

# Big Data for Labour Market Intelligence

**Capacity development programme 2022** 

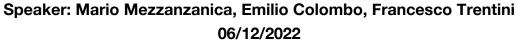
Module 2: dissemination and analysis

# **Session 8**

OJV and labour market statistics: key elements for a methodological approach









#### **Overview**

- What do OJA represent?
- How are they related to labour market variables?
- Can they help to predict labour market variables?
- Linking Big data and traditional survey data using AI to study skills mismatch
- An application: PIAAC2ESCO and analysis of skill gaps in Europe
- Conclusions



# **OJA** and representativeness

Are OJA representative of the LM structure?

- We need to define what we observe
- We need to select a benchmark
- We need to compare OJA with the benchmark



# **OJA** and the population

We observe some (large) number of OJA

Are they representative of the universe of OJA?

We do not know as the true population of OJA is unobservable

We need to estimate it

- Capture recapture models: need ads to be posted in different sites
- Multilevel modelling: individual response is a function of different covariates



# **OJA** finding the right benchmark

Which labour market variable and statistics is closest to the concept of OJA?

#### 1. Job Vacancy Statistics

Advantage: JVS is the closest definition to OJA. Allows for sectoral dimension Problem: JVS are constructed from a survey of firms which has a relatively small sample size and has not information about occupations nor it allows a regional division. Moreover JVS measure the stock of vacancies OJA the flow

#### 2: LFS

Advantage: by far the most detailed and granular survey on Labour Market in Europe Problem: LFS is a supply side variable. Measures the stock of employment, (not vacancies) We can restrict the analysis to job changes <3months but different concept wrt OJA

3. Administrative data: very promising not available across countries

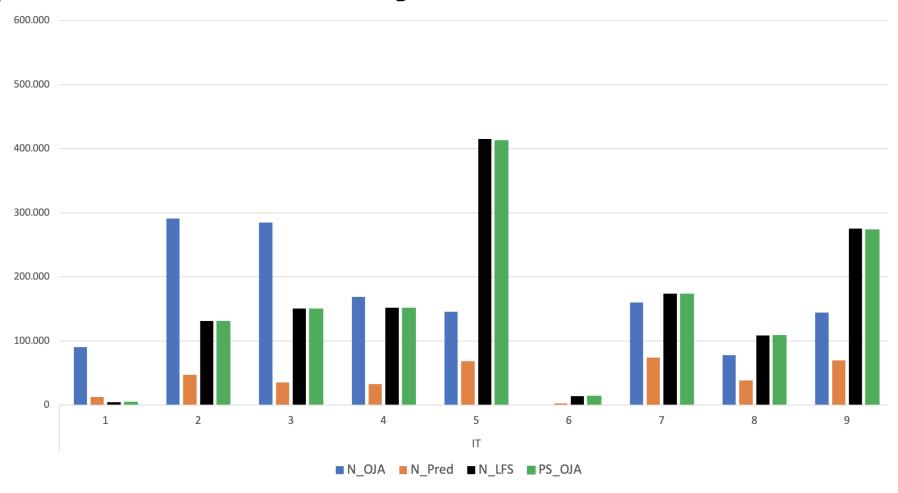


# **OJA representativeness**

If OJA represents a labour market phenomena and they are a biased description of it they can be "corrected" to match the distribution of some benchmark variable Post-stratification delivers just that, i.e. it reweights the population of OJA with weights obtained from the reference sample.



# **OJA representativeness: PS by ISCO 1D**





# **Do OJA predict LM variables?**

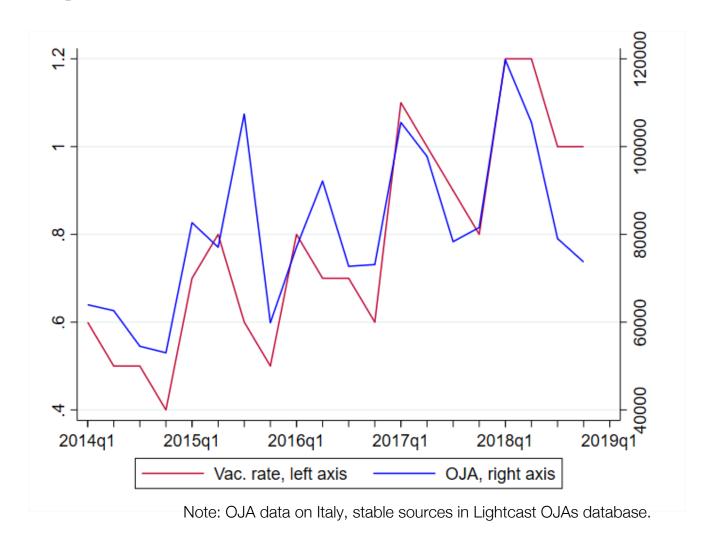
Even if OJA are a biased description of some LM phenomena/variable they may be a good predictor of the same phenomena

For a good fit both in sample and out of sample representativeness is not needed, what is needed is predictive ability (fit or MSE)

This is true both at cross sectional and at time series analysis

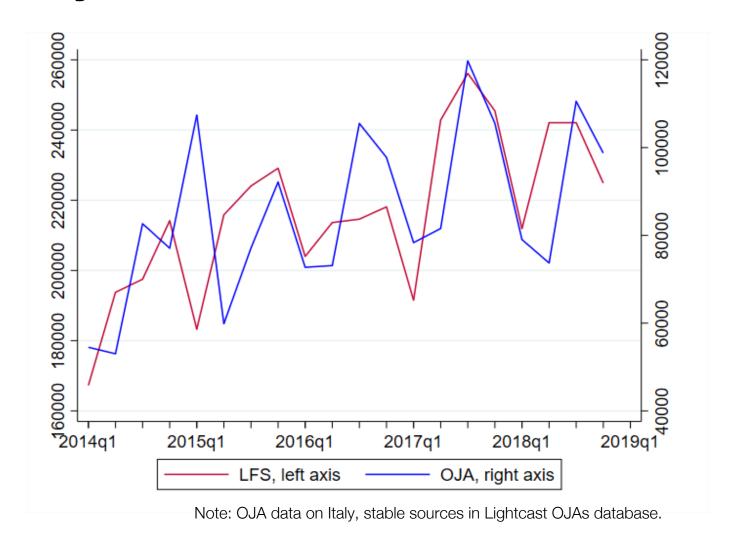


# **Predictive ability: OJA vs JVS**





# **Predictive ability: OJA vs LFS**





#### **Combined use of official statistics and LMI**

Web data can enrich and complement existing datasets.

#### Enrichment

Al-methods using web data can simplify complex tasks and support human experts

- Pros: cost reduction and bounded risk of non-systematic errors
- Caveat: need of transparent design and account of the process

#### Complementarity

Web data can provide detailed information about phenomena otherwise not observed

- Pros: uniqueness (frequency, detail)
- Caveat: representativeness, find robust measures



# PIAAC – Program for the International Assessment of Adult Competences (OECD)

- Cycle I, all rounds: 2012, 2014 and 2017. Cycle II, currently ongoing; data release expected in 2023.
- Representative samples of working-age individuals
- Background questionnaire + test of the performance in some activities (e.g. writing an email, search for a job online)
- Background questionnaire: self-declared intensity (frequency or extent) of skill use in different domains:
  - at work (Module F)
  - Literacy, Numeracy and ICT at work (Module G)
  - Literacy, Numeracy and ICT in everyday life (Module H).
  - And a module on «[...] about how you deal with problems and tasks you encounter.» (Module I)
- PIAAC is used in the literature to measure skill content in a global perspective (Lewandowski et al. 2022) and risk of automation (Nedelkoska and Quintini, 2018)

Lewandowski, P., Park, A., Hardy, W., Du, Y., Wu, S. (2022) Technology, Skills, and Globalization: Explaining International Differences in Routine and Nonroutine Work Using Survey Data, *The World Bank Economic Review*, https://doi.org/10.1093/wber/lhac005

Nedelkoska, L. and G. Quintini (2018). Automation, Skills Use and Training. Technical report, OECD, Paris.



# **PIAAC** background questionnaire

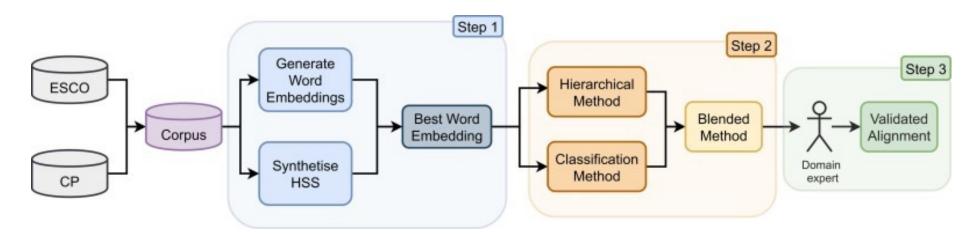
#### **F\_START** Skills used at work

Layout	Item group table  How often ^DoesDid your ^JobLastjob usually involve	
<b>F_Q02b</b> (JRA) (B)	instructing, training or teaching people, individually or in groups?	
	<ul> <li>1 Never</li> <li>2 Less than once a month</li> <li>3 Less than once a week but at least once a month</li> <li>4 At least once a week but not every day</li> <li>5 Every day</li> <li>DK</li> <li>RF</li> </ul>	



### **How to link PIAAC and OJA? Use ESCO and Al**

- The linkage is done using AI in a framework that combines various methods: embeddings, selection of the best embedding, taxonomy alignment and experts' validation
- PIAAC questions are processed to tag the most similar ESCO Skills.
- The embedding is trained on OJA UK data and the matching is done on the English language.



Giabelli, A., Malandri, L., Mercorio, F., & Mezzanzanica, M. (2022). WETA: Automatic taxonomy alignment via word embeddings. Computers in Industry, 138, 103626.



#### **PIAAC2ESCO** validated dataset

• The validated dataset covers **21 PIAAC questions** and the **mapped ESCO skills, enriched** with alternative labels

F\_Q02b: instructing training or teaching people individually or in groups?

- coach young people
- coach youngsters
- educate others
- educate young people
- facilitate young peoples education
- facilitate young peoples mentoring
- instruct colleagues
- instruct others

- instruct young people
- instructing others
- teach others
- teach young people
- train others
- train young people
- train youngsters
- tutoring



# **PIAAC2ESCO** mapping

PIAAC Question Id	PIAAC Question Description	Label	Group	Type of variable
F_Q02b	instructing training or teaching people individually or in groups?	Teaching people	General	Frequency (time units)
F_Q02d	selling a product or selling a service?	Selling	General	Frequency (time units)
F_Q04a	persuading or influencing people?	Influencing people	General	Frequency (time units)
F_Q05a	[] «Problem solving» []. How often are you usually faced by relatively simple problems that take no more than 5 minutes to find a good solution?	Simple problems	Problem solving	Frequency (time units)
G_Q01b	read letters memos or e-mails?	Read letters memos or mails	Literacy	Frequency (time units)
G_Q01g	read bills invoices bank statements or other financial statements?	Read financial statements	Literacy	Frequency (time units)
G_Q01h	read diagrams maps or schematics?	Read diagrams maps or schematics	Literacy	Frequency (time units)
G_Q02a	write letters memos or e-mails?	Write letters memos or mails	Literacy	Frequency (time units)
G_Q03b	calculate prices costs or budgets?	Calculating costs or budgets	Numeracy	Frequency (time units)
G_Q03c	use or calculate fractions decimals or percentages?	Use or calculate fractions or percentages	Numeracy	Frequency (time units)
G_Q03d	use a calculator - either hand-held or computer based?	Use a calculator	Numeracy	Frequency (time units)
G_Q03g	use simple algebra or formulas?	Use simple algebra or formulas	Numeracy	Frequency (time units)
G_Q03h	use more advanced math or statistics such as calculus complex algebra trigonometry or use of regression techniques?	Use advanced math or statistics	Numeracy	Frequency (time units)
G_Q04	you use a computer in your job?	Experience with computer in job	ICT	Yes (1) / No (2)
G_Q05a	use email?	For mail	ICT - Internet	Frequency (time units)
G_Q05d	conduct transactions on the internet for example buying or selling products or services or banking?	Conduct transactions	ICT - Internet	Frequency (time units)
G_Q05e	use spreadsheet software for example Excel?	Spreadsheets	ICT - Computer	Frequency (time units)
G_Q05f	use a word processor for example Word?	Word	ICT - Computer	Frequency (time units)
G_Q05g	use a programming language to program or write computer code?	Programming language	ICT - Computer	Frequency (time units)
I_Q04d	I like learning new things	Like learning new things	Learning strategie	s Extents ***
I_Q04I	I like to figure out how different ideas fit together	Figure out how different ideas fit together	Learning strategie	s Extents

# **PIAAC2ESCO – Open data**

https://crisp-unimib.github.io/PIAAC2ESCO/

#### Open access to:

- Dataset enriched mapping
- Methodological annex

# PIAAC2ESCO - An Al-driven classification of the PIAAC Background questionnaire onto the ESCO Skills Pillar

View on GitHu

# PIAAC2ESCO - An Al-driven classification of the PIAAC Background questionnaire onto the ESCO Skills Pillar

#### What is PIAAC2ESCO?

PIAACZESCO provides a characterisation of the PIAAC background questionnaire on the base of the ESCO Skills Pillar. In practice it associates a list of ESCO skills (v1) to questions of the PIAAC background questionnaire (version 2010), based on their similarity. We use the section F to I of the PIAAC background questionnaire, from which we select the relevant questions (73 questions out of 84) and all the ESCO skills (13600 items). The validated dataset covers 21 PIAAC questions and the mapped ESCO skills, which are enriched using alternative labels.

#### How does PIAAC2ESCO work?

The linkage is done using Ai in a framework that combines various methods: embeddings, selection of the best embedding, taxonomy alignment and experts' validation. A description of the adopted methodology is available in the Technical Annex.

The training dataset of the embedding is the representative sample of the job ads collected by Eurostat and Cedefop as part of the Web Intelligence Hub - Online Job Advertisements (WIH-OJA)





# **Skill mismatch across Europe**

Descriptive insights on skills mismatch in 17 European countries in 2019. Relation with automation and training.

Yuchen Guo & Christina Langer & Fabio Mercorio & Francesco Trentini, 2022. "Skills Mismatch, Automation, and Training: Evidence from 17 European Countries Using Survey Data and Online Job Ads," EconPol Forum, CESifo, vol. 23(05), pages 11-15, September.



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the EU."
However, empirical evidence on the prevalence of skills mismatch between the skills requested by employers and the skills provided by employers and the skills provided by employees across Europe is scarce. We contribute to the understanding skill domains such as digital skills. Finally, we investigate the skill domains such as digital skills. Finally, we investigate the skill domains such as digital skills. Europe is caracte. We contribe to the understanding of skills firmland, we will obtain such as digital skills. Finally, we invest- of skills sparse arcs can be to be to be skill sparse arcs can be skill sparse arcs can ad skill domains such as depart and the skill sparse arcs can ad skill domains under investment of the skill sparse arcs can ad skill domains under investment of the skill sparse arcs can ad skill domains under investment of the skill sparse arcs can be to skill sparse arcs can be skill sparse arcs can be to skill sparse arcs can be skilled sparse arcs ca

POLICY DEBATE OF THE HOUR

■ Linking survey data and online job ads offers new insights into skills gaps in the EU

workers have skill supply surpluses

Matching labor market needs and skill supply remain

On-the-job training might be a potential measure to meet future skills needs

2022

#### job vacancy data on skills requested by employers and survey data on skills supplied by workers, We ADS OFFERS NEW INSIGHTS INTO SKILLS

and survey data on skills supplied of yworders. We AUS UPPERS NEW INSIGHTS IN IT O SALLS document four key findings: first, skill gaps in the European Union exist, but the extent and direction vary across occupation types workers in cognitive intensive occupations provide more skills than are skills demanded by employers and the skills supdemanded (skill surplus), whereas workers in man-ual intensive occupations face higher skill demand online job vacancies (OJV) data from the European compared to the skills they have (skill shortage). Center for the Development of Vocational Train compared to the skills they have (skill shortage). Genter for the Development of Vocational Trainin Second, this pattern is consistent across annost all LT countries that are part of our analysis. This sug-gests that overall patterns of skills minimath do not reflect country-specific factors but are arther subgroupm-under phonemone.



of labor market and skills needs as a top priority fo



#### **Data**

#### **Online Job Ads (WIH-OJA, Eurostat and Cedefop)**

- Collection of online job ads from 27 European countries + UK and EFTA countries. Since 2018Q4
- Data on occupations and related skills as they emerge from online job postings
- Skills are extracted based on the ESCO Skill Pillar

#### **Samples**

- PIAAC comprises 250,000 observations (4,000 8,000 per country). 2012 and 2014. Projected to 2019 using changes in US (observed in 2014 and 2017) as inflation parameters.
- WIH-OJA includes 17,966,812 observations in 2019.



### **Skill mismatch - measure**

For each skill in demand and supply, the RCA is ranked among all occupation and mapped to the percentile of belonging.

Our **mismatch measure** at the occupation level is the mean RCAs-percentile-rank gap between demand and supply. Negative values indicate over-skilling, vice versa positive values indicate under-skilling.

An example:

F\_Q02b: "[...] teaching people individually or in groups?" → ESCO skill: "Teaching others"

RCA<sub>oja</sub> percentile rank: 0.95

RCA<sub>piaac</sub> percentile rank: 0.97

Skill gap:  $pRCA_{oia} - pRCA_{piaac} = -0.02$ 

<0 Over-skilling

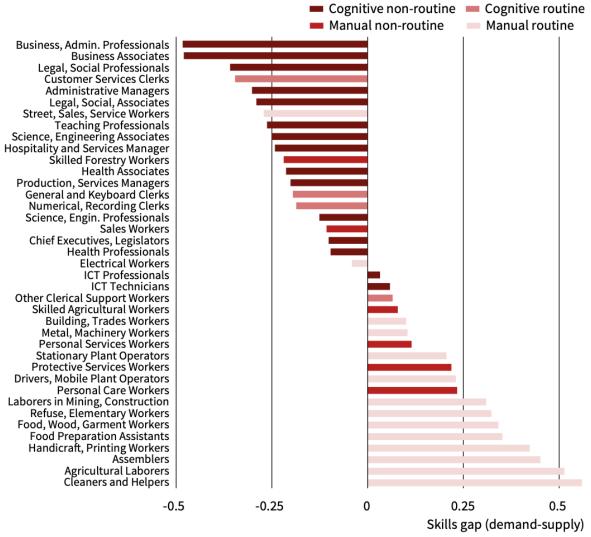


# **Findings**

Skill gaps indicates under-skilling is more pervasive among manual workers than cognitive workers.

Autor, D. H., F. Levy, and R. J. Murnane (2003). The Skill Content of Recent Technological Change: An Empirical Exploration. The Quarterly Journal of Economics 118 (4), 1279–1333.

#### Average Skill gap by occupation ISC008 II digit. 2019 pooled 17 countries.



Note: Pooled for 17 European countries. Source: CEDEFOP; PIAAC.

**European Training Foundation** 

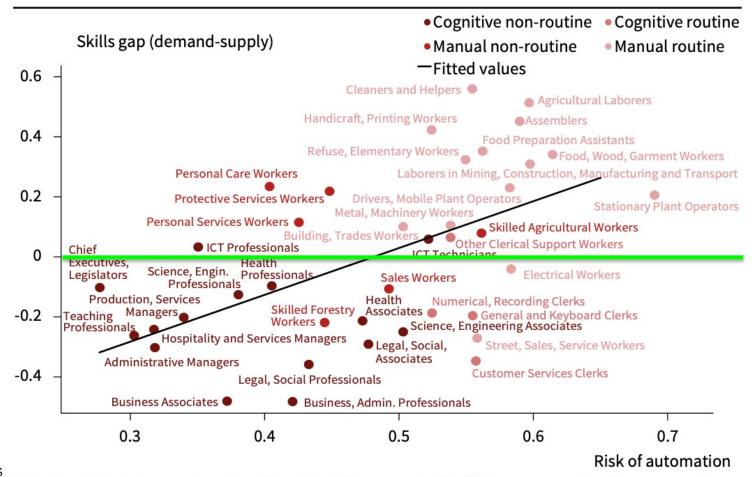
# **Automation changes job content**

#### Automation Risk (2012) and Skill gaps (2019). Pooled, 17 countries

Risk of automation from Nedelkoska and Quintini (2018)

Positive relation between underskilling and risk of automation.

Jobs are changing in terms of tasks and the skill composition of jobs changes.



Nedelkoska, L. and G. Quintini (2018). Automation, Skills Use and Training. Technical report, OECD, Paris.

Note: Correlation between automation risk and the skills gap, pooled for 17 European countries. Our measure of automation risk stems from Nedelkoska and Quintini (2018), who constructed the automation probability for all occupations and countries in our sample using PIAAC data.

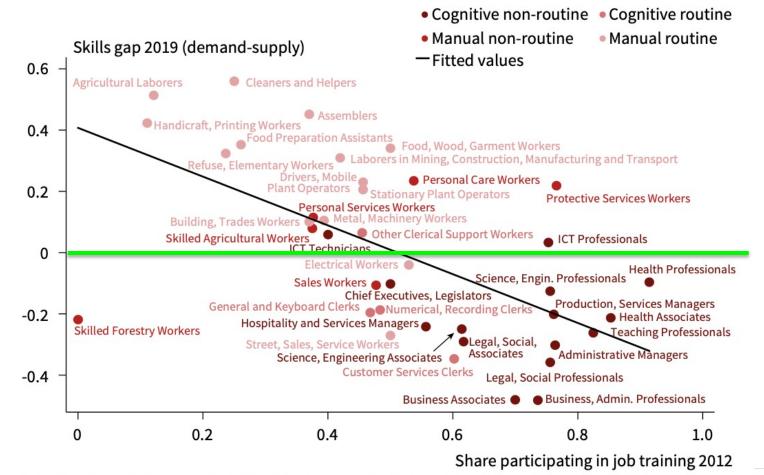
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# **Training**

On aggregate, on-the-job training is negatively related to skill gaps.

Occupation with high levels of underskilling also present low participation in on-the-job training.

#### On-the-job training (2012) and skills gaps (2019). Pooled, 17 countries



Note: Correlation between on-the-job training (measured in 2012) and the skills gap (measured in 2019), pooled for 17 European countries.

Source: CEDEFOP; PIAAC.



# **Findings**

Descriptive figures show high heterogeneity among European regions.

Within-country heterogeneity is also relevant

- Over-skilled (on average): BE, DE, FR, IE, SE
- Mix: EL, ES, IT, LT, PL

High heterogeneity at the country and regional level calls for more depth in understanding the role of institutional features of the countries labour markets.

