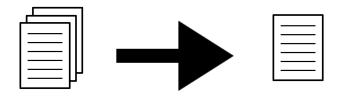
# Automatic Detection of Linguistic Quality Violations

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Bachelor Thesis Defense Universität des Saarlandes 21.08.2014



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- ▶ Multi-Document: Multiple documents on the same topic

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	Single-document	Multi-document
Abstractive		
Extractive		

Summarization systems should produce coherent and grammatical output.

Summarization systems don't produce coherent and grammatical output. Why?

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 $\Rightarrow$  LQVSumm (Friedrich et al., 2014)

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  - pronoun with missing antecedent
  - acronym without explanations
  - **.**..

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- Entity level:
  - definite noun phrase without reference to previous mention
  - pronoun with missing antecedent
  - acronym without explanations
  - **.**..
- Clause level:
  - incomplete sentence (INCOMPLSN)
  - inclusion of datelines (INCLDATE)
  - other ungrammatical form (OTHRUNGR)
  - no semantic relatedness (NOSEMREL)
  - redundant information (REDUNINF)
  - no discourse relation (NODISREL)

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- few occurences of some types
- corpus preprocessing with CoreNLP (Manning et al., 2014)
- unit of annotation (clauses vs. sentences)
- OTHRUNGR has different violation subtypes (annotated by us)

### Development and Test Sets

#### 2 development sets:

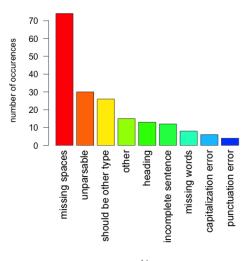
- ► dev-1: 20% (D1101-D1108, 351 summaries)
- dev-2: 20% (D1109-D1116, 352 summaries)

#### 1 test set:

test: 60% (D1117-D1144, 1232 summaries)

# Ungrammaticality (OTHRUNGR+INCOMPLSN) on dev-2





### Detecting missing spaces

"A strong earthquake measuring 7.8 magnitude struck Wenchuancounty of Sichuan Province on Monday, leaving at least 12,000people died and thousands more injured."

"Virginia Tech reported a campus shooting Monday and told studentsto stay inside their residences and away from windows."

"A gunman opened fire on classrooms at Virginia Tech University onMonday morning, killing at least 30 people before turning his gunon himself in the bloodiest school shooting in US history."

#### **UnknownTokens**

#### Idea:

Sentence contains violation iff any word  $\not\in$  known tokens

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#### known tokens?

- Source documents available? → all tokens in source documents = UnknownTokens-source
- ▶ Otherwise → UnknownTokens-general

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- + NER (Finkel et al., 2005) = UnknownTokens-gw+heur+ner
- + Wikipedia = UnknownTokens-gw+heur+ner+wiki
- = UnknownTokens-general

	Missing spaces		
	PRF		
UT-source	95.9	94.6	95.2

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	P R F		
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UT-gw+heur	30.5	97.3	46.5

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UT-gw+heur+ner	35.5	97.3	52.0

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UT-gw+heur+ner+wiki	70.3	96.0	81.2

RandomForest (Breiman, 2001) to train decision trees

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#### Features:

classification from UnknownTokens

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- ▶ 3 features from HPSG parser output (Packart, 2011; Copestake, 2002):
  - number of readings
  - RAM usage
  - status

#### RandomForest: Evaluation on test

#### 10-fold cross validation:

	Precision	Recall	F-Score	Accuracy
Baseline	_	0	_	77 7
(Most frequent class)	_	U	1	11.1
Random Forest	72.8	49.1	58.6	84.5

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#### **Ablation Study:**

Feature	Decrease in F-Score
UnknownTokens Output	0.8
Language Model Perplexity	1.4
Number of Words	2.2
Parser RAM	1.3
Parser Readings	5.7
Parser Status	0.4

# Datelines (INCLDATE)

**BLACKSBURG, Virginia 2007-04-16 18:34: 44 UTC** A gunman opened fire in a dorm and classroom at Virginia Tech on Monday, killing at least 30 people in the deadliest shooting rampage in U.S. history.

**BERLIN, May 13( Xinhua)** The German government announced on Tuesday that it is to provide 500, 000 euros( around 770, 000 U.S. dollars) in aid for earthquake victims in Sichuan Province of China.

**00** a.m. People are panicking.

# Detecting Datelines

```
Regular expression:

UTC|

^\d{4}-\d{2}-\d{2}|

^[A-Z]{3,}|

^(Jan|Feb|Mar|Apr|May|Jun|Jul|Aug|Sep|Oct|Nov|Dec)
```

## **Detecting Datelines**

Regular expression:

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#### **Evaluation on** test:

Precision	Recall	F-Score
86.0%	89.7%	87.8%

# Redundancy (REDUNINF)

Cyclone Sidr, described as the worst storm in years to hit low-lying and disaster-prone Bangladesh, crashed into the southwestern coast Thursday night before sweeping north over the capital Dhaka.

The cyclone hit the southwestern coast of Bangladesh on Thursday before sweeping north to the capital Dhaka.

Mary saw the 5 "elephants". She saw the horses.

```
\{\mathit{Mary}, \mathit{saw}, \mathit{the}, 5, \mathit{elephants}\}, \{\mathit{She}, \mathit{saw}, \mathit{the}, \mathit{horses}\}
```

 Remove non-alphanumeric characters and split into set of words

$$|\{\mathit{saw},\mathit{the}\}|=2$$

- Remove non-alphanumeric characters and split into set of words
- Cardinality of intersection between sets

$$\textit{score} = \frac{2}{|\{\textit{She}, \textit{saw}, \textit{the}, \textit{horses}\}|} = 0.5$$

- Remove non-alphanumeric characters and split into set of words
- Cardinality of intersection between sets
- Normalize by maximum overlap

#### 0.5 > threshold?

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Variations: Bigrams, Combined

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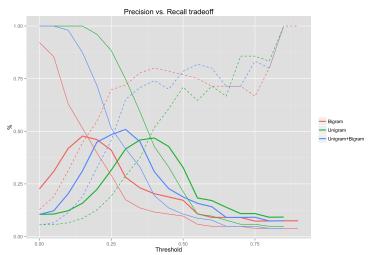
Variations: Bigrams, Combined

Threshold?



# Finding a threshold





# Evaluation of **Unigrams**, ... on *test*

	Unigrams	Bigrams	Combined
Threshold	0.5	0.4	0.4

	Precision	Recall	F-Score
Baseline	4.5%	100%	8.7%
Levenshtein	15.8%	17.3%	3.1%
Unigrams	58.0%	28.2%	37.0%
Bigrams	55.6%	14.5%	22.9%
Combined	56.8%	24.3%	34.0%

## Redundancy: False Negatives

According to a survey by the State Food and Drug Administration, 65 percent of the respondents worried about the food safety situation in China.

**Food and drug safety** has become a major concern of Chinese people.

## Conclusion

#### Best methods:

	Precision	Recall	F-Score
Ungrammaticality	72.8	49.1	58.6
Datelines	86.0	89.7	87.8
Redundancy	58.0	28.2	37.0

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Adapted annotation scheme, better for automatic processing

Tool will be made available to annotate with our methods

#### Other violations

- pronouns: coreference resolution
- acronyms: finding full form near first unexpanded form
- mentions & noun phrases: NER + ?
- ▶ no semantic relatedness: semantic parsing? Wordnet distance?
- no discourse relation: discourse parsing, does connective match relation?

### Ungrammaticality

detection methods for other subtypes

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### Redundancy

- include contextual information
- include source document information
- semantic approaches

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detection methods for other subtypes

### Redundancy

- include contextual information
- include source document information
- semantic approaches

### Corpus

- annotate a corpus with subtypes, sentence based
- evaluate methods on other data sets/corpora/domains

### References

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## Bonus Slide: Full **UnknownTokens** Evaluation

	Missing spaces		No missing spaces			
	Р	R	F	Р	R	F
Baseline	0.0	0.0	0.0	94.8	100	97.3
UT <sub>gw</sub>	15.0	98.7	26.0	99.9	69.1	81.7
UT <sub>gw+heur</sub>	30.5	97.3	46.5	99.8	87.8	93.4
UT <sub>gw+heur+ner</sub>	35.5	97.3	52.0	99.8	90.3	94.8
UT <sub>gw+heur+ner+wiki</sub>	70.3	96.0	81.2	99.8	97.8	98.8
UT <sub>source</sub>	95.9	94.6	95.2	99.7	99.7	99.7

## Bonus Slide: RandomForest: Evaluation of all classes

	Precision	Recall	F-Score
Ungrammatical	72.8	49.1	58.6

## Bonus Slide: RandomForest: Evaluation of all classes

	Precision	Recall	F-Score
Ungrammatical	72.8	49.1	58.6
Not ungrammatical	86.6	94.7	90.5
Weighted Average	83.5	84.5	83.4