta R$ : depreciation rate of robots
- $\delta R = 10\%$  per year, corresponds to IFR hypothesis  
This rate is higher than 4% to 7% proposed by Klump *et al.* (2021)  
Evaluations robust to other values ( $\delta R = 5\%$  or  $\delta R = 15\%$  or  $20\%$ )
- Calculation of  $KR$  is made in two steps
  - We calculate the stock of robots in current value:  $KRCV = NR * UVR$   
 $NR$ : number of robots,  $UVR$ : unitary value of robots, from IFR completed by us
  - We divide  $KRCV$  by a quality-adjusted robot price index ( $QARP$ )  
 $QARP$ : US '*information processing equipment*' price index computed by BEA  
From 1990 to 2005, the annual decrease of this index is 8%, almost the same as G&M (2015)

## 4. The two methodologies

### ➤ Methodology 2

- Same approach as Cette, Devillard & Spiezia (2021, 2022)
- $CR_t$  is derived from cost-minimisation based on G&M (2015, 2018) elasticities of labor productivity to the number of robots
- From this methodology, the elasticity of labor productivity to the number of robots per hour worked comes
  - For  $\frac{3}{4}$  via the TFP channel
  - For  $\frac{1}{4}$  via the capital deepening channel

## 5. Results

### ➤ Results with methodology 1

#### ○ Annual robots' contribution to growth via capital deepening (in pp)

	1960-1975	1975-1995	1995-2005	2005-2022	1960-2022
<b>United States</b>	0.0010	0.0010	0.0020	0.0011	0.0011
<b>Euro Area</b>	0.0001	0.0012	0.0046	0.0021	0.0020
<b>United Kingdom</b>	-0.0001	0.0008	0.0015	0.0004	0.0008
<b>Japan</b>	0.0000	0.0001	0.0001	0.0000	0.0000
<b>Canada</b>	0.0000	0.0002	0.0002	0.0010	0.0004
<b>Germany</b>	0.0003	0.0020	0.0088	0.0036	0.0033
<b>France</b>	0.0001	0.0008	0.0025	0.0009	0.0011
<b>Italy</b>	-0.0003	0.0015	0.0052	0.0024	0.0025
<b>Spain</b>	0.0003	0.0006	0.0034	0.0019	0.0017
<b>The Netherlands</b>	0.0001	0.0004	0.0008	0.0011	0.0006

- Robots' contribution to growth is very small according to methodology 1

## 5. Results

### ➤ Results with methodology 2

#### ○ Annual robots' contribution to growth via capital deepening and TFP (pp)

		1960-1975	1975-1995	1995-2005	2005-2022	1960-2022
United States	<i>Cap. Deep.</i>	0.00	0.01	0.03	0.04	0.02
	<i>TFP</i>	0.01	0.03	0.08	0.12	0.06
Euro Area	<i>Cap. Deep.</i>	0.00	0.02	0.07	0.06	0.04
	<i>TFP</i>	0.01	0.07	0.25	0.19	0.12
United Kingdom	<i>Cap. Deep.</i>	0.00	0.01	0.02	0.01	0.01
	<i>TFP</i>	0.01	0.03	0.05	0.03	0.03
Japan	<i>Cap. Deep.</i>	0.01	0.19	-0.03	0.00	0.07
	<i>TFP</i>	0.03	0.59	-0.09	0.00	0.19
Canada	<i>Cap. Deep.</i>	0.00	0.00	0.00	0.05	0.00
	<i>TFP</i>	0.00	0.00	0.01	0.17	0.06
Germany	<i>Cap. Deep.</i>	0.01	0.04	0.15	0.14	0.08
	<i>TFP</i>	0.02	0.13	0.48	0.43	0.25
France	<i>Cap. Deep.</i>	0.00	0.02	0.04	0.02	0.02
	<i>TFP</i>	0.01	0.05	0.13	0.07	0.06
Italy	<i>Cap. Deep.</i>	0.00	0.03	0.09	0.04	0.05
	<i>TFP</i>	0.01	0.09	0.29	0.13	0.13
Spain	<i>Cap. Deep.</i>	0.00	0.01	0.07	0.03	0.03
	<i>TFP</i>	0.01	0.03	0.23	0.11	0.08
Netherlands	<i>Cap. Deep.</i>	0.00	0.01	0.01	0.06	0.02
	<i>TFP</i>	0.00	0.02	0.04	0.18	0.07

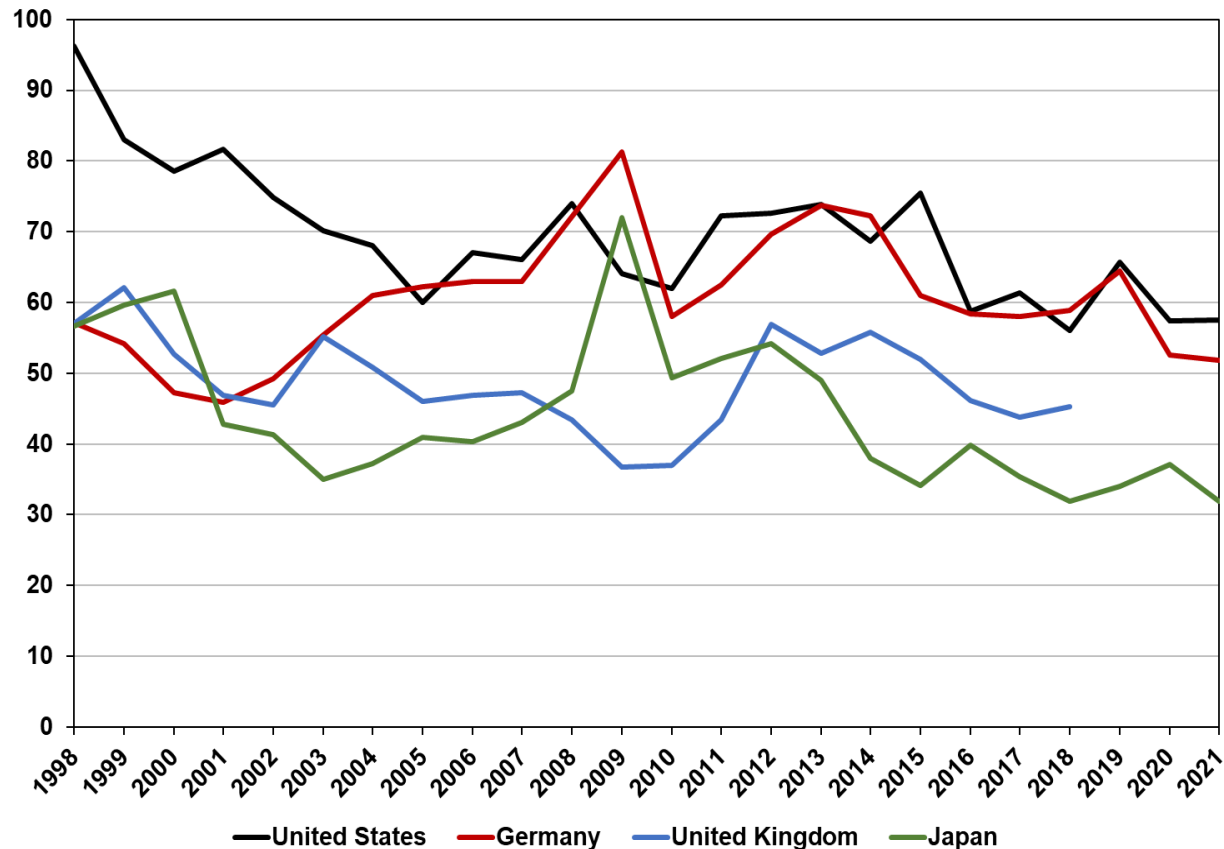
- Higher contribution than with methodology 1, consistent with other evaluations (G&M, 2015 & 2018, and Cette, Devillard & Spiezia, 2021 & 2022 ...)
- Average yearly robots' contribution to productivity growth appears the largest in Germany and Japan

## 6. Why so low evaluation with methodology 1?

### ➤ More likely explanation:

IFR data may underestimate the value of the stock of robots because of an undervaluation of the unitary value of robots (*UVR*)

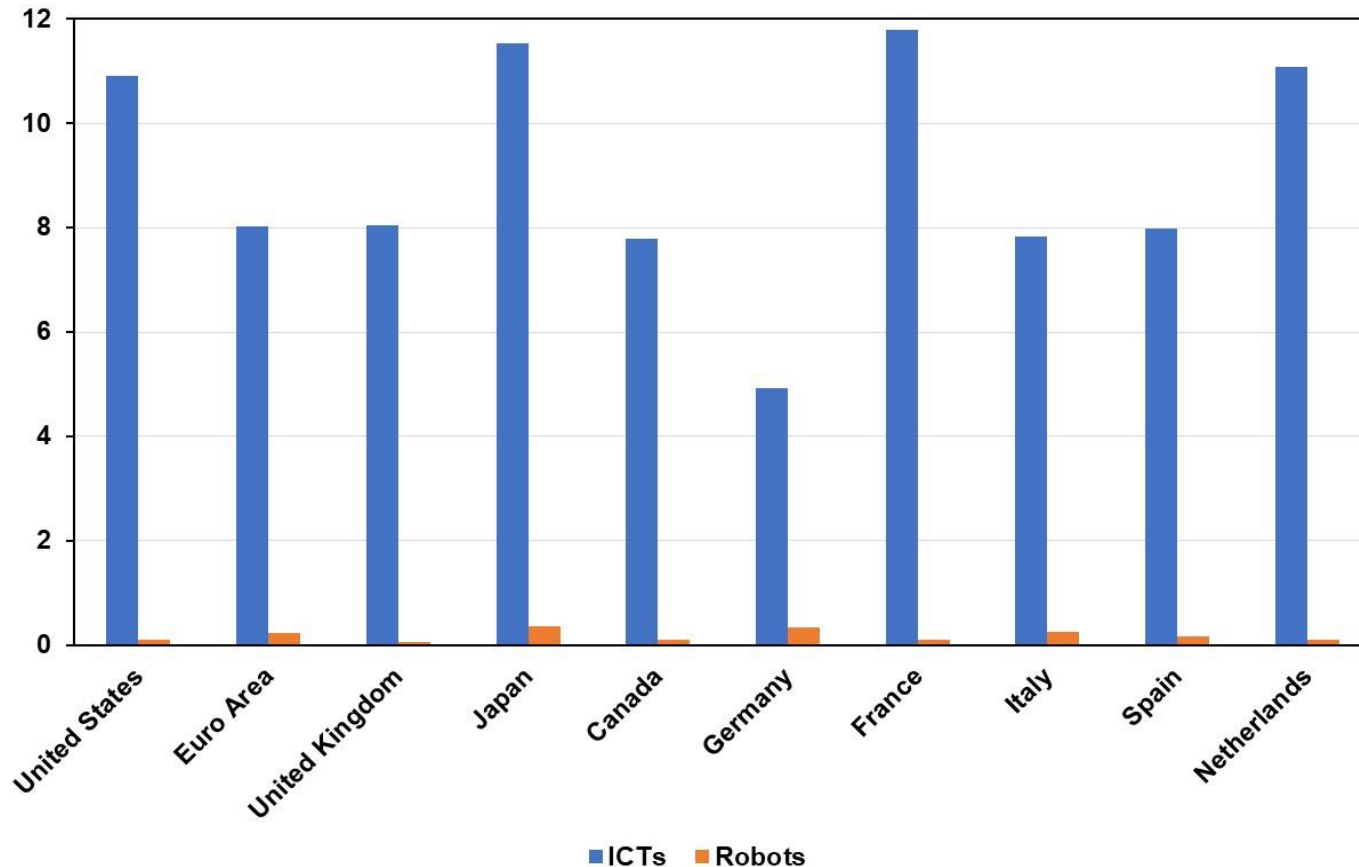
#### ○ Unitary value of robots (*UVR*) (USD thousand)



#### ○ Low *UVR* from IFR data

## 6. Why so low evaluation with methodology 1?

- Low value of robot stock
- Ratios of the ICT capital stock and the robot capital stock to GDP, current value, in %, 2021



- And consequently, low value of the robot capita stock compared to ICTs for instance

## 7. To conclude

- **Few evaluations** of robot contribution to growth at the country level
- We have proposed **two evaluations**, giving contrasted results
  - **Methodology 1 must underestimate** robot contribution, due to a low evaluation of the Unitary Value of Robots (*UVR*) by IFR
  - **Methodology 2: higher impact**, but still based on G&M (2015, 2018) estimates
    - But G&M (2015, 2018) may overestimate robot contribution, see for instance Bekhtiar *et al.* (2024) or Almeida and Sequeira (2024)
    - => Robot contribution to productivity growth would be large only in some manufacturing industries where the use of robots is itself large
- **Need for further research** on the robots' contribution to growth at the country level