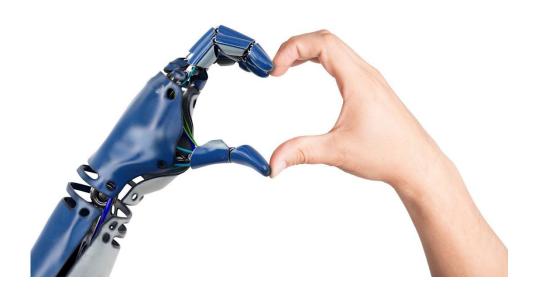


Human-Al Interaction in Supply Chain Decision-Making

12.05.2023

Christina Imdahl Assistant Professor for Machine Learning in OM Eindhoven University of Technology



SOME EXAMPLES OF SC PLANNING DECISIONS

Component sourcing



"How much to order today to ensure we can build the vehicles next week?"

Supermarket replenishment



"Which quantity will be sold of each product in each store each day?"

Truck routing



"Which routing do I suggest to my truck drivers?"

AI TO THE RESCUE

"We have found that around 15 percent of the global workforce, or about 400 million workers, could be displaced by automation in the period 2016–2030"

From: "AI, automation, and the future of work: Ten things to solve for", McKinsey, June 2018

"With sufficient memory and computation, AI-based solutions can easily look across millions of parts and projects and billions of relationships to identify opportunities [...]"

From: "The future is now: Unlocking the promise of AI in industrials", McKinsey December 2022

AI HAS A LOT OF POTENTIAL, BUT...



Predicting risk of welfare fraud

DIFFERENT TRAITS, DIFFERENT RISK SCORES

	DECRE	ASED RISK	INCREASED	RISK
AGE 60	63.0%			~
AGE 30				45.9%
NON-PARENTS	1	4.4%		
PARENTS			21	9%
MEN		9.5%		
WOMEN		•	13.6%	
STRONG DUTCH SKILLS	50.7%			
WEAK DUTCH SKILLS			9.7%	

Source unknown (reddit), not replicable with the current version of <u>https://chat.openai.com/chat</u> as of March, 16th 2023

WHY MACHINES? WHY HUMANS?



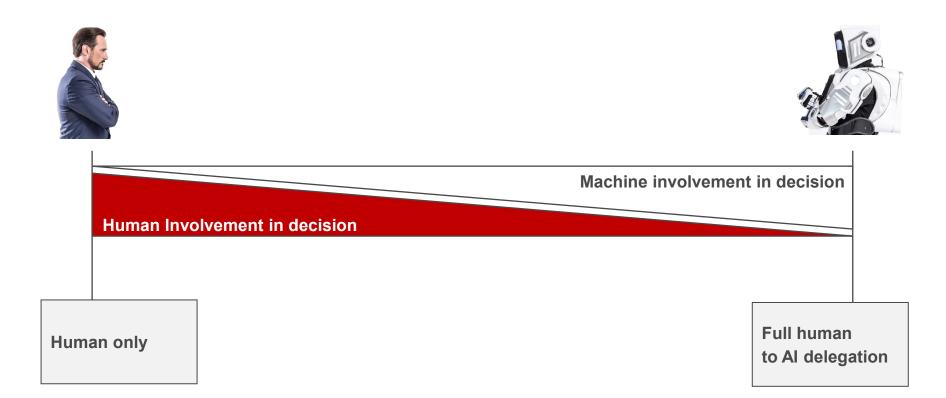
- Computing power
- Cost
- Rational behavior
- ...



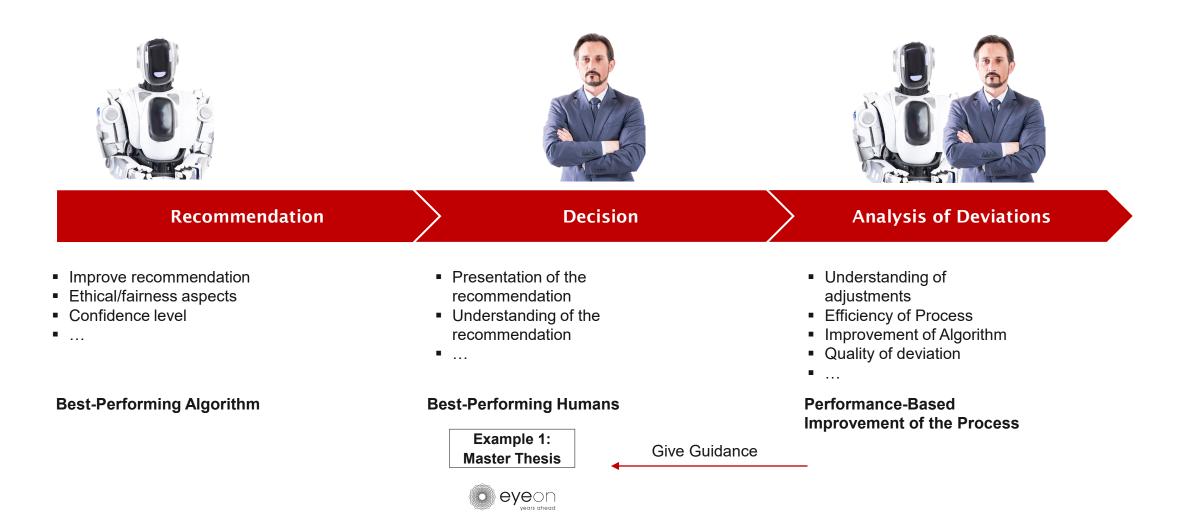
- Human judgment
- Improvisation
- Tacit knowledge

...

HUMAN AND MACHINE INVOLVEMENT IN DECISION MAKING



MACHINE, HUMANS, AND HUMAN-MACHINE INTERACTION



ENHANCE FORECASTERS DECISION-MAKING





Experiment

Training task 2

You will now receive some historical information about the item you are going to forecast. Remember the base price is €5.00 and the average temperature 20°C, which are both used for the statistical forecast.

Product ID Ite		m classif	ication					
Salted Car	ramel Y							
Historical information								
Period	Forecast	Price	Temperature	Demand				
1	78	€5,50	16°C	43				
2	55	€4,70	25°C	86				
3	50	€6,00	26°C	26				

Below you see the specific information about the period you are going to forecast.

Keep in mind that the statistical forecast also has a certain error.

You are expected to review the forecast of period 6 and decide if an adjustment is necesarry.

Please give the final forecast in the text box below.

Forecast						
Period	Forecast	Price	Temperature	Demand		
6	63	€5,40	19°C	?		

Guidance

Informative Guidance

!Extra information! You adjusted the forecast to 126. On earlier situations with similar conditions the system analyzed an increase of 65% on the forecast based on the given information. Please provide your final forecast below.

Suggestive Guidance

!Extra Information!

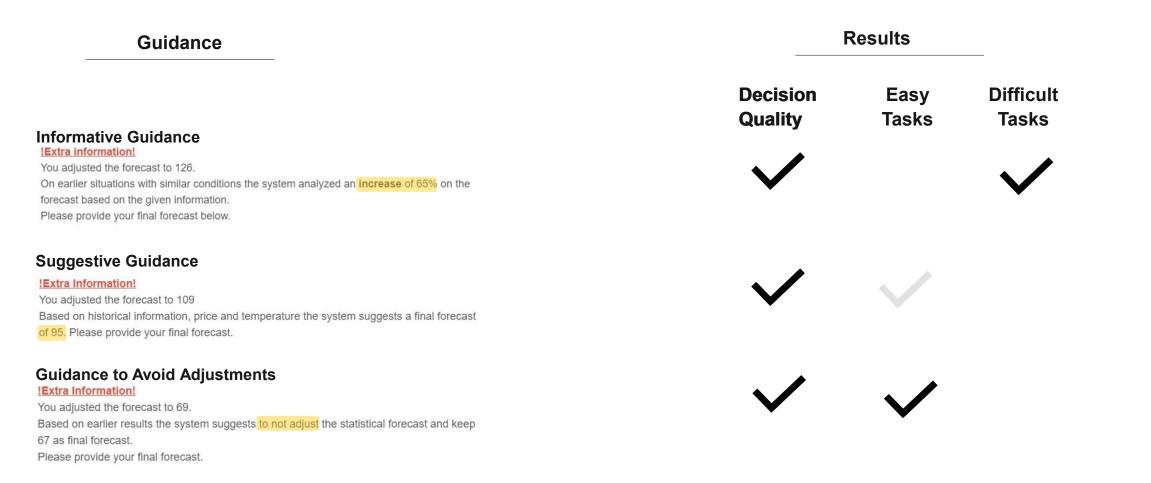
You adjusted the forecast to 109 Based on historical information, price and temperature the system suggests a final forecast of 95. Please provide your final forecast.

Guidance to Avoid Adjustments <u>!Extra Information!</u>

You adjusted the forecast to 69. Based on earlier results the system suggests to not adjust the statistical forecast and keep 67 as final forecast. Please provide your final forecast.

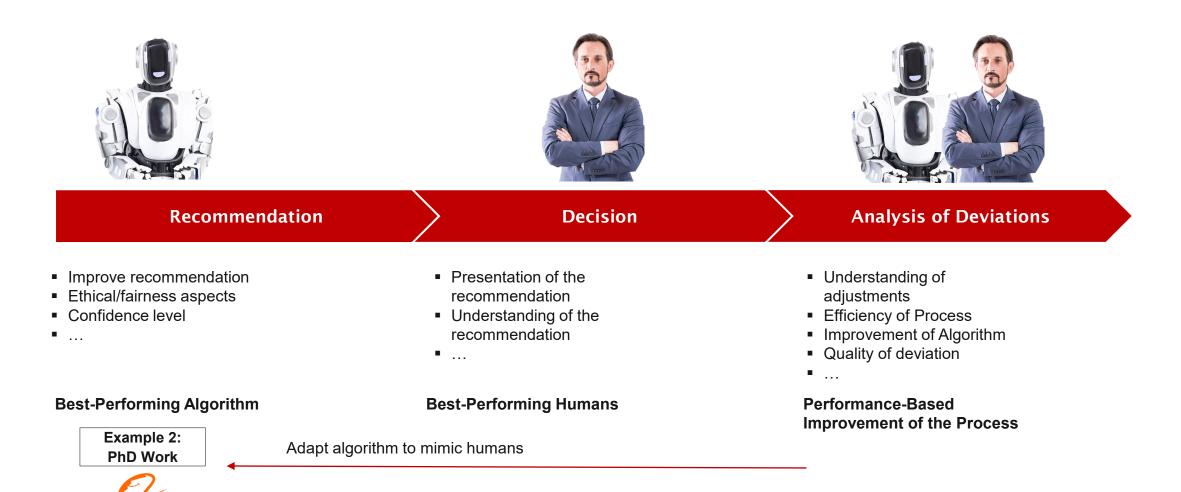
ENHANCE FORECASTERS DECISION-MAKING





MACHINE, HUMANS, AND HUMAN-MACHINE INTERACTION

Alibaba.com



PREDICTING HUMAN DISCRETION TO ADVICE



Alibaba.com

Example 2:



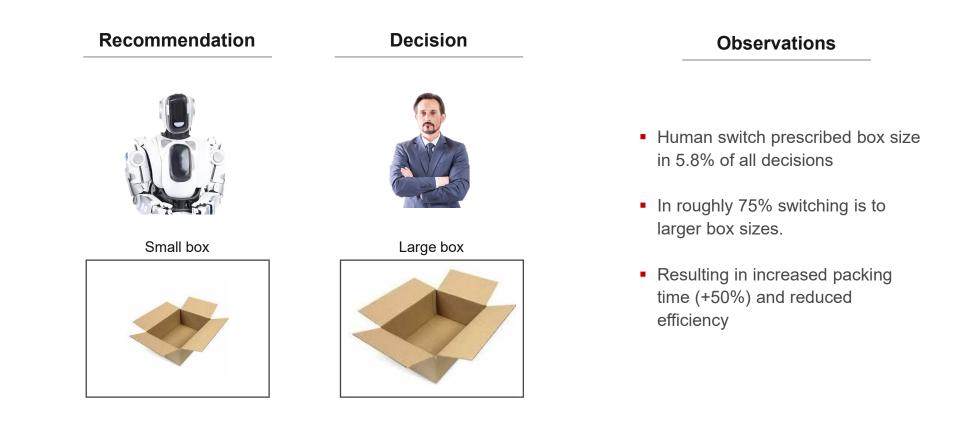
Based on: Sun, J., Zhang, D. J., Hu, H., & Van Mieghem, J. A. (2022). Predicting human discretion to adjust algorithmic prescription: A large-scale field experiment in warehouse operations. *Management Science*, *68*(2), 846-865.

PREDICTING HUMAN DISCRETION TO ADVICE



Example 2:





Based on: Sun, J., Zhang, D. J., Hu, H., & Van Mieghem, J. A. (2022). Predicting human discretion to adjust algorithmic prescription: A large-scale field experiment in warehouse operations. *Management Science*, 68(2), 846-865.

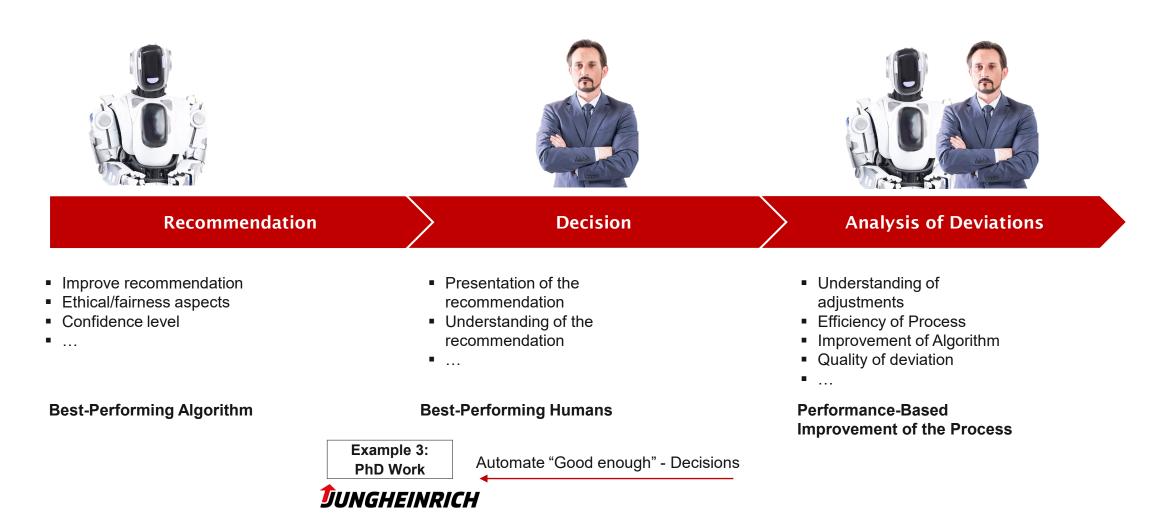
PREDICTING HUMAN DISCRETION TO ADVICE



Example 2:



MACHINE, HUMANS, AND HUMAN-MACHINE INTERACTION

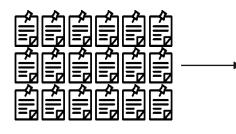


WHICH DECISION SHOULD BE AUTOMATED?





Recommendation







Adjustments

Adjustment Decision (3.5%)

Motivation / Setting

- 1. In many settings, decision makers rarely adjust system recommendations
- 2. Decision makers may still be required to manually review every recommendation
- 3. Some adjustments *worsen* performance outcomes

Performance



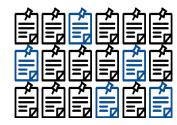
Decision Evaluation (40%)

TARGETED AUTOMATION IMPACTS WORKLOAD AND PERFORMANCE

Adjustment

Await planner's decision

Automate



Performance

Await planner's decision

Automate



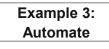
Objectives

Can automation reduce the workload without impacting planner's decision?

Objectives

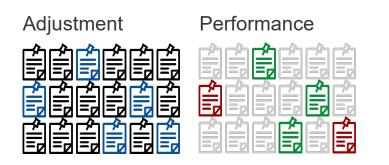
Can automation improve the planner's decision?





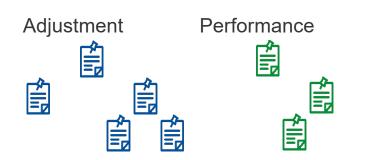


Step 1: Prediction



 Identify orders for which an adjustment is <u>likely</u> to add value

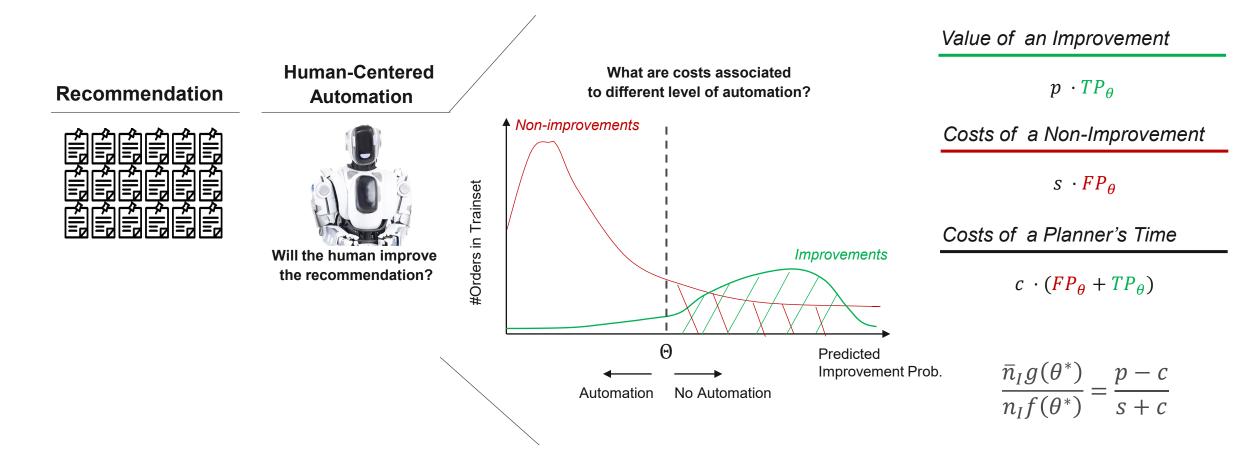
Step 2: Automation / Optimization



 Automate the orders, for which an adjustment is <u>not likely</u> to add value

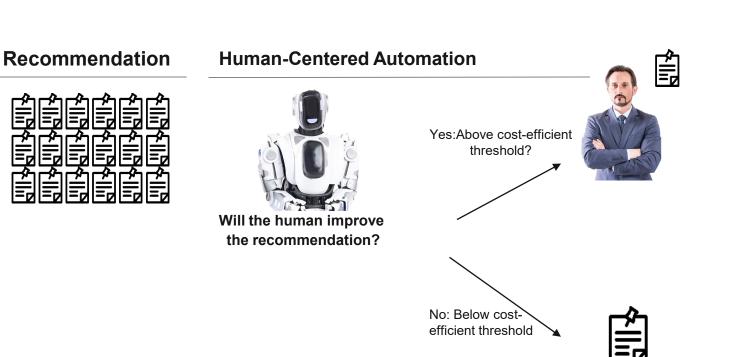
WHICH DECISION SHOULD BE AUTOMATED?





WHICH DECISION SHOULD BE AUTOMATED?







 More than 70% of the orders are automated

Of non-automated orders:

- 5.9% are adjusted
- 47% are improvements

SUMMARY

Human-AI interaction can be used to improve:



The human's adjustments (e.g. by personalized feedback).

2 The Al's performance (e.g. by augmenting with human knowledge).



³ The whole process (e.g. by automating).

Data on human decisions and interactions with the AI are rich sources for improvement

TIME FOR Q&A

Feel free to contact me for discussions/ideas:

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