

Supporting the Legal Reasoning Process by Classification of Judgments Applying Active Machine Learning

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Motivation

Legal Reasoning Process

Approach

Implementation

Data Set

Evaluation

Conclusion

Motivation

Legal Documents

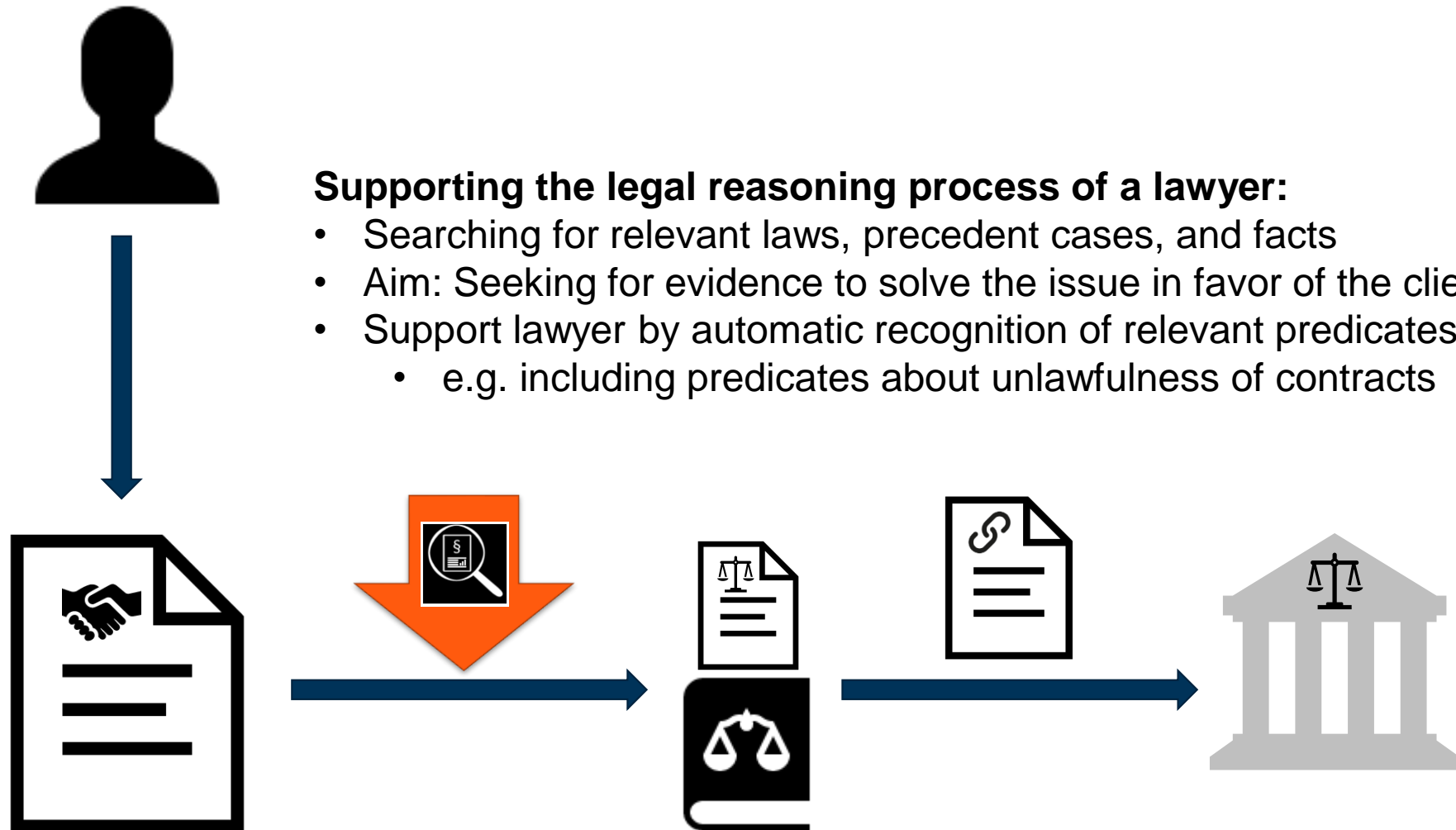
- are growing and changing: *2013-2017 period more than 550 adopted laws [1]*
- getting more complex

Document Discovery

- often done manually (knowledge-intensive)
- very time-consuming and costly [2]

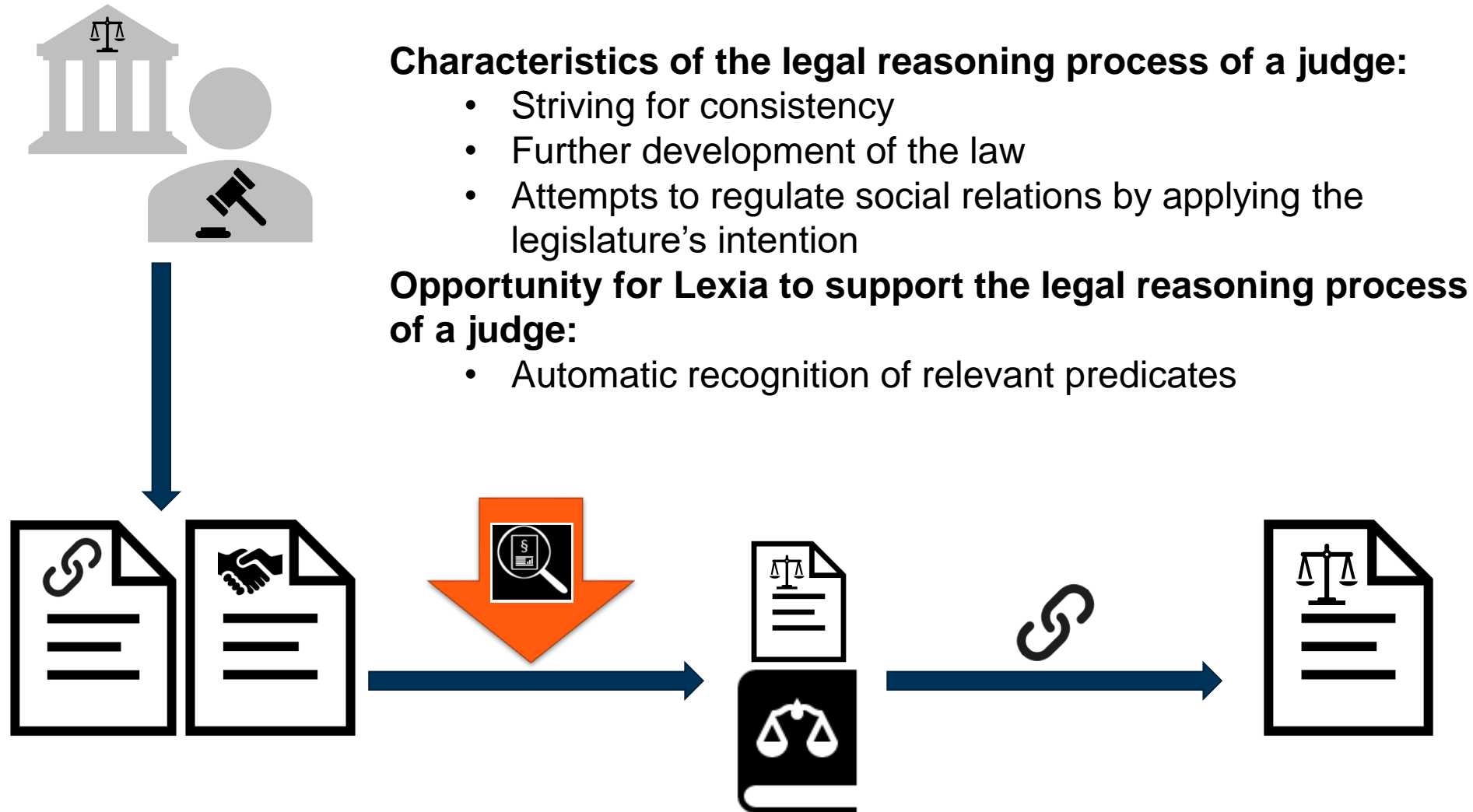
Supporting the legal reasoning process by classification of Judgments

- common law relies heavily on precedents
- judgments relevant in (German) civil law?
 - further development of the law by judges
 - Case law is often applied in dynamic areas of life
 - German landlord and tenant law is largely shaped by case law



Supporting the legal reasoning process of a lawyer:

- Searching for relevant laws, precedent cases, and facts
- Aim: Seeking for evidence to solve the issue in favor of the client
- Support lawyer by automatic recognition of relevant predicates
 - e.g. including predicates about unlawfulness of contracts



Manuel Annotation of
Judgments

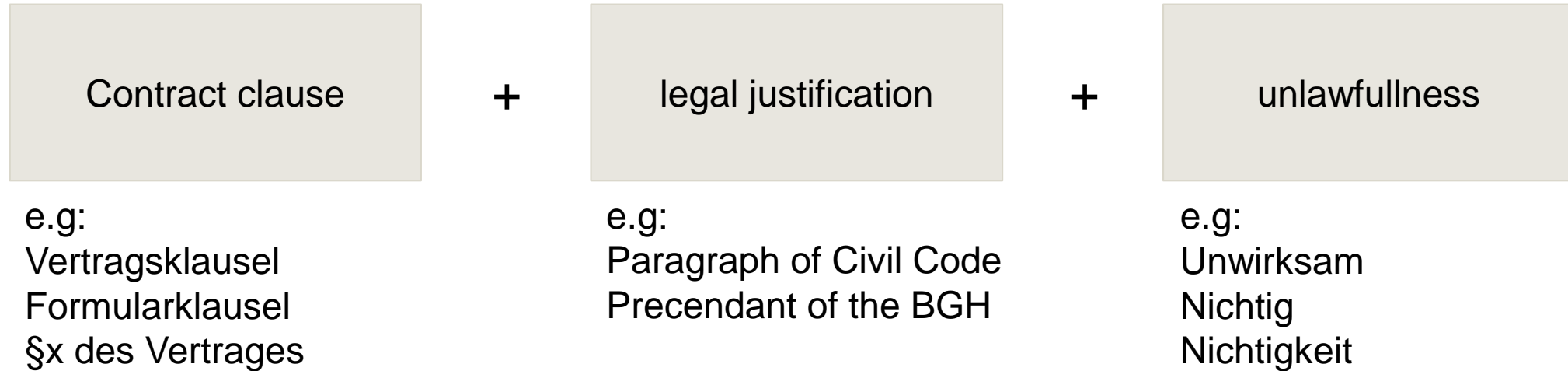
Sentences about the
unlawfulness of contract clauses

Implementation of a BTC for
judgments

Adaption of Lexia and LexML

Evaluation

Evaluating common
performance indicators like
f1score, ROC, recall and
precision



Example:

[...] entsprechende Formularklausel wegen unangemessener Benachteiligung des Mieters nach § 307 BGB unwirksam [...]

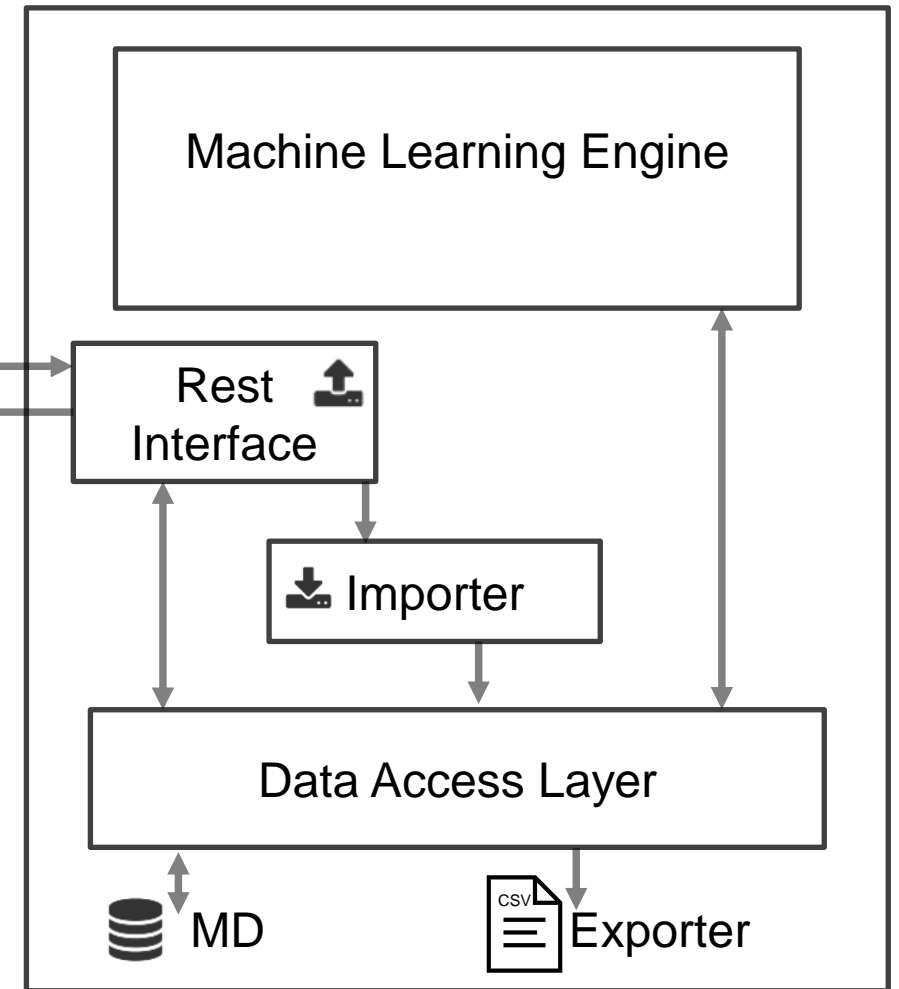
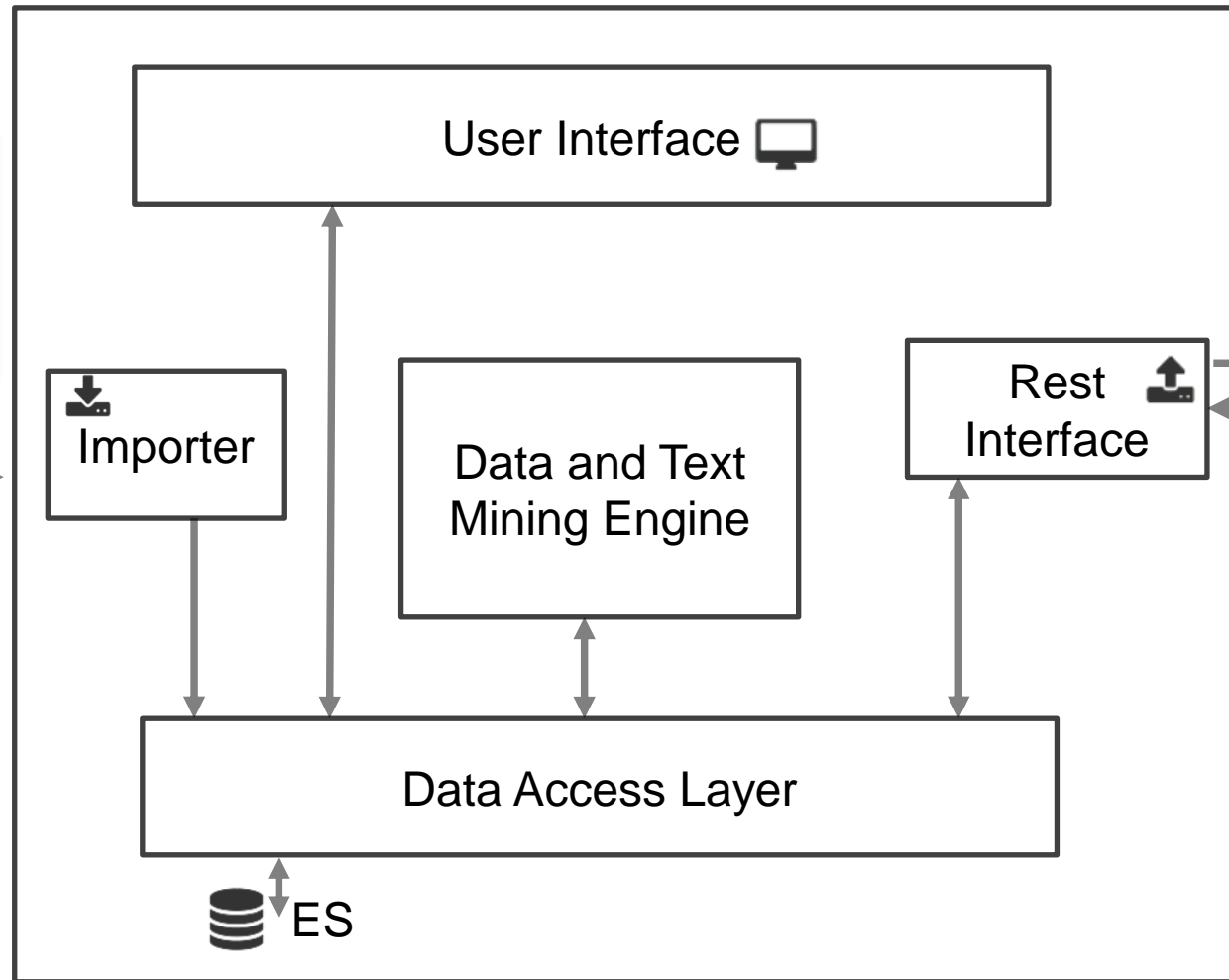
Problems with the Annotation „Rule“:

- Sometimes one Part is mentioned in following or previous sentence -> partly causing the False Positives

Implementation

Lexia

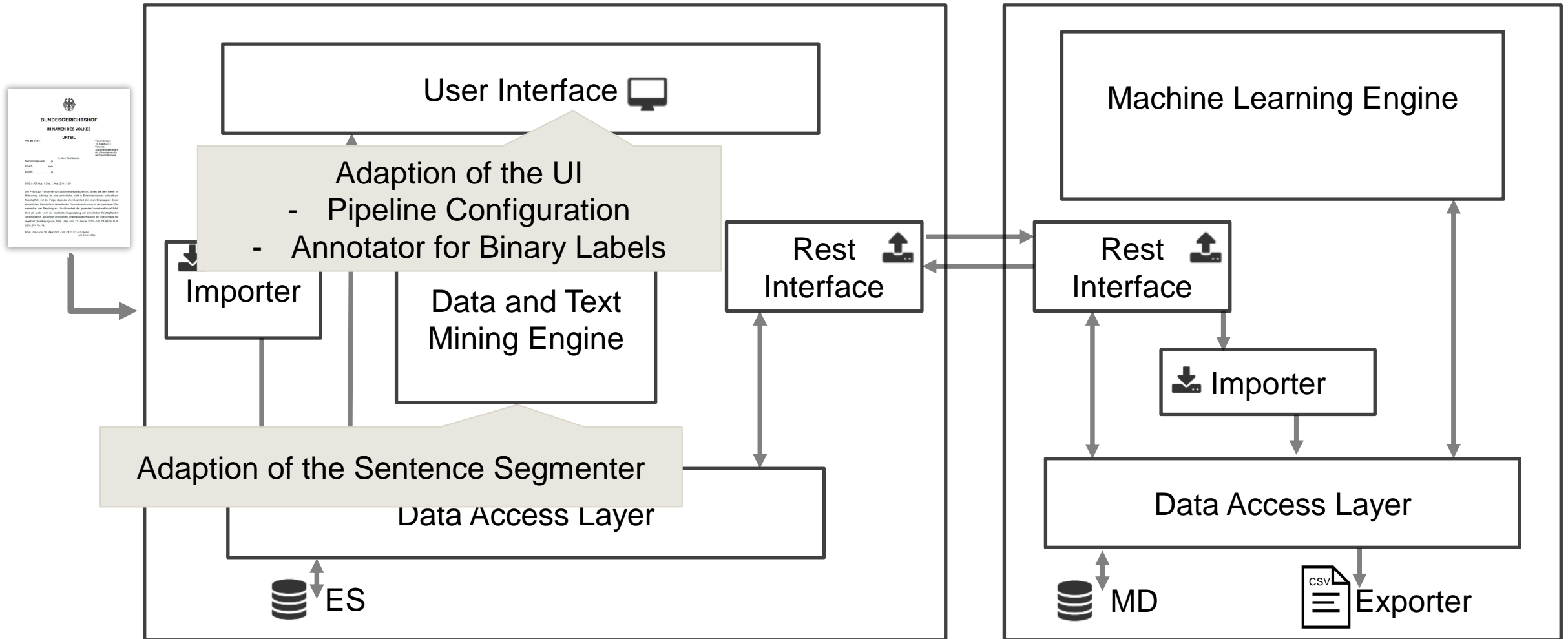
LexML



Own illustration based on Muhr [3]

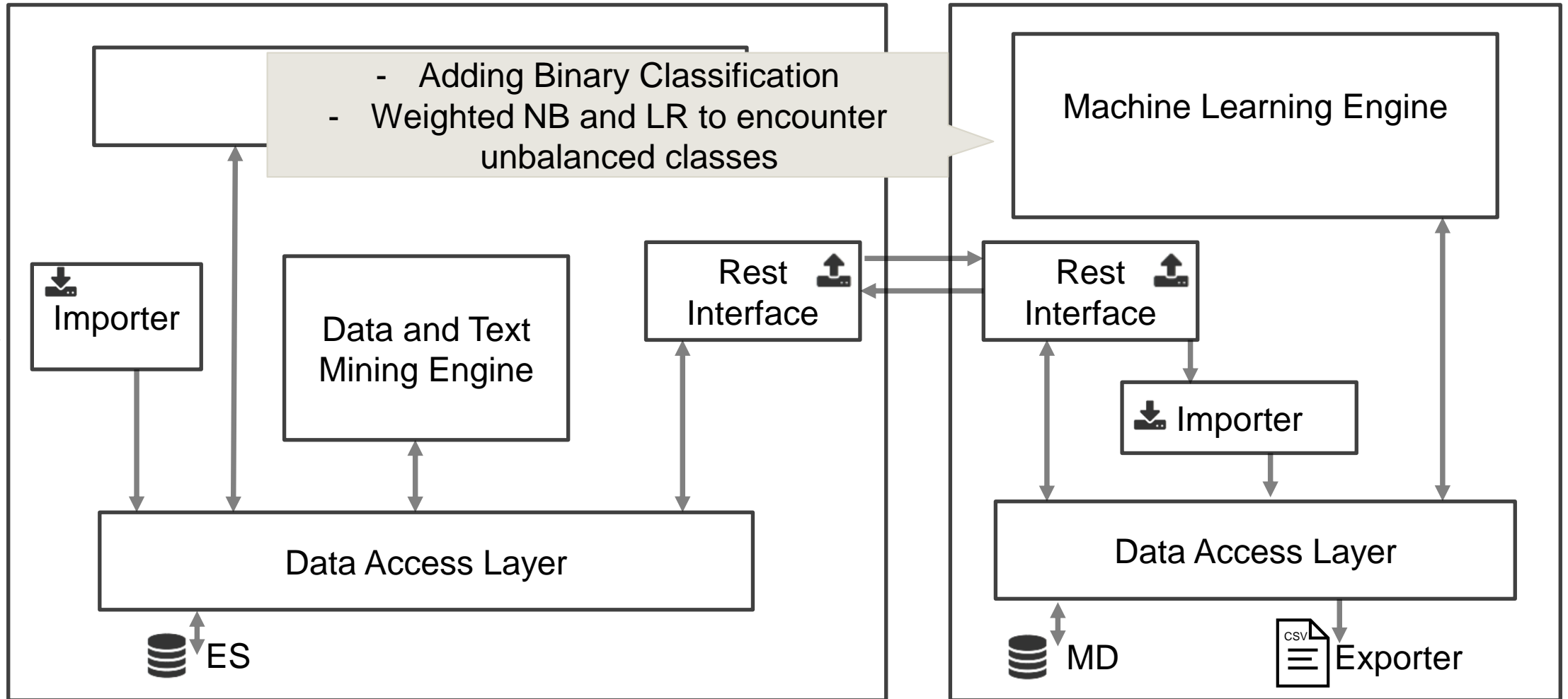
Lexia

LexML



Lexia

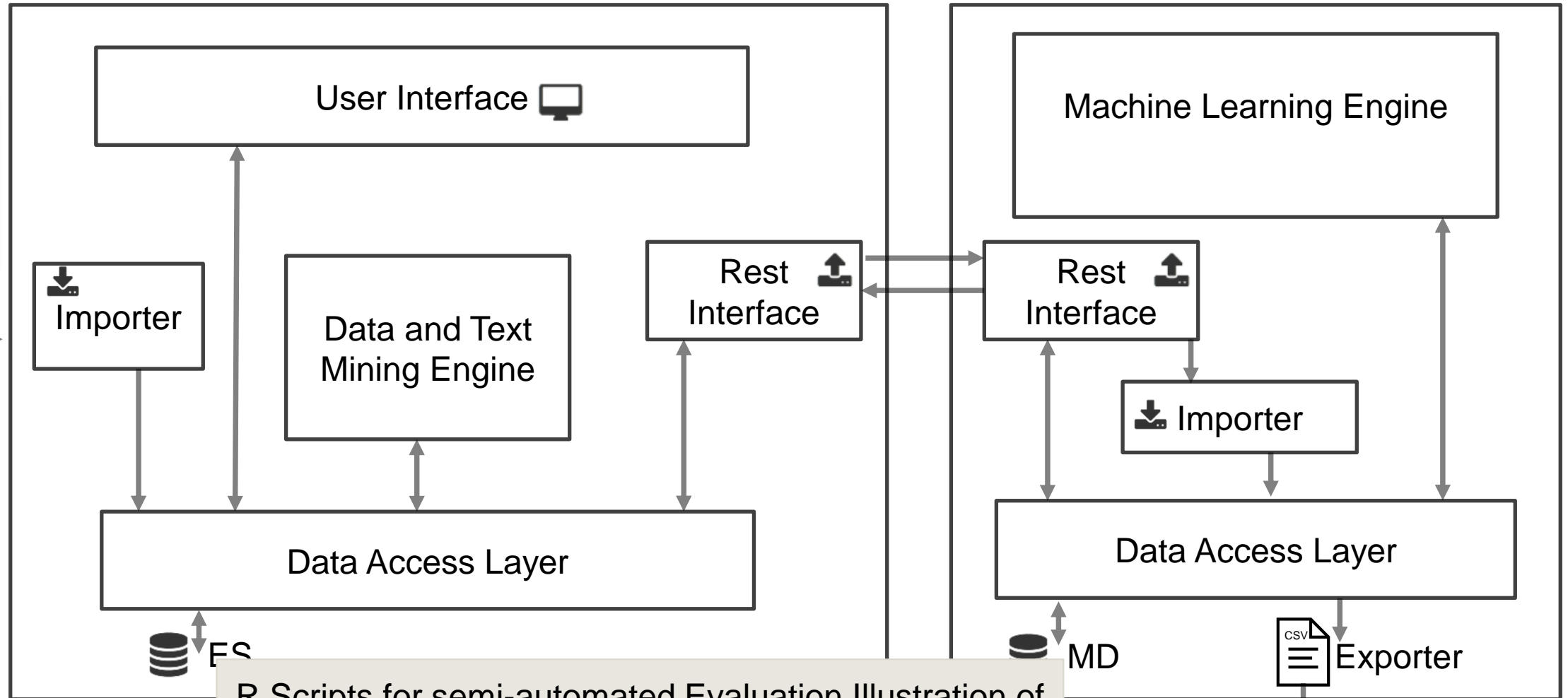
LexML



Exporter for Binary Evaluation Metrics

Lexia

LexML



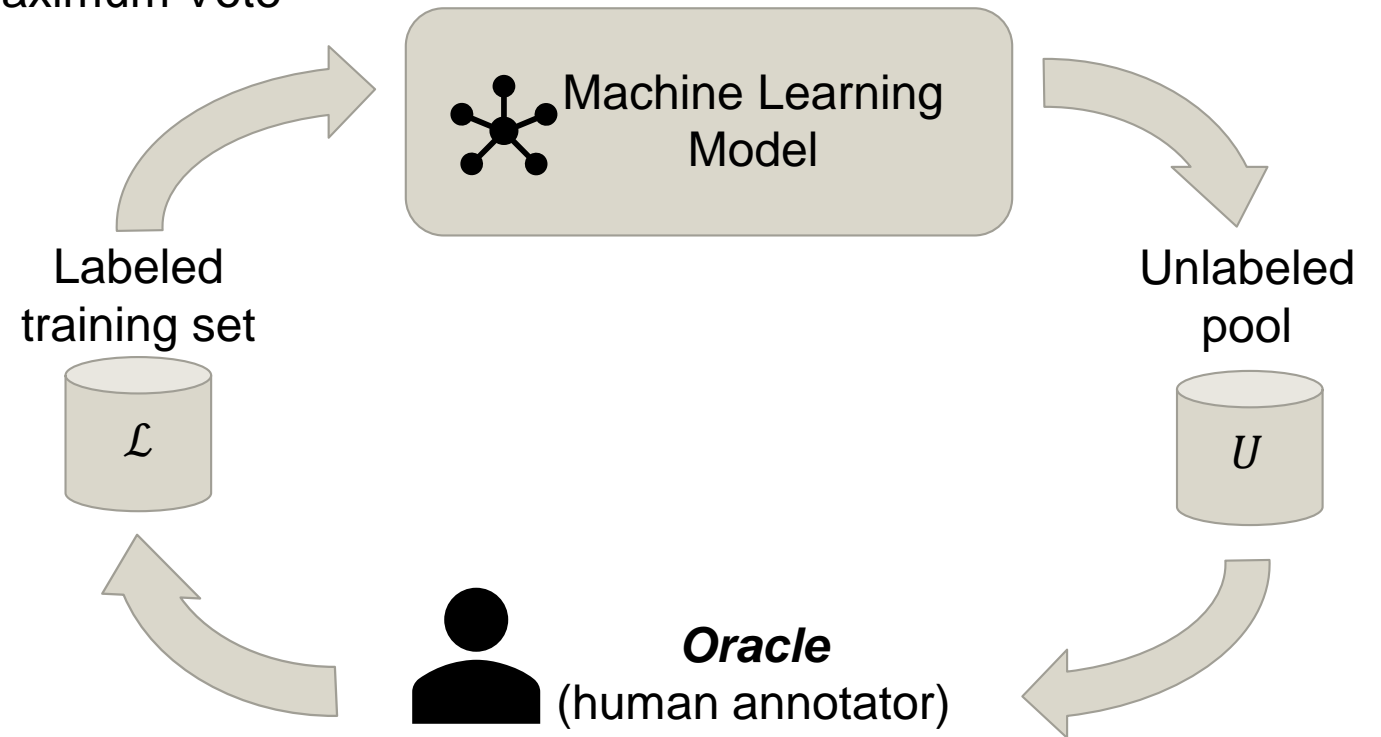
R Scripts for semi-automated Evaluation Illustration of different performance Measurements like f1 score, PR-Curve, ROC, etc.



Active Learning – Setting

AML Pipeline Stages:

- Preprocessing: Tokenizer, Stopwords Removing
- Vector Representation: Binary Word Representation
- Classifier: Multinomial Naive Bayes (Weighted)
- Query Strategy: Uncertainty Sampling (Maximum Vote Entropy)



Data Set

Raw Documents

- Over 800 imported and annotated judgments
- BGH judgments of the VIII. Civil Senate
 - specialized in law on the sale of goods, landlord and tenancy law

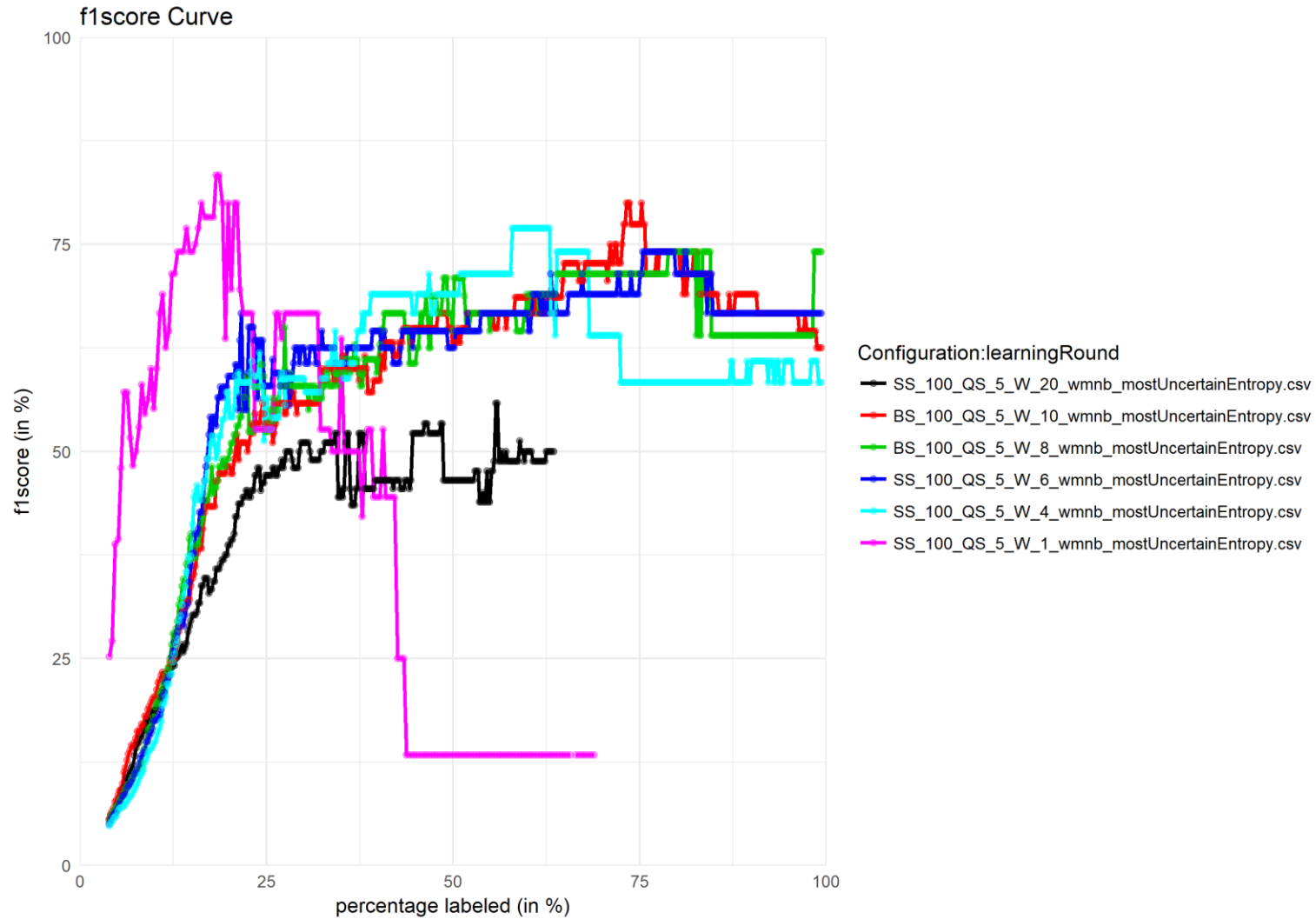
Training & Test Set

- Using a subset of 82 documents
- Only Tenor and Reasoning part imported
- 3135 Sentences of which are 72 True (2,3%)
- 80/20 partition

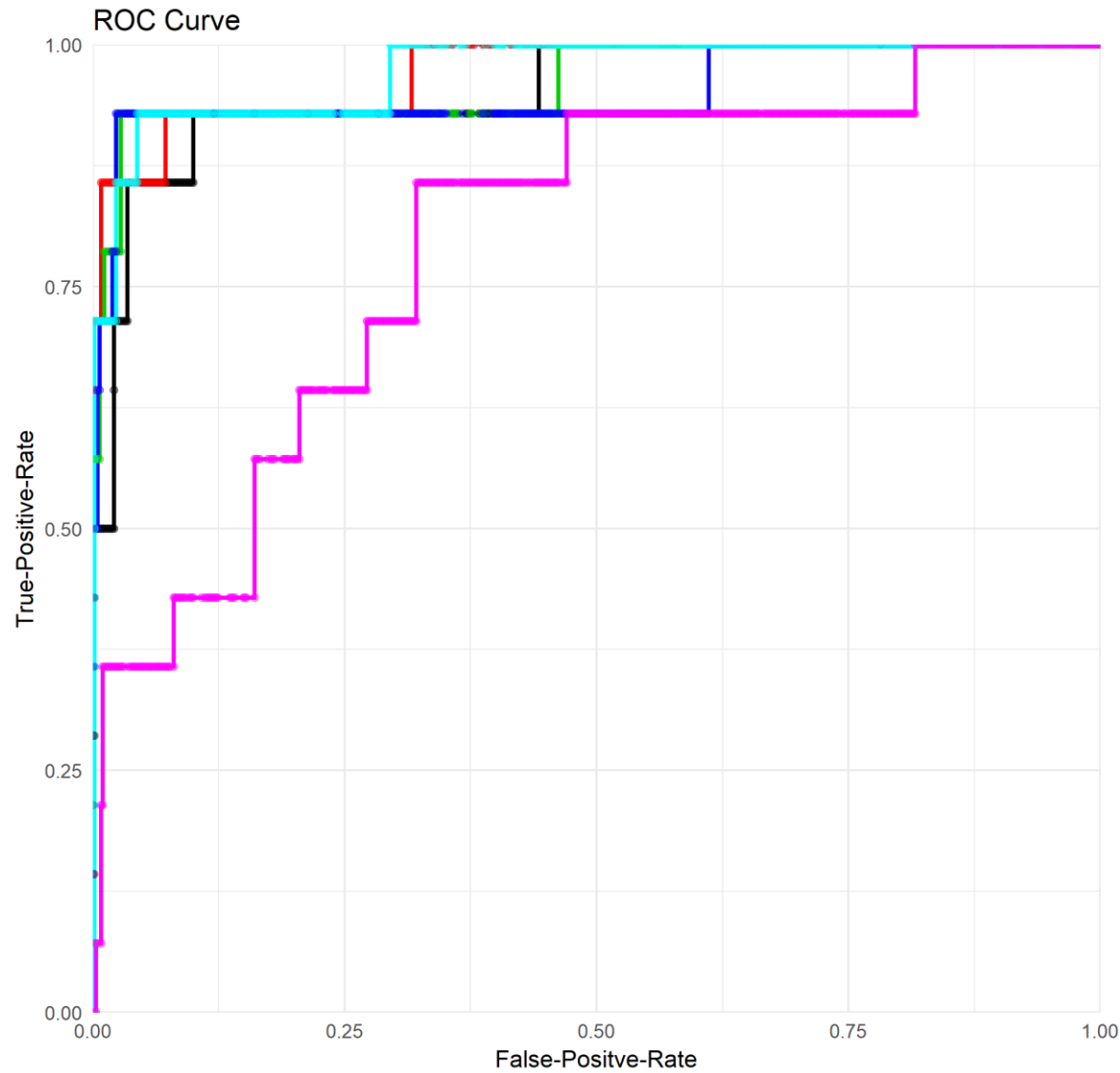


Evaluation

Discover weight factor



Discover weight factor

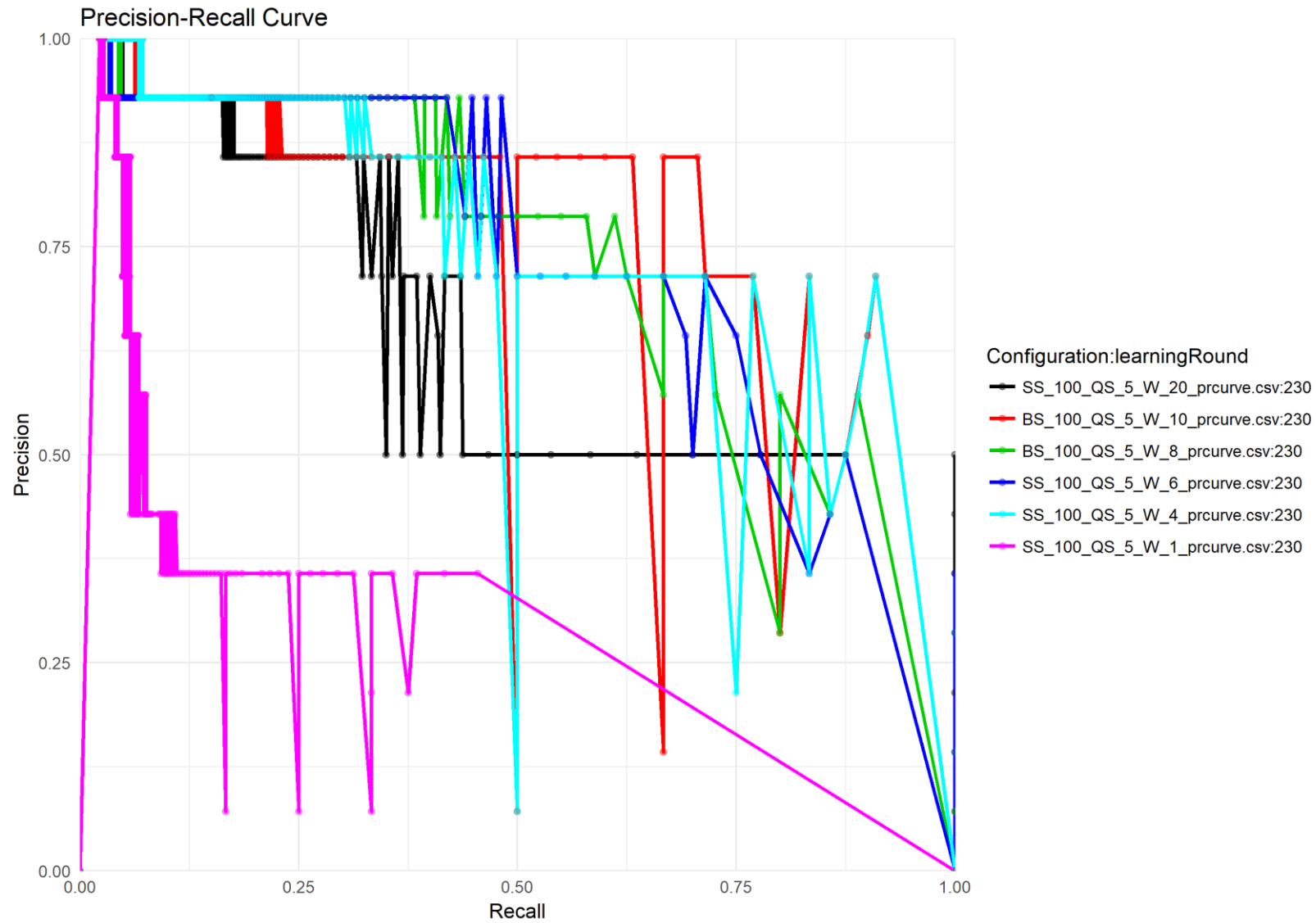


Weight 20/0.05 seems to perform well

Configuration:learningRound

- SS_100_QS_5_W_20_roccurve.csv:230
- BS_100_QS_5_W_10_roccurve.csv:230
- BS_100_QS_5_W_8_roccurve.csv:230
- SS_100_QS_5_W_6_roccurve.csv:230
- SS_100_QS_5_W_4_roccurve.csv:230
- SS_100_QS_5_W_1_roccurve.csv:230

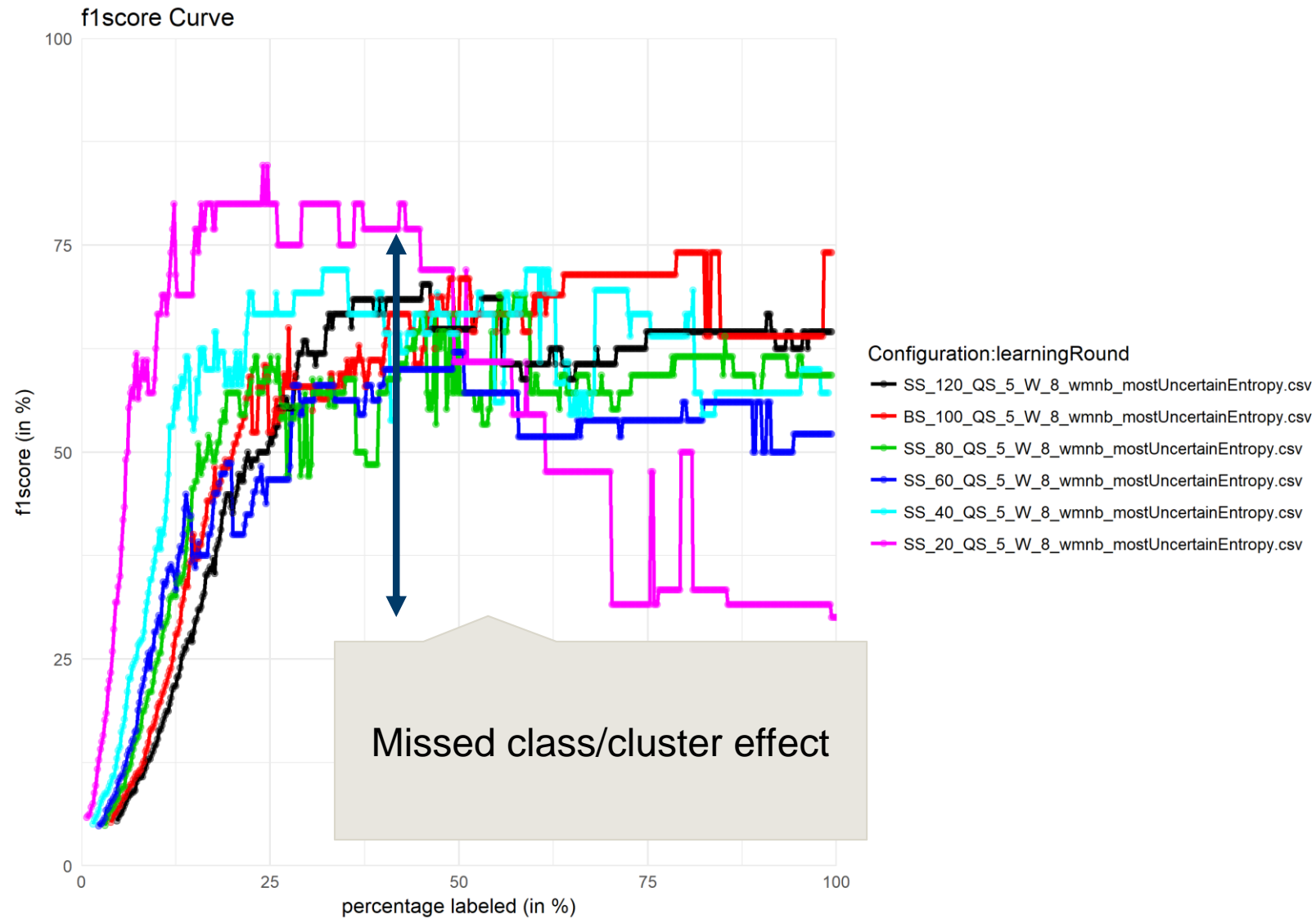
Discover weight factor



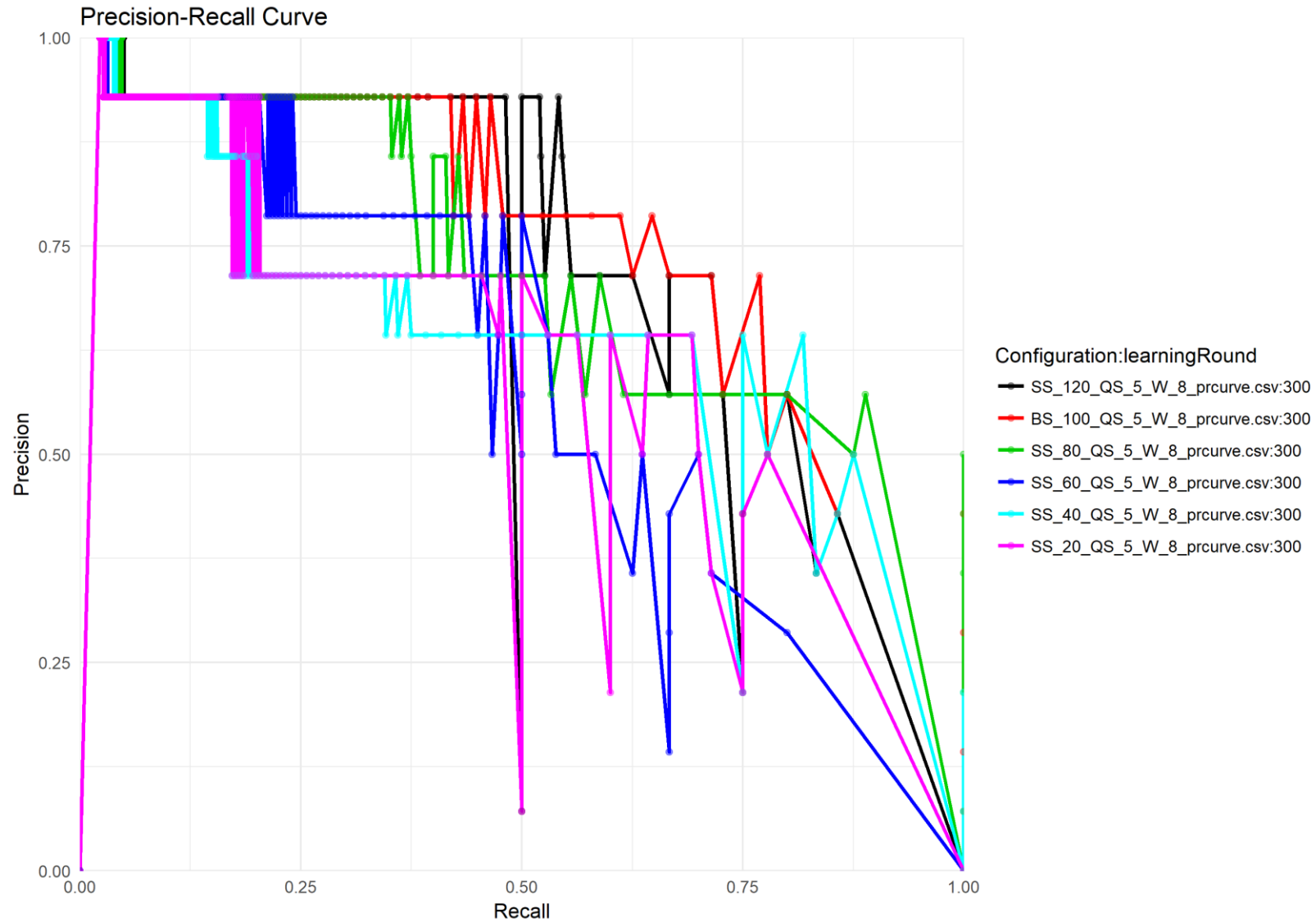
Comparison of Seed Set Size and Query Size Influence

- Weight true with factor 200 (w8)
 - Seed Set random sampled

Influence of Seed Set



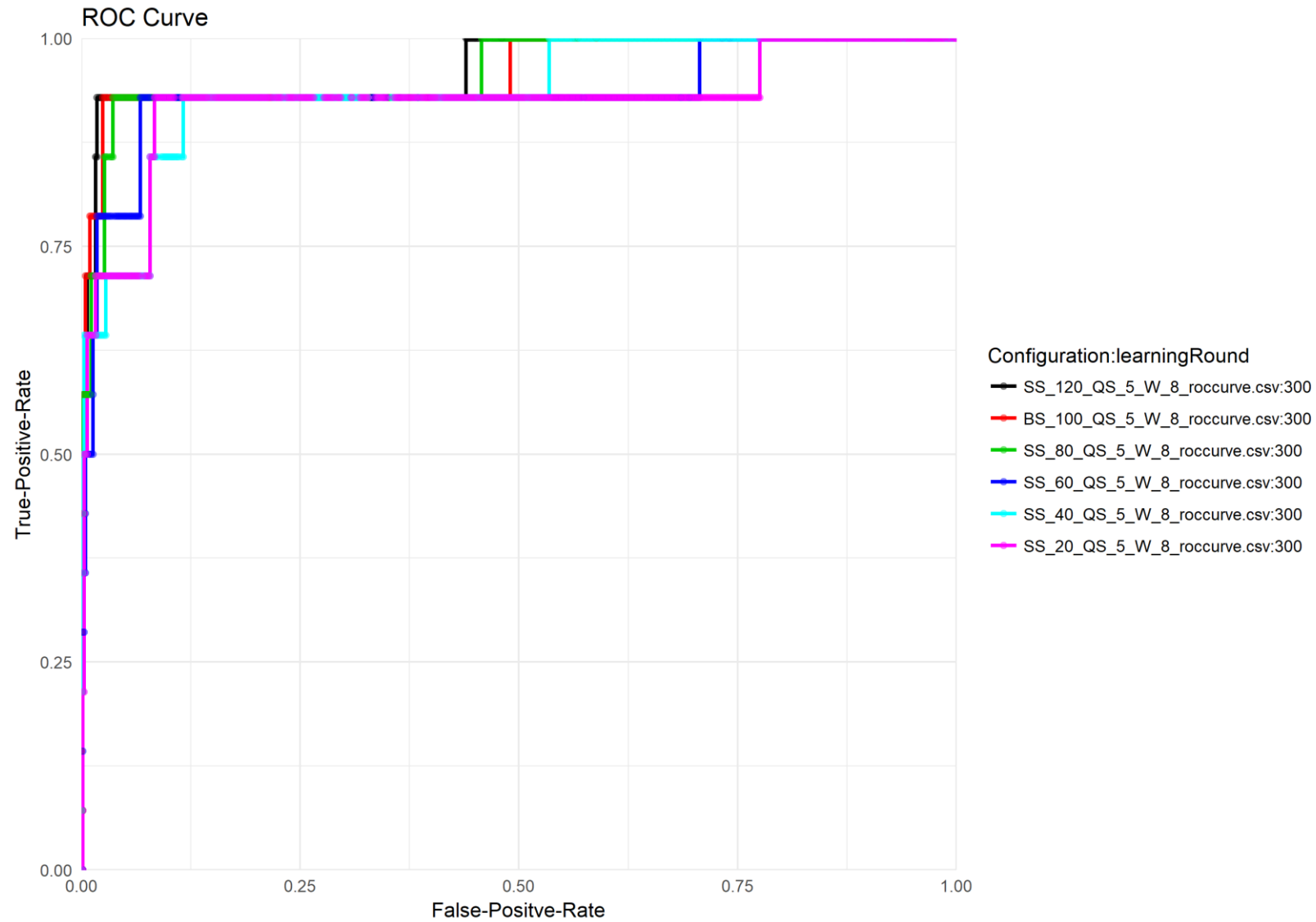
Influence of Seed Set



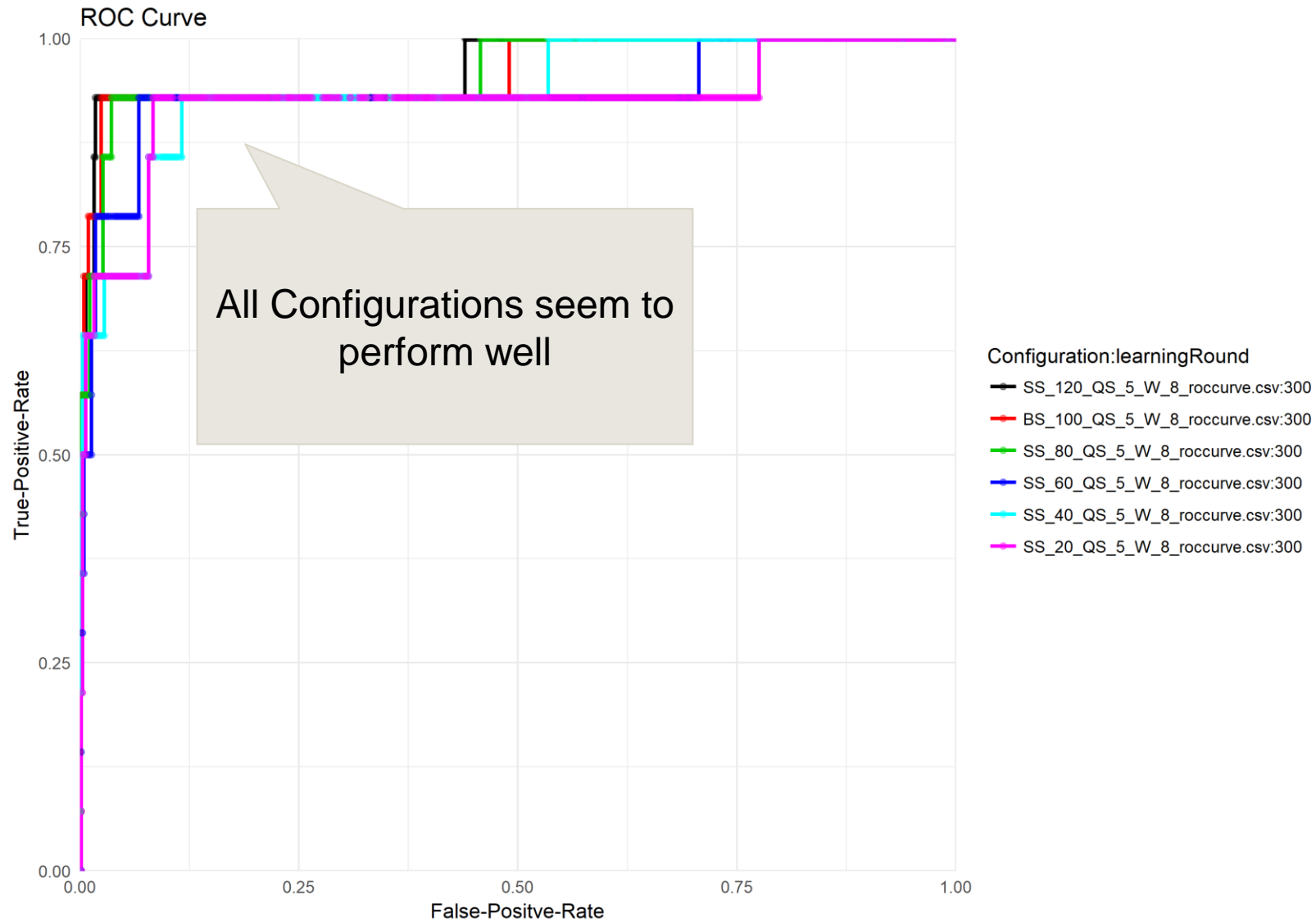
Reasons

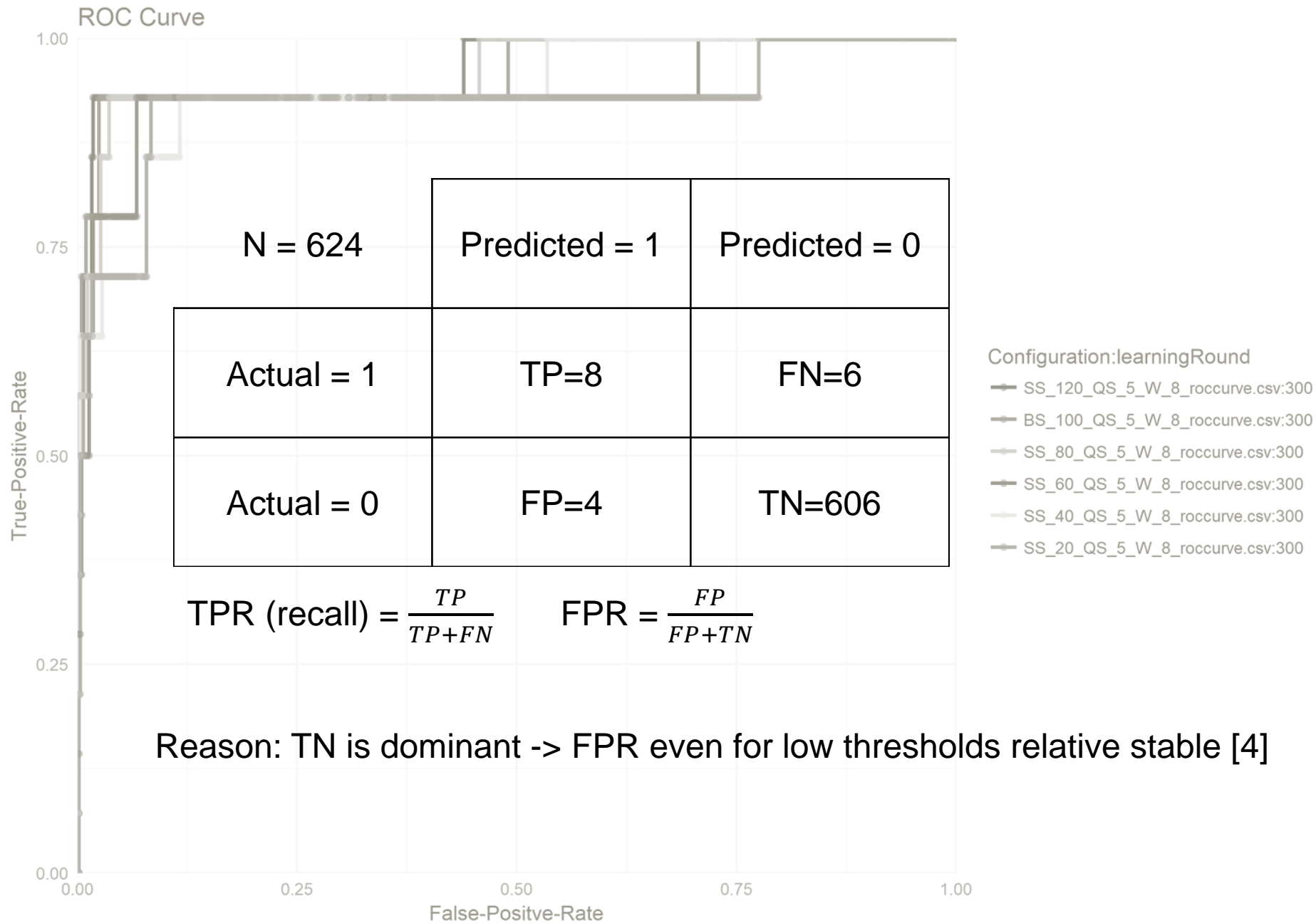
- Small Random Sampled Seed Sets are often not representative
- Assumption BC 3% labeled as True:
 - $P(\text{no True in Seed Set with } n=20) \approx 54\%$
- Unbalanced classes can lead to missed class/cluster effect
- => sufficient exploration phase during the seed set generation is crucial

ROC with unbalanced classes

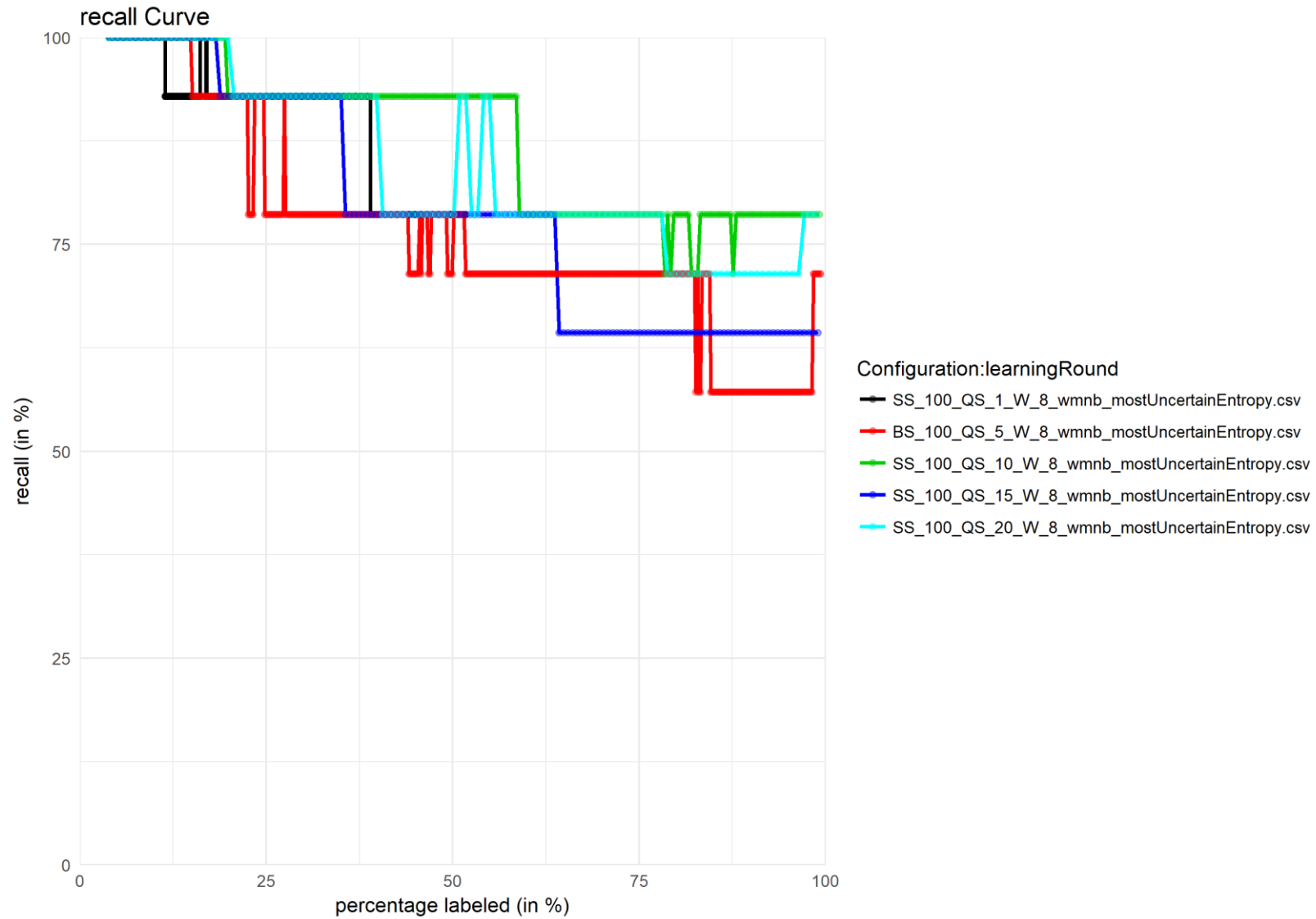


ROC with unbalanced classes

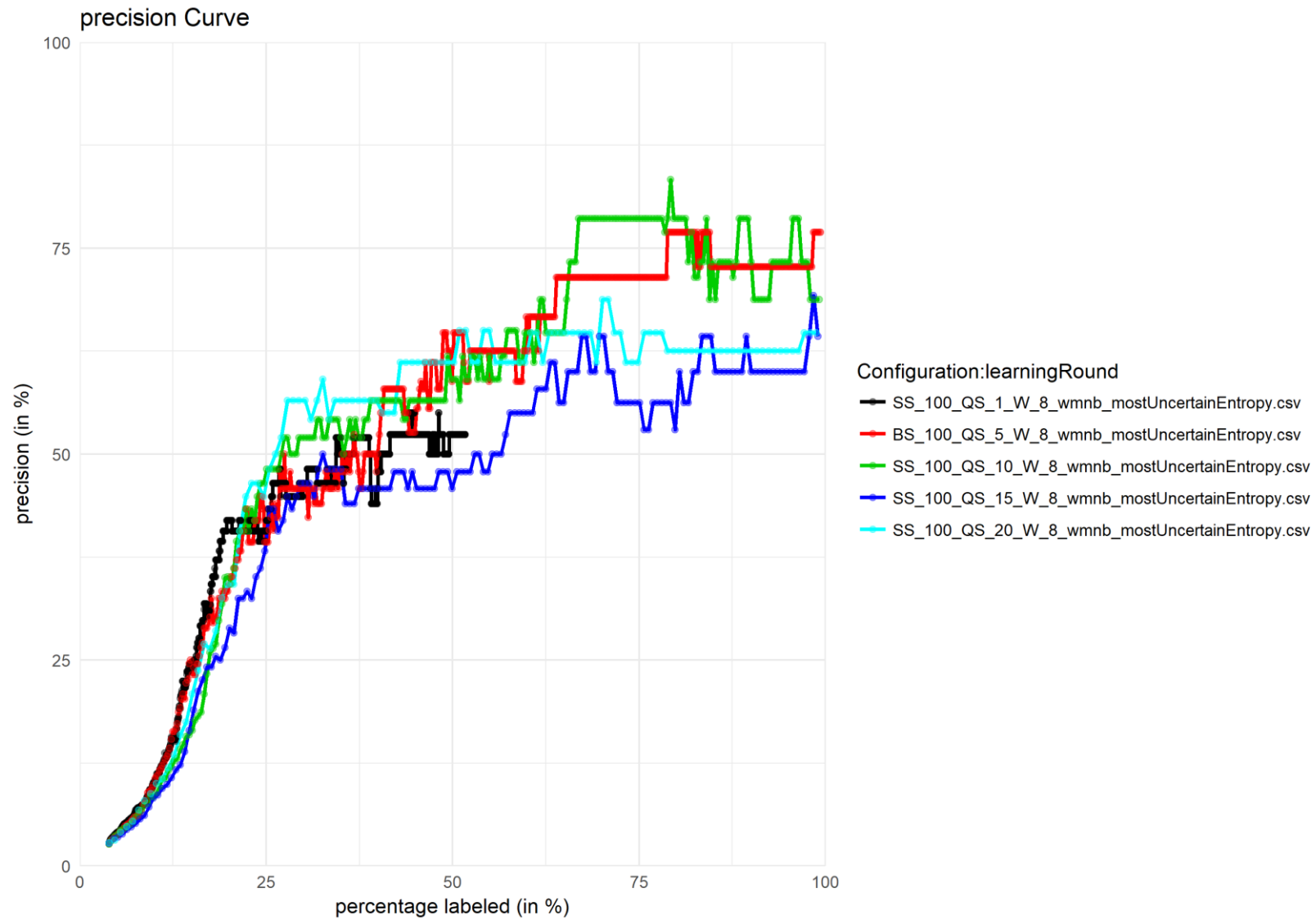




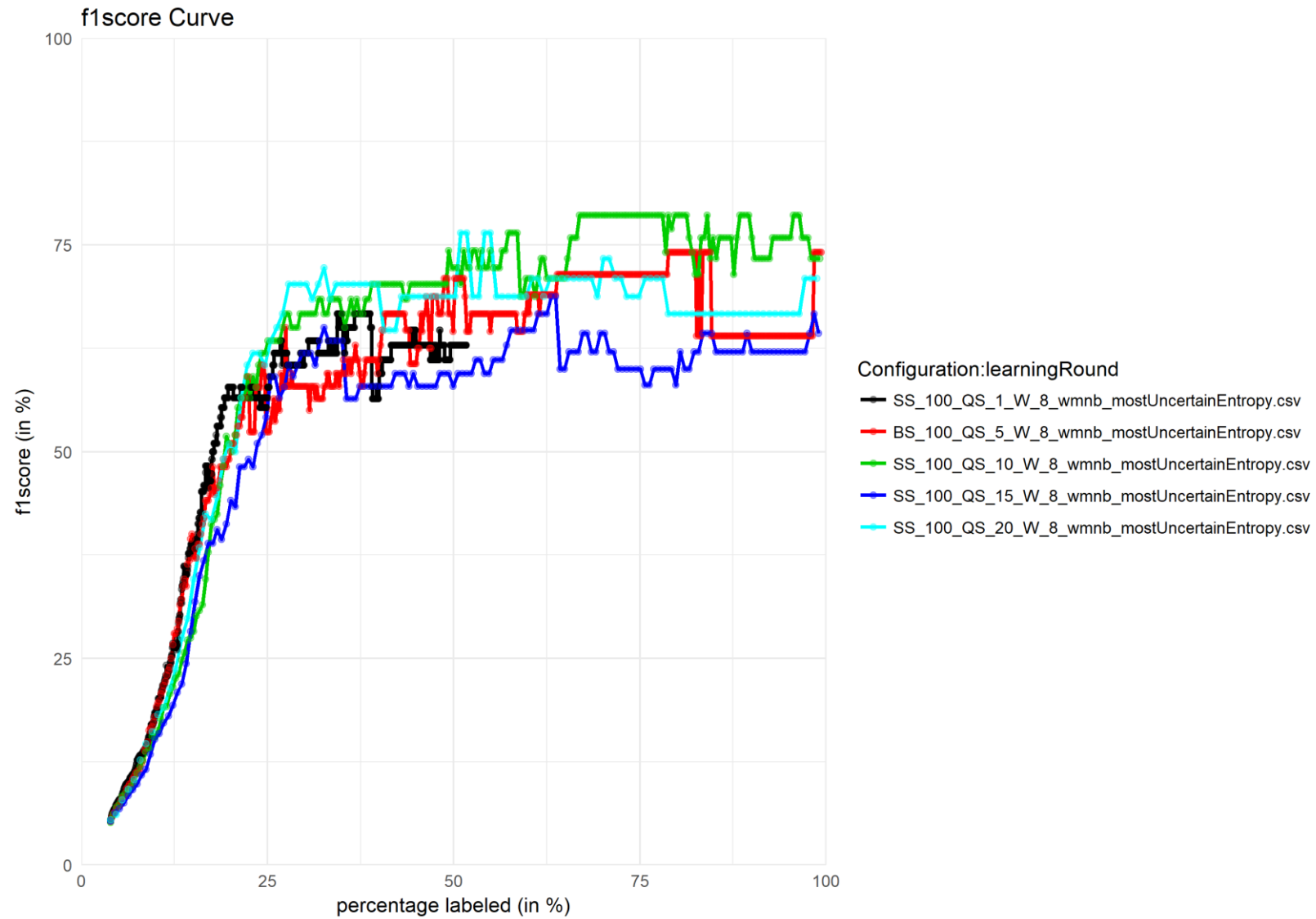
Influence of Query Size



Influence of Query Size

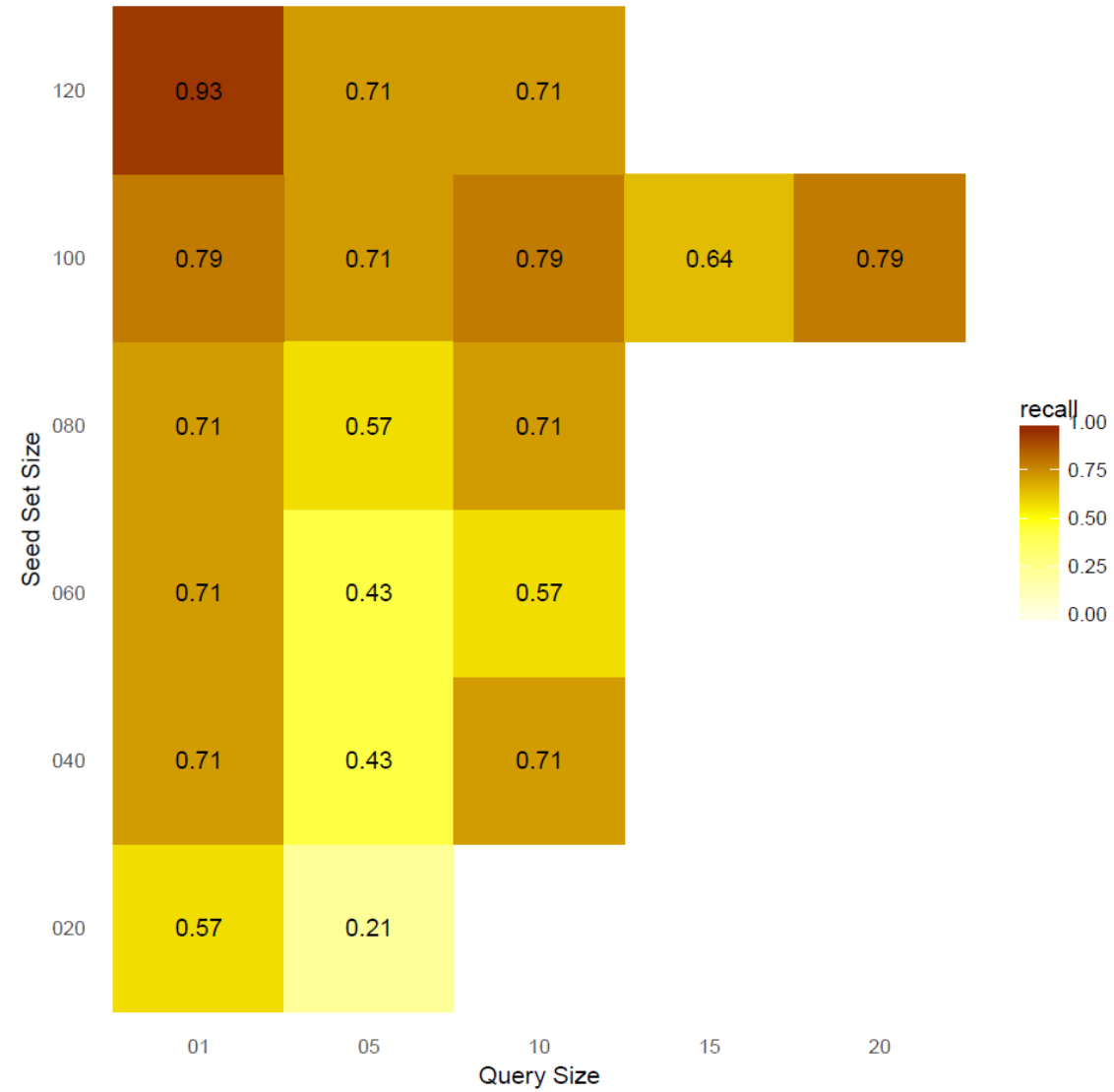


Influence of Query Size



Recall Heatmap

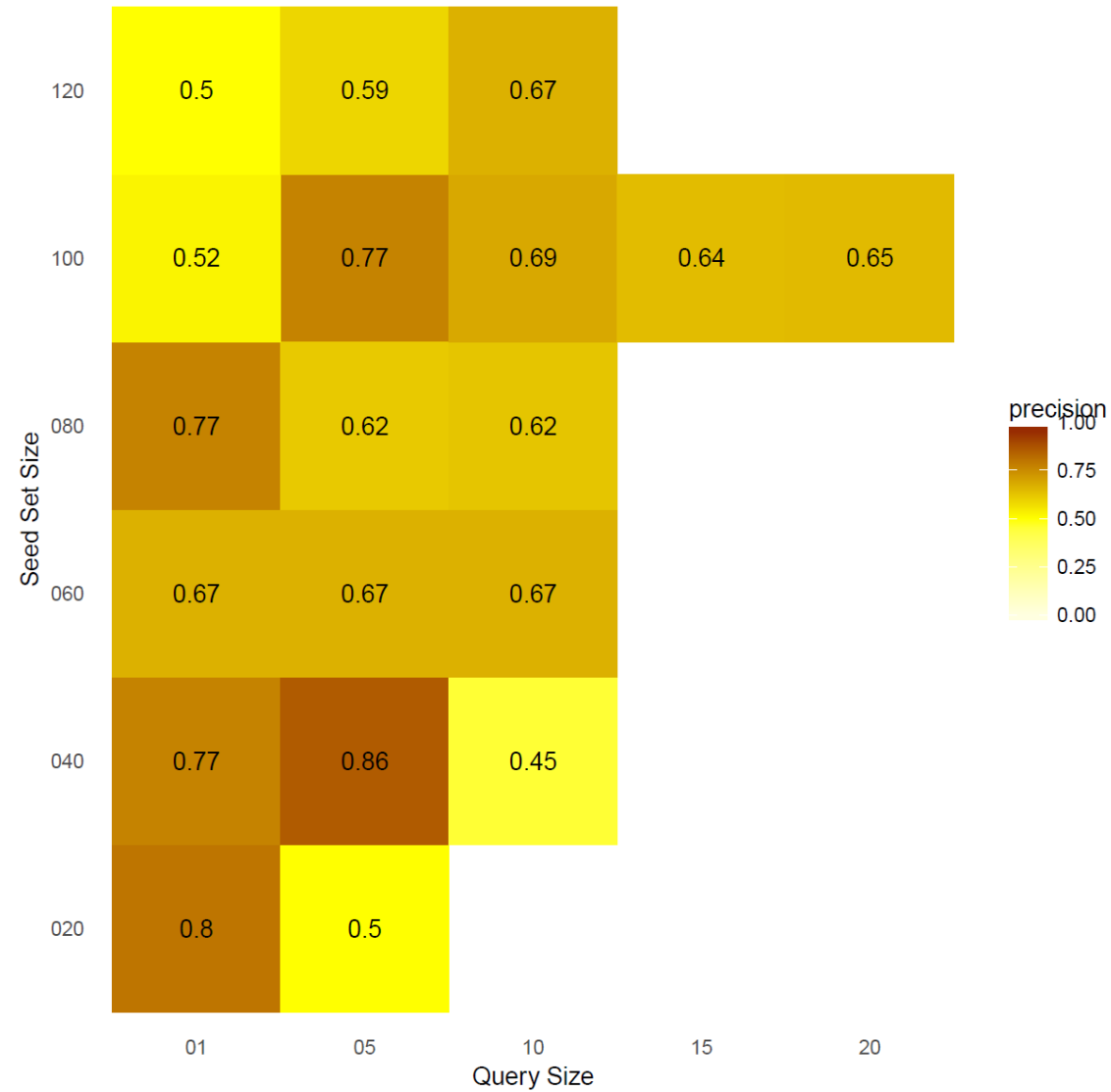
$$\text{Recall} = \frac{TP}{TP+FN}$$



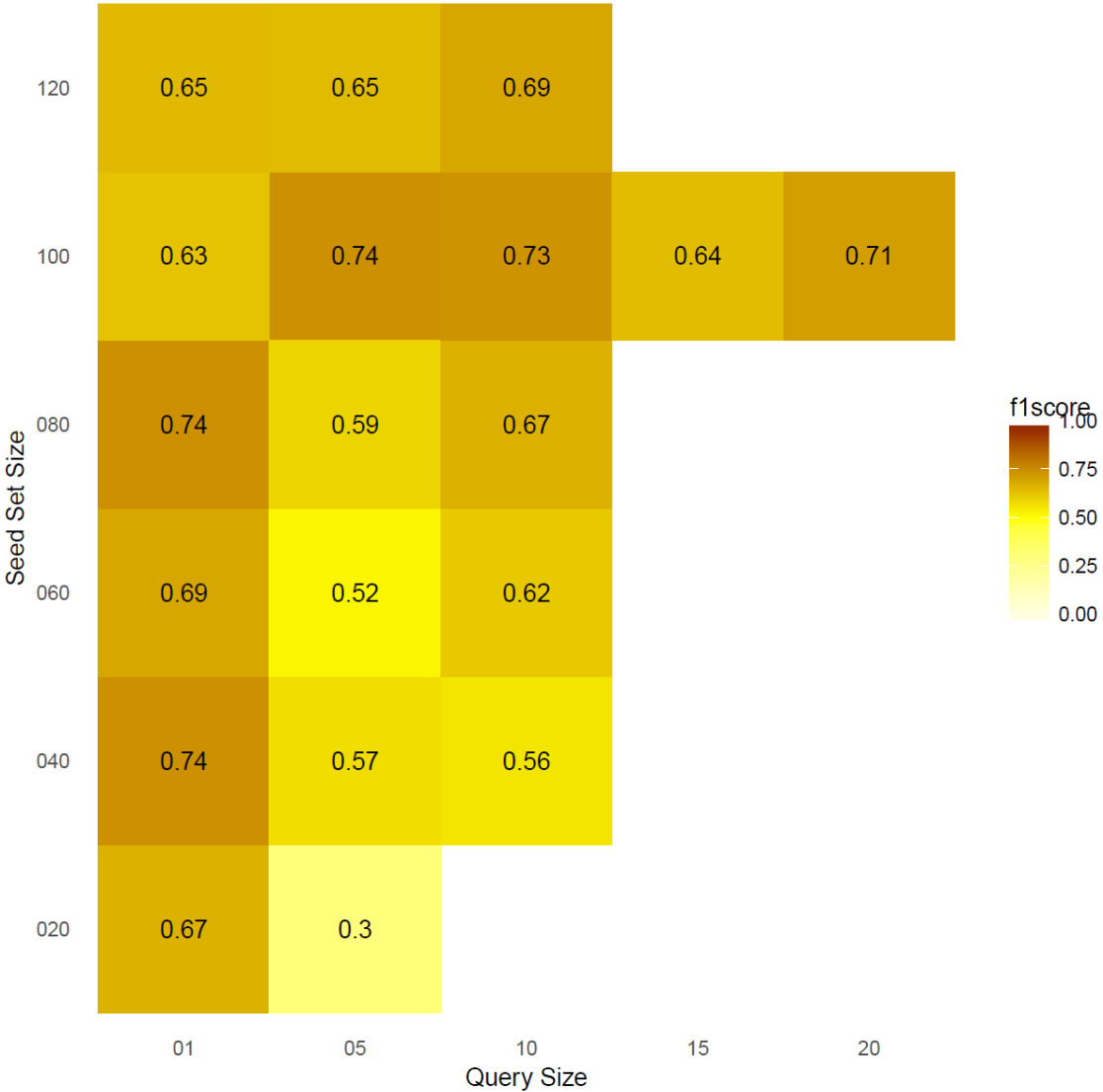
Precision Heatmap

$$\text{Precision} = \frac{TP}{TP+FP}$$

Low FP cause
high precision
even when



F1Score Heatmap



Demo

Conclusion

- Binary Classification is a promising approach for identifying predicates in judgments
- Seed Set Generation and Size is crucial

Limitations

- Evaluation with only one preprocessing configuration
- No Cross-Validation
- Strict Annotation rule misses some sentences



Future Work

- Evaluation of different preprocessing configurations in context of judgments
- Validate results with Cross-Validation
- Try different methods to encounter unbalanced classes
- Use a combination of AML and rule-based learning for a better class balance
- Using different predicates than Sentences like paragraphs or “n-grams” of sentences





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References

- (1) https://www.bundestag.de/blob/194870/7c8a01e16c98fc9c32ddb203d7bd88e0/gesetzgebung_wp18-data.pdf
- (2) Gruner, Ronald H. "Anatomy of a lawsuit." (2008).
- (3) Muhr, Johannes, Master Thesis, Design, Prototypical Implementation and Evaluation of an Active Machine Learning Service in the Context of Legal Text Classification (2017)
- (4) <https://www.kaggle.com/lct14558/imbalanced-data-why-you-should-not-use-roc-curve>

Design Science as most fitting method

- Focus is on the creation of an effective IT solution
- Description of the design problem
- Development an new IT artefact
- Strict evaluation of the IT artefact to measure the utility

