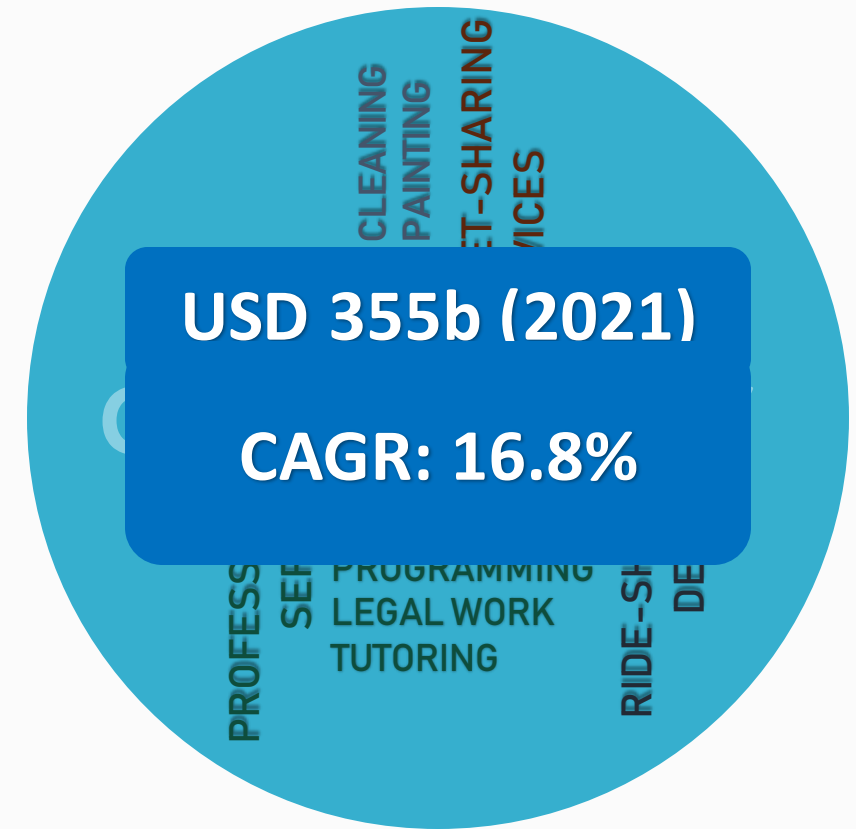


# The Gig Economy

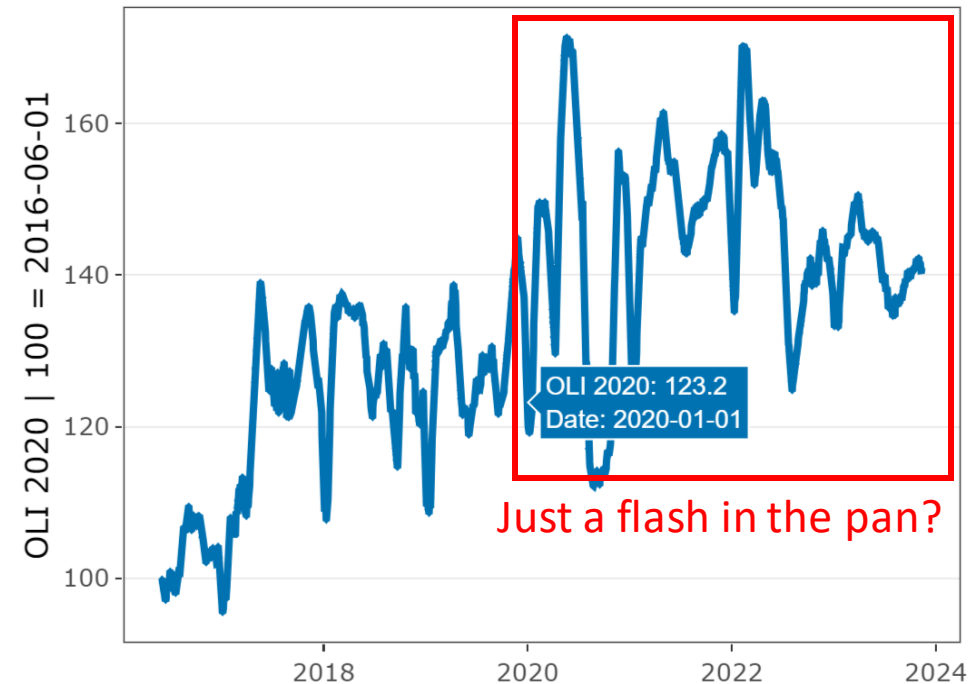
- “people take up several part-time jobs to make a living or engage in project-based or consulting work” (Brown, 2009)
- On-demand, freelancing activities, temporary employment or even the less formal segments of the labour market
- Includes on-site and off-site jobs



# Trends of the Gig Economy - Increased Employment

- Online Labour Index (OLI): 40% of total jobs, 15m skilled professionals
- Attractive to unemployed (youth) / underemployed population
  - ✓ Young people: opportunity/incentive
  - ✓ Increased participation for women
  - ✓ Work opportunities in smaller towns and villages (6/10)
- A form of unemployment insurance
- Companies: Lower costs of accessing specialized workers
  - Benefits businesses — remain competitive
- An over-glorified gig economy?

## The race to the bottom phenomenon



Source: OLI 2020 | onlinelabourobservatory.org

# However, Gig work is not very generous to its workers.

- Lower wage among gig workers
  - In the US: 29% gig workers earn less than minimum wage
  - 20% have difficulty to fulfil daily needs
- Gig work is much more vulnerable to exogenous shocks
  - During COVID, gig economy drivers reported a 20% larger decrease in income than traditional workers (Apouey et al., 2020)
- Gig workers receive virtually zero social benefits
  - Movements and lawsuits: Uber, Lyft
- So why are gig workers so vulnerable?
  - Inelastic supply due to low wages, low investments in human capital, and a lack of worker protection

# Theory of gig platforms: A two-sided market

- Gig platforms: intermediary for both the worker and the employer.
- Considerations for platforms: network effect
  - Cross-side network effect is positive while same-side network effect is negative.
- Platform commission depends on...
  - demand for tasks
  - costs of those workers
  - **elasticity of response on the other side** (Rysman, 2009)
- As mentioned: inelastic labor supply, low bargaining power for workers in platforms lead to the race to the bottom phenomenon
  - The number of gig workers has seen a record increase during Covid, possibly due to unemployment insurance effect (OLI, 2020)
- Implications: lower wage, less incentive, slower growth / increase exit



# Case Study: the transportation industry

- JP Morgan: between 2013-2017, earnings from rideshare fell by 53%
  - Attributed to rapid growth in the number of drivers on the roads
- During COVID, in the US:
  - All types of workers in the transportation industry work less
  - More people joined the ride-hailing app category
  - The hourly wage of gigs were more affected compared to conventional workers

Variable	Category	Public transport	Ride-hailing apps	Delivery apps
Weekly hours worked	Pre-pandemic: mean	48.5	44.5	37.8
	Pre-pandemic: variation coefficient	26%	45%	53%
	During pandemic: mean	33.3	25.4	34.9
	During pandemic: variation coefficient	36%	86%	59%
Hourly average wage in USD	Pre-pandemic	4.9	5.1	3.6
	During pandemic	5.3	4.9	3.4
	Change	0.4	-0.2	-0.2

Source: Abraham et al., 2021

# Case Study: Amazon's Mechanical Turk Platform



*A micro-crowdsourcing platform where both industry and individual users can submit tasks (such as content creation, reviews and evaluations, information finding etc.) for "Turkers" to take up.*

- Mean wage: **\$3.13/h**
- **~4%** of workers earn more than \$7.25/h (US minimum wage)
- Requesters pay **\$11.58/h** on average
  - Those who pay lower tend to post more
- Factors inducing low hourly wages
  - Searching for tasks
  - Working on rejected tasks (task spamming)
  - Working on tasks that aren't submitted

*Source: Hara et al., 2018*

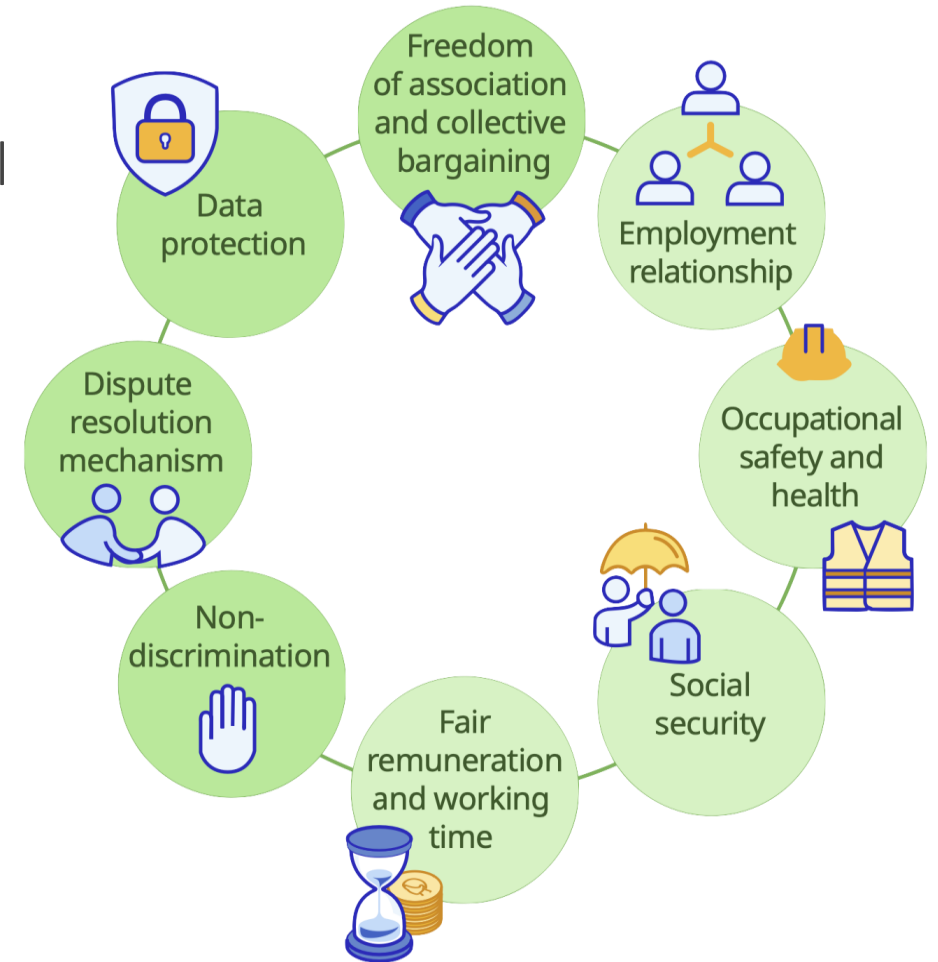
# Implications and the future of the Gig Economy

## 1. Government regulation

- Growing trend of jurisdictions addressing digital labour working conditions
- Classification of gig workers as employees
- Updated labour policy and inspections

## 2. Platform design and regulation

- Several prominent platforms signed the World Economic Forum Charter of Principles (2020)
- The Fairwork Foundation translated a code of principles into measurable thresholds
- Transparency of platform algorithms



Source: International Labour Organization, *World Employment and Social Outlook 2021*

# Key Takeaways

- Gig economy: on-demand, project-based work
- Growth of the gig economy is over-estimated
- Wages of gig workers are falling/plateauing
- Sufficient regulation is crucial for the gig economy to be fully utilised

# Evolution of Routine Jobs vs. Non-Routine Jobs —The Impact of Technology Advancement

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Technological  
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Example**

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**Typical Non-routine  
Job as Example**

**05**

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**Suggestions and  
Conclusion**

# Definition & Technological Background

## Routine Jobs

- **Definition:**
  - Jobs with repetitive tasks that can be easily automated or outsourced.
  - Common in manufacturing, clerical sectors.
- **Technological Background**
  - Job displacement caused by technology
  - Efficiency and productivity improvement
  - Standardization of product quality

## Non-Routine Jobs

- **Definition:**
  - Irregular jobs usually requiring problem-solving, creativity, and social interaction, making them less susceptible to automation and outsourcing.
  - Common in technology and innovation related sectors.
- **Technological Background**
  - Enhancement of cognitive and analytical capabilities
  - Efficiency and productivity improvement
  - Changing skill requirements

# General Trend in Labor Market

- Occupations that are most susceptible to automation:
  - office support occupations
  - some customer interaction jobs
  - jobs carried out in predictable settings
- High-skilled, non-routine jobs represent almost exclusively- the increase in share of total UK employment since 2001, while exclusively low-skilled, routine jobs have diminished in share.

➔ **Technology-driven shift from low skill, routine jobs to higher-skill, non-routine jobs.**

**ROUTINE JOB WORKERS**

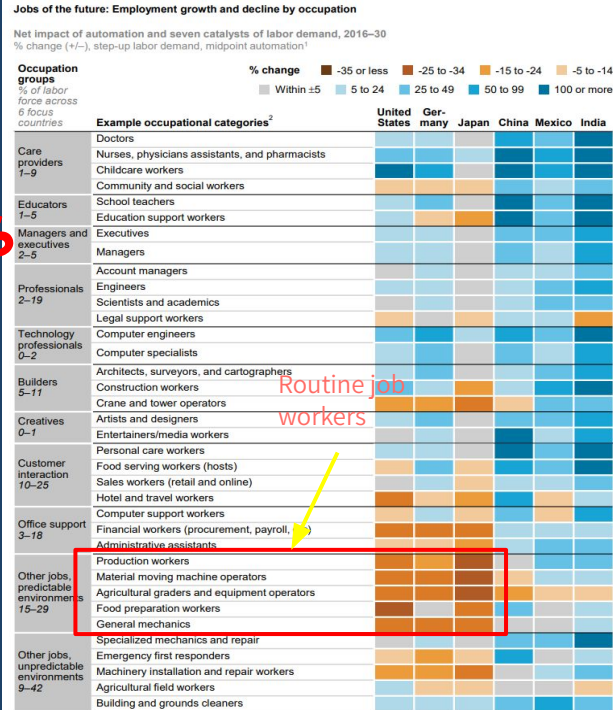
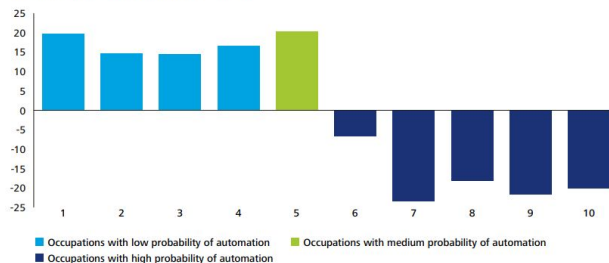


Figure 4. Change in employment by probability of computerisation

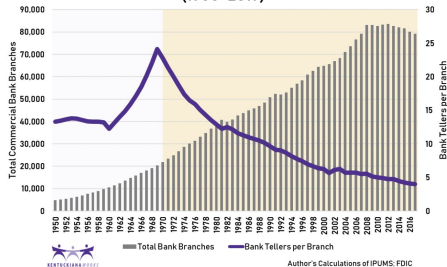
Percentage change in employment share (2001–2015)



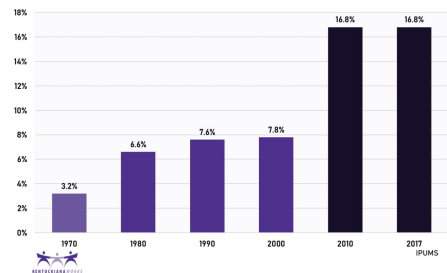
Note: Each decile represents approximately ten per cent of the UK's workforce, with jobs ordered by probability of computerisation.  
Source: Frey and Osborne, Deloitte analysis 2015



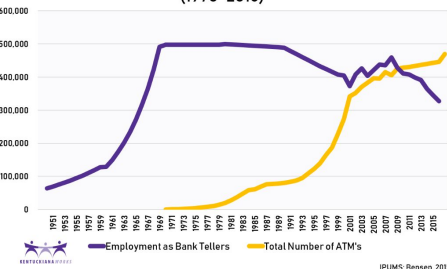
Bank Teller Employment per  
Commercial Bank Branch  
(1950-2017)



Percentage of Bank Tellers  
with 4 years of College or more



Total Employment of Bank Tellers and  
Total ATM's  
(1970-2016)

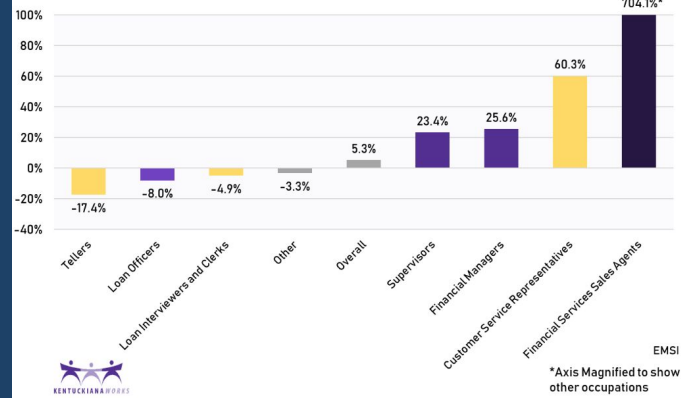


# Typical Routine Job —Bank Tellers

Why the introduction of ATMs didn't immediately reduce the need to hire bank tellers?  
—ATM allowed banks to operate more efficiently→ more branches were opened → Hired more tellers although fewer tellers per branch.

- As ATM technology becomes more sophisticated, in the next 20 years up to **90%** of bank teller's original duties are expected to be automated by ATMs.
- Two results:
  - 1) "skill-biased technological change"- as a result of automation, **more education is needed** for bank teller jobs.
  - 2) **Job duties** of bank tellers have **changed**—more customer service and sales training.

Occupational Change at Commercial Banks  
(2001-2018)



- Number of bank tellers in Commercial Banks decreased during 2001-2018 (**-17.4%**)

# Typical Non-Routine Job —Tech Workforce

- Six occupations comprise the tech workforce. Only “Computer and Mathematical Science Occupations”, is a broad occupation category (part of the 22 broad ones)
- The tech workforce has been **growing steadily between 2001-2020**, with an average annual rate of 2.2%.
- The pace of expansion in the tech workforce increases in recent years.
- Even during the pandemic, the tech workforce continued to expand, rising by 0.7% in 2020.
  - Remote-working mode needs tech talent to support
  - The digitization of businesses and economy-wide adoption of AI and the IoT likely intensified → rising demand for talent with advanced analytical skills
- Why demand for core engineering-related talent decreases???
  - Increased digitization and more sophisticated systems may not require a proportionate increase in the engineering and technician talent pool.



FIGURE 1

The tech workforce comprises six key occupations

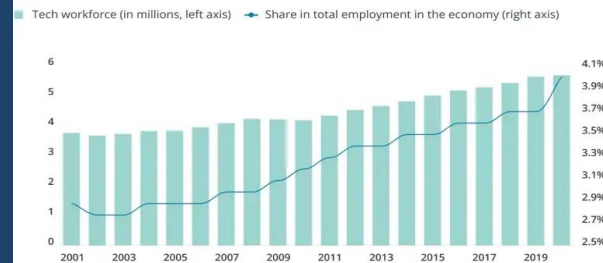
Occupation name	Part of broad occupation (yes/no)?	If yes, name of broad occupation category under which the sub-occupation falls
Computer and information systems managers	Yes	Management occupations
Computer and mathematical science	No	
Computer hardware engineers	Yes	Architecture and engineering occupations
Electrical engineers	Yes	Architecture and engineering occupations
Electronics engineers, except computer	Yes	Architecture and engineering occupations
Electronic and electronic engineering technicians	Yes	Architecture and engineering occupations

Source: United States Bureau of Labor Statistics; Deloitte analysis.

Deloitte Insights | deloitte.com/insights

FIGURE 3

The tech workforce has expanded steadily over the years, with the pace increasing in recent years

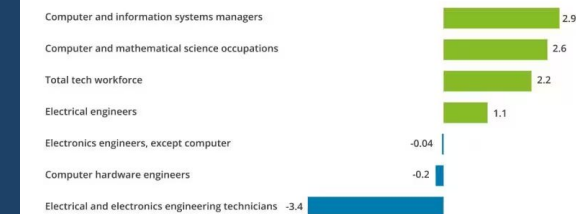


Source: United States Bureau of Labor Statistics (sourced through Haver Analytics); Deloitte analysis.

Deloitte Insights FIGURE 4

Employment of computer and information systems managers grew the fastest during 2001–2020

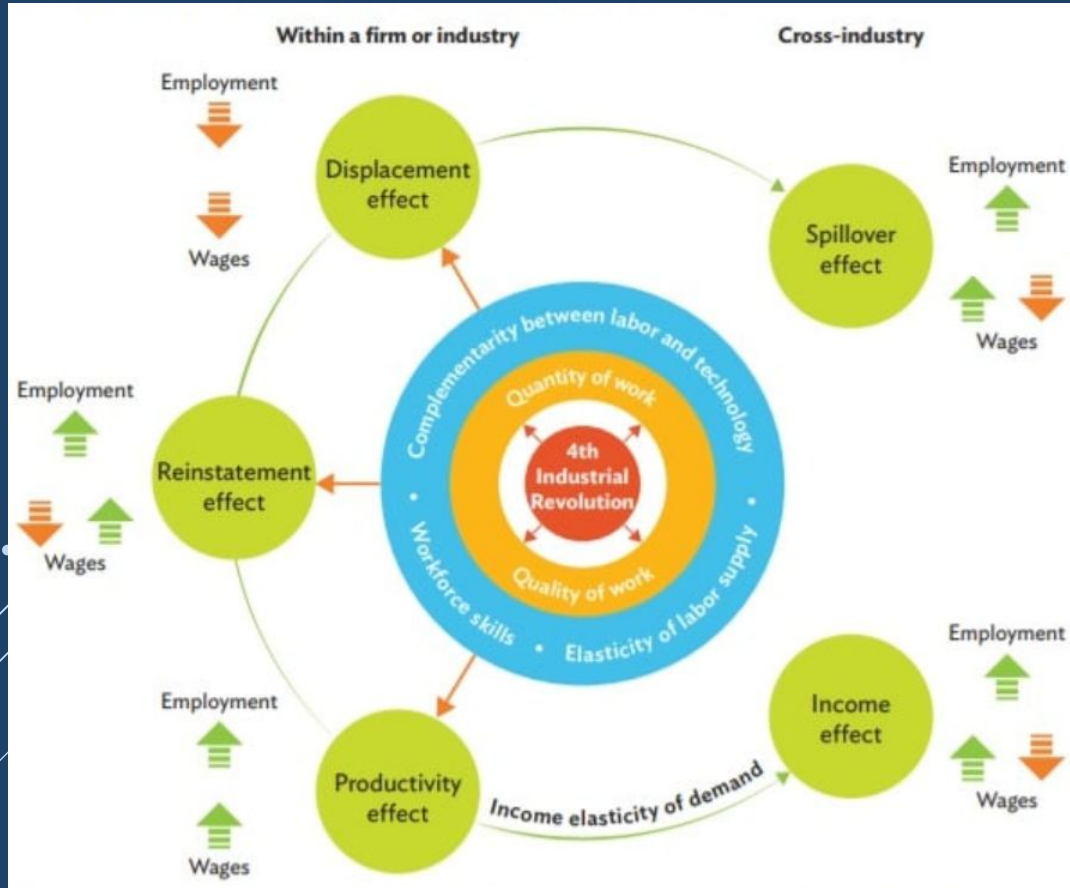
Average annual growth in employment during 2001–2020 (%)



Source: United States Bureau of Labor Statistics (sourced through Haver Analytics); Deloitte analysis.

Deloitte Insights | deloitte.com/insights

# Analytic Framework



- **Displacement Effect:** Robots and computers are good at routine tasks, demand and wages will fall for labors doing routine tasks.
- **Reinstatement Effect:** Automation can spawn new labor-intensive tasks and jobs, raising demand for labor. The effect on wages is depends.
- **Productivity Effect:** Automation improves productivity and lowers production costs. Larger product demand increases labor demand.
- **Income Effect:** When technology complements labor, workers' higher incomes will increase their demand for goods and services.
- **Spillover Effect:**
  - Firms in downstream industries benefit from cheaper and/or better-quality inputs, while firms in upstream industries benefit if the output of the automating industry expands.
  - Other industries learn the benefits of adopting the new technology.
  - Workers with new skills move between industries, spreading technological knowledge.

Note: Arrows indicating a rise or fall in employment or wages reflect empirical findings from existing studies, but they do not necessarily mean the result is obtained each time the effects are studies

Source: ADB based on Autor (2015) and Acemoglu and Restrepo (2018)



# Suggestions

## Routine Job Workers

- Upskilling programs
- Soft skills development

## Non-Routine Job Workers

- Enhancing creative and analytical abilities
- Maintaining compatibility between skills and technology

## Companies

- Training programs like mentorship and leadership programs
- Career pathing support

## Government

- Increasing subsidies for training programs
- Educational reform

# Conclusion



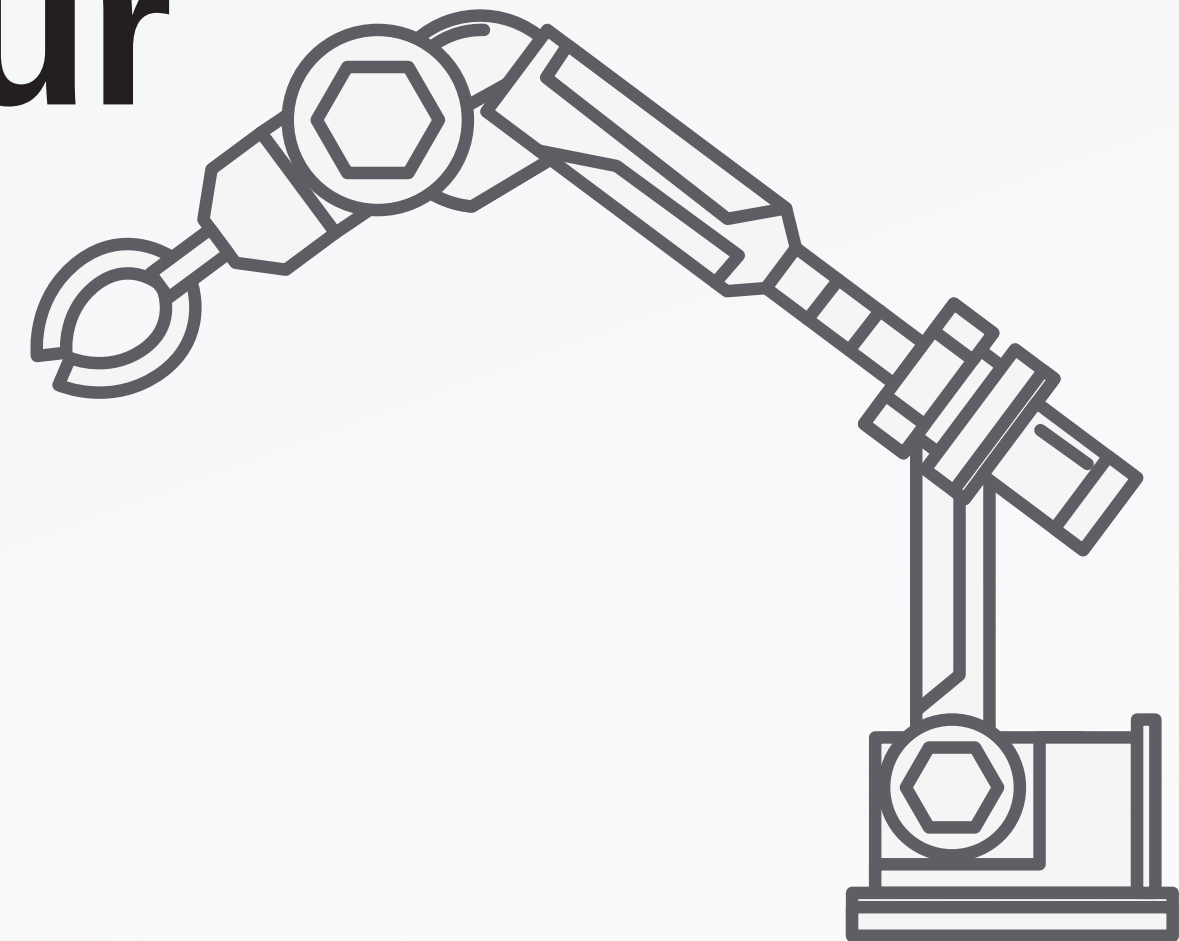
- Embrace technology
- Balance technological advancement with the preservation of meaningful employment opportunities through joint efforts to boost economic development

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# The Effects of AI & Robotic Automation on Human Labour



## **Group4:**

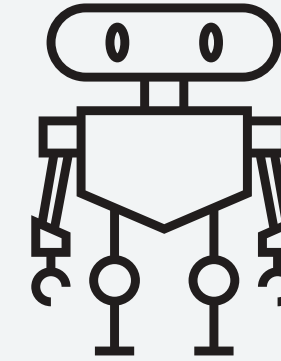
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Wang Tianyang 3035945857

# CONTENT



**01**

**Background**——The application of AI & robotic automation

**02**

The **effect** of AI & robotic automation on workers:  
Employment/ Wages/ Satisfaction/ Productivity

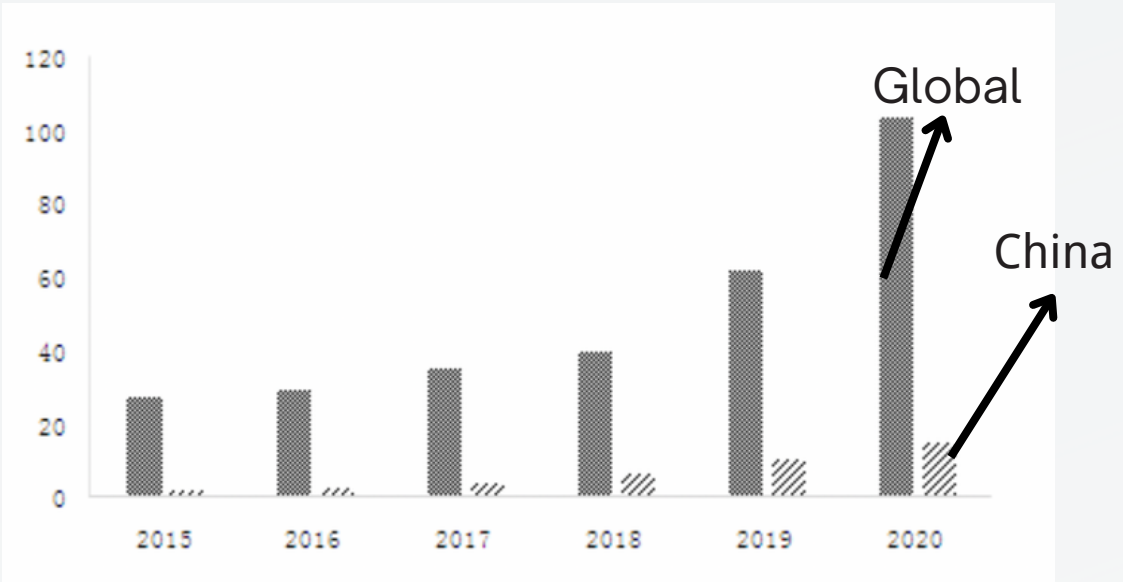
**03**

**Conclusion**——How to deal with the effect on human labour

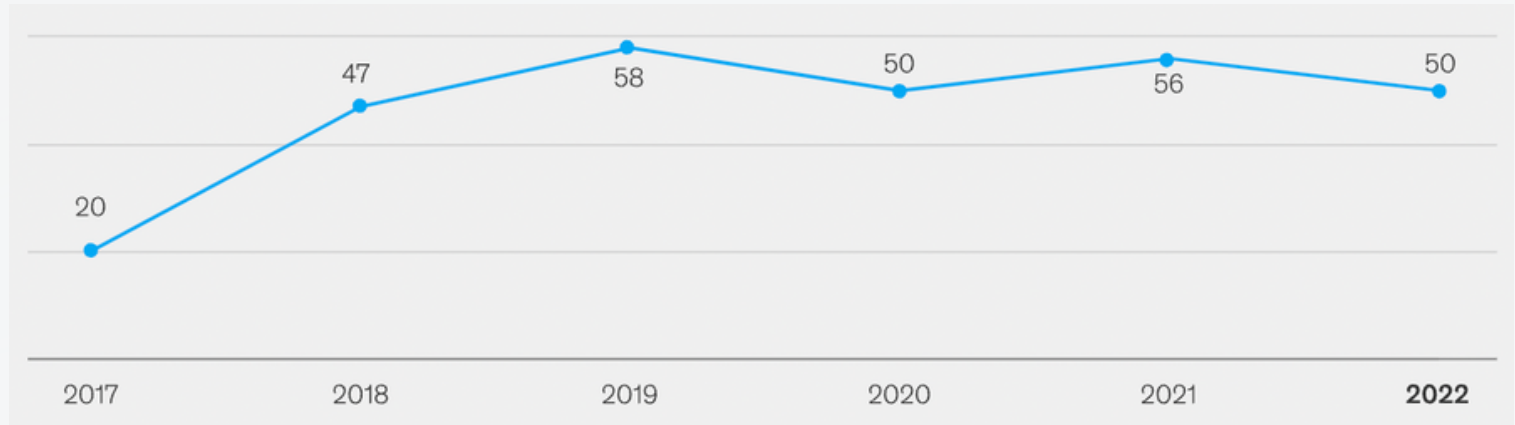


# BACKGROUND——THE APPLICATION OF AI& ROBOTIC AUTOMATION

AI Market Size (in billions of dollars)



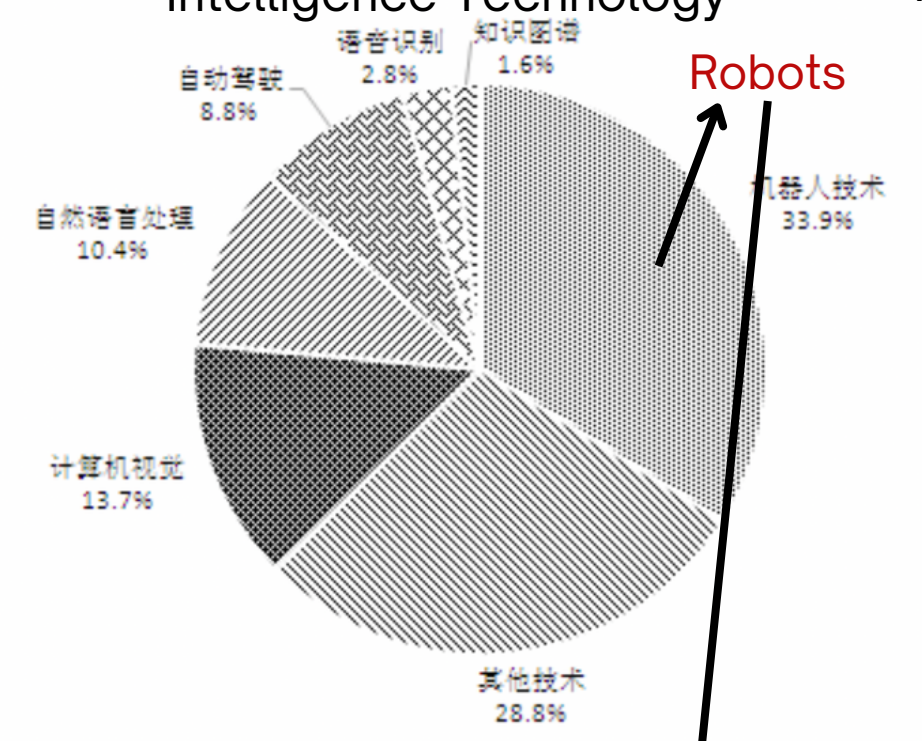
Share of respondents who say their organizations have adopted AI in at least one business unit or function, %



“The state of AI in 2022”——Michael Chui

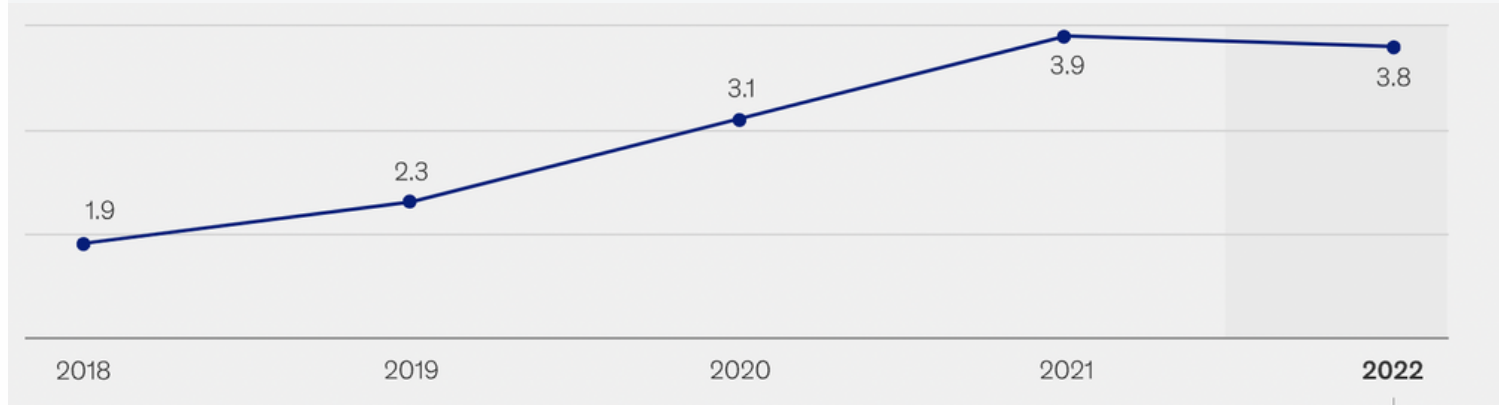
- **AI adoption** has grown rapidly and more than **doubled** over the past 5 years

Global Funding Volume in Artificial Intelligence Technology

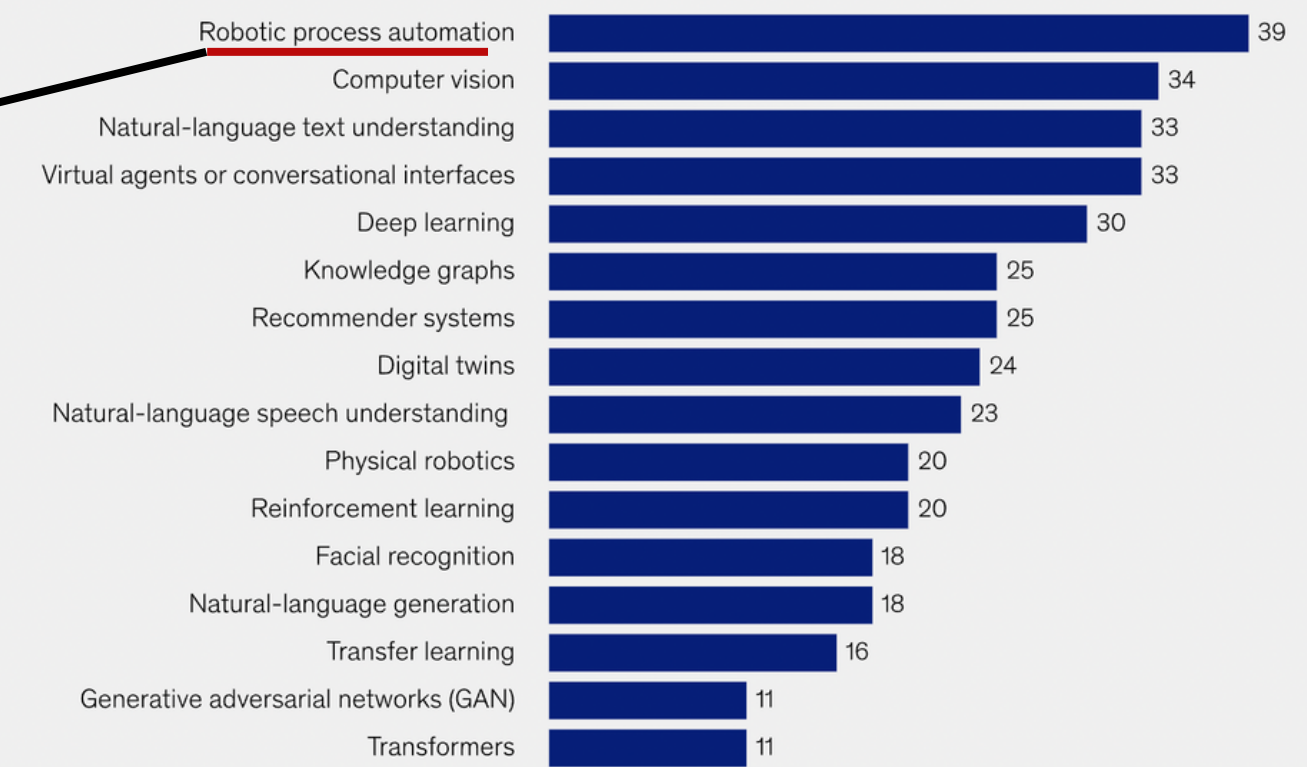


**Robotic process automation** remained the most commonly deployed

Average number of AI capabilities that respondents' organizations have embedded within at least one function or business unit



Percentage of respondents who say given AI capability is embedded in products or business processes in at least one function or business unit<sup>2</sup>



- The average number of **AI capabilities that organizations use** has also doubled—from 1.9 in 2018 to 3.8 in 2022

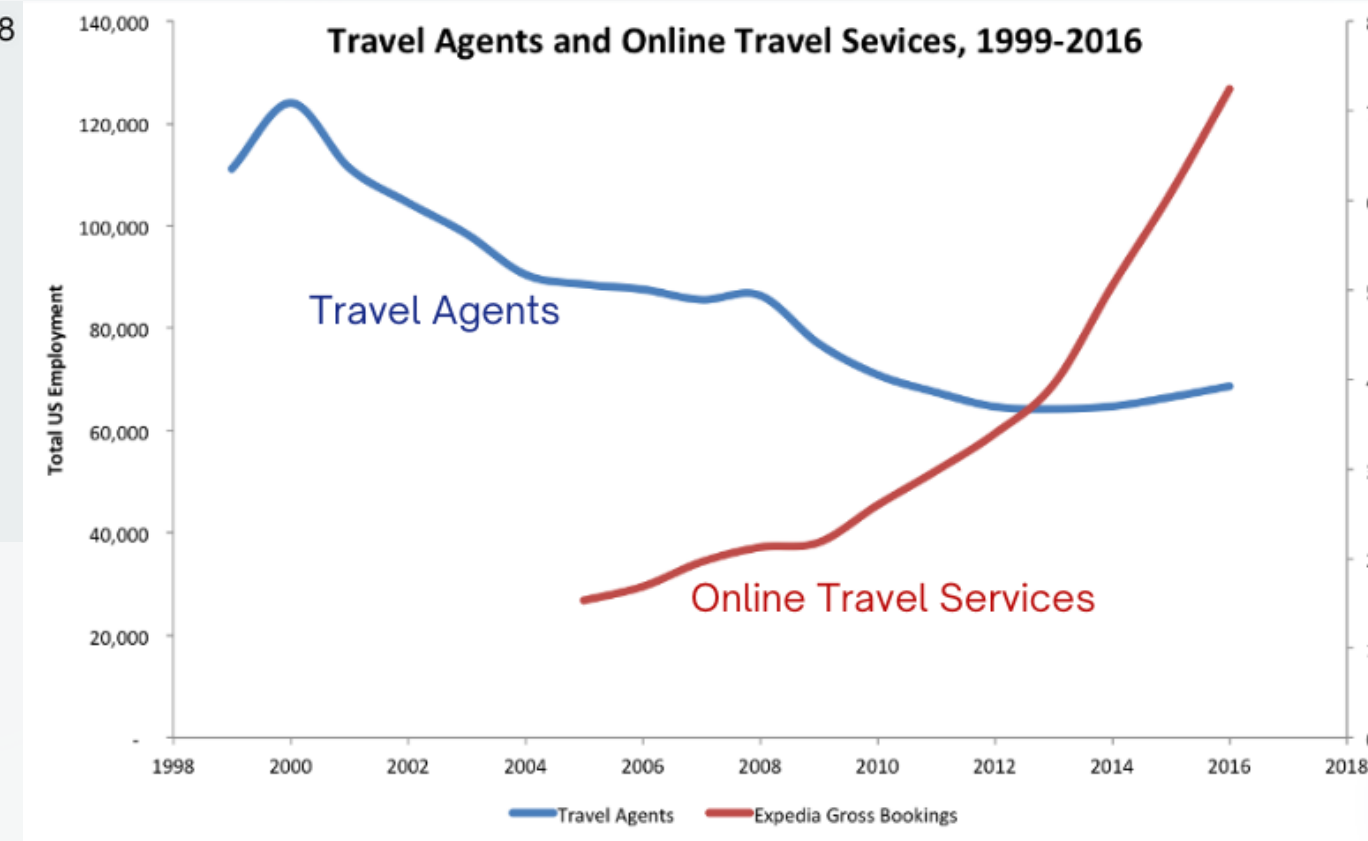
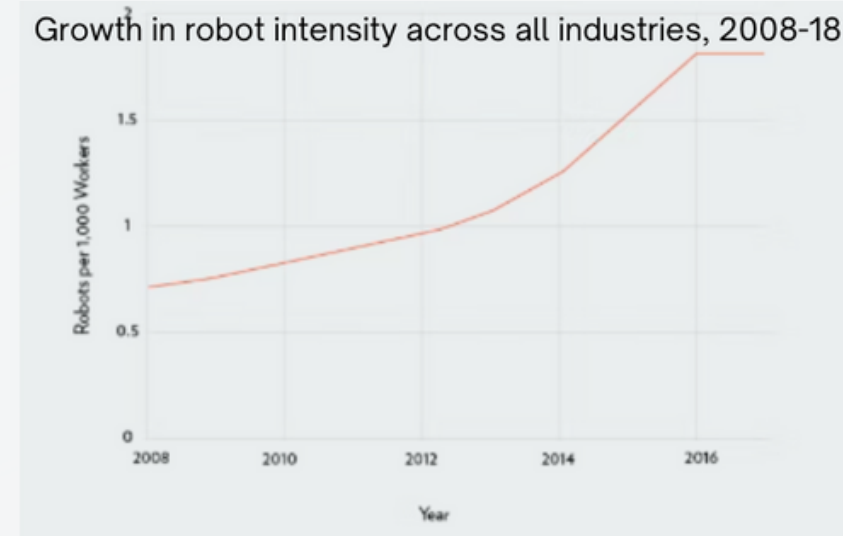
<https://www.yicai.com/news/101088220.html>  
<https://www.mckinsey.com/capabilities/quantumblack/our-insights/the-state-of-ai-in-2022-and-a-half-decade-in-review#/>



# THE EFFECT OF AI & ROBOTIC AUTOMATION ON WORKERS

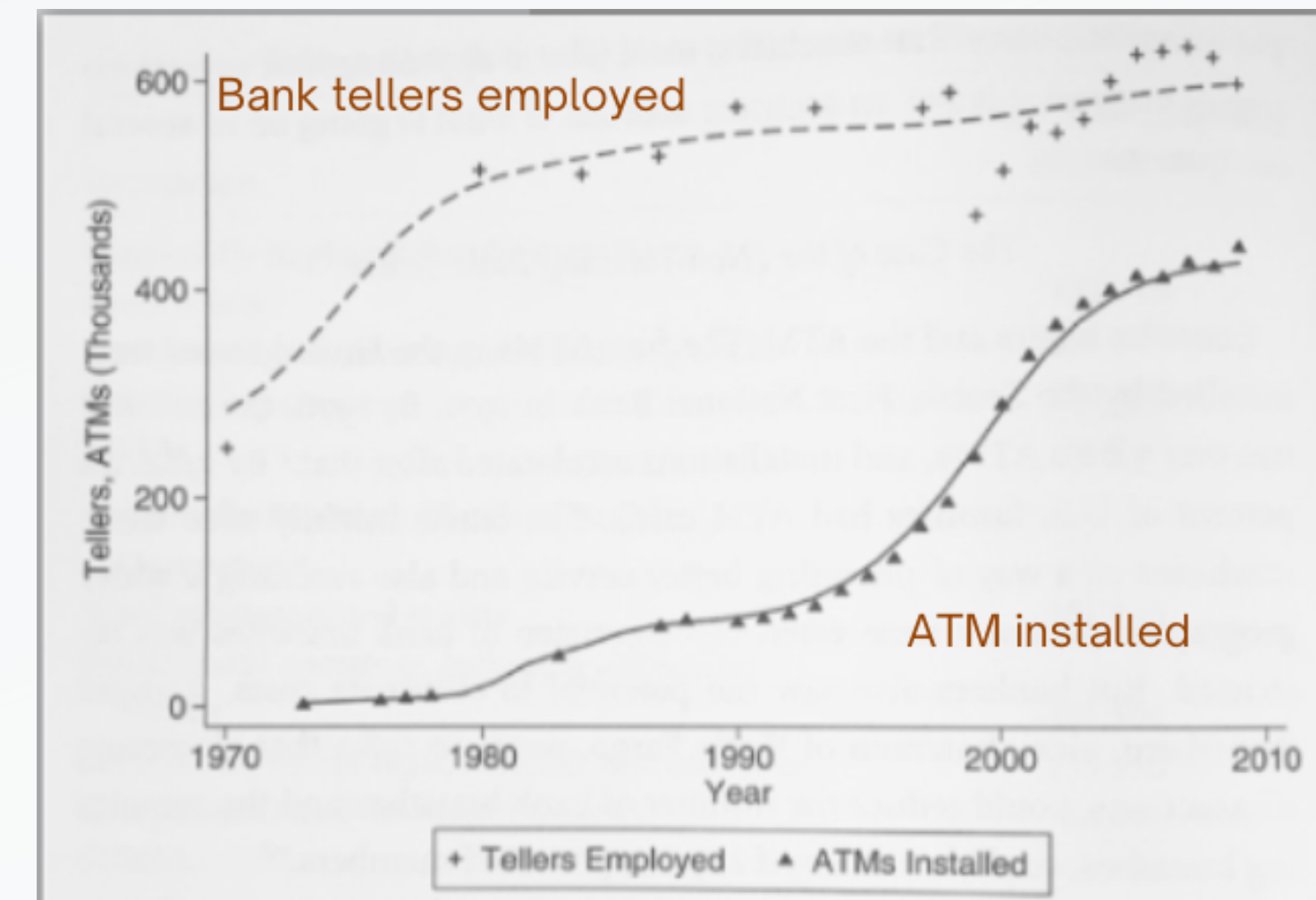
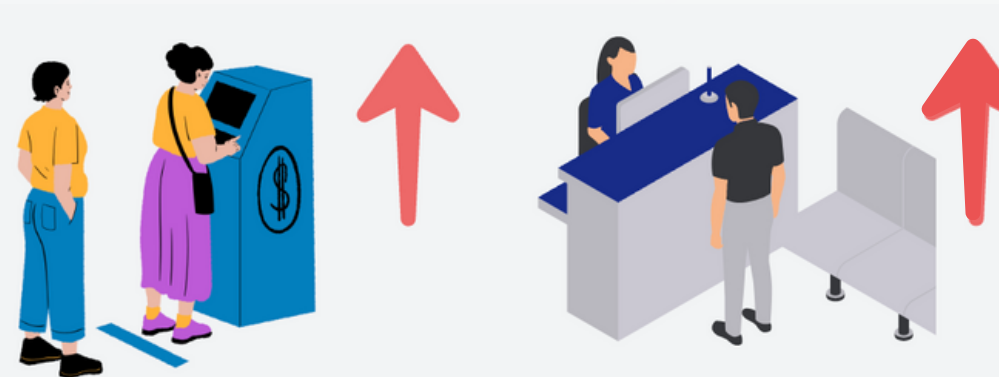
## DISAPPEARING INDUSTRIES:

- **Manufacturing industries**
  - e.g 'in the U.S., where researchers found a **reduction** in employment by **0.2 percentage points** for every robot per 1,000 workers'
- **Travel agencies**
  - With the rise of online booking platforms, the need for traditional travel agents has diminished.
  - 'At the beginning of the dot-com bubble, there were **111,130** travel agents in the U.S. In 2016, there were only **68,680**, a **38 percent decrease**.'



## INCREASED POSITION:

- **Services industries**
  - e.g. Bank teller
    - The number of bank teller jobs actually **increased** after the introduction of ATMs, as the **reduction in operating costs** led to an **increase in the number of bank branches**
- Similarly, scanning technology and e-discovery software have led to an increase in related positions such as **tellers** and **paralegals**.



# DIRECT IMPACT OF TECHNOLOGICAL PROGRESS ON WAGES

Automation replaces  
low educated workers



income gap between  
workers increases

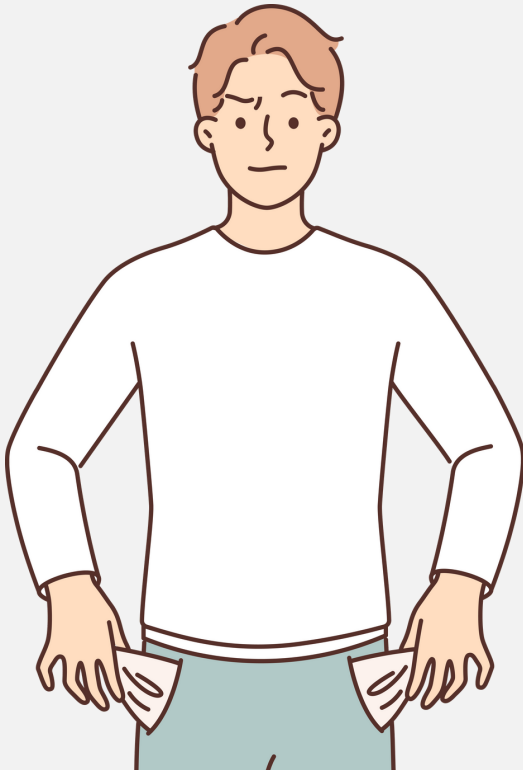


- ‘Researchers estimate that since 1980, **automation** has led to an **8.8 percent decline** in wages for men without a high school degree and a **2.3 percent decline** for women, adjusted for inflation’
- ‘**The BLS** reported that wages in tech industries have **risen substantially over the past decade**’



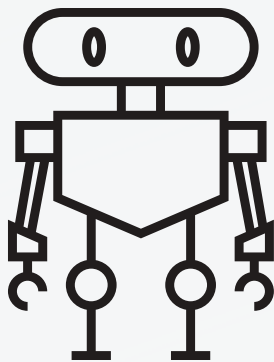
Industries	computer and mathematical occupation	production occupations
Median Anual Wages	\$88,340	\$36,000

<https://www.celdred.com/blog/2017/7/2/what-travel-agents-could-mean-for-the-rest-of-us>  
<https://news.mit.edu/2022/automation-drives-income-inequality-1121>



# THE EFFECT OF AI & ROBOTIC AUTOMATION ON WORKING SATISFACTION

## EXPERIMENTS ON SATISFACTION



### Manual control

Only human workers during the task

### Semi- Autonomous control

- Human workers
- Robots co-leader/ Human co-leader
- Human assistants

### Autonomous control

Only robots workers during the task

### Results:

- **Question 5 6 8:** Human co-leader More liked, appreciated and better-understood workers.
- **Question 9 7:** Human co-leader understood, trusted, and respected each other.
- **Question 19 21:** Both workers and human co-worker were necessary

### Finding:

**More satisfied** and **perceive** the team as more fluent when working with a **HUMAN-ONLY team** rather than a **HUMAN-ROBOTS team**.

(Gombolay et al., 2015)

Table 1 Subjective measures—post-trial questionnaire

Robot teammate traits	
1. The robot was intelligent	
2. The robot was trustworthy	
3. The robot was committed to the task	
Working alliance for human–robot teams	
4. I feel uncomfortable with the robot (reverse scale)	
5. The robot and I understand each other	
6. I believe the robot likes me	
7. The robot and I respect each other	
8. I feel that the robot worker appreciates me	
9. The robot worker and I trust each other	
10. The robot worker perceives accurately what my goals are	
11. The robot worker does not understand what I am trying to accomplish (reverse scale)	
12. The robot worker and I are working towards mutually agreed upon goals	
13. I find what I am doing with the robot worker confusing (reverse scale)	
Additional measures of team fluency	
14. I was satisfied by the team’s performance	
15. I would work with the robot the next time the tasks were to be completed	
16. The robot increased the productivity of the team	
17. The team collaborated well together	
18. The team performed the tasks in the least time possible	
19. The robot worker was necessary to the successful completion of the tasks	
20. The human worker was necessary to the successful completion of the tasks	
21. I was necessary to the successful completion of the tasks	

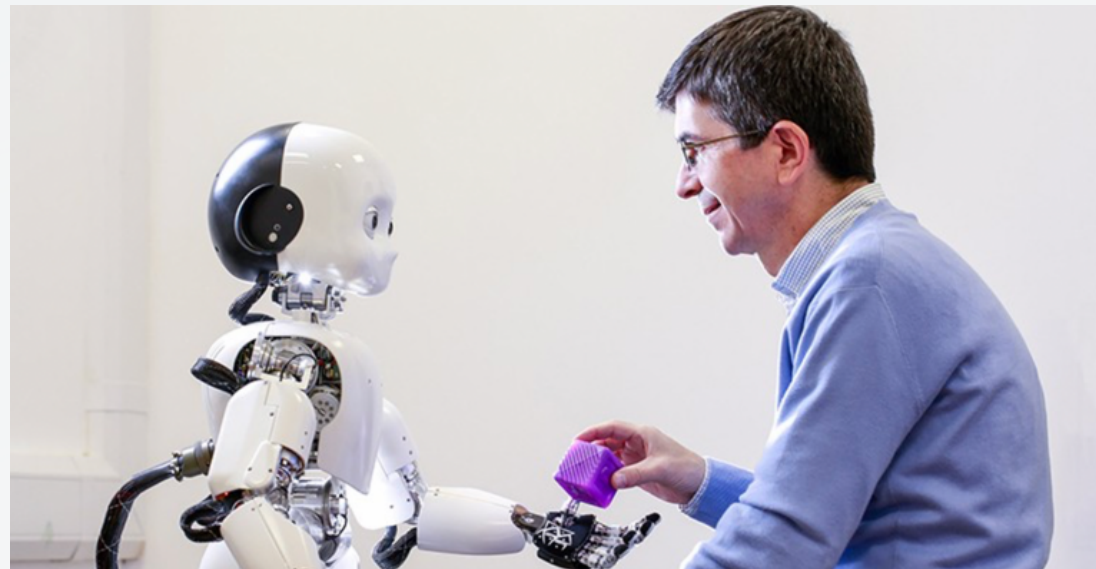
Table 6 Differences between Likert-scale responses for the autonomous and manual conditions when working with a robot versus a human co-leader

Question	Difference	Robot co-leader versus human co-leader
1	$\Delta_R > \Delta_H$	$\chi^2 = 112.232, p < \mathbf{0.001}$
2	$\Delta_R > \Delta_H$	$\chi^2 = 4.672, p < \mathbf{0.031}$
5	$\Delta_R > \Delta_H$	$\chi^2 = 7.291, p = \mathbf{0.007}$
6	$\Delta_R > \Delta_H$	$\chi^2 = 14.070, p < \mathbf{0.001}$
8	$\Delta_R > \Delta_H$	$\chi^2 = 5.036, p = \mathbf{0.025}$
9	$\Delta_R > \Delta_H$	$\chi^2 = 17.831, p < \mathbf{0.001}$
10	$\Delta_R > \Delta_H$	$\chi^2 = 39.287, p < \mathbf{0.001}$
11	$\Delta_R > \Delta_H$	$\chi^2 = 5.000, p = \mathbf{0.025}$
12	$\Delta_R > \Delta_H$	$\chi^2 = 7.170, p = \mathbf{0.007}$
13	$\Delta_R > \Delta_H$	$\chi^2 = 15.515, p < \mathbf{0.001}$
14	$\Delta_R > \Delta_H$	$\chi^2 = 51.564, p < \mathbf{0.001}$
15	$\Delta_R > \Delta_H$	$\chi^2 = 104.836, p < \mathbf{0.001}$
16	$\Delta_R > \Delta_H$	$\chi^2 = 100.000, p < \mathbf{0.001}$
17	$\Delta_R > \Delta_H$	$\chi^2 = 83.571, p < \mathbf{0.001}$
18	$\Delta_R > \Delta_H$	$\chi^2 = 84.366, p < \mathbf{0.001}$
19	$\Delta_R > \Delta_H$	$\chi^2 = 105.780, p < \mathbf{0.001}$
22–24	$\Delta_R > \Delta_H$	$\chi^2 = 68.702, p < \mathbf{0.001}$
4	$\Delta_H > \Delta_R$	$\chi^2 = 24.923, p < \mathbf{0.001}$
20	$\Delta_H > \Delta_R$	$\chi^2 = 68.702, p < \mathbf{0.001}$
21	$\Delta_H > \Delta_R$	$\chi^2 = 5.838, p = \mathbf{0.016}$

Bold values are statistically significant ( $p < 0.05$ )



# THE EFFECT OF AI & ROBOTIC AUTOMATION ON WORKERS' PRODUCTIVITY



People working with a robot co-leader tend to assign a **disproportionate** amount of work to themselves

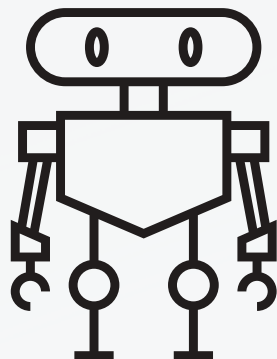


Planning fallacy: people **underestimate** the amount of time they need to complete a set of tasks while **overestimate** the amount of time that others need to complete the same sets of tasks

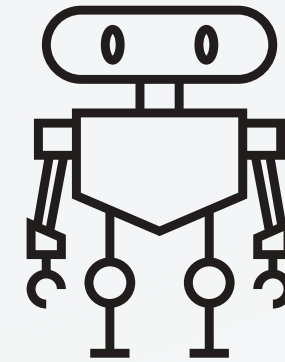
Workers perceived the human co-leader **more favorably** than the robotic co-leader



Allocate **less work** to themselves when working with the human in the semi-autonomous control



# CONCLUSION



The Effects of AI & Robotic Automation on Human Labour

Current situations  
of AI and Robotic  
Automation  
application

1

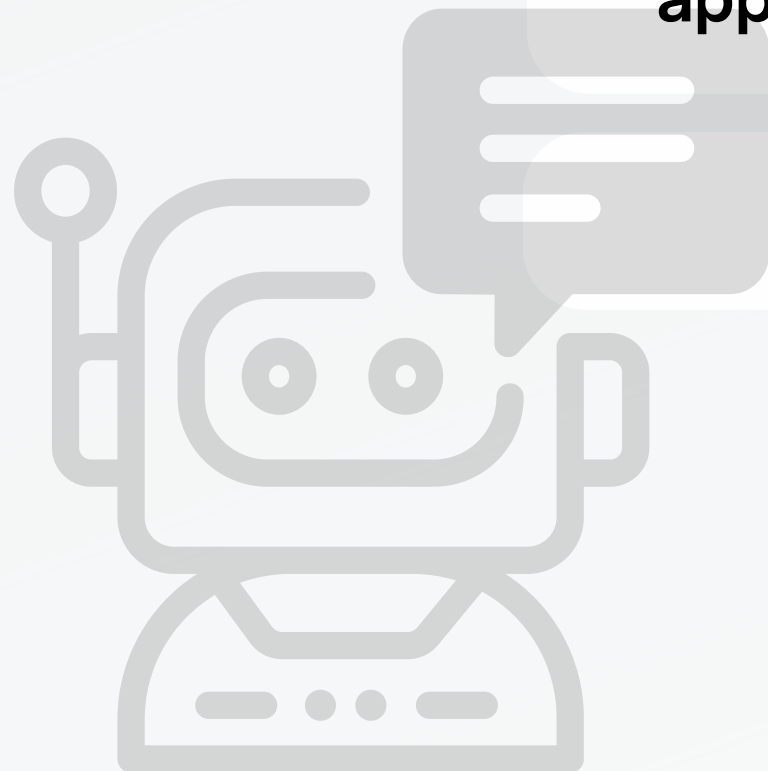
Effects on workers'  
employment, wage,  
satisfaction and  
productivity

2

## Promotions:

- Creative jobs
- Optimal ratio of AI and human workers

3





# How Does Technology Affect Employment and Wages: Evidence from the Emergence of Digital Nomad

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Tsui Fung Cheung	3035858860
Yan Yang	3035949358
Tian Tian Ye Zi	3035948342



## What is digital nomad?

Individuals that leverage technologies to work remotely

## Digital nomads' life style



- pursuit of a work-travel equilibrium
- prioritize autonomy and adaptability

## How do digital nomads earn?



- Engaging in diverse professions: freelance writing, software development, graphic design





## **ubiquity of high-speed internet**

providing digital nomads with constant connectivity essential for remote work



## **collaborative tools and communication platforms**

to interact with global teams, and ensuring the fluidity of project



## **Cloud-based storage solutions**

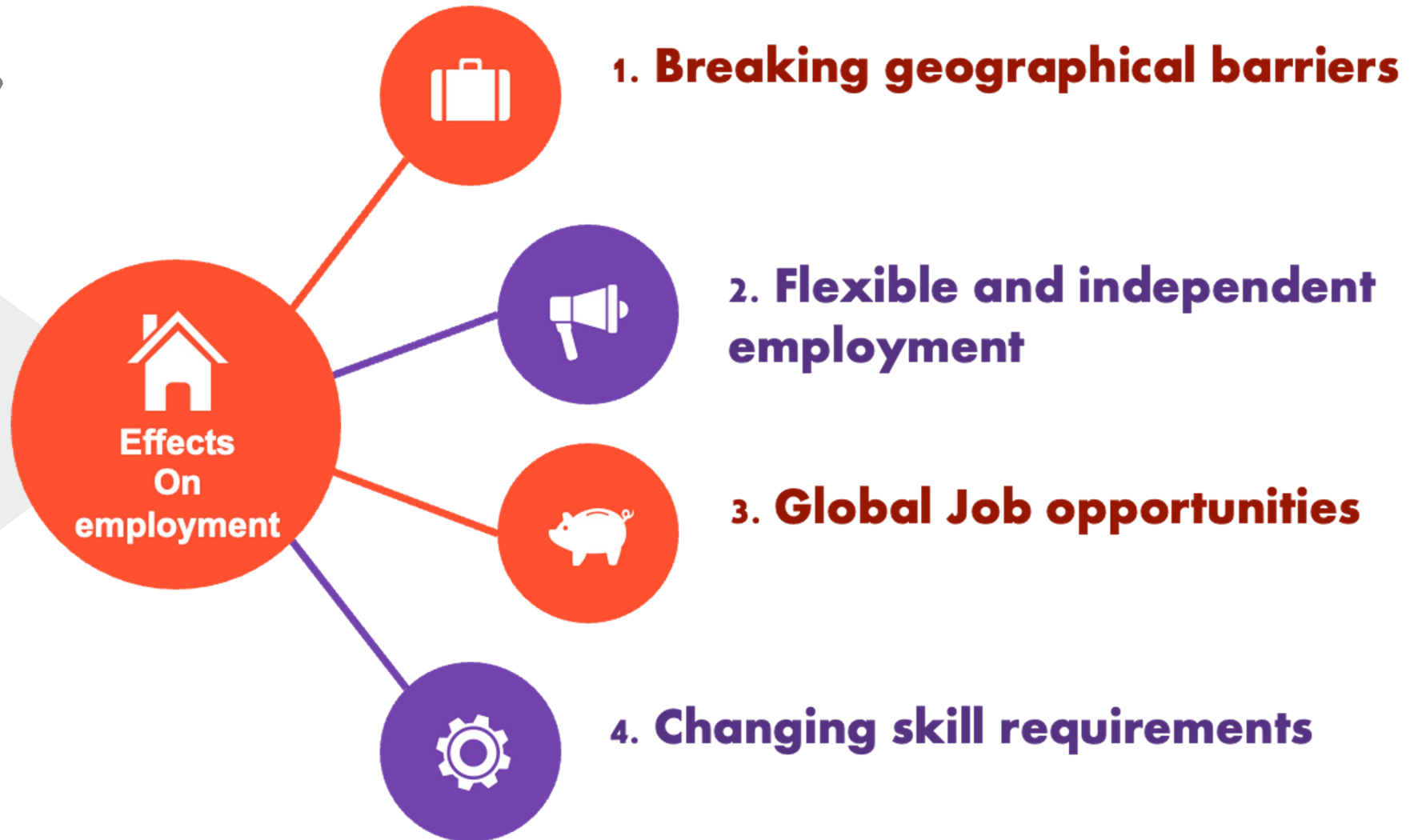
enabling the secure and accessible storage of files and resources



## **powerful and portable computing devices**

make sure that digital nomad can work anywhere (remotely)



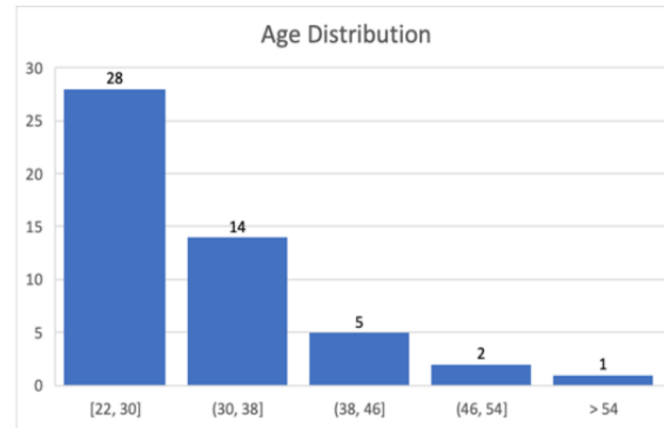


02

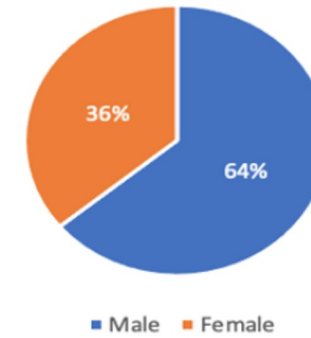
## Factors affect the employment of Digital Nomads



### 1 Age & gender

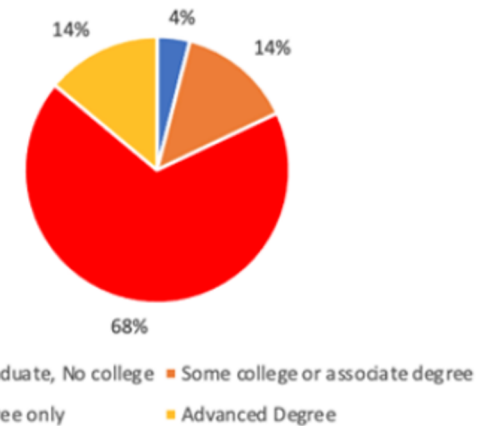


Gender Distribution



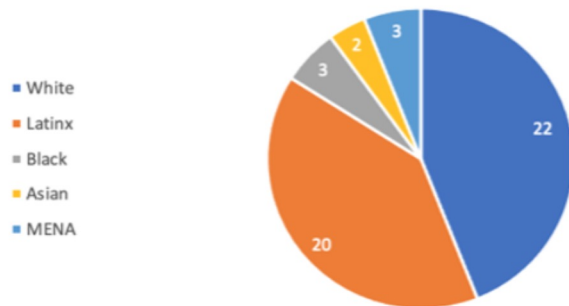
### 2 Education Level

Education Level Breakdown

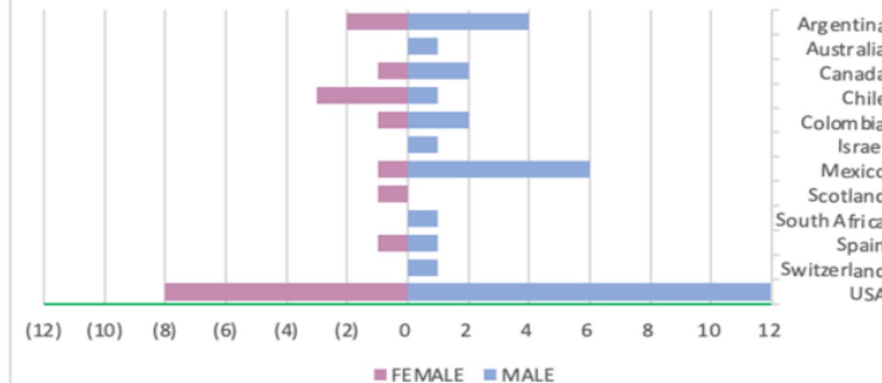


### 3 Nationality & Racial

Racial Distribution

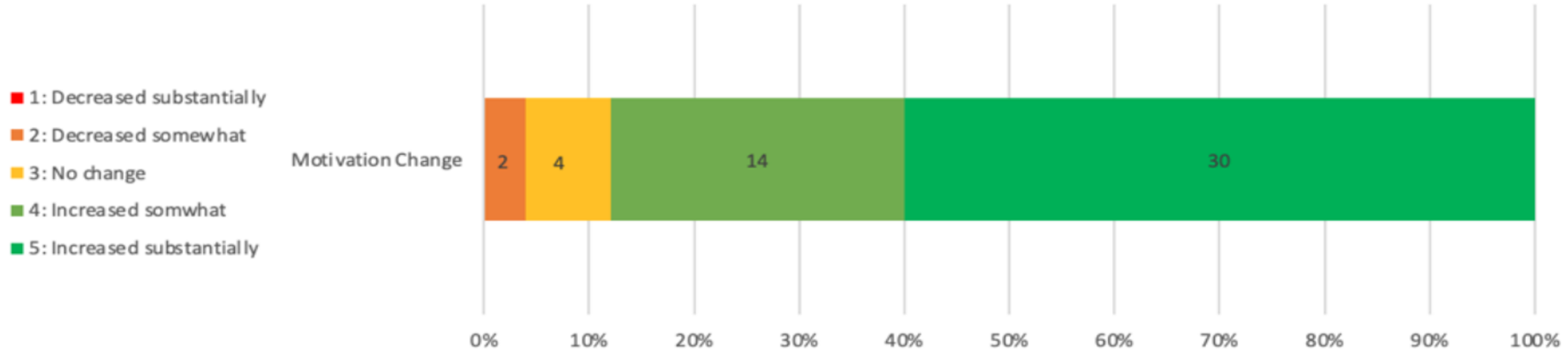


Nationality by Gender Breakdown

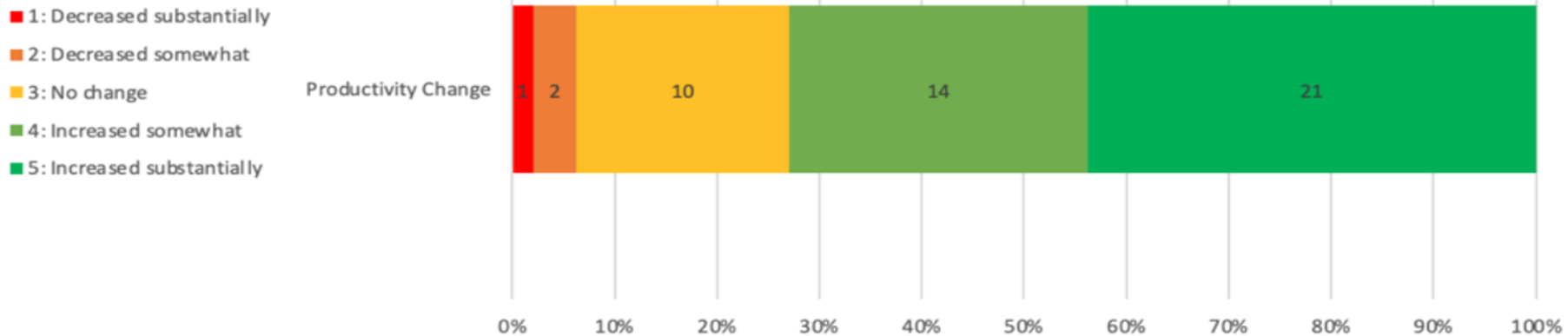


## Factors affect the employment of Digital Nomads

### Change to Motivation to Work



### Change to Productivity



# 03 Work Incentive as DN: Difference in Living Cost

## 1. Nationality

US (51%)  
UK  
Russia  
Canada  
Germany

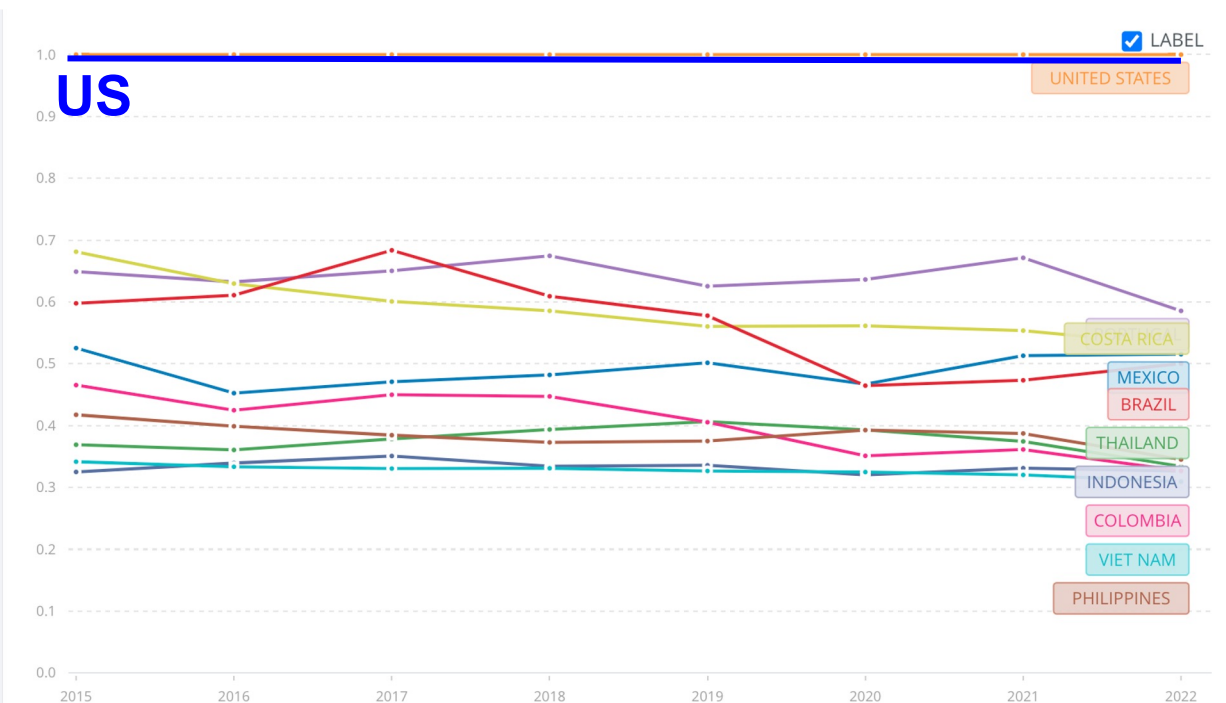
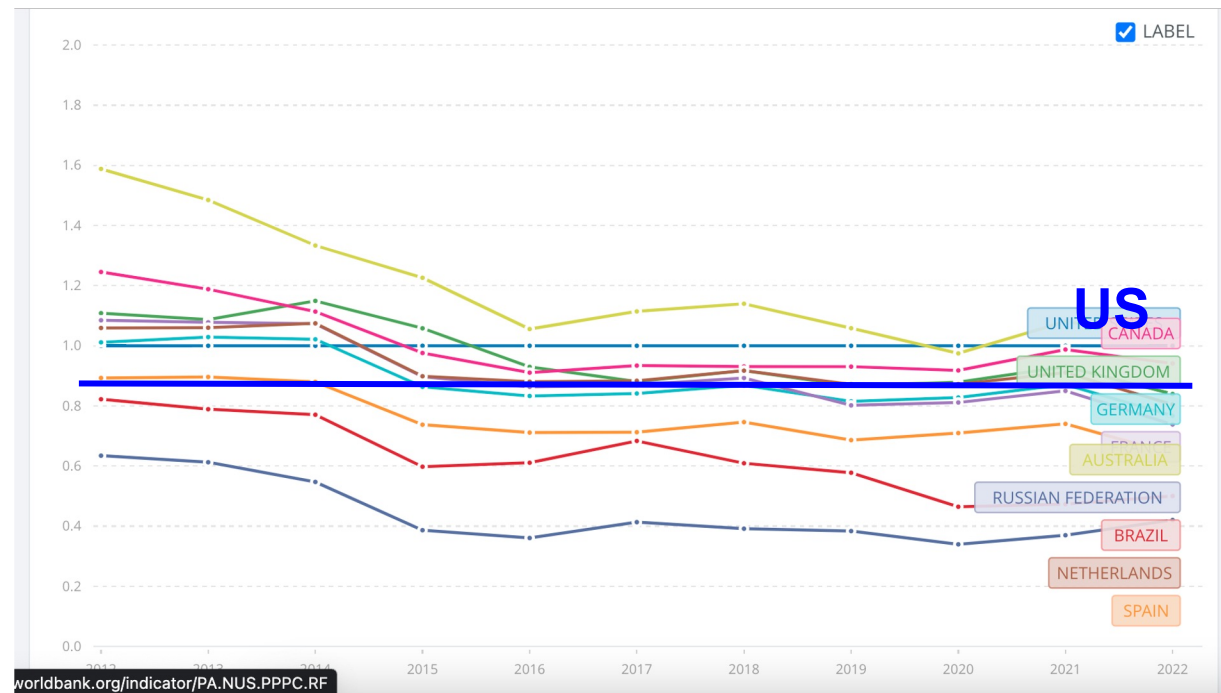
France  
Australia  
Brazil  
Netherlands  
Spain

## 2. Destination

Mexico  
Thailand  
Indonesia  
Colombia  
Vietnam



Portugal  
Turkiye  
Costa Rica  
Brazil  
Philippines

Price Level Ratio of PPP conversion factor (GDP) to market exchange rate



## Choice of DN: Labor-Leisure Choice Model

Consider a DN from US:

1. Baseline model of Labor-Leisure Choice
2. Model in the context of DN
  - a. Uneared income : flexibility, beautiful working environment, travel while working etc
  - b.  $P$  : living cost is lower in destination countries
  - c. Still get wage in the US level
  - d. Implication: a
    - i. higher level of leisure and utility
    - ii. Part-time: reserved wage increase

$$PC + WL = WT + y_0$$

$$C = -\frac{W}{P}L + \frac{WT + y_0}{P}$$

