

A Machine Learning based approach for the Competitor Information Analysis in the Automotive Industry

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1. Introduction

- Competitor Analysis
- Problems
- Research questions

2. Requirements

3. System Design

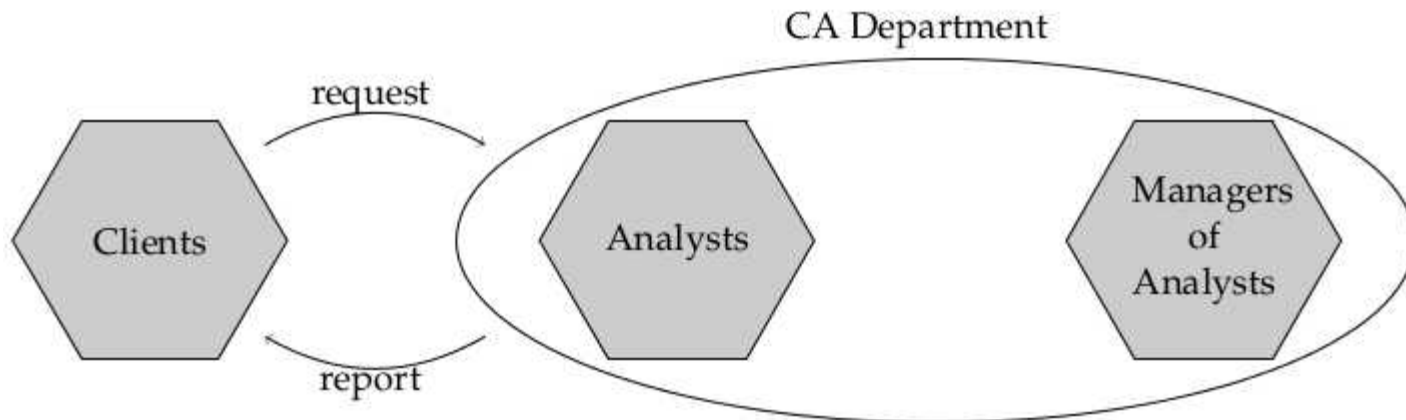
4. Evaluation

- System Usability
- Feedback

5. Future Work

Competitor Analysis (CA)

- Identify Strength, Weakness, Opportunities, Threats (SWOT)
- Identify
 - All competitors
 - Basic competitors
 - Primary competitors
- Support decisions in management and development
 - Spread Competitive Intelligence (CI) within the company

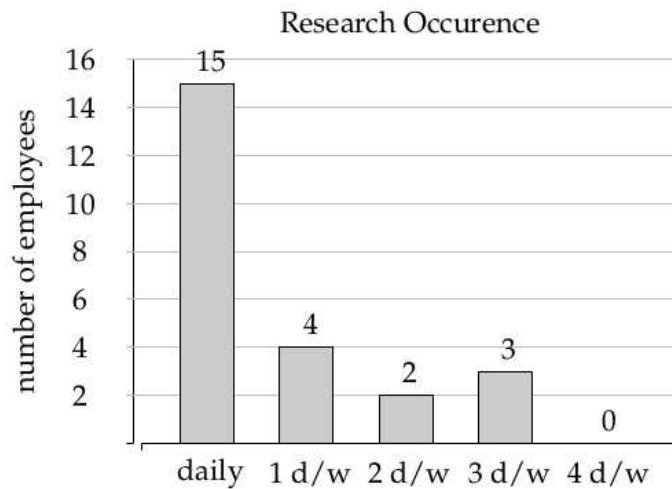


Competitive Intelligence Process

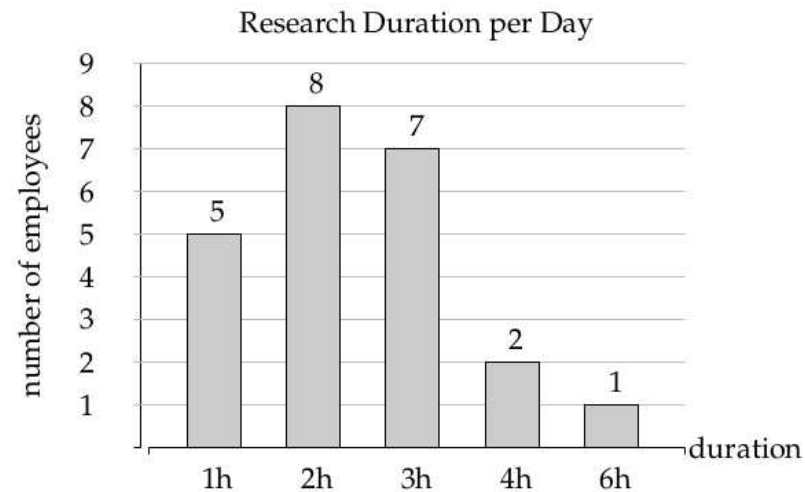
50 % of process time



Source: Rainer Michaeli. Competitive Intelligence: Strategische Wettbewerbsvorteile erzielen durch systematische Konkurrenz-, Markt- und Technologieanalysen. Springer-Verlag Berlin Heidelberg, 2006.

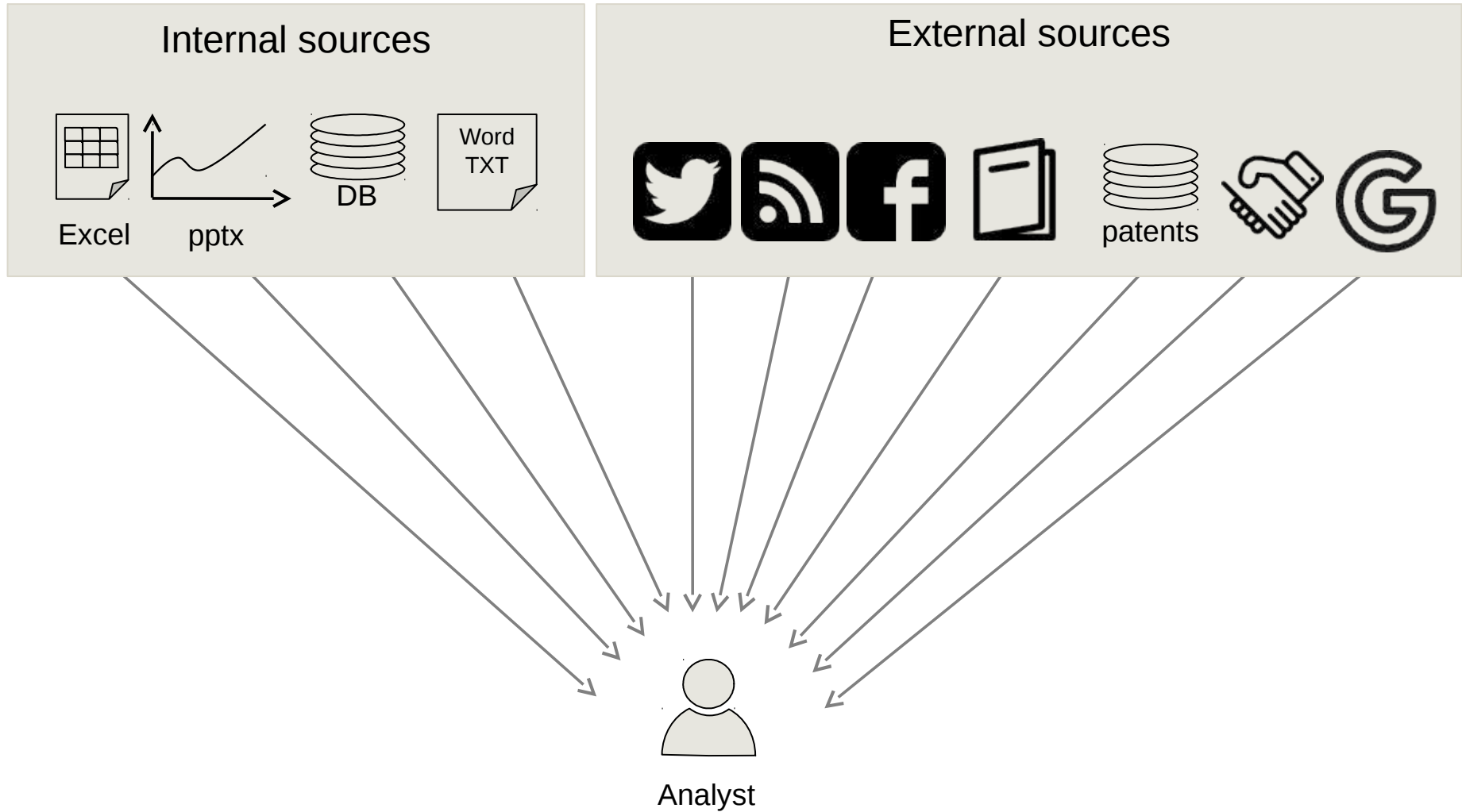


(a) Research Occurrence per Week



(b) Research Duration per Day

Competitive Intelligence Process – Data Acquisition



Competitive Intelligence Process – Problems

- Too many information sources for manual analysis
- Long duration and high occurrence of analysis
- Distinguish between relevant and irrelevant information
- Retrieve processed information

Research Questions

- How to improve the existing competitor information analysis process in the automotive industry?
- What kind of documents and document sources are used for competitor information analysis in the automotive domain?
- Which categories/topics do these documents belong to?
- Which supervised classification algorithm is suitable for automatic document classification to support competitor information analysis in the automotive domain?

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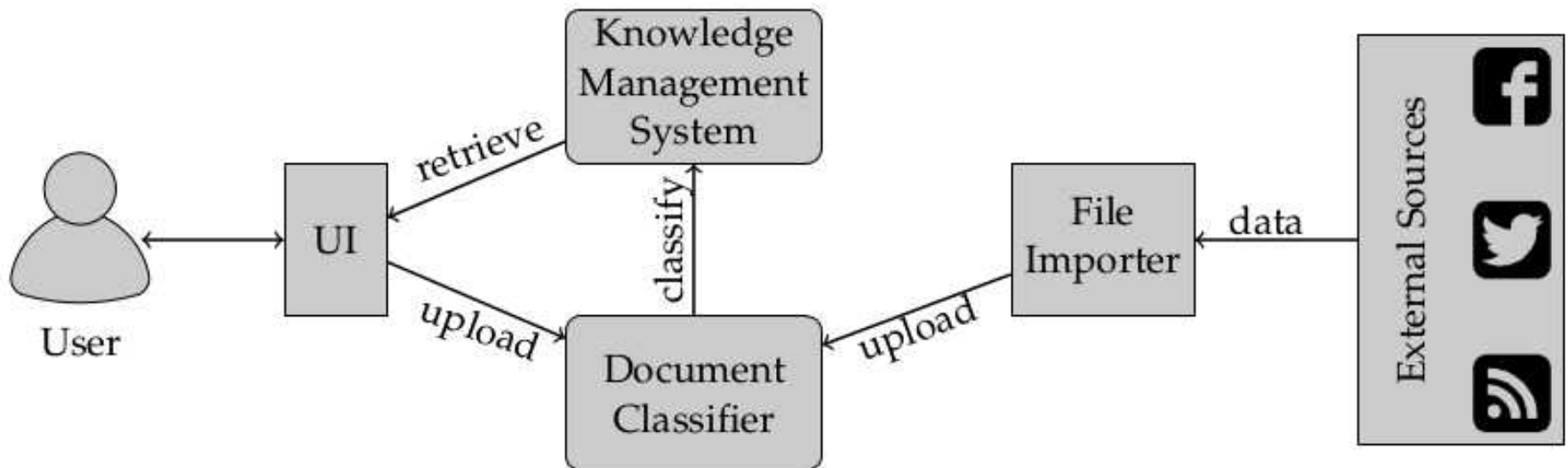
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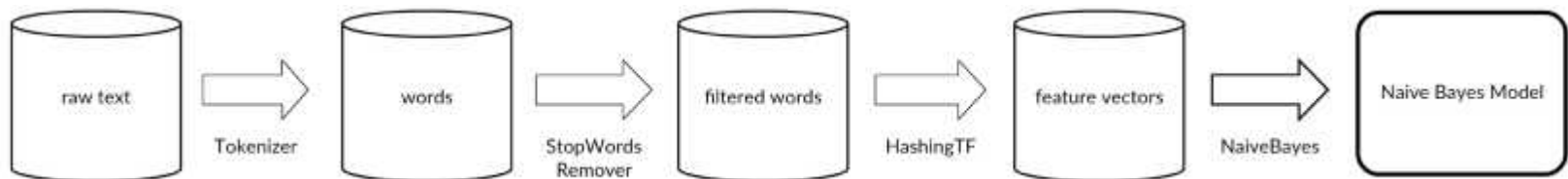
Requirements

- One stream of information
- Search engine as UI
- Automated import of data from external sources
- Knowledge management system with the support for storing both the internal and external data
- Topic structure within knowledge management system



Automatic Document Classification

- Labels defined by department's topics
 - Prognosis
 - Exhibitions
 - General Information
 - Initial Evaluation
 - Press Evaluation
 - Garbage
- Training documents assigned manually to each topic
- Supervised Machine Learning (ML) algorithms compared
 - Support Vector Machine (SVM): low prediction accuracy
 - K-Nearest Neighbours (k-NN): low prediction performance
 - Naive Bayes (NB): good prediction performance and accuracy

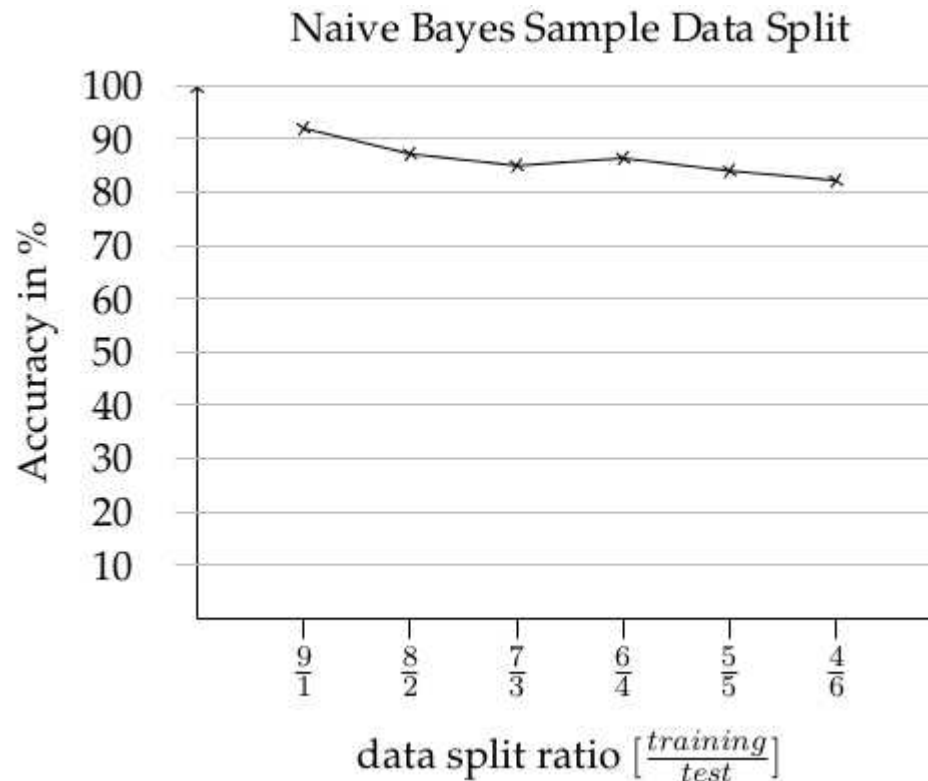


Automatic Document Classification

Naive Bayes

Training Phase:

- n-Fold cross validation (n=10)
- Different data split ratios (training : test) were tested



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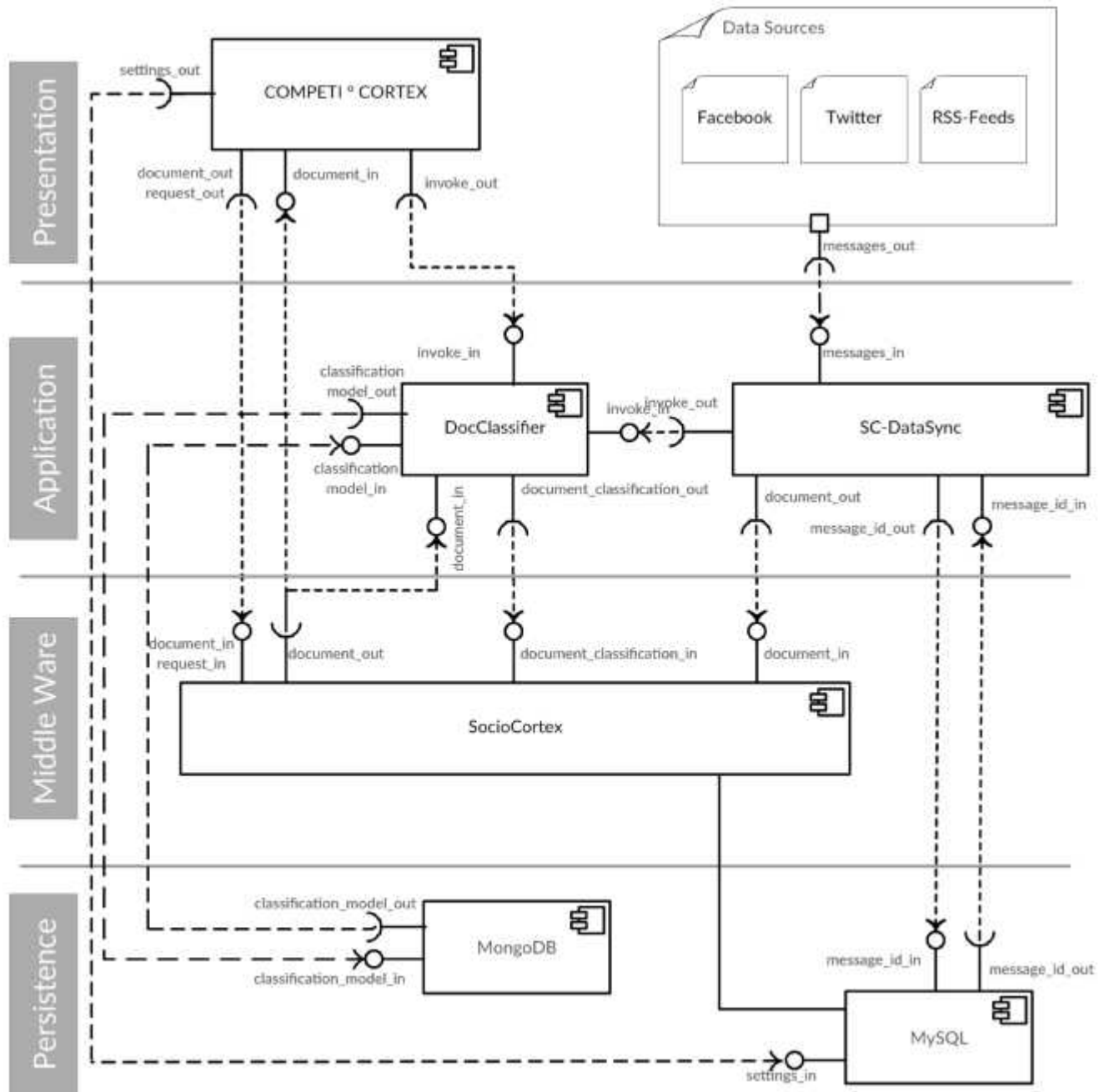
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COMPETI • CORTEX

Demo

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- A subset of internal documents and messages of external sources used
- 9 test subjects (managers and analysts)
- Introduce system to every person individually
- Test the system
- Evaluation questionnaire (System Usability and Feedback)

System Usability Scale (SUS)

- Ten-item scale [1]
- Illustrating subjective assessments of system usability
- Positional value for each question (1 – 5)
- Calculating SUS-score (0% – 100%)

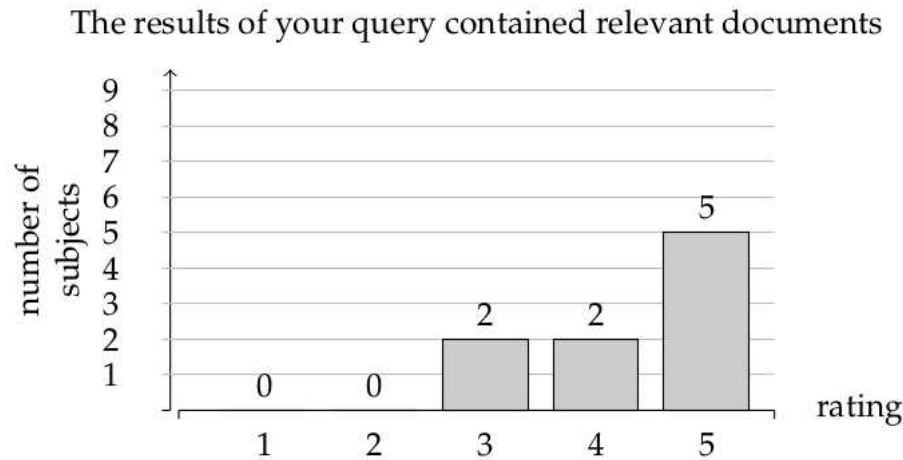


→ Exceeds the recommended value of 68% [2]

Source: [1] John Brooke. Sus: A quick and dirty usability scale, 1996.
[2] John Brooke. Sus: A retrospective. Journal of Usability Studies, 8(2):29–40, 2013.

Feedback (1)

- What do you like most about the application?
 - Usability, simplicity & attractiveness of user views (intuitive system handling)
- What should be improved in the current system?
 - Transparency in results' relevance
- Did the result of your query contain relevant documents?



Feedback (2)

- Could you retrieve your uploaded file?
 - + PDF, Word, Powerpoint
 - Excel

- Assess whether the efficiency of the information acquisition can be optimized by using the system
 - + 7 subjects agree, if document pool is extended and external message currentness is granted
 - 1 subject: Only a subset of internal information sources can be included

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- Extend information sources → improve automatic document classification
- Provide individuality (user accounts)
- Dictionary
 - Synonyms
 - Translations
- Provide Optical Character Recognition (OCR) for scanned documents



B.Eng.

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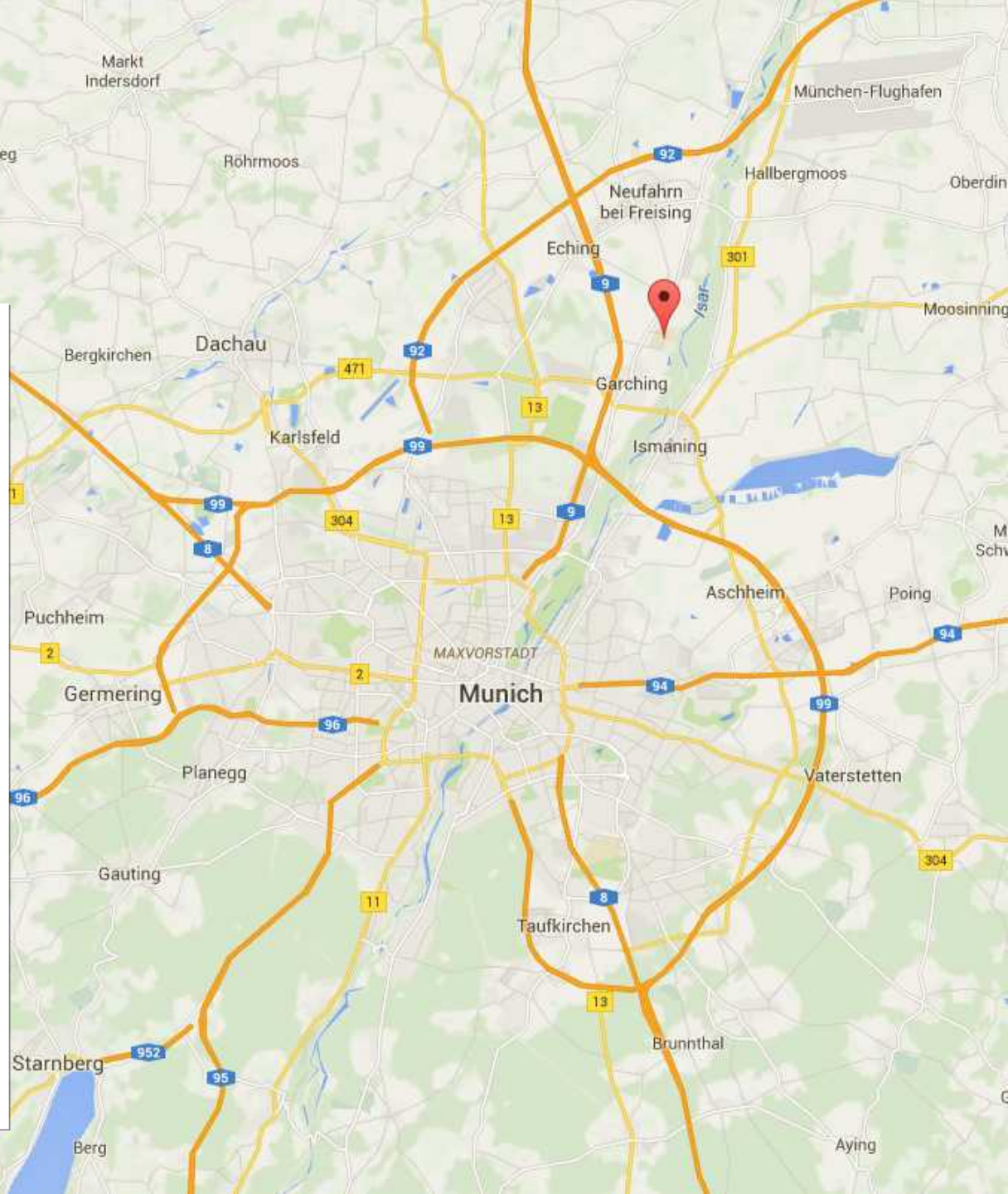
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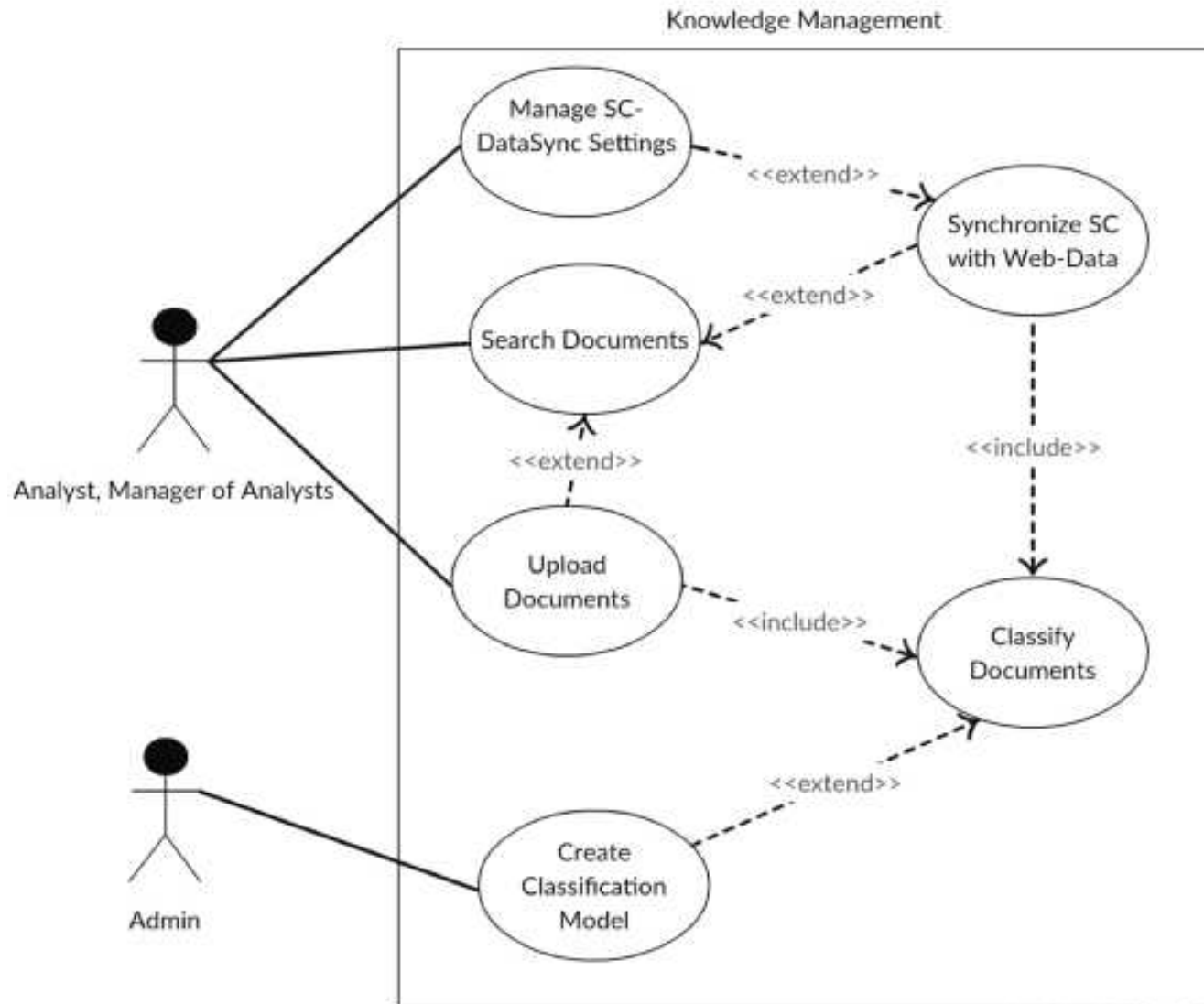
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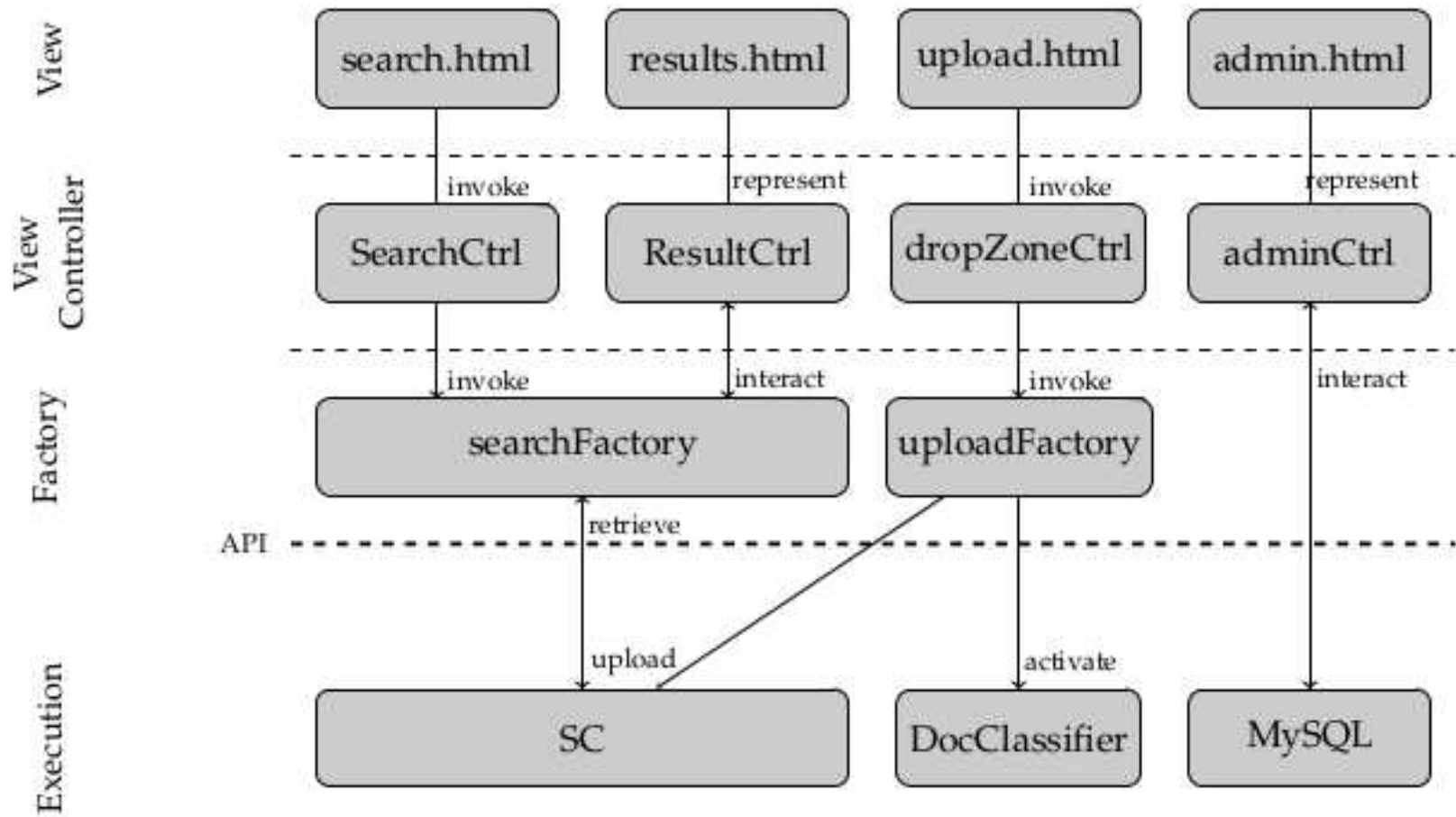
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Backup



CompetiCortex Layers



Comparison Between Classification Algorithms

| Classification | accuracy | ø-precision | ø-recall |
|----------------|----------|-------------|----------|
| kNN (k=3) | 91,28 % | 88,12 % | 85 % |
| Naive Bayes | 92,77 % | 89,46 % | 88,02 % |
| SVM | 47,66 % | 7,94 % | 16,67 % |