

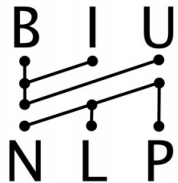
# Factuality Prediction over Unified Datasets

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# Factuality

## Task Definition

*Author's commitment towards a proposition*

- **Factual**
  - It is *not surprising* that **the Cavaliers lost the championship**
- **Uncertain**
  - She still *has to check* whether **the experiment succeeded**
- **Counter-factual**
  - Don was *dishonest* when **he said he paid his taxes**
- Useful for
  - Knowledge base population
  - Question answering
  - Recognizing textual entailment

# In this talk

- **Problem: Limited Generality**

- Previous work focused on *specific* flavors of factuality

- **Approach**

- Build a unified dataset
- Train a new model

- **Contributions**

- Normalized annotations
- Large aggregated corpus
- Improving performance across datasets

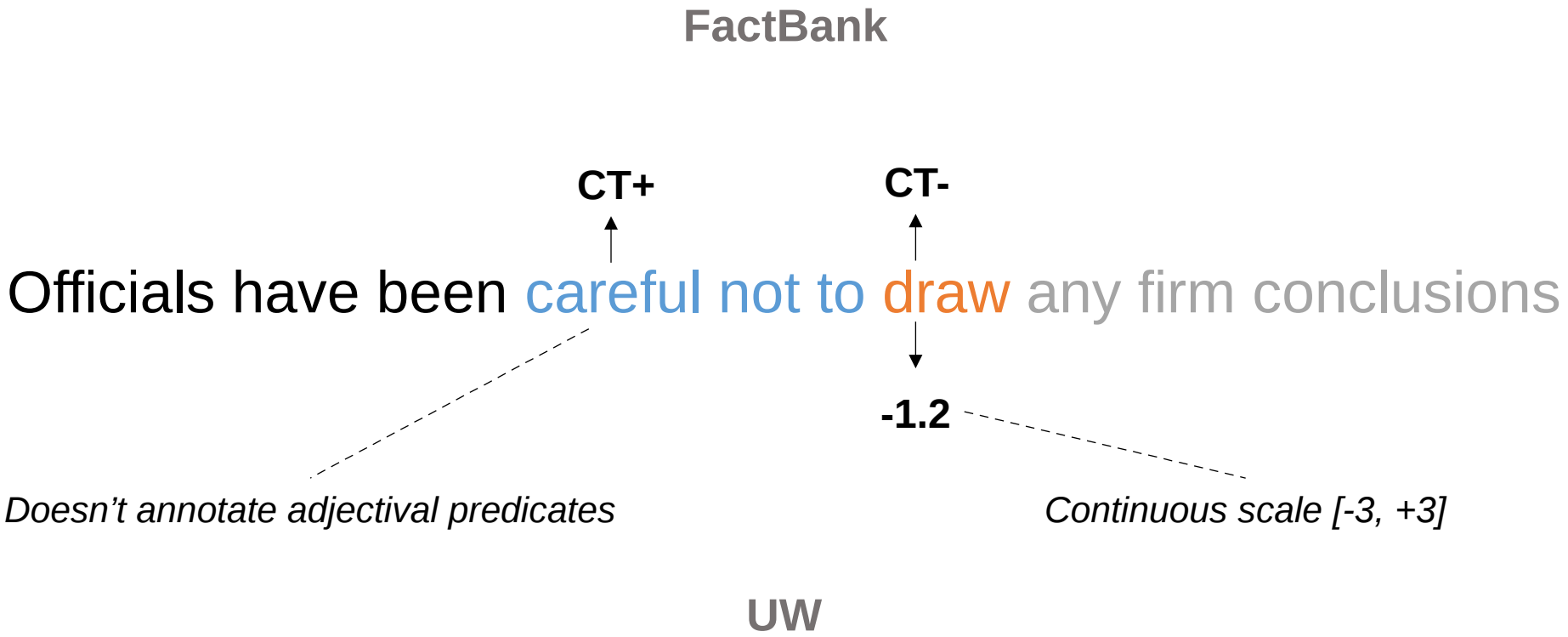
**Problem:**  
Limited Generality

# Datasets

- Many annotation efforts
  - FactBank (Saurí and Pustejovsky, 2009)
  - UW (Lee et al., 2015)
  - Meantime (Minard et al., 2016)
  - ... and more
- Datasets differ in various aspects
  - **Discrete vs. continuous** values
  - **Expert vs. crowdsourced** annotation
  - **Point of view**

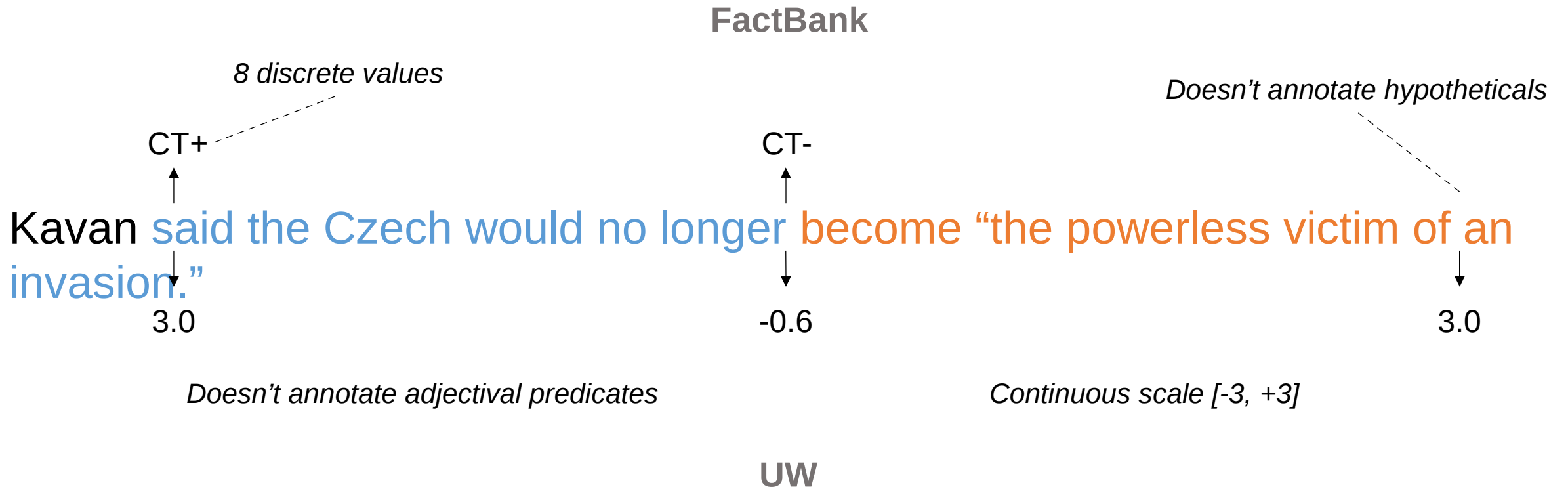
# Annotated Examples

FactBank vs. UW



# Annotated Examples

FactBank vs. UW



# Previous Work: Factuality Prediction

- Models were designed and evaluated on *specific* datasets
- For example, Lee et al. (2015):
  - Used SVM on syntactic features
  - lemma, POS, dependency paths
  - Tested on the UW corpus

→ **Non-comparable results**

→ **Limited portability**

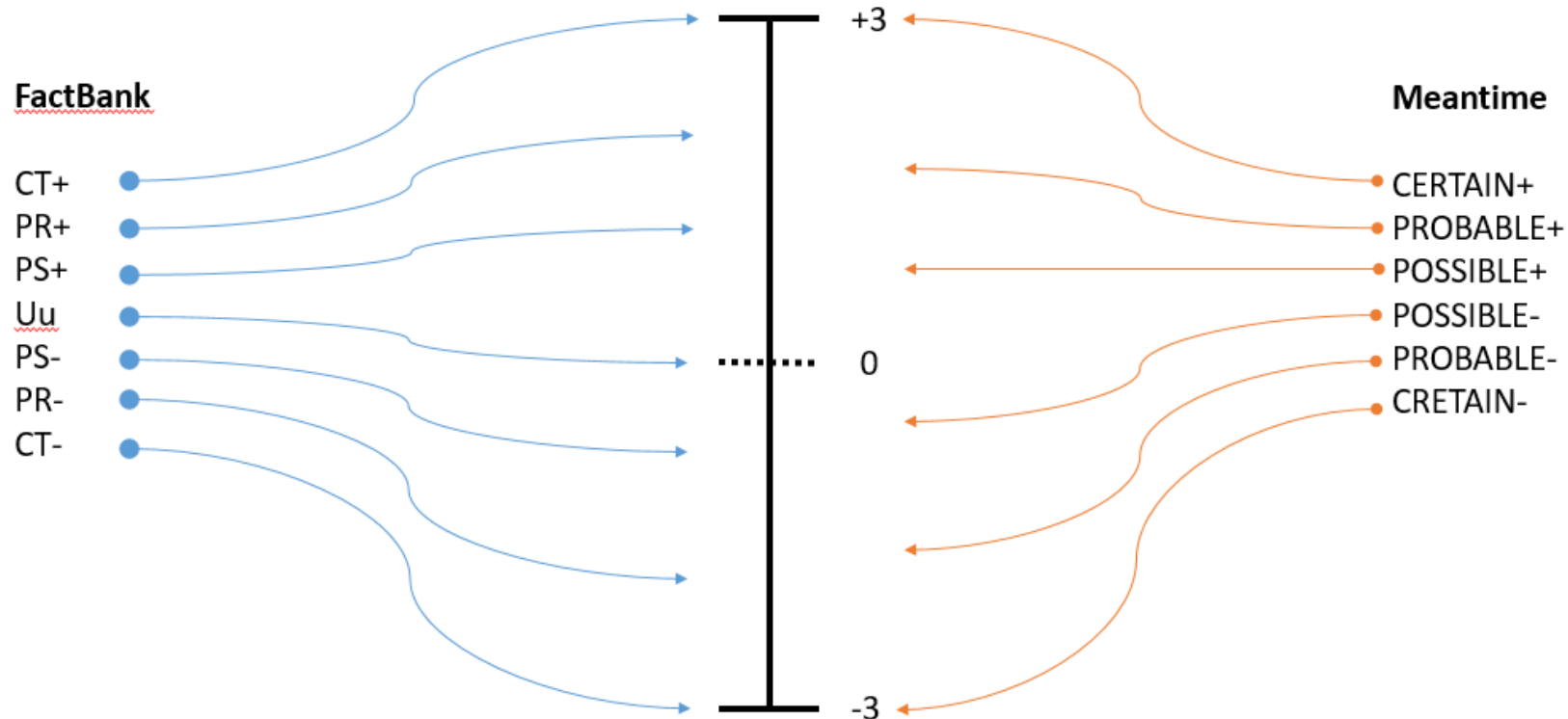


# **Solution:** **Unified Corpus**

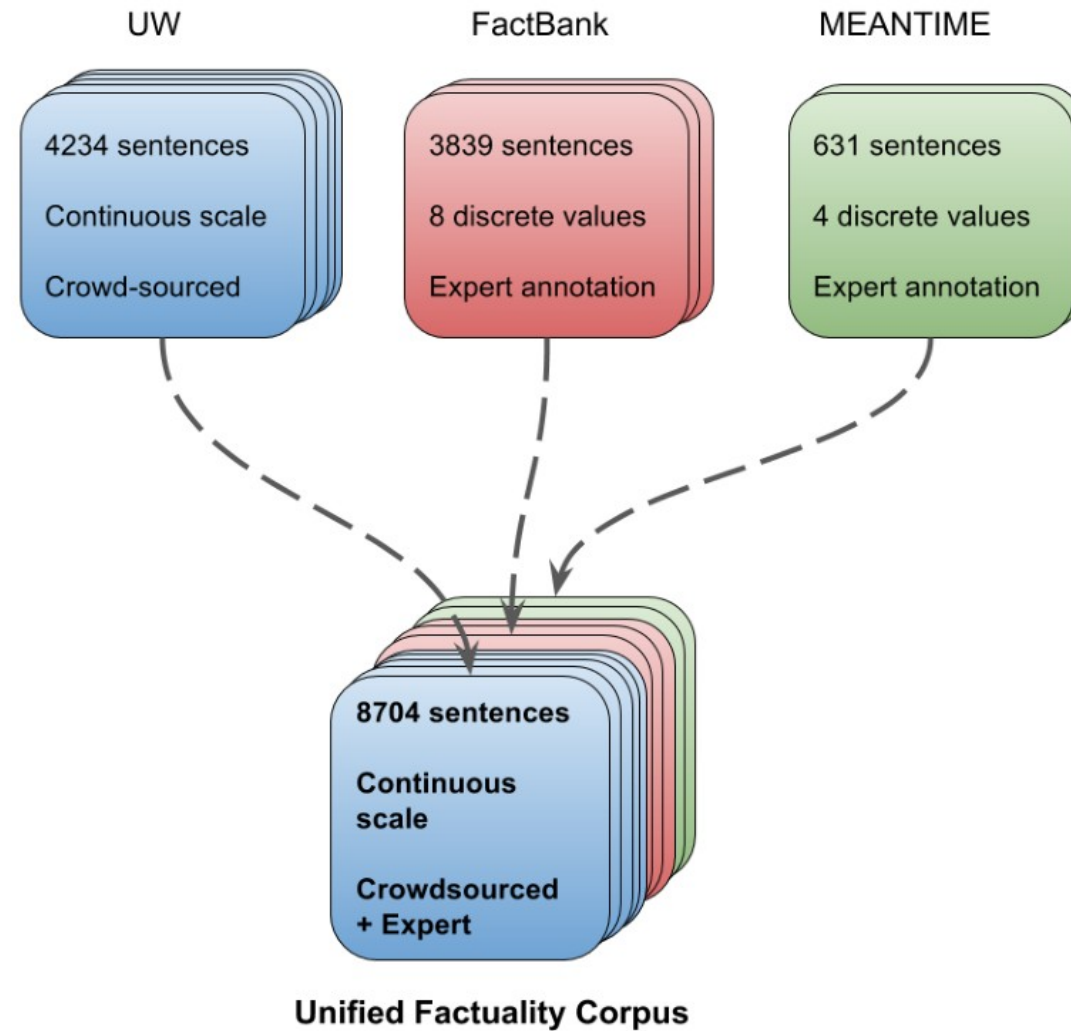
Extending TruthTeller  
Evaluation

# Simple Normalization

- Mapping discrete values to the continuous UW scale
  - Simple mapping based on overlapping annotations

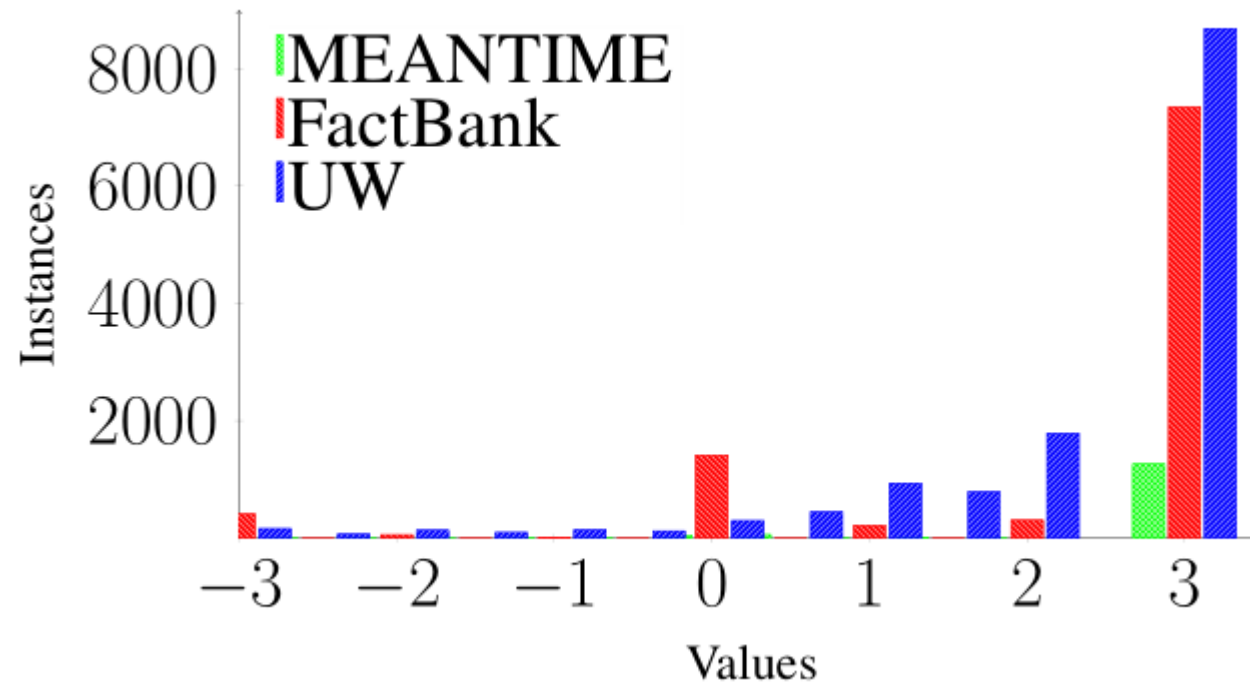


# Unified Factuality Corpus



# Biased Distribution

- Corpus skewed towards factual
- Inherent trait of the news domain?



**Solution:**

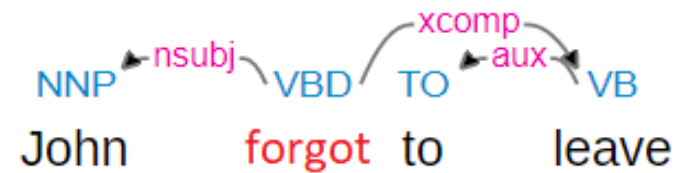
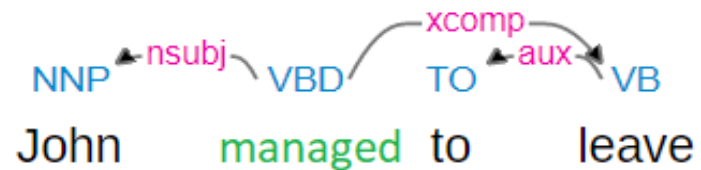
Unified Corpus

**Model: Extending TruthTeller**

Evaluation

# TruthTeller (Lotan et al., 2013)

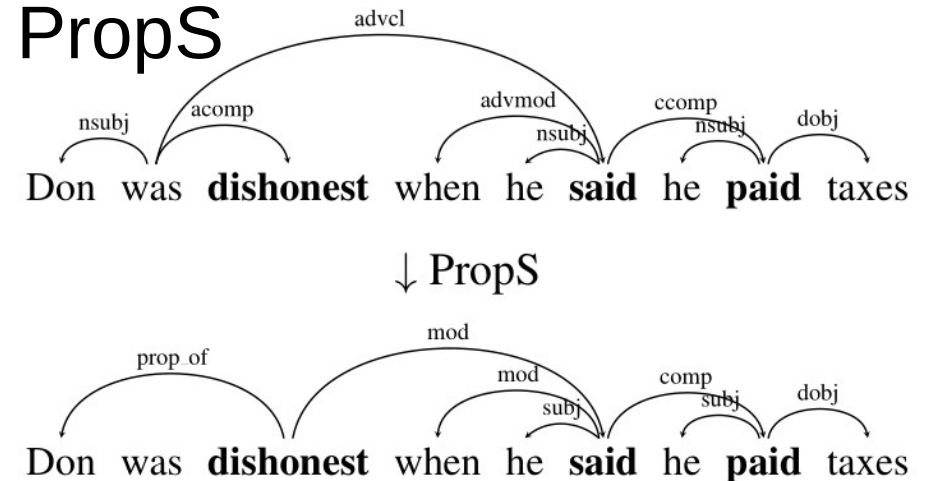
- Rule based approach on dependency trees
  - Karttunen implicative signatures
  - Syntactic cues (modality, negation, etc.)



- Hand-written lexicon of 1,700 predicates

# Extending TruthTeller

- Semi automatic extension of lexicon by 40%
  - Translated from German verb classes (Eckle-Kohler, ACL 2016)
- Supervised learning: **TruthTeller as signal**
- Application of implicative signatures on PropS



# **Solution:**

Unified Corpus

Extending TruthTeller

# **Evaluation**



# Metrics (lee et al., 2015)

## 1. Mean Absolute Error

- Range: [0, 6]
- *Smaller is better!*

## 2. Pearson Correlation

- How good is a system in recovering the variation
- Well-suited for the biased news domain

# Evaluations

Dataset	FactBank		UW		MEANTIME	
	MAE	<i>r</i>	MAE	<i>r</i>	MAE	<i>r</i>
<b>All-factual</b>	.80	0	.78	0	<b>.31</b>	0
<b>UW feat.</b>	.81	.66	.51	<b>.71</b>	.56	.33
<b>AMR</b>	.66	.66	.64	.58	.44	.30
<b>Rule-based</b>	.75	.62	.72	.63	.35	.23
<b>Supervised</b>	<b>.59</b>	<b>.71</b>	<b>.42</b>	.66	.34	<b>.47</b>

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Marking all propositions as factual  
Is a strong baseline on this dataset

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Dependency features correlate well

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Applying implicative signatures on AMR did not work well

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Hard coded rules aren't robust  
Enough across datasets

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Our extension of TruthTeller gets good results across all datasets

# Conclusions and Future Work

- Resources made publicly available
  - Unified Factuality corpus
  - Conversion code and trained models
- Future work
  - Annotate diverse domains
  - Integrate TruthTeller with more lexical-syntactic feats.
- Try our online demo:  
<http://u.cs.biu.ac.il/~stanovg/factuality.html>

**Thanks for listening!**