# A live model for product type mapping

Michael Hardegger 27. 1. 2020

Contributions by Elvis Murina, Beate Sick and Helmut Grabner, all from ZHAW

### About Digitec Galaxus AG

Founded as a start-up in 2001, with two of the founders still present

1'14 Mio.

**Revenues 2019** 

32 Years

**Average Age of Employees** 

16%

Growth 2019

> 1300 Employees

Logistics, Category Mangagement, Customer Service, etc.

> 150 Software Engineers

**Focus on Online Shop and ERP** 

> 1.5 Mio.

**Active Customers** 

5 ML Engineers

Team founded in 2019

> 3 Mio Products

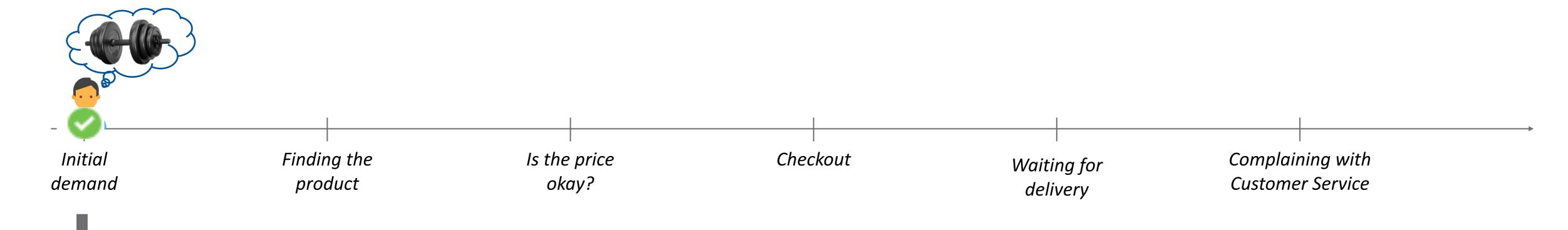
~100'000 on stock



Digitec Galaxus AG

7

# Part 1: ML Use Cases at Digitec Galaxus AG

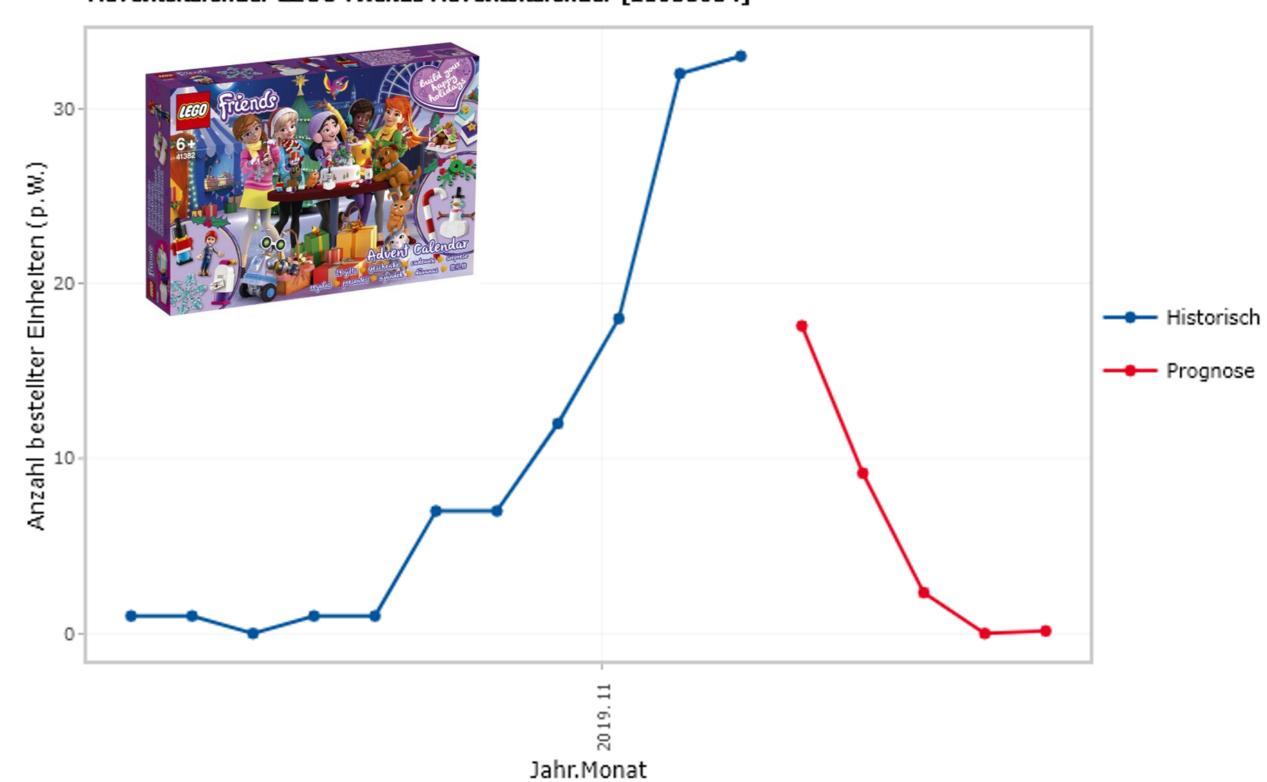


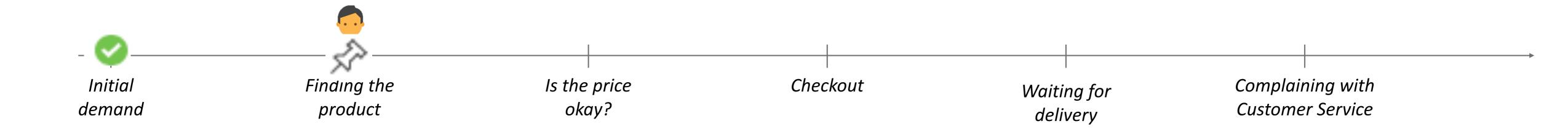
To fulfill this demand, we have to have the products on stock!

80% of the articles we have on stock were bought based on demand forecast models!

Method: Time-Series-Models + ML-Stacker

#### Adventskalender LEGO Friends Adventskalender [11038834]





#### Oft zusammen gekauft mit



175.83 Bowflex Selecttech



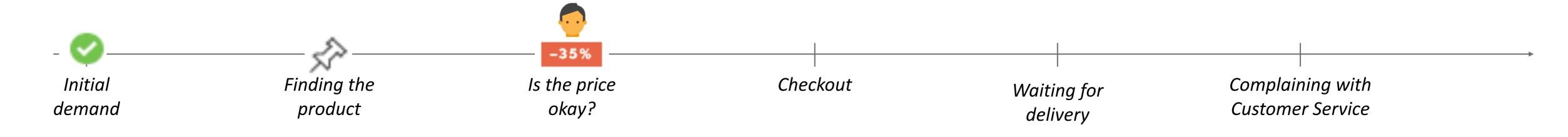
**225.05** statt 280.– **Bowflex** 4.1 Bench



21.26 Tunturi Fitnessmatte (15mm) Recommenders speed up navigation and make customers aware of unconscious demands.

#### Models in productions:

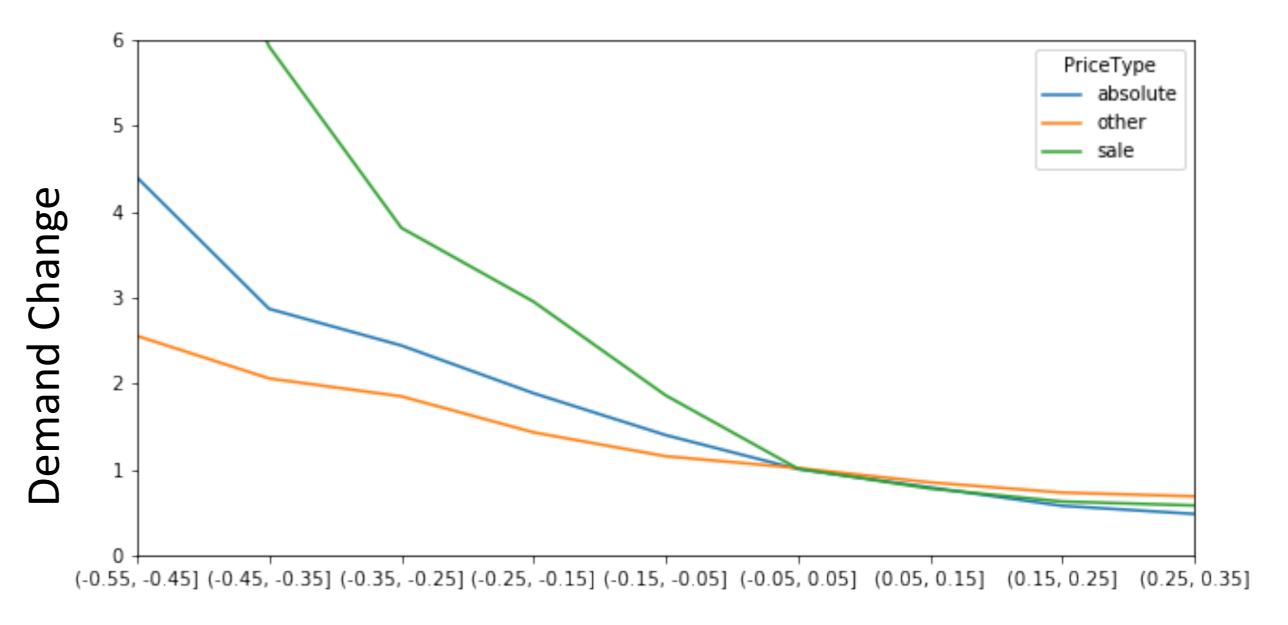
- Meta Prod2Vec
- Affinity Profiles
- Many heuristic models



-35%

187.69 statt 289.-

ML-based price elasticity models tell us by how much we need to reduce a product to generate sales.

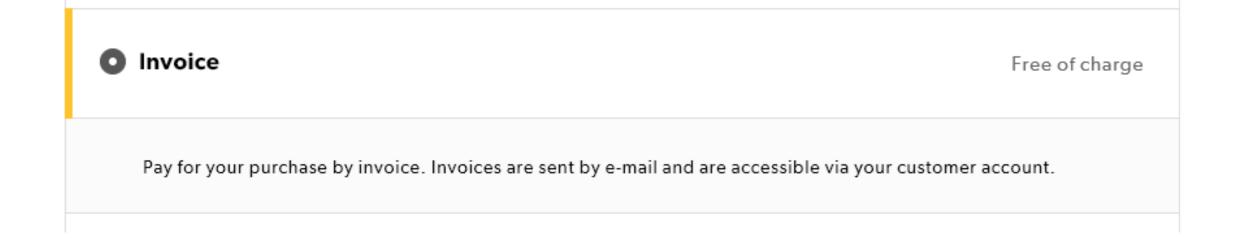


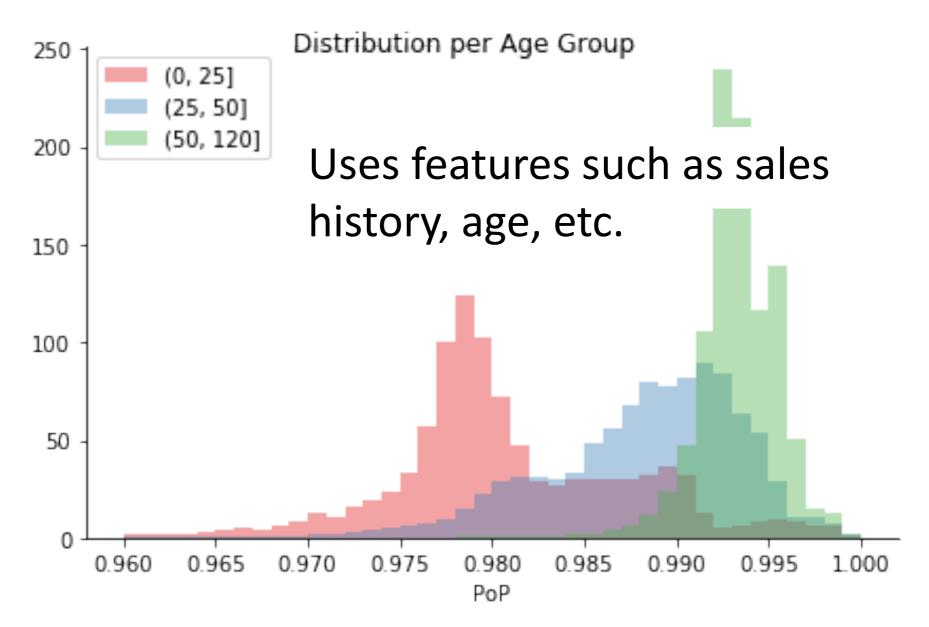
Price Change



ML-based credit rating

«Probability of Payment» learned from previous orders and defaults



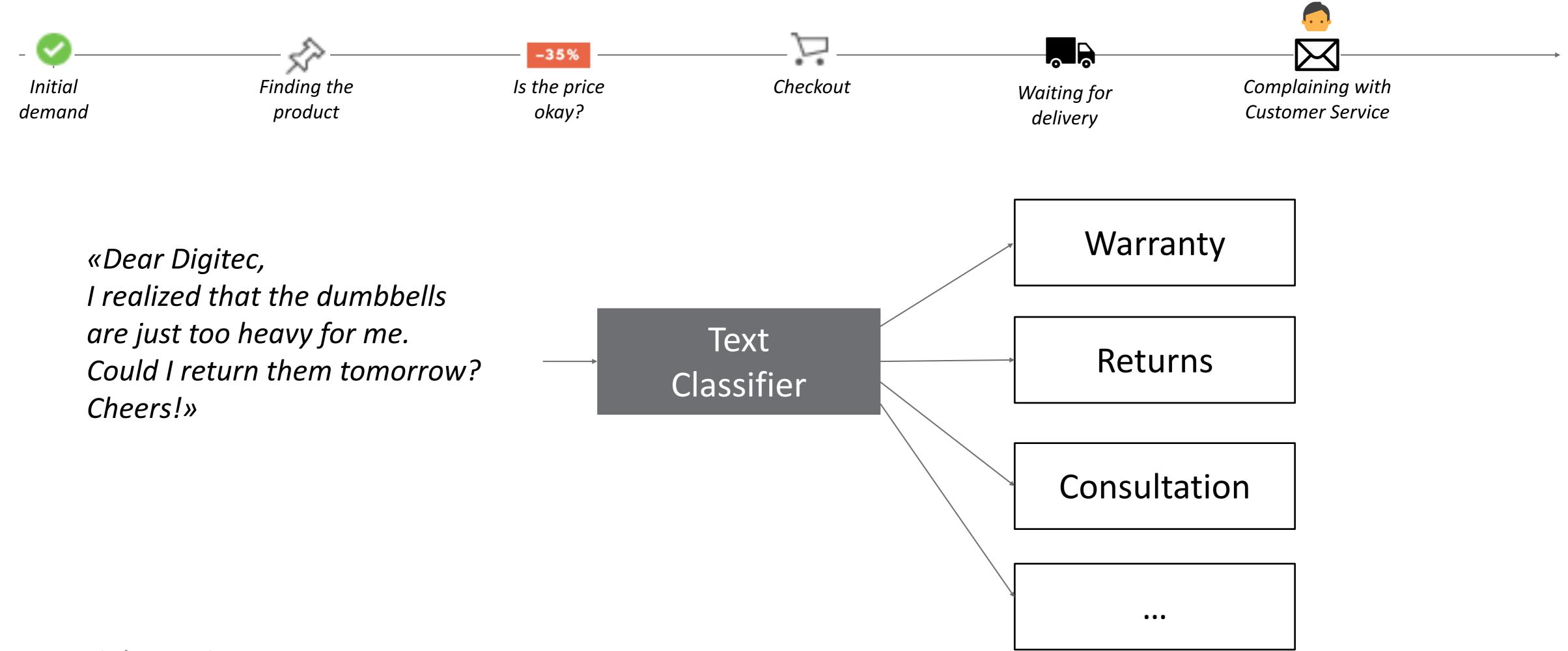




#### Some logistics use cases:

- Predicting the delivery date
- Process Mining
- Placement of articles in the warehouse
- -



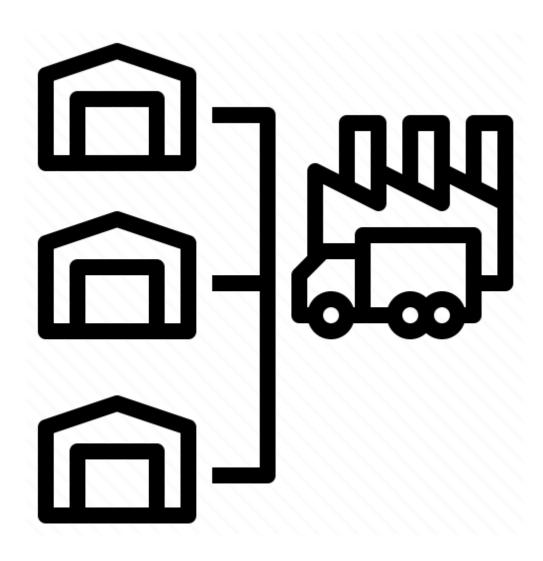


# Part 2: Product Type Mapping

# Where do we have our product data from?

Product data quality is super important for us ...

... but we are not really in control of it



- >100 suppliers send us data about new products
- Data formats and quality vary widely
  - Some suppliers don't care enough about us to improve data quality
  - Others are not able to do it
- As a result, dozens of employees clean up data manually
- Even that is just barely enough to handle bestsellers, the long tail of products is never touched

#### Data Cleaning Process

#### Provider Record

Image:



Name:

Flux S Smart T2900

Brand:

Takx

Category:

...;Sport;Radsport;Zubehör

#### Product Description:

«Der Flux S Smart hat alles, was Sie sich von einem Smart-Trainer mit ...»

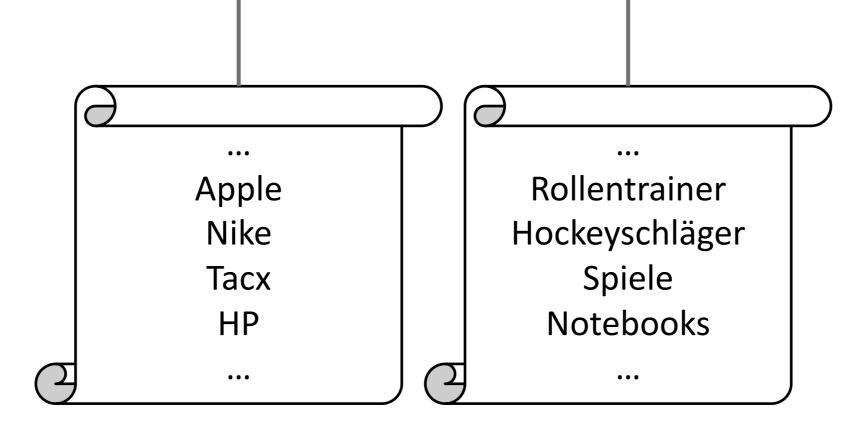
#### Product Properties

{Weight: 21 kg, Connectivity: [BT, ANT, ANT+, ...], Max. Resistance: 1.5kW, ...}

What arrives at our interfaces ...

... and how it ends up in the shop

Brand/ProductType/Property
Mapping Process



#### ProductType

Rollentrainer









Der Flux S Smart hat alles, was Sie sich von einem Smart-Trainer mit Direktantrieb wünschen. Zuverlässig, genau, geräuscharm, leistungsstark, und er vermittelt ein gutes Fahrgefühl...

#### Die wichtigsten Spezifikationen auf einen Blick

Max. Widerstand	1500 W
Gewicht	21 kg
Rollenbremsentyp	Elektromagnetischer Widerstand
Konnektivität	Bluetooth Smart, ANT+, ANT+ FE-C

↓ Mehr anzeigen

Property & Value

#### Data Cleaning Process

#### Provider Record

Image:



Name:

Flux S Smart T2900

Brand:

Takx

Category:

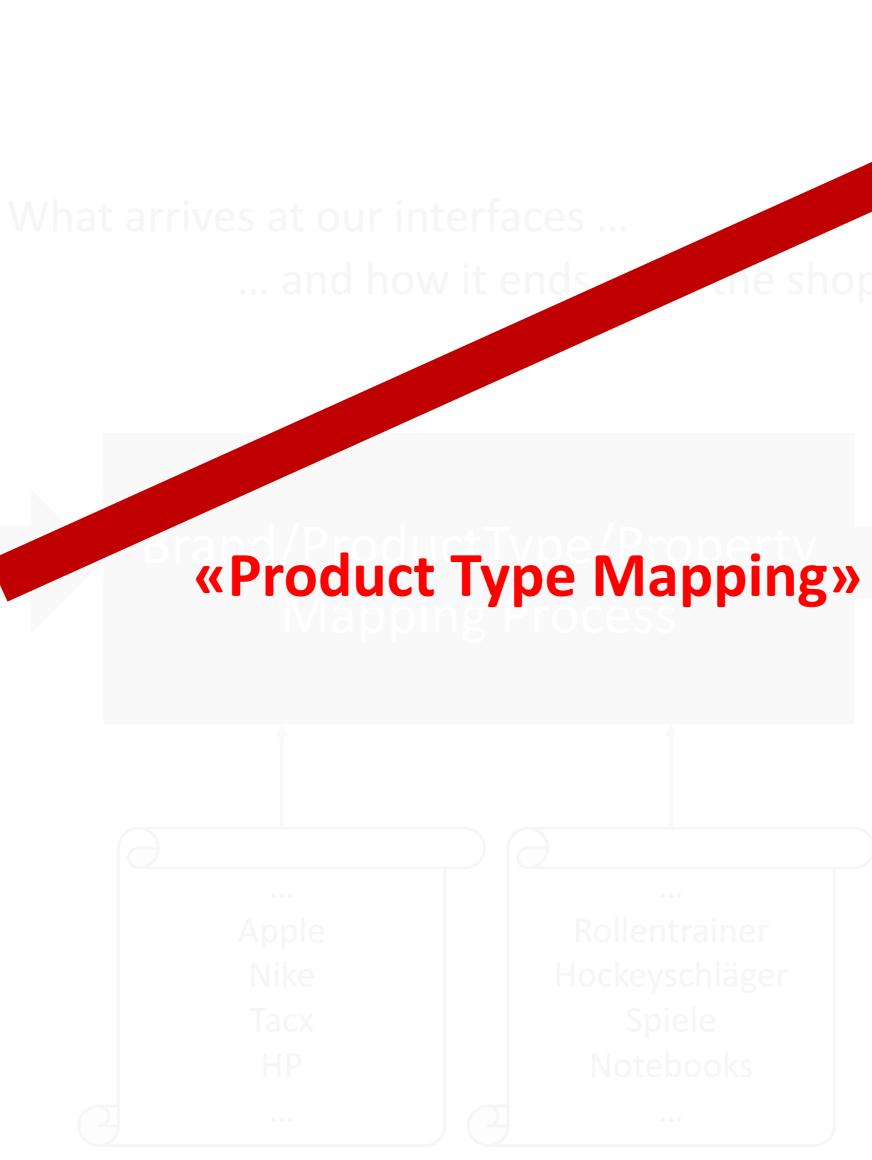
...;Sport;Radsport;Zubehör

#### Product Description:

«Der Flux S Smart hat alles, was Sie sich von einem Smart-Trainer mit ...»

#### Product Properties

{Weight: 21 kg, Connectivity: [BT, ANT, ANT+, ...], Max. Resistance: 1.5kW, ...}



#### ProductType

Rollentrainer





623.-Tacx Flux S Smart T2900





Der Flux S Smart hat alles, was Sie sich von einem Smart-Trainer mit Direktantrieb wünschen. Zuverlässig, genau, geräuscharm, leistungsstark, und er vermittelt ein gutes Fahrgefühl...

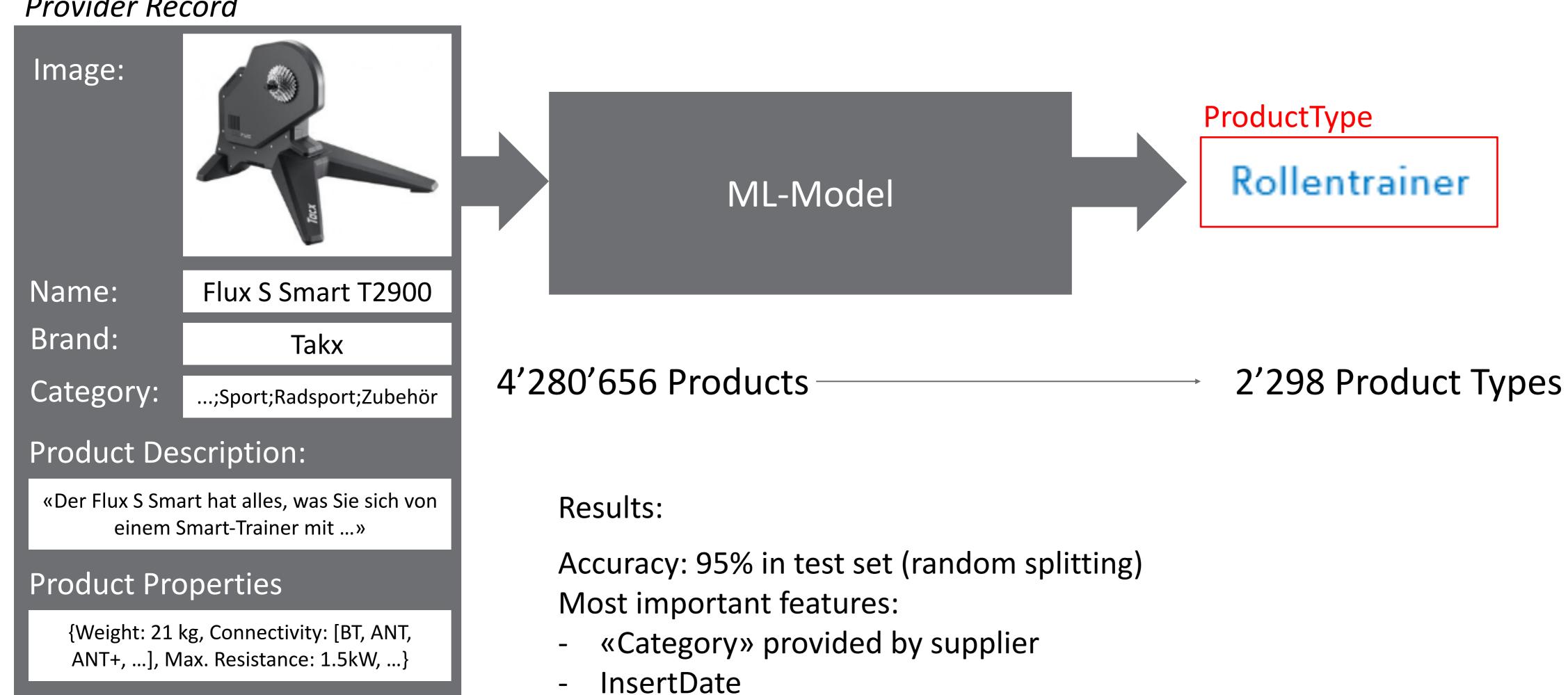
#### Die wichtigsten Spezifikationen auf einen Blick

Max. Widerstand	1500 W
Gewicht	21 kg
Rollenbremsentyp	Elektromagnetischer Widerstand
Konnektivität	Bluetooth Smart, ANT+, ANT+ FE-C

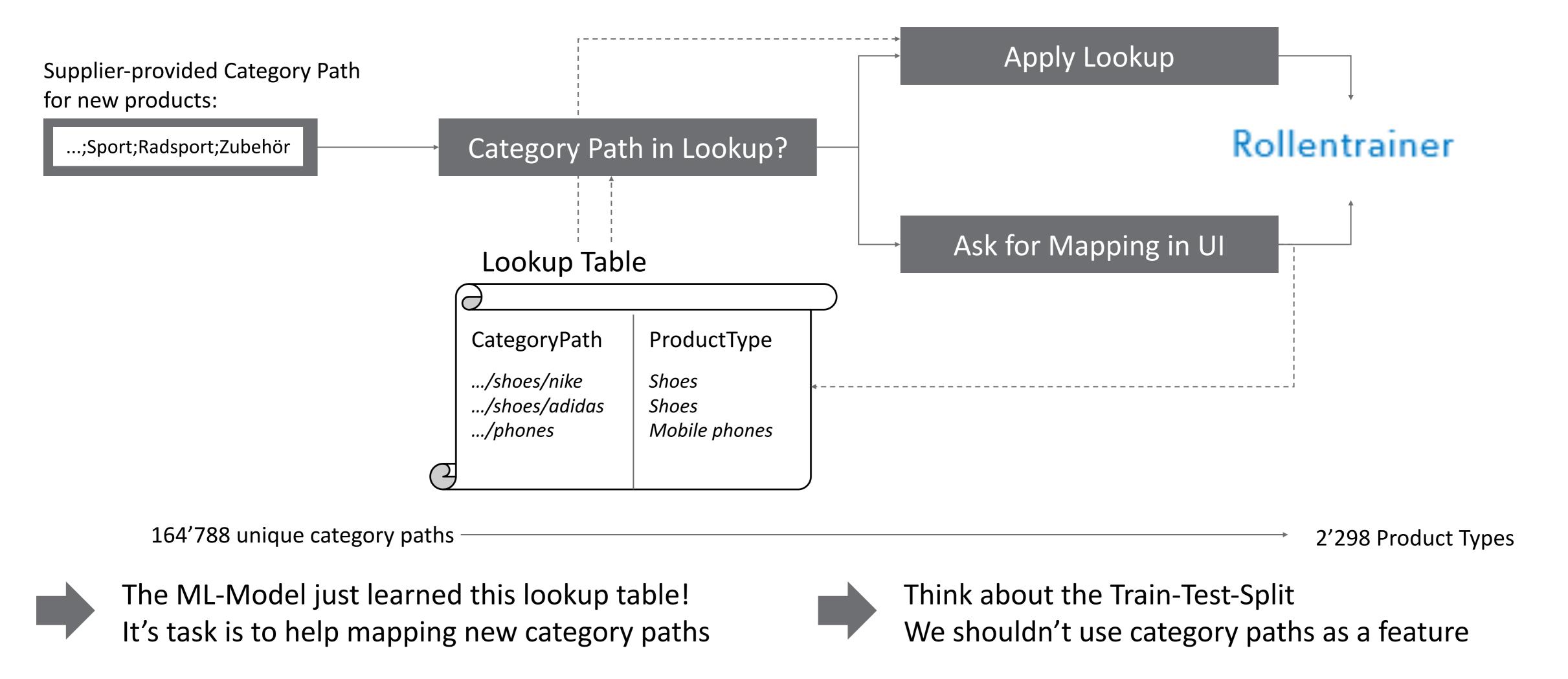
↓ Mehr anzeigen

# Naive Approach to PT-Mapping

#### Provider Record



# Current Business Process explains the Result



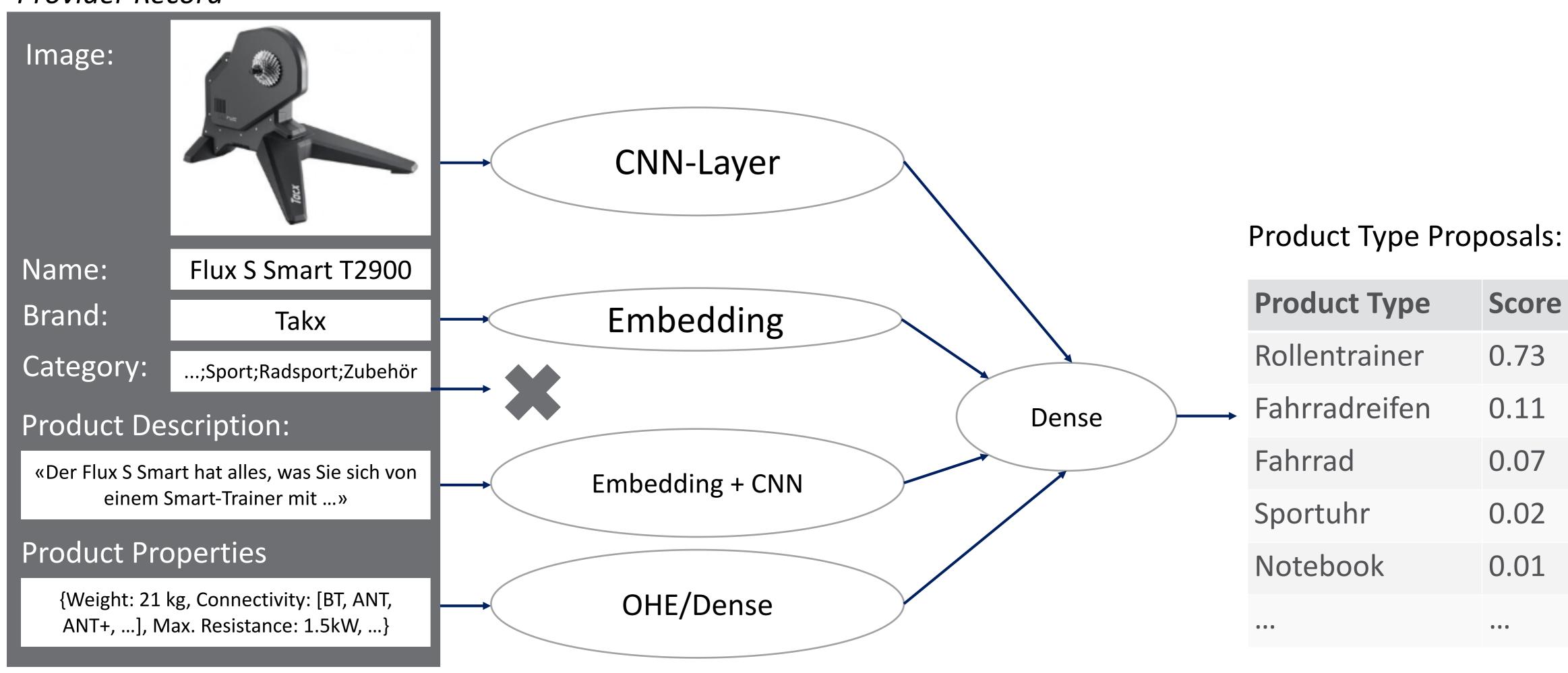
# Challenges with the PT-Mapping Problem

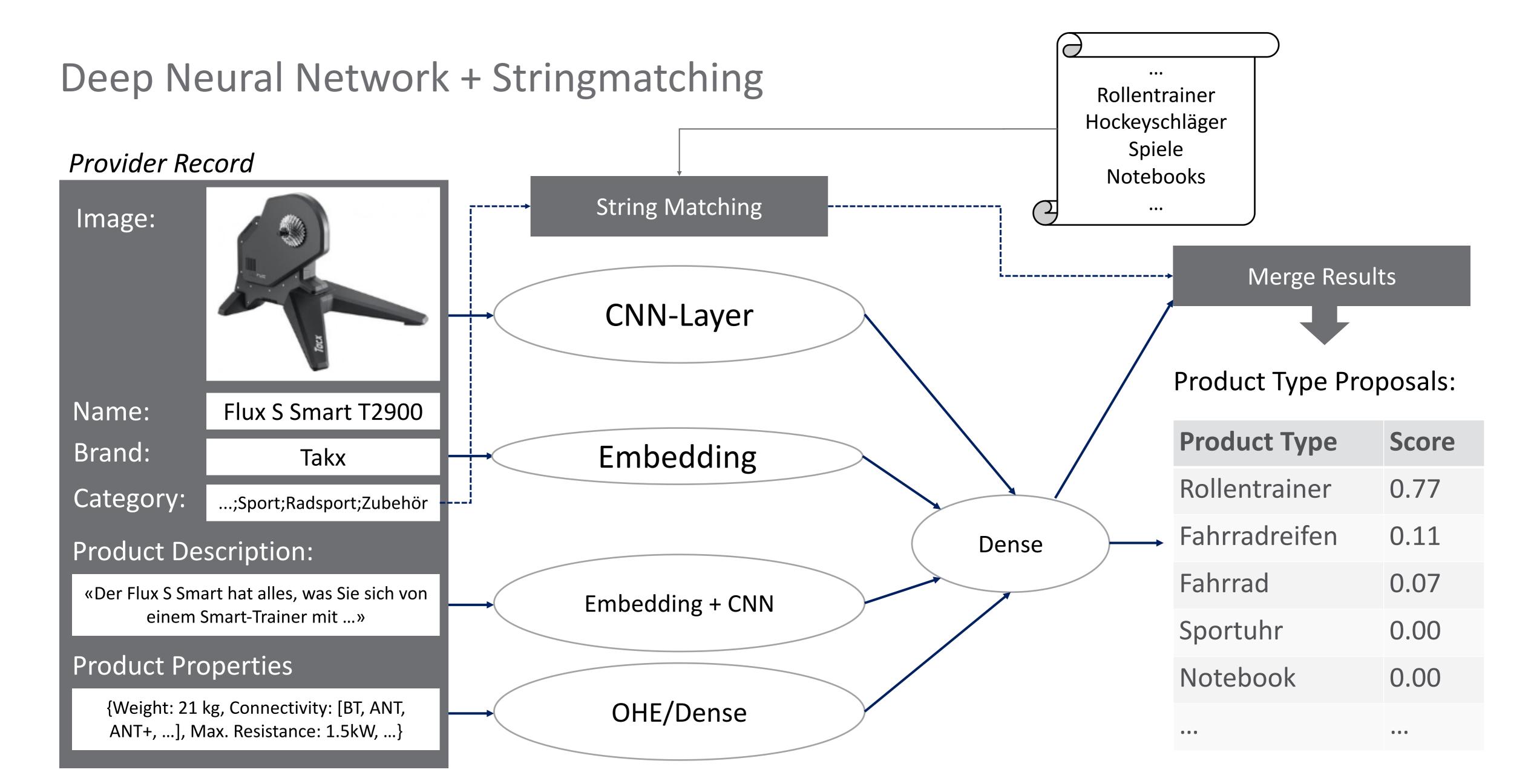
#### This is a really «dirty» problem!

- Many incorrect labels
- Very imbalanced classes (and features)
- Lots of NAs
- Mappings change over time (new PTs, PTs are combined, etc.)
- Evaluation / Train-Test-split by date of mapping
  - > Test distribution can be very different from training
  - > However, what we care about is real-life performance

# Deep Neural Network for Product Type Mapping

#### Provider Record





# Controller for Assigning Product Type Mappings: High Confidence

#### CategoryPath provided by supplier

#### Proposed Product Type

KategoriePfad	Akzeptieren	Zuweisen	Ignorieren	Kategorien
Ausrüstung;Pflege & Hygiene;Schuhpflege;Schuhpflege;Schuhpflege	•			<u>Schuhpflege</u>
n;Baby;Babypflege & Bad;Babybadewannen & Zubehör;Badespielzeug	•			Badewannenspielzeug
pielwaren;Baby;Schlafen & Nuggi;Spieluhren & Nachtlichter;Spieluhren	•			<u>Spieluhr</u>
Baby & Spielwaren;Baby;Spielsachen;Babyspielzeug;Holzspielzeug	•			Bauklötze + Stapeln
aren;Kinderfahrzeuge;Laufrad & Kinderfahrrad;Laufrad & Kinderfahrrad	•			Kindervelo
by & Spielwaren;Spiele & Puzzle;Spiele;Zubehör Spiele;Zubehör Spiele	•			Sammelaufbewahrung
aren;Spieltische & Spielwelten;Multimedia-Spielwaren;Roboter;Roboter	•			Robotik Kit
Bekleidung;Hosen;Ski- & Snowboardhosen	•			<u>Skihose</u>
ng;Gewerbe- und Industrieleuchten;Bürostehleuchten;Bürostehleuchten	•			<u>Stehlampe</u>
telligente Beleuchtung;Intelligente Beleuchtung;Intelligente Beleuchtung	•			<u>Leuchtmittel</u>

# Controller for Assigning Product Type Mappings: Lower Confidence – Multiple Choice

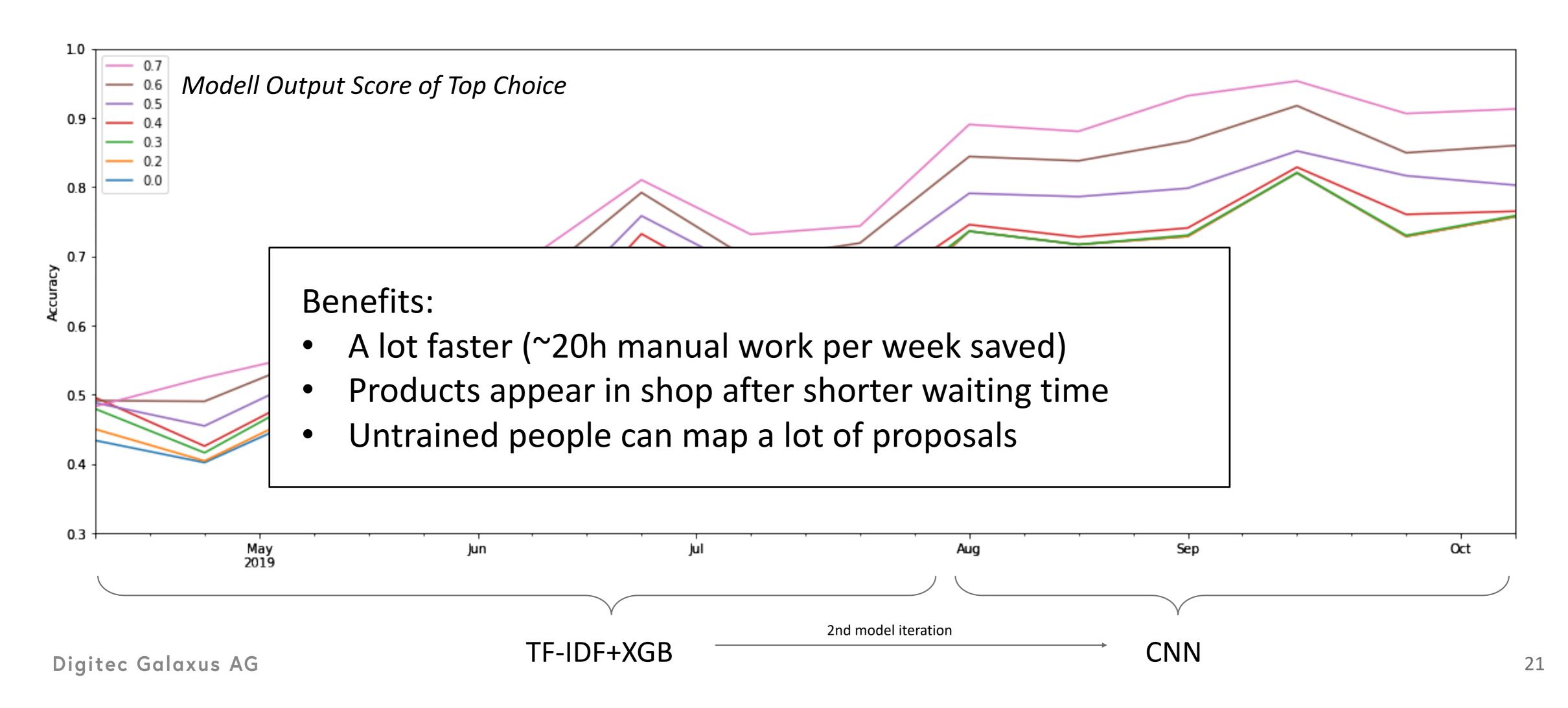
#### CategoryPath provided by supplier

#### Proposed Product Type

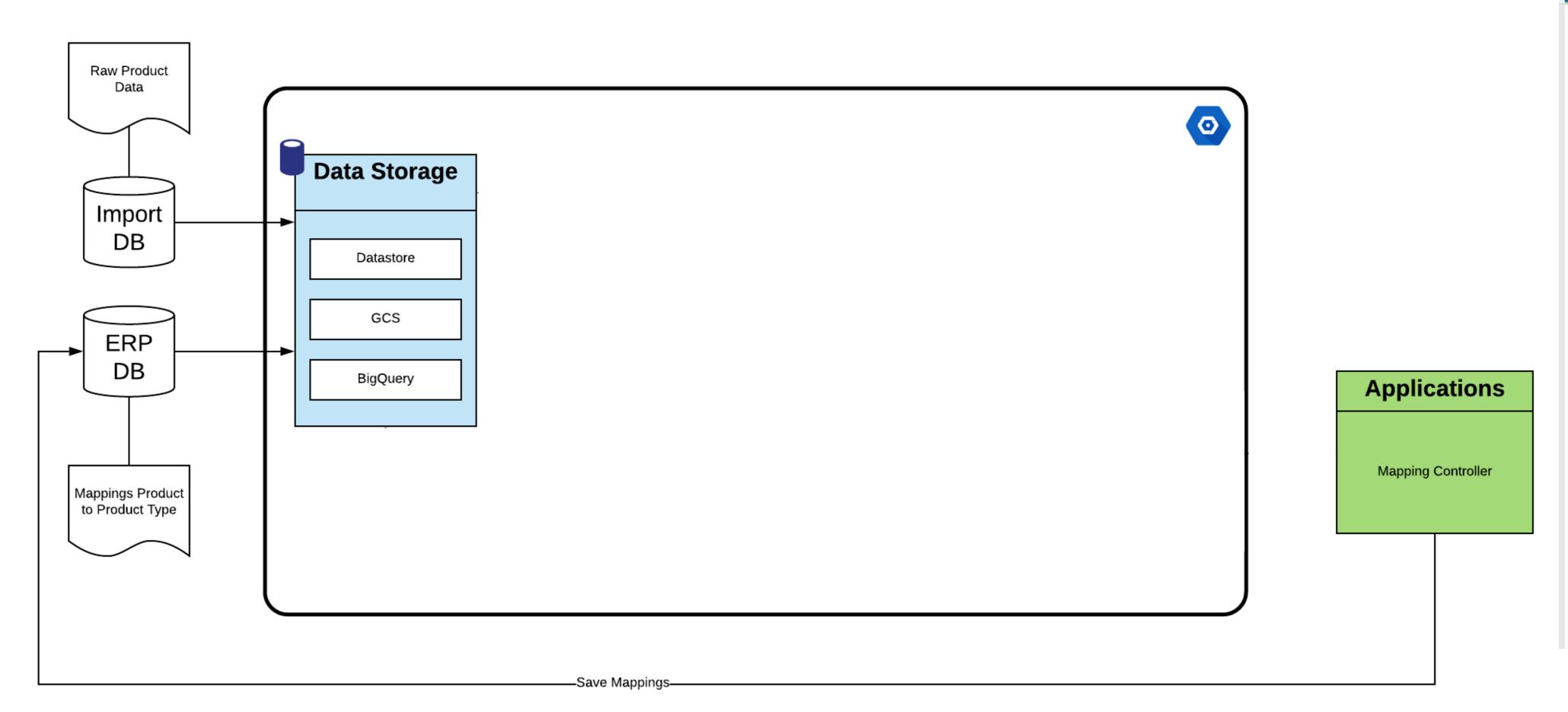
KategoriePfad	Akzeptieren	Zuweisen	Ignorieren	Kategorien
Baby & Spielwaren;Baby;Unterwegs;Kinderwagen;Buggys	•			Kinderwagen ▼
ung;Beleuchtungszubehör;Beleuchtungszubehör;Beleuchtungszubehör	•			Einbauleuchte + LED-P ▼
htung;Gewerbe- und Industrieleuchten;Einbauleuchten;Einbauleuchten	•			Einbauleuchte + LED-P ▼
CE;Audio;Radio & Streaming;Radio-Zubehör;Radio-Zubehör	•			Audio Zubehör ▼
CE;Foto & Videografie;Fotostudiobedarf;Lichtformer;Lichtformer	•			Softbox + Reflektor ▼
rafie;Speichermedien;Speichermedienzubehör;Speichermedienzubehör	•			Bitte wählen Softbox + Reflektor
CE;Foto & Videografie;Videokameras;Professional;Recorder	•			Blitz Zubehör
Videografie;Videokameras;Videokamerazubehör;Videokamerazubehör	•			Hintergrundsystem
CE;Kabel & Adapter;Adapter;Video-Adapter;Video-Adapter	•			Data + Video Adapter ▼
CE;Kabel & Adapter;Audio-Adapter	•			Bitte wählen ▼

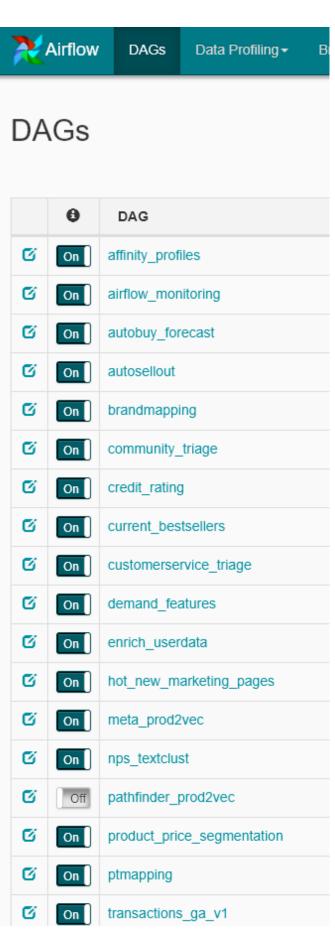
#### Live Performance

#### Top 5-Accuracy: Toppe per for three fill we are puser of page smalls propring gused as mapping



#### An overview of our ML module





# Conclusions

### Take home messages

- Online retailers are a great playing field for data scientists (if you have management support and a good infrastructure)
- > Expect to work on really messy data sets
- Take your time to understand the business processes well in order to find out how ML can bring the most benefit
- ➤ By bringing solutions into production quickly, you learn fast and can iteratively improve

# Appendix

# A few findings from the modelling efforts

Performances in Test-Set (Split by date; without string matching)

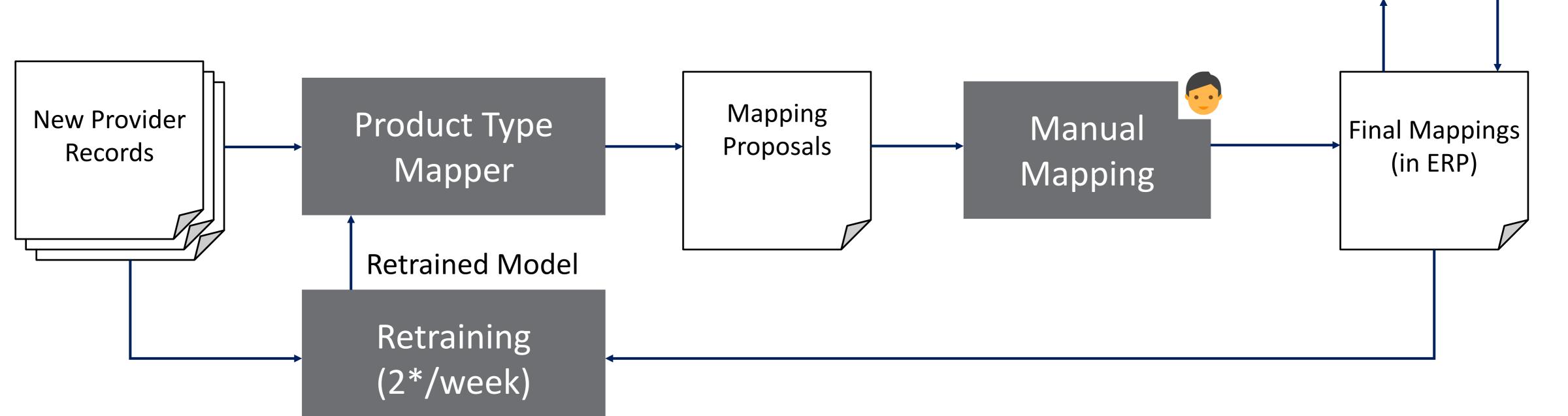
Modell	Top5-Accuracy
Text+Image	73.6%
Text+Img+Meta	72.3%
Text (stemming,)	66.4%
Text (raw)	63.9%
Baseline (TFIdf+XGB)	51.5%
Metadata only	36.3%

- DNNs outperform baseline model
  - ➤ Memory-limitations for XGB (2300 values for each leaf)!
- Image+Text resulted in the best performances
- Other meta data did not improve quality
- Text Model:
  - > LSTM or CNN did not make a difference
  - Preprocessing (stemming, etc.) reduced model complexity and improved performance
- Weekly Retraining required in production:
  - Product hierarchy changes frequently

### Retraining

#### Required because:

- Product hierarchy changes frequently
- Often patterns are very provider-dependent and when adding new providers, performance is poor in the beginning



Manual

Mapping Edits

#### Live Performance

KPIs	Value
#Mappings/Week	~400
Top1-Accuracy	75.2%
Top5-Accuracy	91.3%

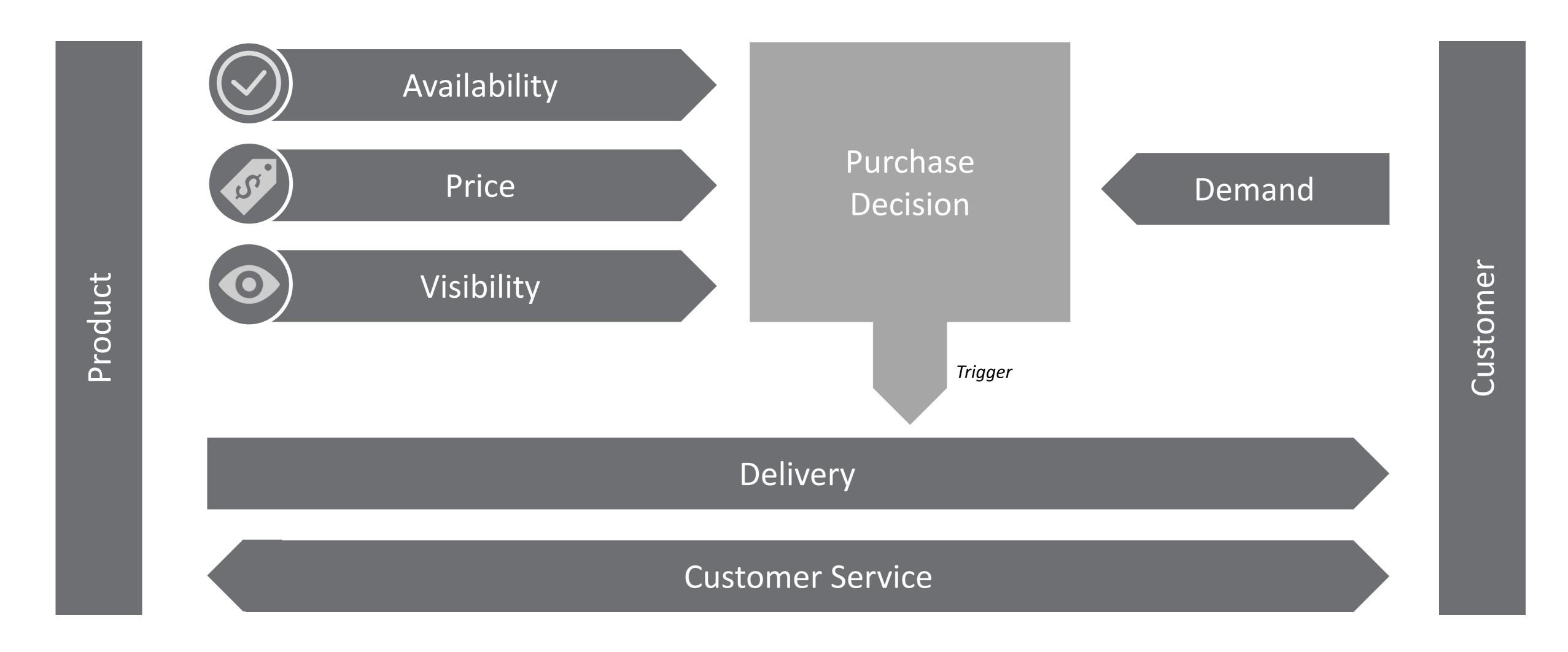
#### Benefits:

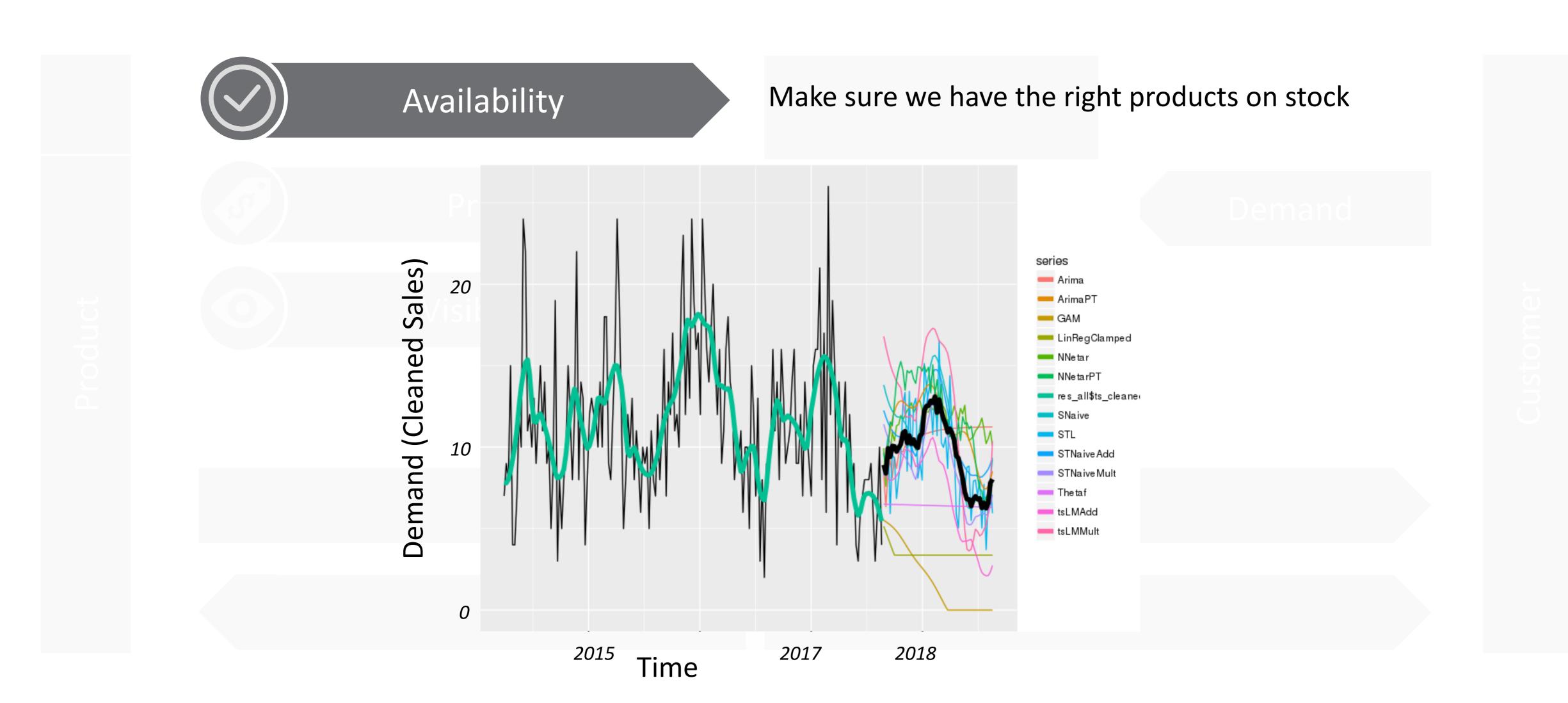
- A lot faster (before, 20-30h per week manual work)
- Products appear in shop after shorter waiting time
- Untrained people can map most proposals
- Full automation for cases with high score possible

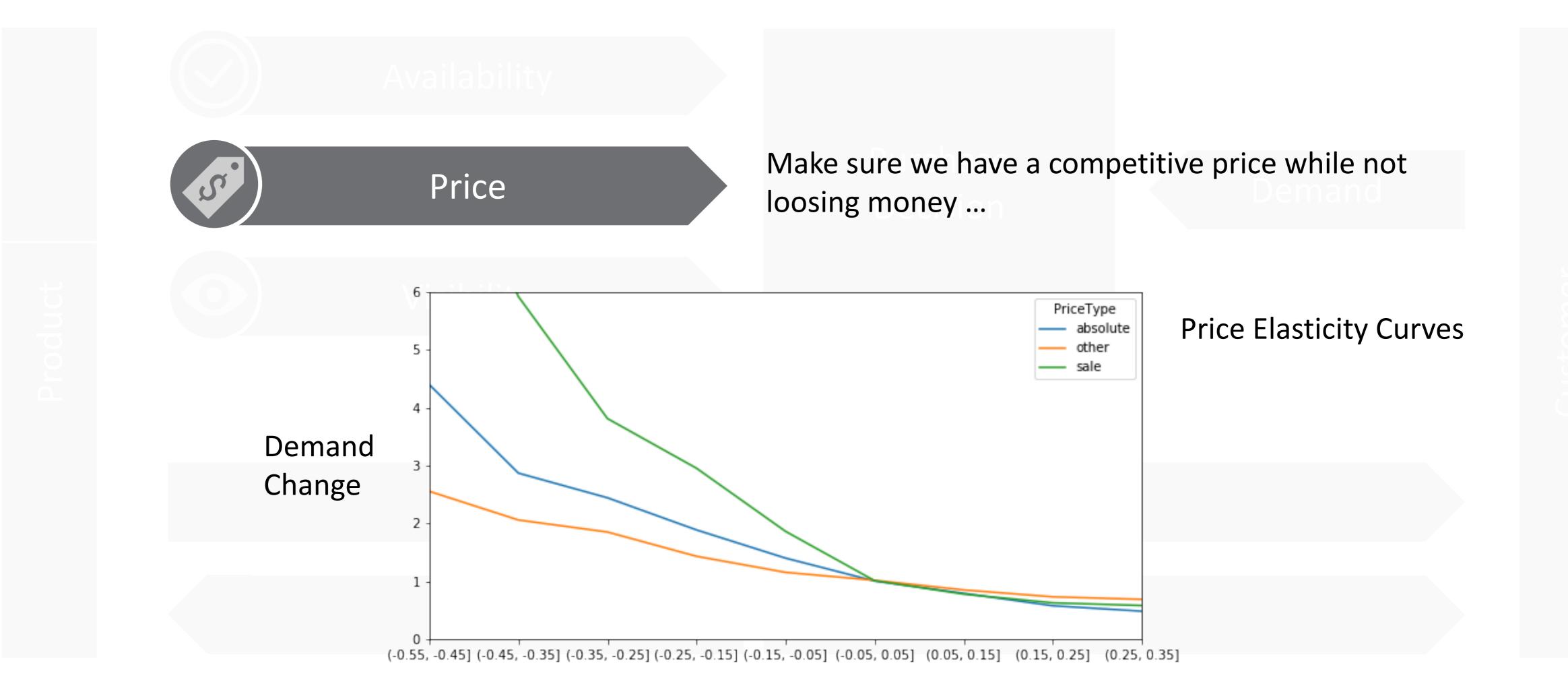
# Text preprocessing

	Step	Preprocessed Text	
1	Raw text	Der Flux S Smart T2900 hat alles, was Sie sich von einem Smart- Trainer mit	
2	Lower case and remove stopwords	flux smart t2900 hat alles smart-trainer	
3	Replace string patterns with tokens	flux smart LETTERNUMBERTOKEN hat alles smart trainer	
4	Stemming	flux smart LETTERNUMBERTOKEN hat all smart train	
5	Tokenizing	171 5231 879 7 66 5231 9	
6	Input to embedding layer (token to vector)		

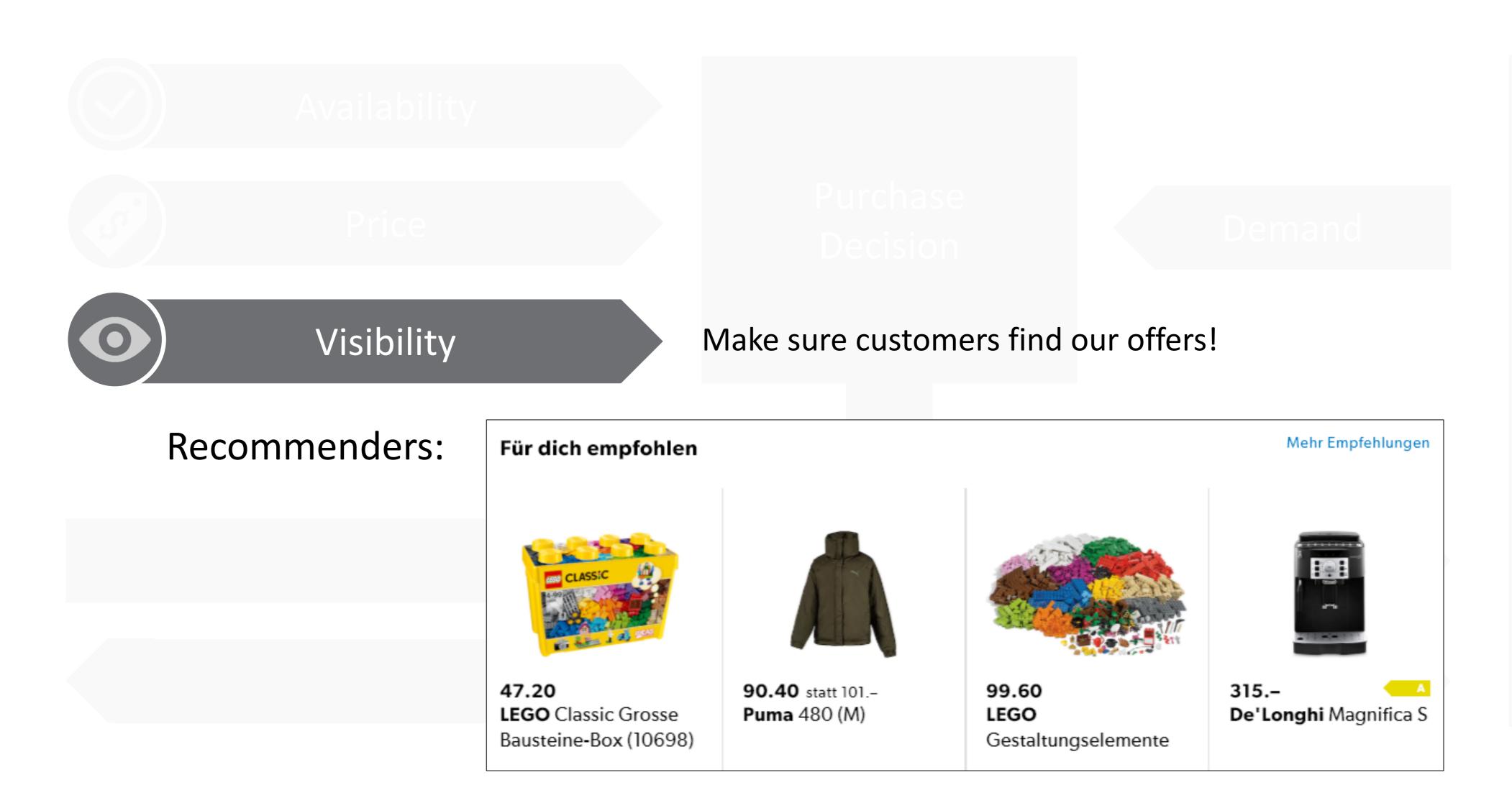
# The core processes of an online shop



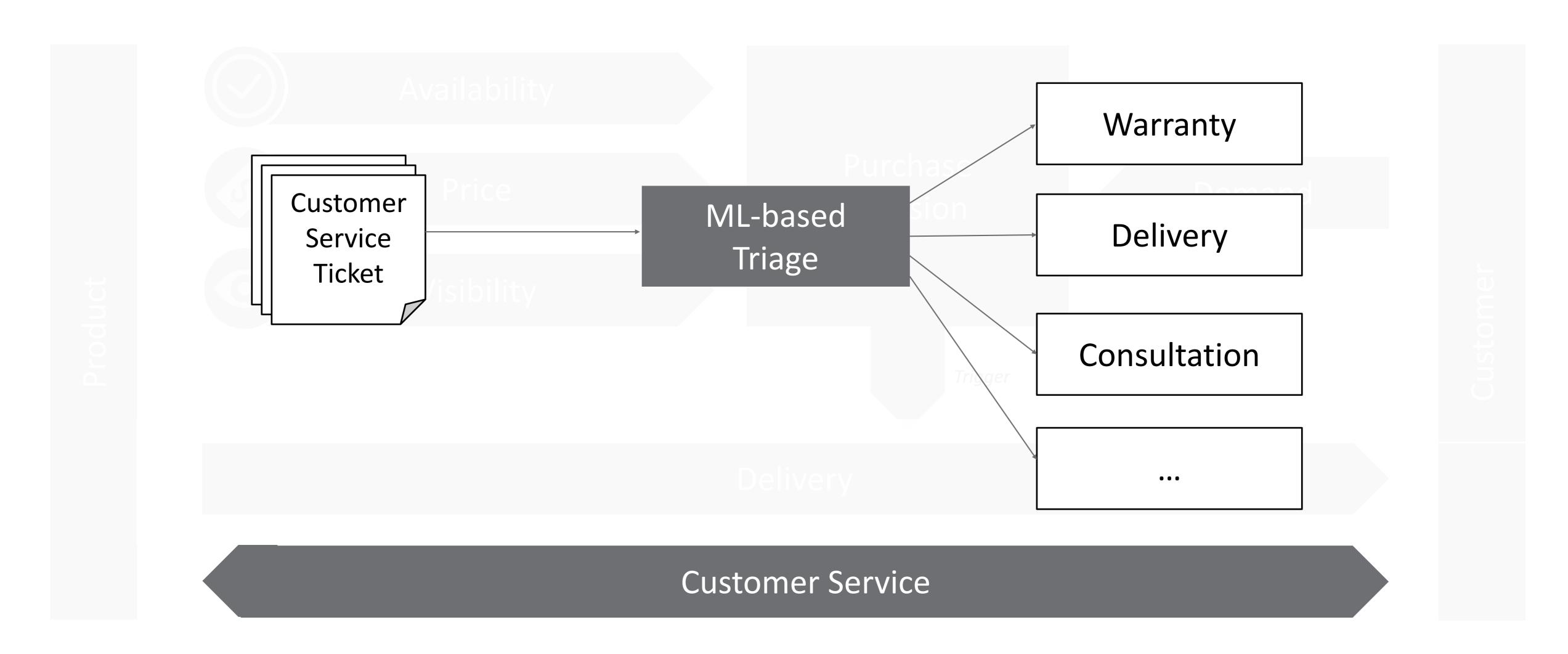




Price Change







# ... what we'll be talking about next ...

