

# Technequality

Understanding the relation between technological innovations and social inequality

Prof. Tomas Korpi SOFI, Stockholm University



#### Outline of talk

- Brief presentation of the Technequality project
- Presentation of results from WP 2
  - Automation risks
  - Cognitive and non-cognitive skills
- Discussion with audience





Project examines links between technological innovation and ...

- The number of jobs
- The nature of our tasks
- Skill needs and education
- Social inequalities

Website: <a href="https://technequality-project.eu/">https://technequality-project.eu/</a>





Six work packages focused on research

- WP 1: The future of work in Europe
- WP 2: Technology, skills and inequality
- WP 3: Educating today for tomorrow's labor market
- WP 4: Reinventing social welfare
- WP 5: Automation, taxation, and public finances
- WP 6: Is this time really different?





# WP2: Technology, skills & inequality

- Task 2.1: Assessing the role of skills, social class, credentials and employment on tech driven labor markets
- Task 2.2: Examining insiders and outsiders on skillbiased labor markets
- Task 2.3: Assessing different consequences for different social classes





Task 2.1 – Description of task

- Focus on automation risk, skills, social class
- Deliverable consisting of
  - Parallel analyses in DE, FI, NL, and SE
    - School-to-work transitions and wages in the short- and long-run (1 yr. vs 10 yr.)
    - Unique individual career perspective on automation risk
  - Three in-depth studies of specific issues
  - Contributions from the EUI, MU, SU, and WZB





#### Task 2.1 – Automation risk

- Technical change may make jobs obsolete
- How is the risk of obsoleteness measured?
  - Expert assessment vs job tasks
- How do they differ?
  - Marked difference in risks; expert assessment yields dramatically higher risks
- What have we done?
  - Apply risks based on task approach





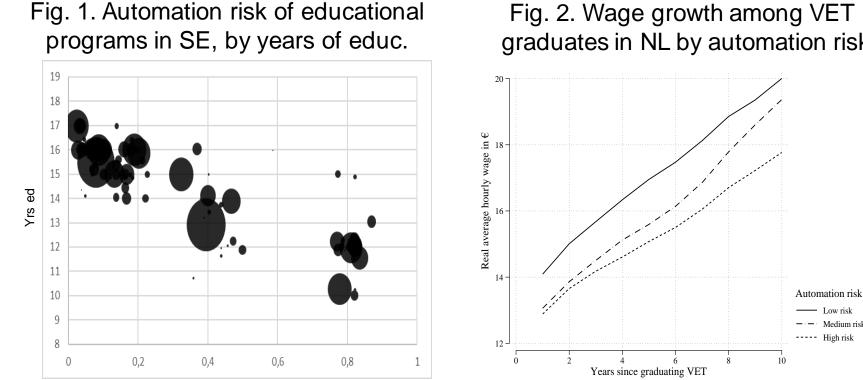
#### Task 2.1 – Automation risk

- Our results
  - Short-term
    - Automation risk may both increase, decrease and be unrelated to earnings
  - Long-term
    - Very similar results





#### Task 2.1 – Automation risk ex. NL and SE

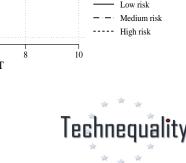




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Automation risk

graduates in NL by automation risk



#### Task 2.1 – Automation risk ex. FI

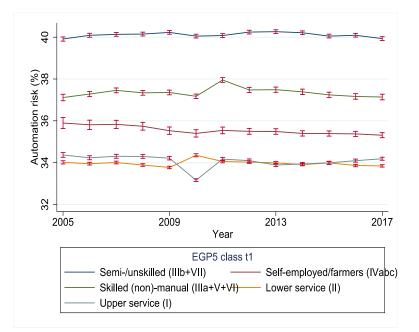


Fig. 3. Automation risk by class in FI

- Higher automation risk increases career class mobility
  - Particularly in the unskilled, skilled and lower service classes
- Also in relation to parental class





Task 2.1 – Automation risk

- Conclusions
  - Automation risk strongly related to education, and maybe dominated by it
  - Automation risk not related to wage growth
    - But related to class mobility (potentially both up-& downward)
  - The consequences of automation likely context dependent, e.g. dependent on
    - Education and training system (youth and adult)
    - Work organization





Task 2.1 – Automation risk

- Questions to audience
  - Is automation risk a useful concept for policy?
  - How should risk be conceived; at the level of industries, occupations, jobs or tasks?





Task 2.1 – Generic skills

- Skills of two kinds: cognitive and non-cognitive (personality)
- Cognitive clearly important for labor market attainment, and non-cognitive most likely as well
- Potentially more important with tech change; cognitive skills to learn new tasks and non-cognitive as personality traits may become more relevant in new jobs
- What have we done?
  - Apply available measures of generic skills (incl. some of the well-know "Big 5" personality traits)





#### Task 2.1 – Generic skills

- Our results
  - Short-term
    - Both cognitive and non-cognitive skills display varied results
  - Long-term
    - Cognitive skills increase earnings
    - Non-cognitive skills more mixed, but tendency to increase here as well





#### Task 2.1 – Generic skills ex. DE and SE

Table 1. OLS regression of skills on earnings among grad. w/ upper secondary vocational degree 1 alt. 10 yrs after grad.

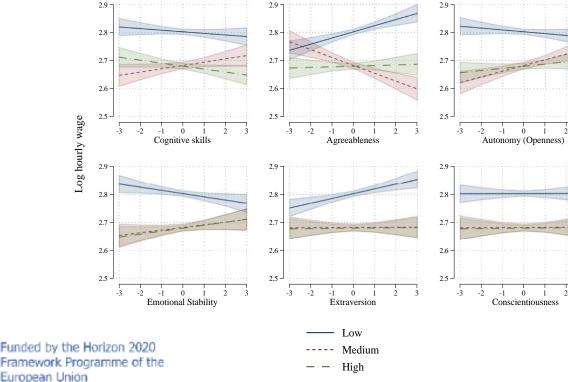
	DE (1 yr.)	SE (10 yrs.)
Cognitive skills	0.050***	0.029***
	(0.012)	(0.009)
Non-cognitive skills		
Extraversion	-0.006	0.035***
	(0.013)	(0.011)
Conscientiousness	0.010	0.010
	(0.012)	(0.010)
Emotional stability	0.038**	-0.017*
*	(0.012)	(0.010)



#### Task 2.1 – Generic skills ex. NL

and wages in the NL 2.9 2.9 2.8 2.8 -2.8 2.7 2.7 2.7 2.6 2.6 2.6 2.5 25. 2.5 Log hourly wage -3 -3 -2 -1 Ó -3 -2 -2 -1 Ó -1 0 Cognitive skills Autonomy (Openness) Agreeableness 2.9 29 2.9 2.8 2.8 2.8

Fig. 4. Automation risk, cognitive and non-cognitive skills





#### Task 2.1 – Generic skills

- Conclusions
  - Both cognitive and non-cognitive skills important
  - Unclear which non-cognitive skill most beneficial
  - Relationship between skills and automation risks varied
    - In particular with regard to non-cognitive skills





#### Task 2.1 – Generic skills

- Questions to audience
  - Should policy focus on generic skills, or on specific?
  - If non-cognitive skills are important, what can policy do and which skills should be in focus?





#### Thank you for your attention!

Prof. Tomas Korpi Swedish Institute for Social Research Stockholm University tomas.korpi@sofi.su.se







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Task 2.1 – Questions to audience

- Automation risk
  - Is automation risk a useful concept for policy?
  - How should risk be conceived; at the level of industries, occupations, jobs or tasks?
- Generic skills
  - Should policy focus on generic skills, or on specific?
  - If non-cognitive skills are important, what can policy do and which skills should be in focus?



