

Office Supplies Recommendation System

Volha Puzikava
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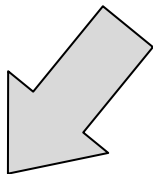
Disclaimer

The described analyses fulfill educational purposes only. The hypothetical business case, made-up data and the results of the performed analyses should not be considered as real recommendations of the seller, and have not been approved by any professional organization or trading company.

DISCLAIMER

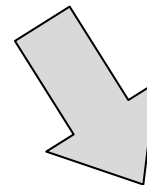
Overview

Office Supply Products



New school year:

- >50% of consumers shop online
- Record amounts are spent to prepare kids for school



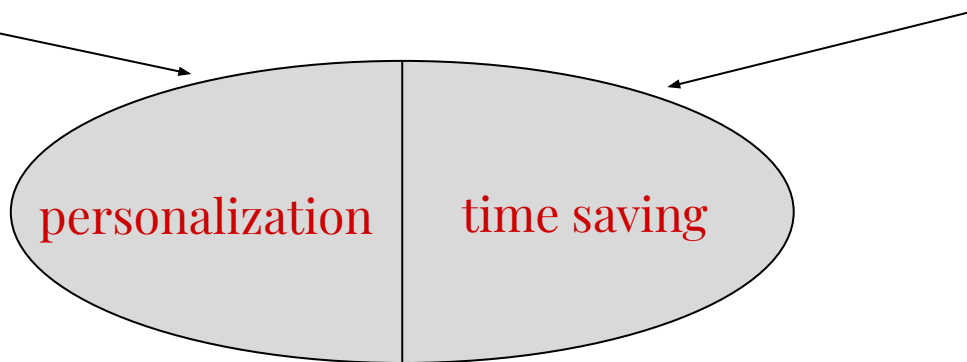
Companies:

- Company with 1-4 employees spends ~ \$1,844 per employee yearly
- Well-stocked office = successful office

Overview Cont'd

This project:

1. **Recommends office supplies** based on the reviews of previously bought products



2. Gives advice, **if it is valuable** to offer products in **the pack of two**

Outline

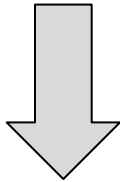
- Business Problem
 - Data
 - Data Preparation and Exploration
 - Modeling
 - Evaluation
 - Conclusions



Business Problem

The *Stationary and Co. Company* asked:

1. For recommendation system of their office supply products
2. If it is worth to offer products in the pack of two



more sales



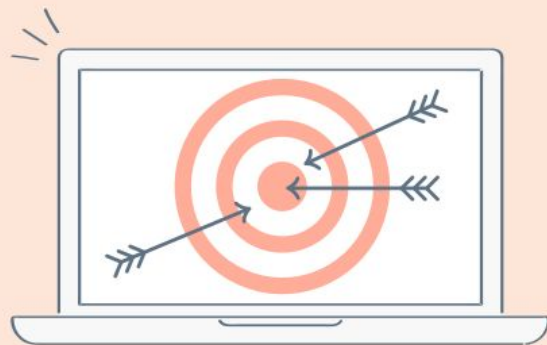
high level of personalization



Business Problem Cont'd

Goals:

1. Find the best recommendation model
2. Perform experiment to test the effectiveness of the new package



Data

- Taken from [Amazon review data](#) (2018) page
- Included 5,581,313 reviews
- Contained meta data for 315,644 office supply products



Data Preparation and Exploration

1.

Ratings

Meta Data

Merged Data

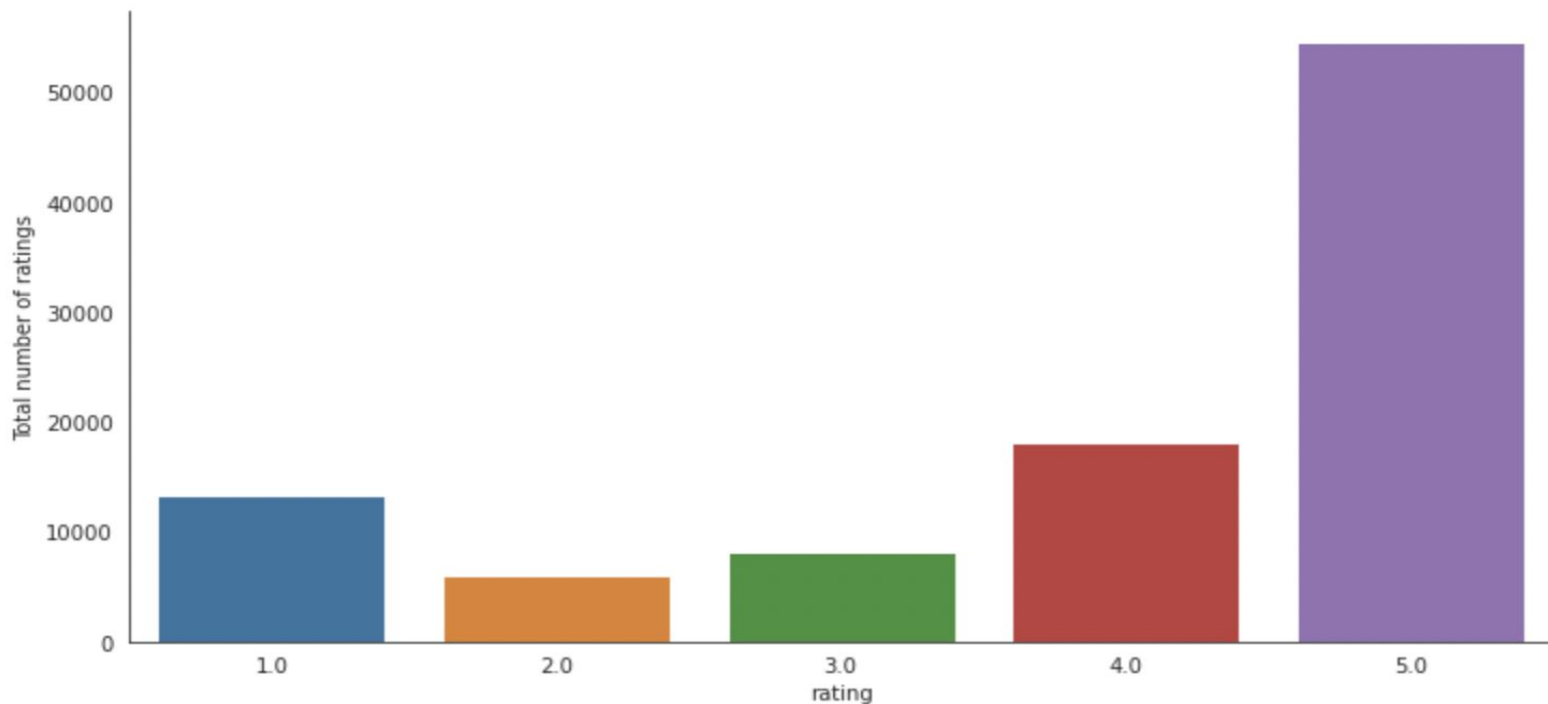
Reviewer ID	Product ID	Rating
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Product ID	Title	Category Type
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100,000 random rows chosen

Data Preparation and Exploration Cont'd

2. The distribution of ratings plotted



Modeling

Models

- **KNN Basis**
RMSE: 1.431
- **KNN Baseline**
RMSE: 1.370
- **KNN with Means**
RMSE: 1.433

**Singular Value
Decomposition
(SVD)**
RMSE: 1.366

**Alternating
Least Squares
(ALS)**
RMSE: 4.209

***RMSE – error made while predicting the data. The lower, the better

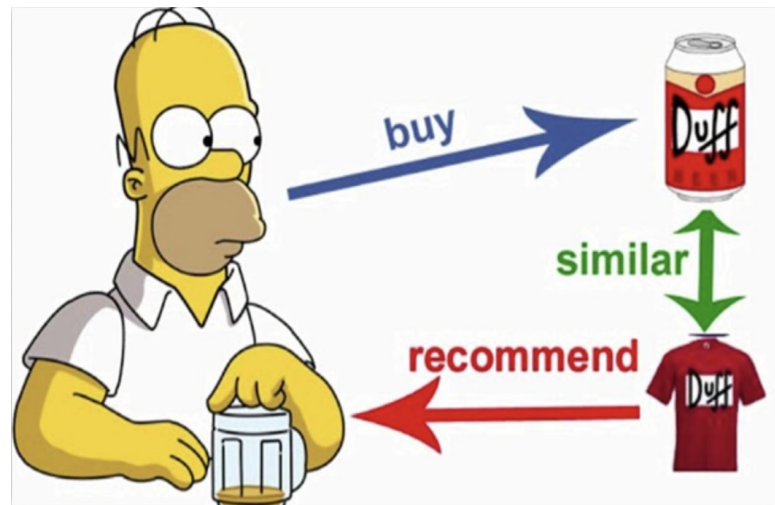
Modeling Cont'd

1. Models built in
 - a. surprise library,
 - b. Spark programming environment
2. *Item-item similarity* performed

Number of users: 94184

Number of items: 27405

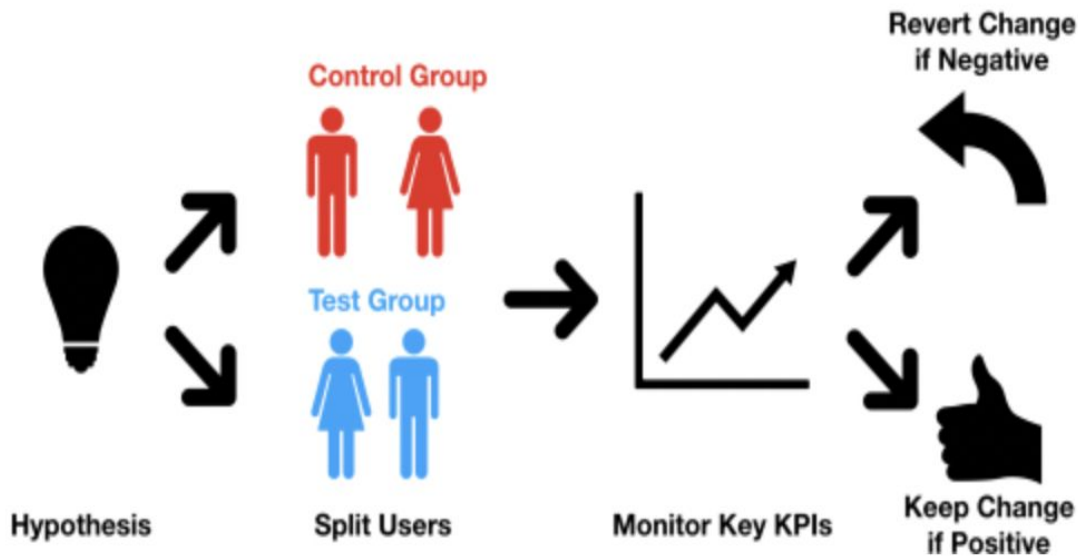
3. For ALS model: columns of labels transformed into columns of label indices (StringIndexer used)
4. *Recommended 10 products* for a user based on his/her reviews



Modeling Cont'd

Experiment

(A/B Testing)



What we know:

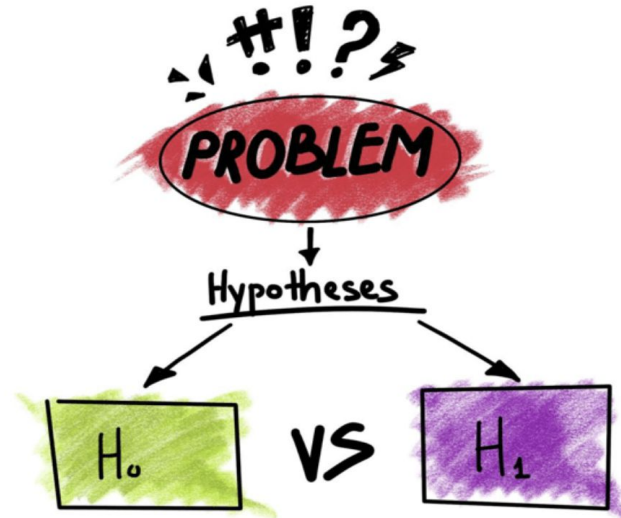
1. Increase in buying rate = 5%
2. ~6.5 million visitors per day
3. ~25% buy products daily

What we do:

1. Two groups: control & experiment
2. Chi-square goodness of fit test

Modeling Cont'd

1. Hypotheses:
 - a. *Null*: the probability of success for the new package < 0.3
 - b. *Alternative*: the probability of success for the new package ≥ 0.3
2. Min # of customers = 60,279
3. Period of 28 days was analyzed



Evaluation

Best model:

SVD ($n_factors=20$, regularization rate=0.02)

Experiment Results:

- a) New package will decrease the buying rate by 6.25%
- b) Not worth implementing



Conclusions

1. Use the SVD model for recommendations
2. Do not implement the new package

NEXT STEPS:

1. Tune the models
2. Come up with other ideas how to increase sales



Thank You!

Email: helga.mikel@gmail.com

GitHub: [@VolhaP87](https://github.com/VolhaP87)

LinkedIn: <https://www.linkedin.com/in/volha-puzikava-2319294a>

