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# Lessons Learned form Studies on Education-Labour Market Matching

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#### **About TIB**



#### Team

#### **Principle Investigators**





Dr. Gábor Kismihók (TIB)

Dr. Stefan Mol (UvA)



Prof. Dr. Maria-Esther Vidal (TIB)

#### **Doctoral and PostDoc Researchers**



Reza Tavakoli (TIB)



Dr. Hannah Berkers (TUE)



Jarno Vrolijk (UvA)



Vladimer Kobayashi (UvA)



Dr. Alan Berg (UvA)







#### University of Amsterdam

#### **Personalization of Learning and Work**

TIB

Learning (and work) is personal, driven by a great number of individual goals and contexts

For a fair selection everybody has to take the same exam: please climb that tree

# **Our Education System**

"Everybody is a genius. But if you judge a fish by its ability to climb a tree, it will live its whole life believing that it is stupid."

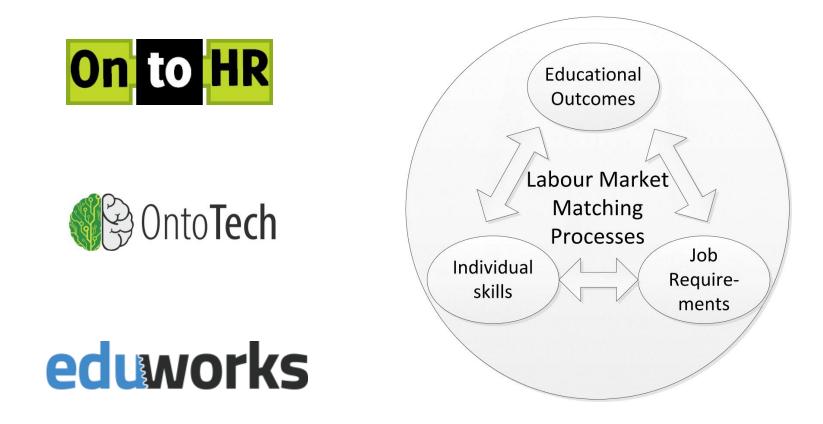
- Albert Einstein

Image:

https://www.psychologytoday.com/us/blog/fin ding-the-next-einstein/201404/do-we-havetrouble-taking-objective-feedback

## **Focus on Individuals and Organisations**





Kobayashi, V. B., Mol, S. T., Berkers, H. A., Kismihók, G., & Den Hartog, D. N. (2017b). Text Mining in Organizational Research. Organizational Research Methods, 1094428117722619. <u>https://doi.org/10.1177/1094428117722619</u>







## What type of data?

#### Data about the context

- Vacancy data
- CV data
- Occupational classification
- Course syllabi
- Surveys
- Qualitative data about work

#### Learning Records

Performance data

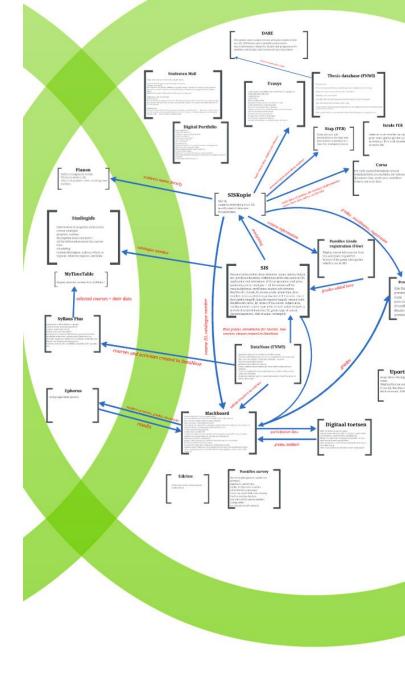
Grades Assignments

Behavioral data

Clicks Content views Social media

#### **Data Providers**

- Textkernel
- Monsterboard
- UWV (NL)
- USG (NL)





#### Redefining nursing education on the basis of labour market changes

- Robotization
- New tasks
- New skills
- Complexity of nursing/care taking occupations

#### Output

- Mapping nursing skills and tasks
- Developing assessment/intervention methods
- Curriculum recommendations

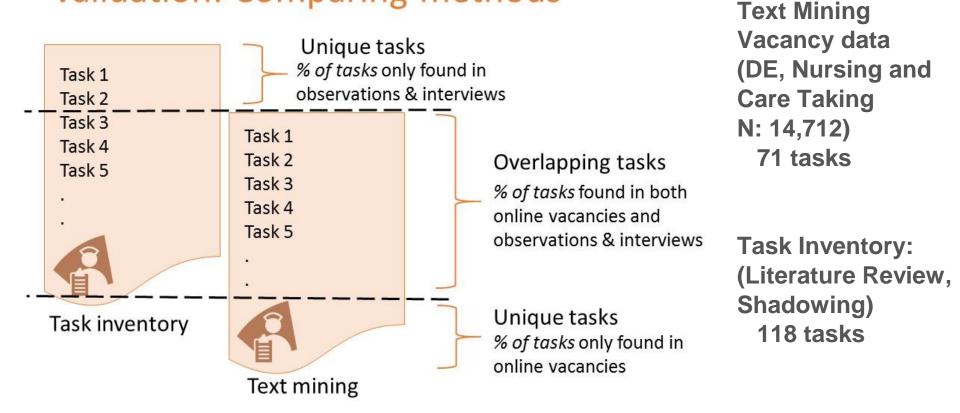


Lifelong Learning Programme



## Can we create nursing job profiles automatically?

## Validation: Comparing methods



Berkers, H., Mol, S. T., Kobayashi, V., Kismihók, G., & Hartog, den D. (2019). Big (data) insights into what employees do—A comparison between task inventory and text mining job analysis methods. In PhD Thesis. What do you do and who do you think you are? (pp. 12–57). Retrieved from <u>https://pure.uva.nl/ws/files/31377407/Chapter\_2.pdf</u>

Kobayashi, V., Mol, S. T., Kismihok, G., & Hesterberg, M. (2016). Automatic Extraction of Nursing Tasks from Online Job Vacancies. In M. Fathi, M. Khobreh, & F. Ansari (Eds.), Professional Education and Training through Knowledge, Technology and Innovation (pp. 51– 56). Retrieved from <u>http://www.pro-nursing.eu/web/resources/downloads/book/Pro-Nursing\_Book.pdf</u>

Order medications		
Order material		
Restock material	Organizing the orders of material	
Restock medication		
Tablet system	Preparing and disposing of materials	
Expire dates		
Fridge temperature	Realising interventions of quality assurance	
Closed medication		
Prepare documentation		
Documentation general	Providing nursing documentation	
Check documentation	Keeping the data up to date	
	Auditing medicial transcriptions	
Insurance memo	Description	
Administration	Documenting of services	
Schedule	Creating shift schedules	
Work division		
Dividing patients		
Scheduling beds	Organizing the work on the ward	
Plan operations		
Scheduling appointments		
Prepare documentation new patients	Prodiving nursing documentation	
Take-over patient		
Welcome new patients	Admissioning and discharging patients	
Admittance new patients		
Prepare tests	Collecting estigate data	
Documentation new patients	Collecting patients' data	
Inform patients	Giving patients and family advice	
Hand out prescriptions and medications	Working on doctors' prescriptions	
End documentation	Prodiving nursing documentation	
Arrange transfer	Transitioning of patients into home care	
Arrange transport	mansiconing of patients into nome care	
Transfer patients	Accompanying patients	
Family	Giving patients and family advice	
Host visits	Participating in visits	
Host head doctor	Assisting at examination	
Inform dootor		
Answer telephone	Communicating with patients, family, and staff	
Unscheduled meetings		
Cross-team meeting	Participating in meetings	



Text

Mining

#### Task Inventory

Check documentation	Keeping the data up to date		
	Auditing medicial transcriptions		
Insurance memo	Descent action of a series		
Administration	Documenting of services		
Schedule	Creating shift schedules		
Work division			
Dividing patients			
Scheduling beds	Organizing the work on the ward		
Plan operations			
Scheduling appointments			
Prepare documentation new patients	Prodiving nursing documentation		
Take-over patient			
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Answer telephone	Communicating with patients, family, and staff		
Unscheduled meetings			
Cross-team meeting	Participating in meetings		
Team meeting			
Supervise students			
Show students	Training and advising students		
	Participating in training programs		
Education session	Participating in hygiene training		
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## Analysis



#### Method:

Panel of German nurses evaluated the task list (N=65) for inclusion, frequency, importance

#### **Results:**

- 64.6% of overlap, 22.7% unique in task inventory and 12.7% were unique in text mining
- The two lists are not interchangeable
  - Level of detail is different
  - > TM is more context sensitive
  - > TI is more fundamental



TM tasks were more abstract and less detailed, but arguably provided a sufficient overview of what nurses do

TM generally yielded higher inclusion and importance ratings

TM is more suitable to address the nonstandard nature of work and complement current forms of job analysis

#### **Fresh from the printery**



Dr. Pablo de Pedraza (JRC)



Dr. Stefano Visintin (UCJC)



Prof. Kea Tijdens (UvA)

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#### Survey vs Scraped Data: Comparing Time Series IZA Journal of Properties of Web and Survey Vacancy Data ... Labor IZA Journal of Labor **Economics** X Economics View More + Pablo de Pedraza Martin , Stefano Visintin Martin 2, Kea Tijdens Martin Gábor Kis... IZA sense o ano senses o concentration 🔁 Open access 🚾 🛈 Volume/Issue: Volume 8: Issue 1 < Published online: 11 Sep 2019 PDF DOI: https://doi.org/10.2478/izajole-2019-0004 ■ Volume 8: Issue 1 Search Abstract Article References Article Recommendations Q Search within Journal Issue Journal Abstract Q This paper studies the relationship between a vacancy population obtained from web crawling and vacancies in the Volume Issue economy inferred by a National Statistics Office (NSO) using a traditional method. We compare the time series properties of samples obtained between 2007 and 2014 by Statistics Netherlands and by a web scraping company. We find that the web and NSO vacancy data present similar time series properties, suggesting that both time series Journal information are generated by the same underlying phenomenon: the real number of new vacancies in the economy. We conclude that, in our case study, web-sourced data are able to capture aggregate economic activity in the labor market. ISSN: 2193-8997

Keywords: web crawling: statistical inference: time series: vacancies: Labor demand: data collection: J23: J63: C22:

Pedraza, P. de, Visintin, S., Tijdens, K., & Kismihók, G. (2019). Survey vs Scraped Data: Comparing Time Series Properties of Web and Survey Vacancy Data. IZA Journal of Labor Economics, 8(1). <u>https://doi.org/10.2478/izajole-2019-0004</u>

#### Survey vs Scraped data



#### Objectives

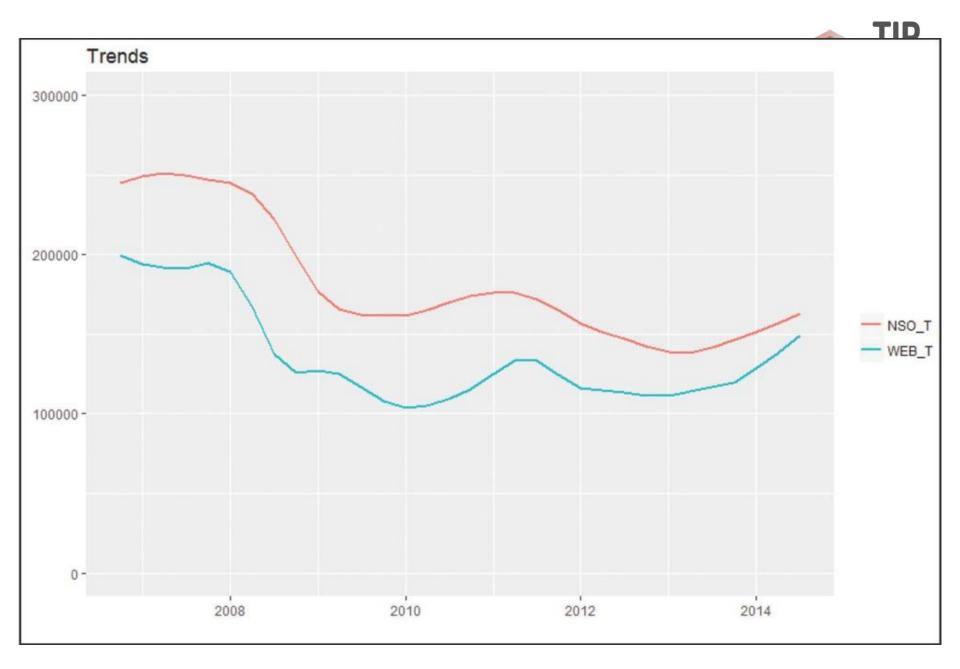
Benchmarking Survey and Web Vacancy datasets

#### Data (2007-14, 8 years 31 quarters)

- NSO of the Netherlands (CBS) survey to measure the number of vacancies at the end of each quarter
- Textkernel data

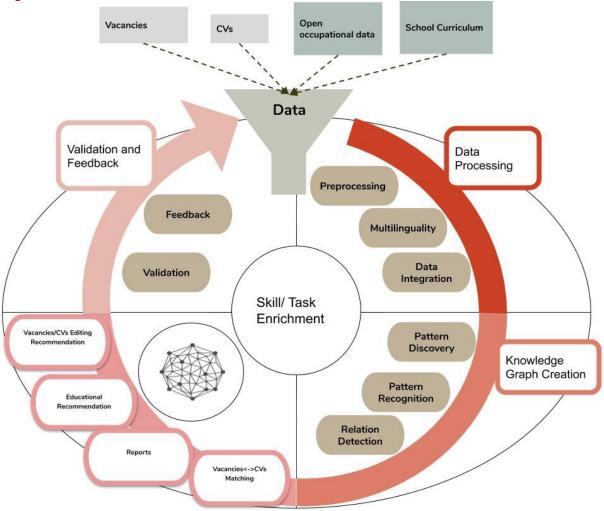
#### **Results**

Evidence of highly correlated co-movements between the time series of Web and NSO



# Work in progress - Dynamic taxonomy development





Kobayashi, V. B., Mol, S. T., Berkers, H. A., Kismihók, G., & Den Hartog, D. N. (2017a). Text Classification for Organizational Researchers: A Tutorial. Organizational Research Methods, 1094428117719322. <u>https://doi.org/10.1177/1094428117719322</u>

#### New skills for Robotization



De-constructing and re-constructing jobs for humanmachine learning and co-working



## **Transferable skills**

The top ten skills, abilities, and knowledge areas associated with rising occupations confirm the overall importance of so-called 21st century skills, and point to those that will be in greatest demand.

	UK		USA
1	Judgment and Decision Making	1	Learning Strategies
2	Fluency of Ideas	2	Psychology
3	Active Learning	3	Instructing
4	Learning Strategies	4	Social Perceptiveness
5	Originality	5	Sociology and Anthropology
6	Systems Evaluation	6	Education and Training
7	Deductive Reasoning	7	Coordination
8	Complex Problem Solving	8	Originality
9	Systems Analysis	9	Fluency of Ideas
10	Monitoring	10	Active Learning

#### Source: FUTURE OF SKILLS, EMPLOYMENT IN 2030 https://futureskills.pearson.com/research/assets/pdfs/media-pack.pdf



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**Example of Other Projects / Application Areas** 



Hybrid Jobs (Teacher Training)

**Refugee skills to enter EU labour markets** 

Predicting the next job of an employee

**Gender bias in selection** 





# rd Annual Learning & Student Analytics Conference An Ethical Vision of Learning Analytics Individuals VS Community #LSAC2019 Université de Lorraine / Loria Lab. Loria, Campus Scientifique, 54506 Vandoeuvre-lès-Nancy, France Save the date : 22-23 October 2019 Loria - Université de Lorraine - The place to be !

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# **MORE INFORMATION**

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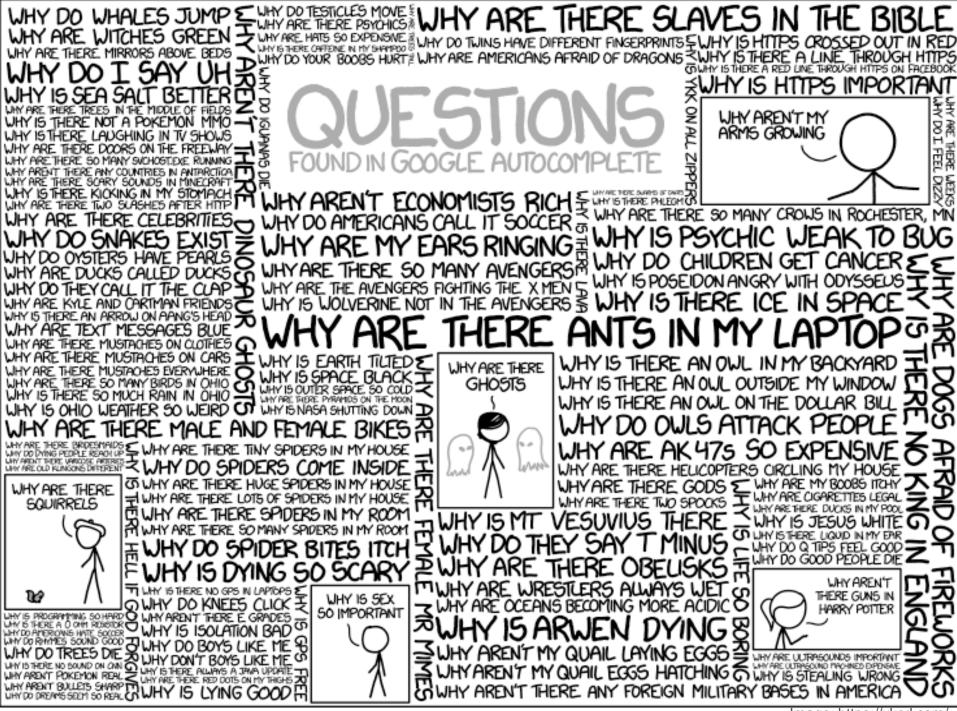


Image: https://xkcd.com/