

Example Projects



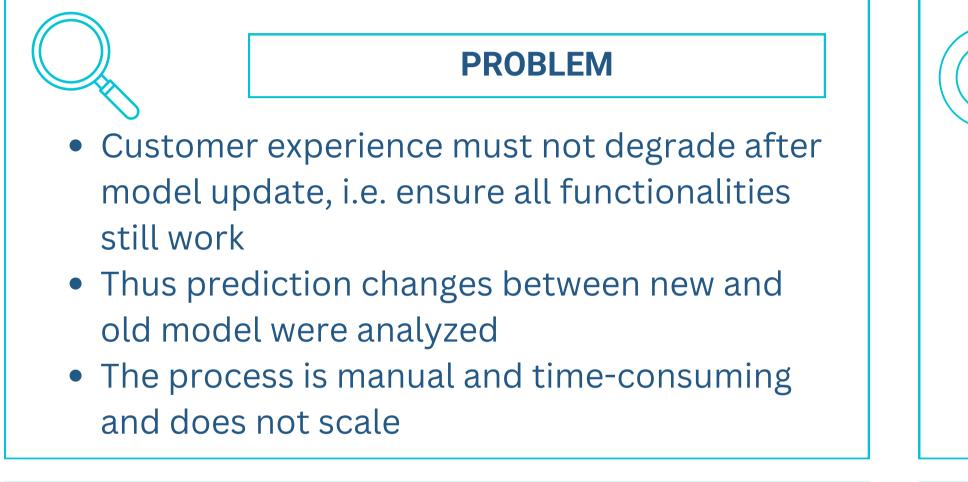
The **paper icon** in the Outcome box means a paper was published about the project, click on the icon to read the paper for more details.







Distillation techniques for model stability







- flips)







SOLUTION

• Bias the new model towards predictions of the old on training samples where the old model gave correct predictions (reduce negative

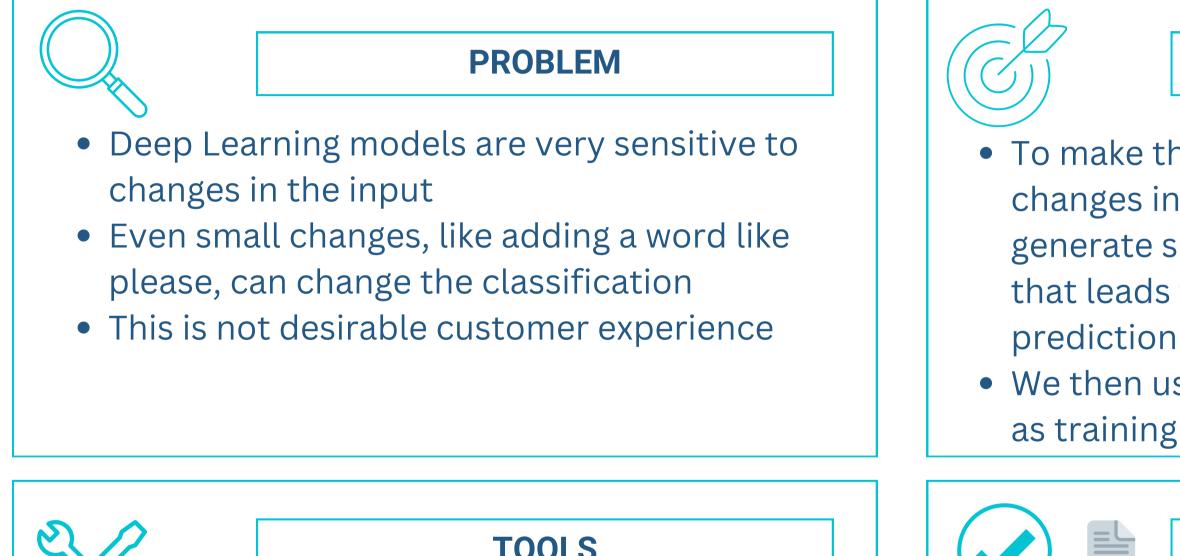
• this reduces amount of prediction changes to be analyzed

• Done through change in the loss function

OUTCOME

• reduced prediction changes by up to 55 % • approach to be rolled out in production • estimated to save 25 % of time in manual analysis

Model robustness and adversarial attacks









SOLUTION

• To make the model more robust towards small changes in the input, we trained a T5 model to generate so called adversarial attacks, i.e. data that leads the target model towards a wrong

• We then used that generated adversarial data as training data for the production model

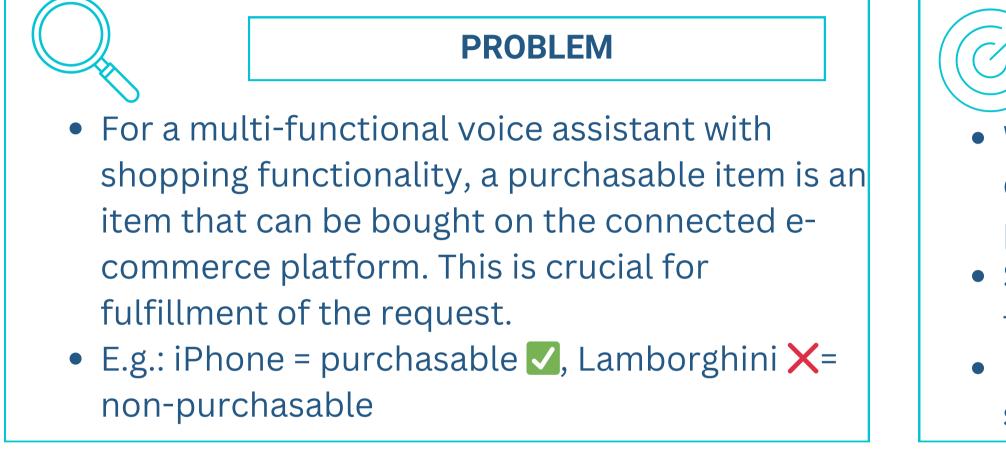


OUTCOME

• 70 % error rate reduction on adversarial test data

 reduction of customer perceived defect in online A/B test, rolled out in production

Purchasable item recognition









SOLUTION

• We designed a classifier run on top of the existing system to detect whether an item is purchasable (signal goes back into system) • Semantic search is used to match the item in the request with data from the product catalog • A twin network decides if item in request is similar enough to data from product catalog

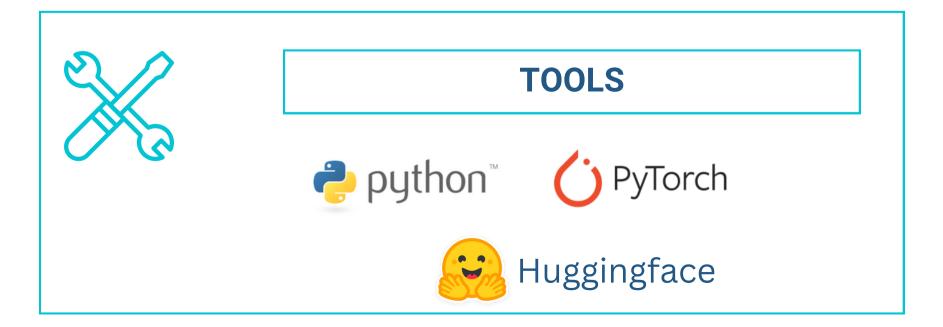
OUTCOME

• Classifier is able to correct the mistakes of existing system, increase in accuracy of 80 % • Simplified approach rolled out in production led to reduction of online defects

Semi-supervised Learning

PROBLEM

- Data annotation by humans is slow and costly.
- Usually just a random sample of data is annotated, regardless of whether the model is able to correctly interpret each instance in the sample (assuming we work with a system) that is regularly updated with new data)









SOLUTION

• To speed up the annotation process, we show the annotator the model interpretation and let them verify or if needed correct it.

• As such, data is only annotated if the sample is incorrectly interpreted by the model

• Correct predictions are directly ingested as training data

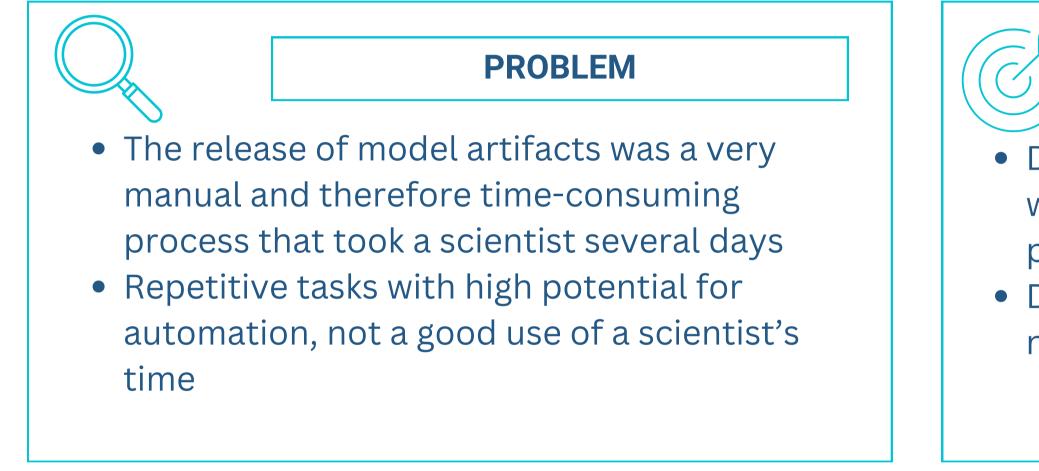
| _ | | | |
|---|---|---|--|
| | L | | |
| | | - | |
| | | | |
| - | - | | |

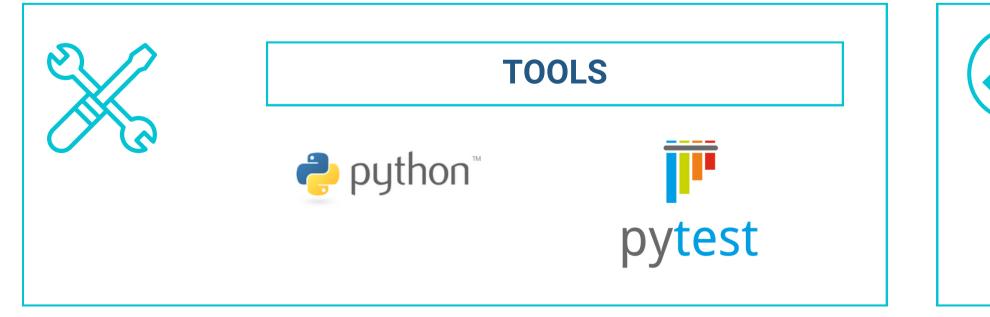
OUTCOME

 Annotation volume is reduced by 97 % and cost by 60 %

• Model performance increases due to reduced annotation inconsistencies

Tool for release of model artifacts







SOLUTION

 Delivered a unit-tested software package written in Python that automates the release process

• Designed and gave training for users of the new package on how to use the software

OUTCOME

Manual work reduced from days to hours
Due to reduced complexity, release could be transferred from scientist to role with less technical expertise





Let's talk.

