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Machine Learning in Production Planning and Control: A Literature Review

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Production Planning & Control (PPC), context in Industry 4.0

Manufacturing context:



Volatile
Markets



Mass
Customization



Demanding
Delivery Dates



Edge over
competitors

Cost killer

- 67% of the companies from high-wages countries claim that the adherence to the delivery dates is their main logistical target
- They struggle to achieve that due to machine failures, missing raw materials or short-term customer changes
- Prerequisite to reach logistical targets → Excellent PPC processes

(Reuter *et al.*, 2016)

Production Planning & Control, challenges in Industry 4.0

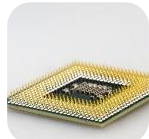
Production Planning & Control (PPC)



Enablers



High volume of data: 1000 Exabytes per year (Tao et al., 2018)



More powerful processors

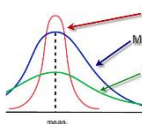


Internet of Things systems and faster internet connexion

Inhibitors



Machine breakdowns



Variable production times



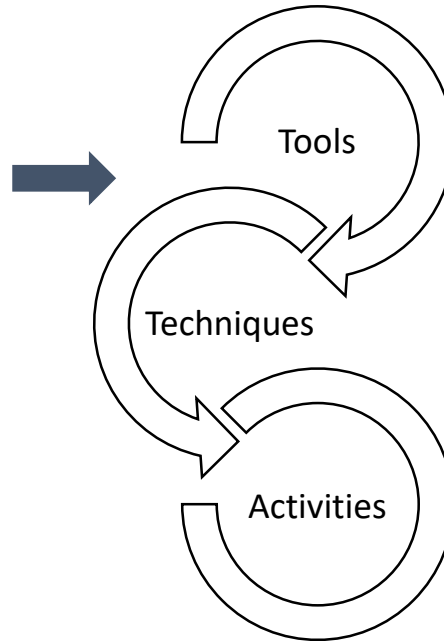
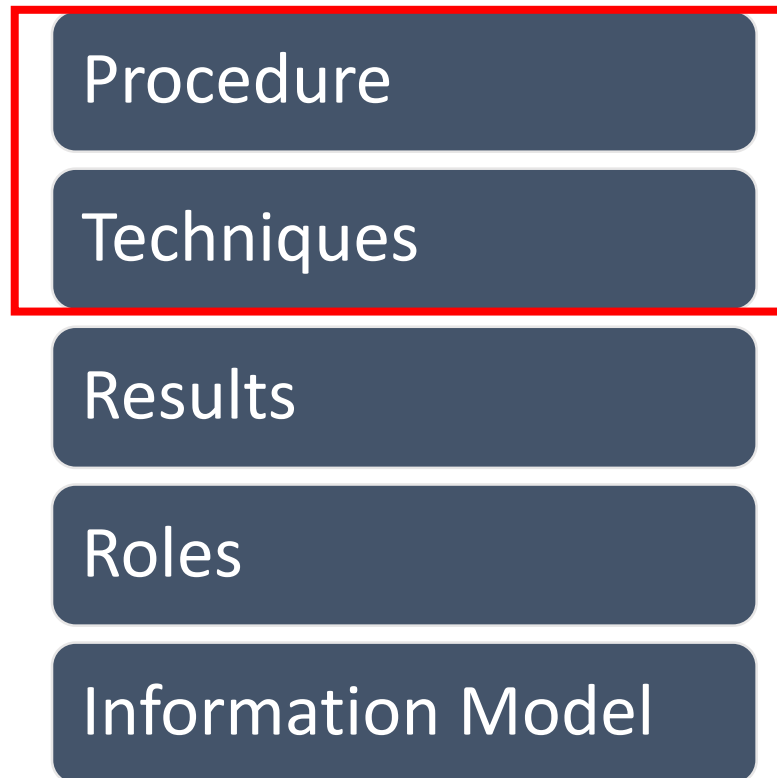
Changes in customer orders

Unfeasible PPC in the execution phase

Need of a more robust PPC

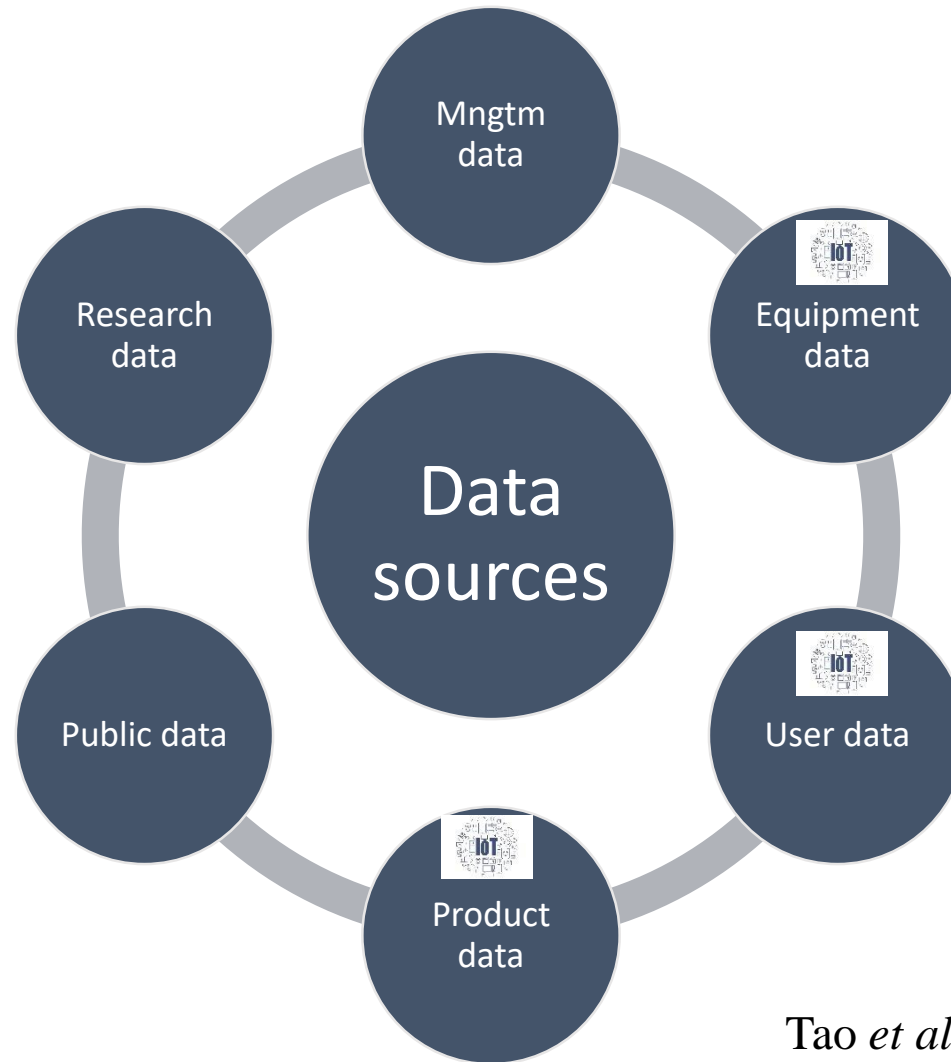
Research objectives (1/2): how to implement ML-aided PPC

Mandatory Elements of a Method (Zellner, 2011)



Which are the activities, techniques and tools used to deploy ML-aided PPC?

Research objectives (2/2): where to find the data?



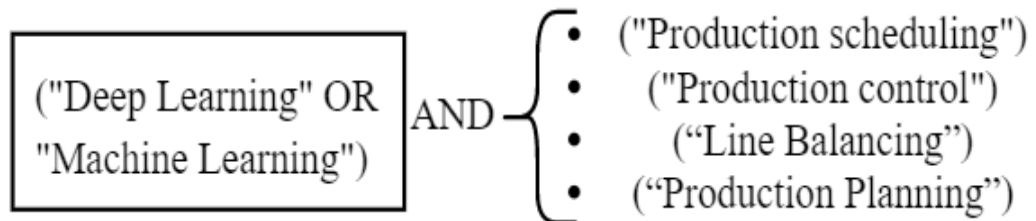
Which are the commonly used data sources when implementing a ML-aided PPC?

Tao *et al.* (2018)

Literature review methodology

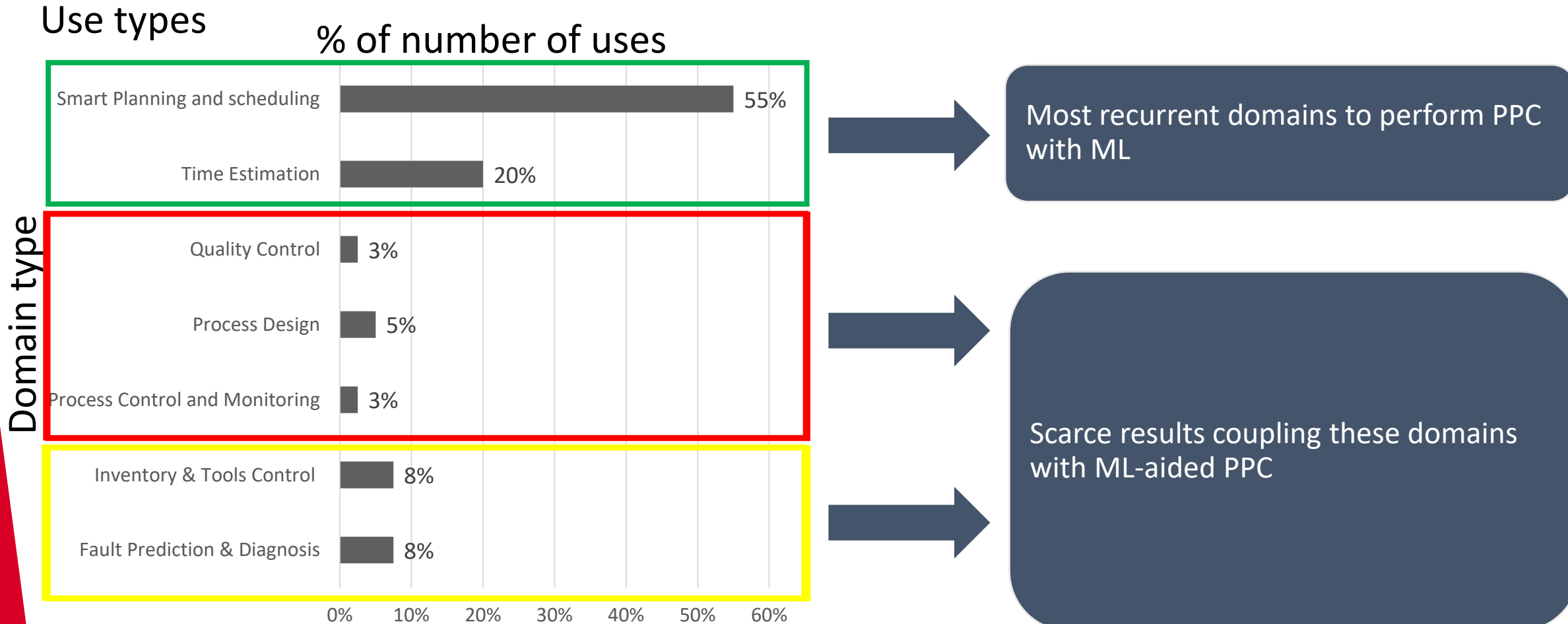
Used keywords

Results



Step	Science Direct	SCOPUS	Kept Articles	Excluded
First results	1538	611	2149	0
Year >= 2011	761	356	1117	1032
Only "Research Articles" (Science Direct) "Conference Paper OR Article" (SCOPUS)	606	299	905	212
Title and Abstract Review	46	24	70	835
Duplicates removal	32	15	47	23
Full text analysis: short list	27	13	40	7

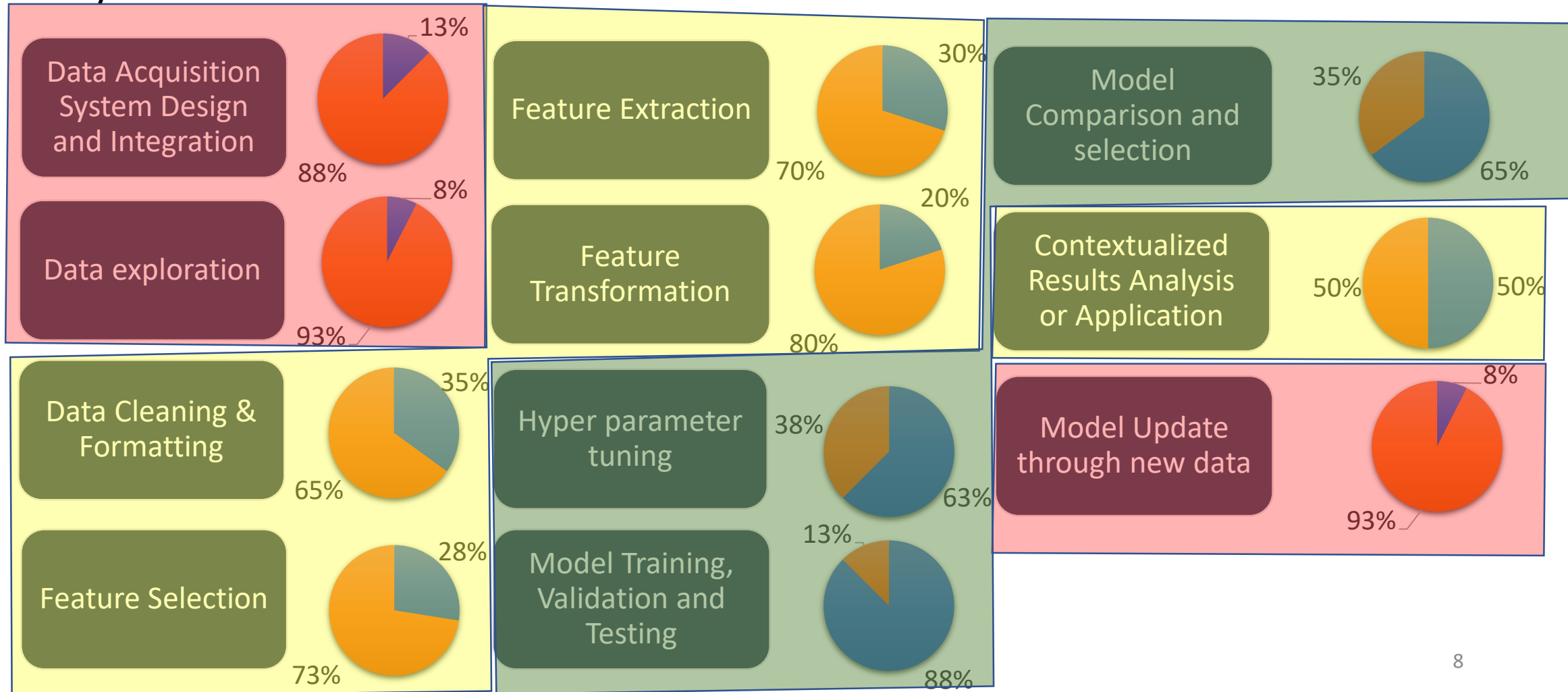
Results (1/5): use types overview



Results (2/5): activities

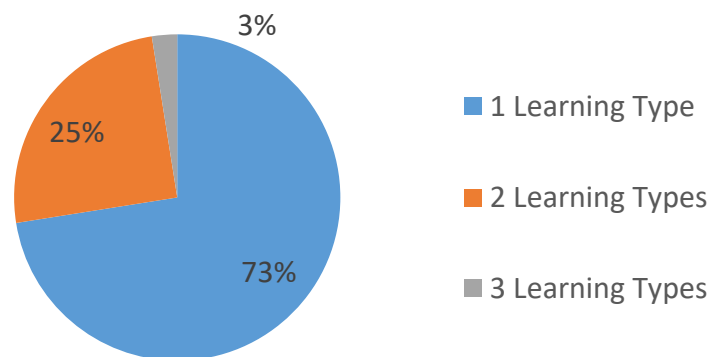
11 key activities were identified

■ Included ■ Not Included

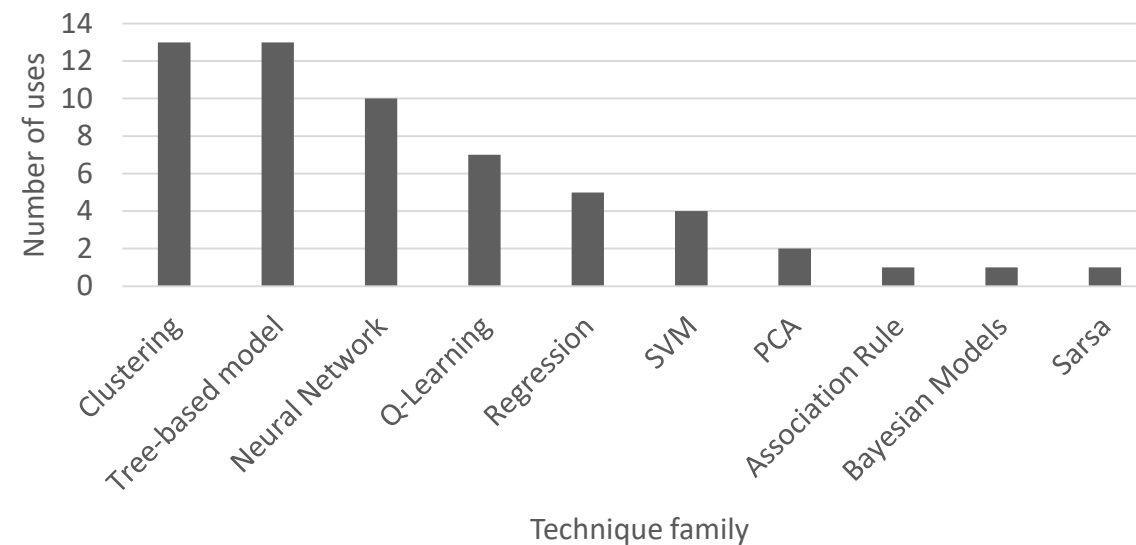


Results (3/5): techniques

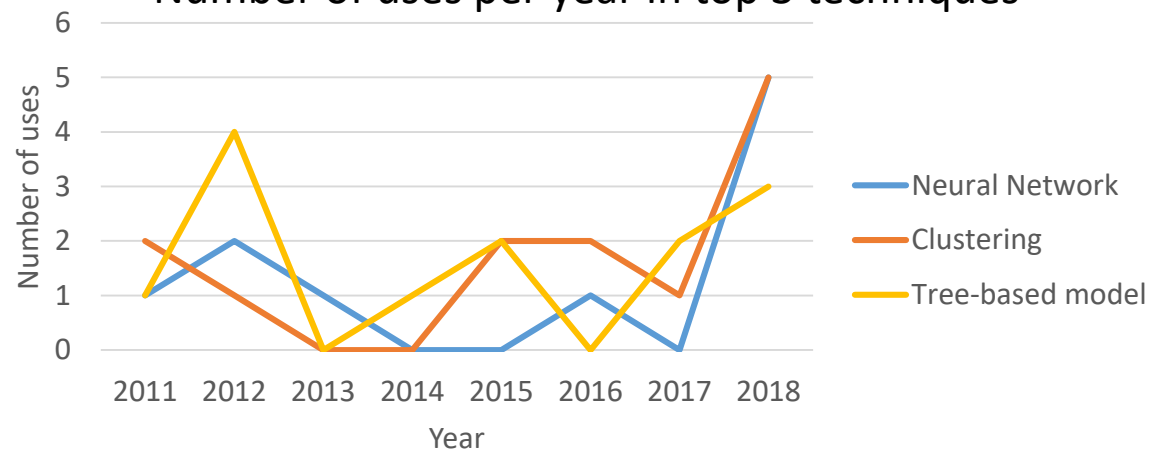
% of publications by used learning types



Number of uses per technique

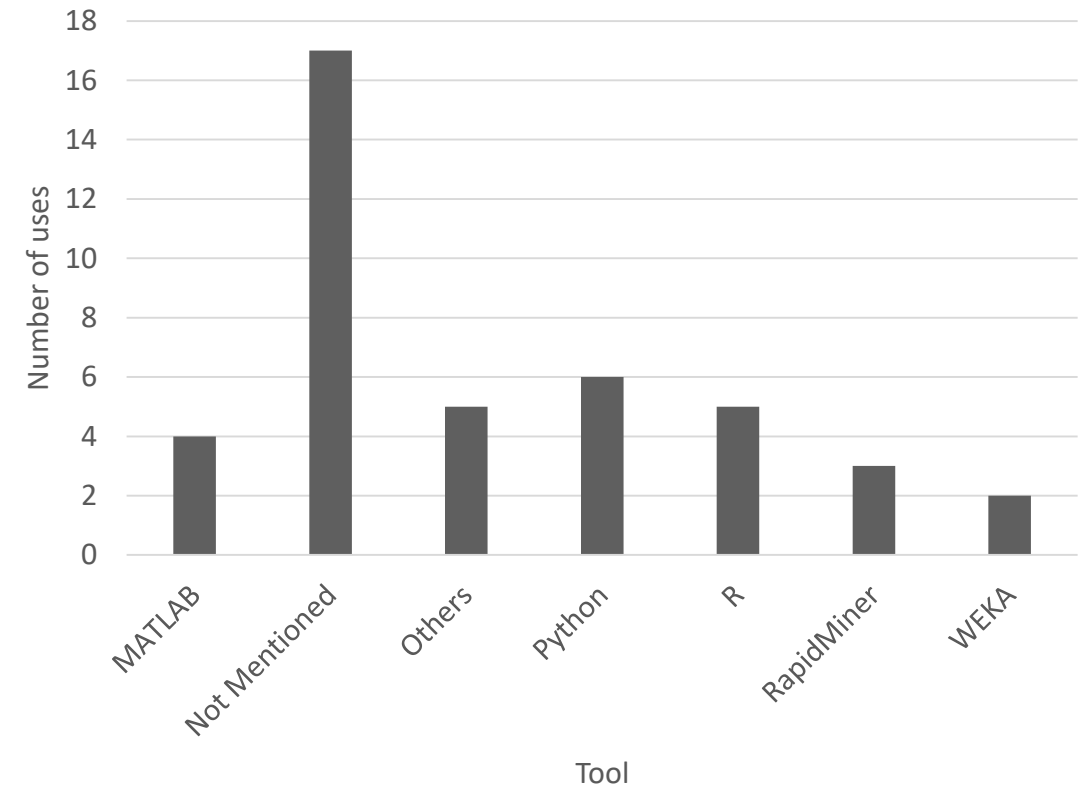


Number of uses per year in top 3 techniques



Results (4/5): tools

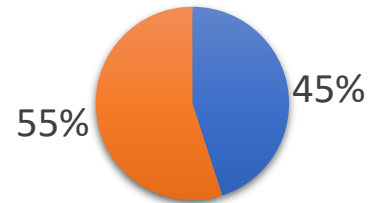
Tools	Number of uses
Not Mentioned	17
R	5
MATLAB	4
Python	3
RapidMiner	3
Tensorflow (Python based)	2
WEKA	2
ACE Datamining System	1
Clementine	1
Keras (Python based)	1
Neural-SIM	1
Visual C++	1
Xelopes Library	1



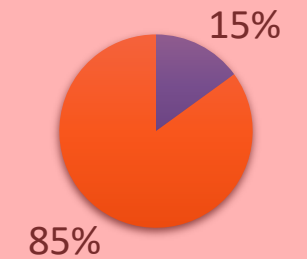
Results (5/5): Data source utilization

■ Used ■ Other data sources

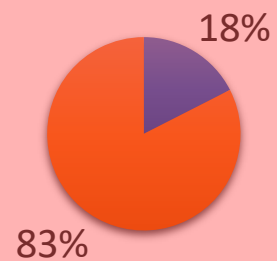
Management data



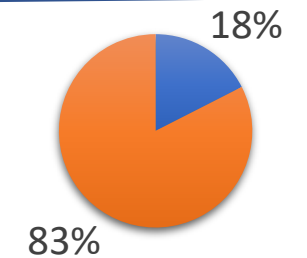
Product data



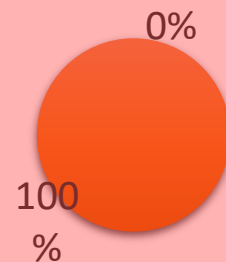
Equipment data



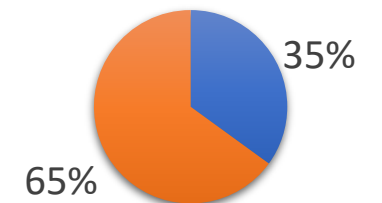
Public data



User data



Research data



Further research

- Develop a Robust Procedure to implement ML-aided PPC → Give an order to different activities
- Link techniques and tools to activities (set the basis to an Information Model)
- Test both previous points with an application

Questions?

References

- Tao, F. *et al.* (2018) 'Data-driven smart manufacturing', *Journal of Manufacturing Systems*. The Society of Manufacturing Engineers, 48, pp. 157–169. doi: 10.1016/j.jmsy.2018.01.006.
- Reuter, C. *et al.* (2016) 'Improving Data Consistency in Production Control by Adaptation of Data Mining Algorithms', *Procedia CIRP*. The Author(s), 56, pp. 545–550. doi: 10.1016/j.procir.2016.10.107.