# Frame-Based Continuous Lexical Semantics through Exponential Family Tensor Factorization and Semantic Proto-Roles

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#### Understanding Participating in Actions

Word embeddings robustly capture word similarity & associativity. What does it mean for "the paper" to "reflect the truth?" And how do we represent it?

The paper reflected the truth.

#### Frame Semantics Provide Structured Word Meaning

Argument

Understanding through defined and structured concepts (Minsky '74; Fillmore '76, '82)

Examples: FrameNet and PropBank

"reflect" triggers Communication frame

ROLE

Meaning

The conveyer of Message The paper COMMUNICATOR A set of beliefs to convey the truth MESSAGE How Message is conveyed MEDIUM

#### Semantic Proto-Roles Decompose Categorical Roles

Dowty '91, Property, re: "reflect" The paper the truth. Reisinger et al. '15: **A**WARENESS Describe semantic MANIPULATED arguments as properties of SENTIENT participating VOLITIONAL in an action

# Frame Embeddings from Tensor Factorization

Further generalize Cotterell et al. '17's 3-tensor factorization

$$p(t_i, w_j, c_1, ..., c_K) \propto \exp(\mathbf{1}^{\mathsf{T}} (t_i \odot w_j \odot c_1 \odot \cdots \odot c_K))$$

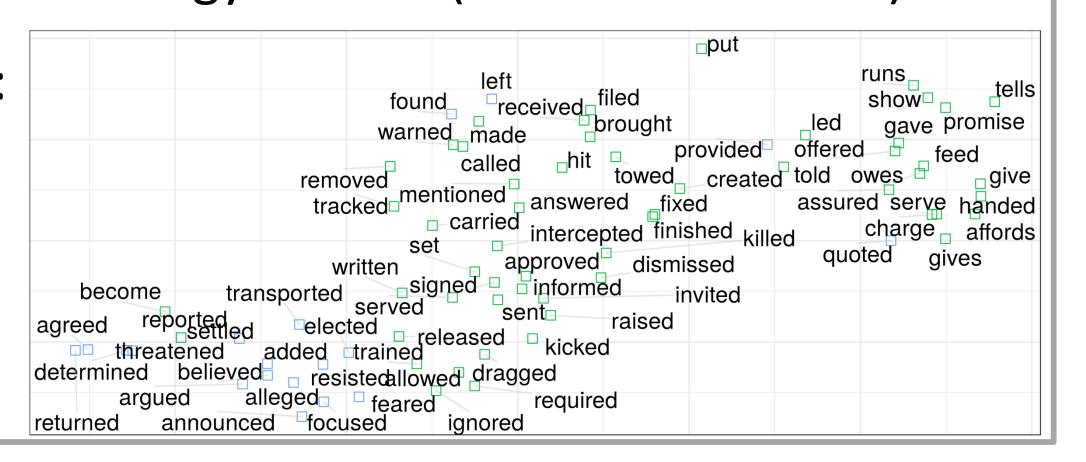
(7)/fmof/tensor-factorization

### **Evaluating Attributive Embeddings**

**QVEC:** Correlate learned and oracle ontology vectors (Tsvetkov et al. '15)

Use SPR annotations as ontology:

- 1. Properties & syntax are coordinates: 80 total
- 2. SPR Likert  $\rightarrow$  0-5 rating
- 3. Sum & normalize over ratings



#### Extracting Frame Counts from Large Corpora

1. Record every word triggering FRAMES... 2. Each ROLE's fillers... 3. FRAMES and ROLES... 4. And context

 $ARG_1$ REFLECT.01 ARG<sub>2</sub> the truth. reflected The paper COMMUNICATOR COMMUNICATION MESSAGE 1 to the *left* 2 to the right

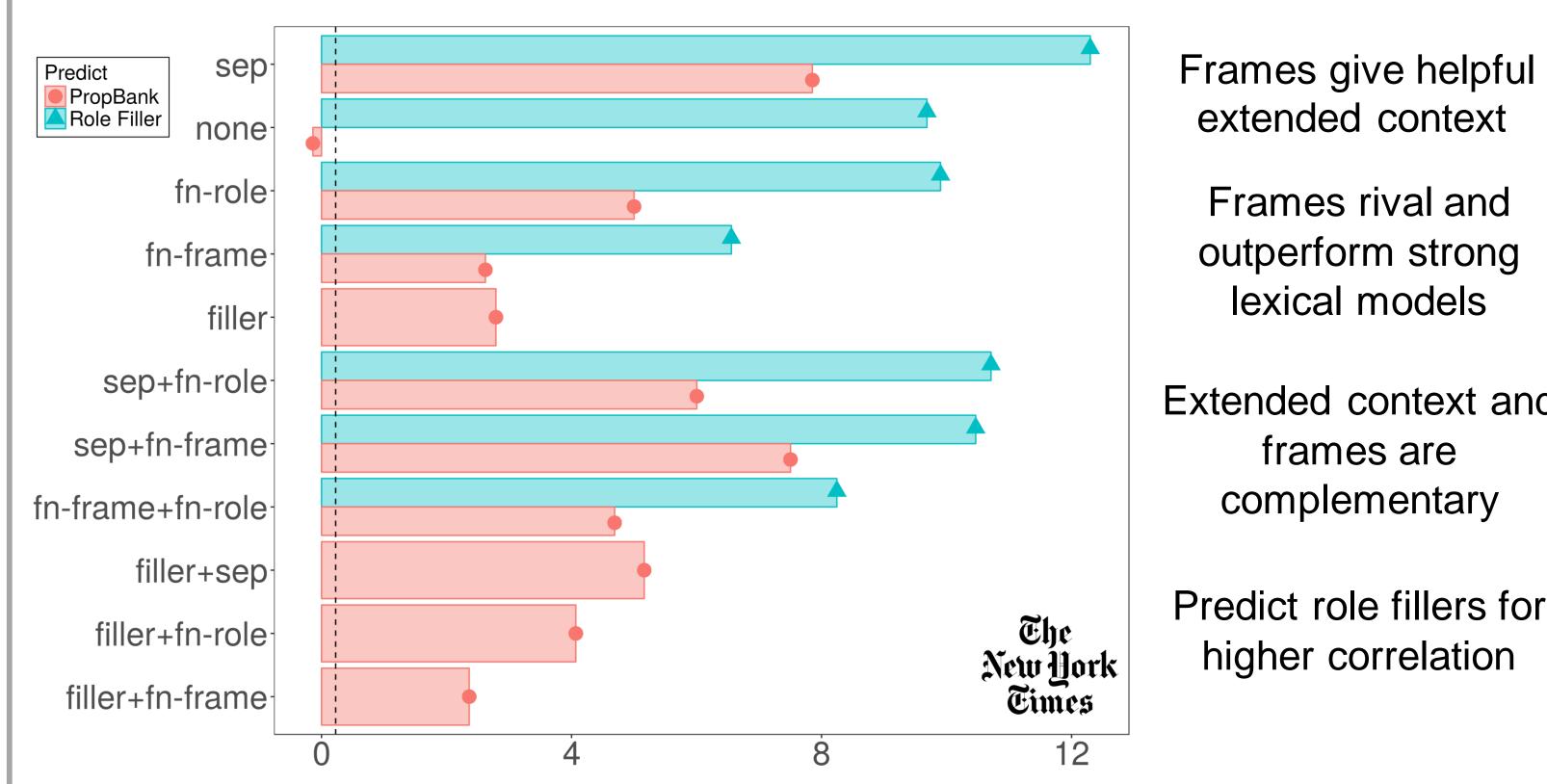
role filler PropBank frames & roles FrameNet frames FrameNet roles Distance context

Extract counts from Concretely Annotated Corpora: 15M+ documents annotated with three semantic parses (Ferraro et al. '14; <a href="http://hltcoe.github.io/">http://hltcoe.github.io/</a>)





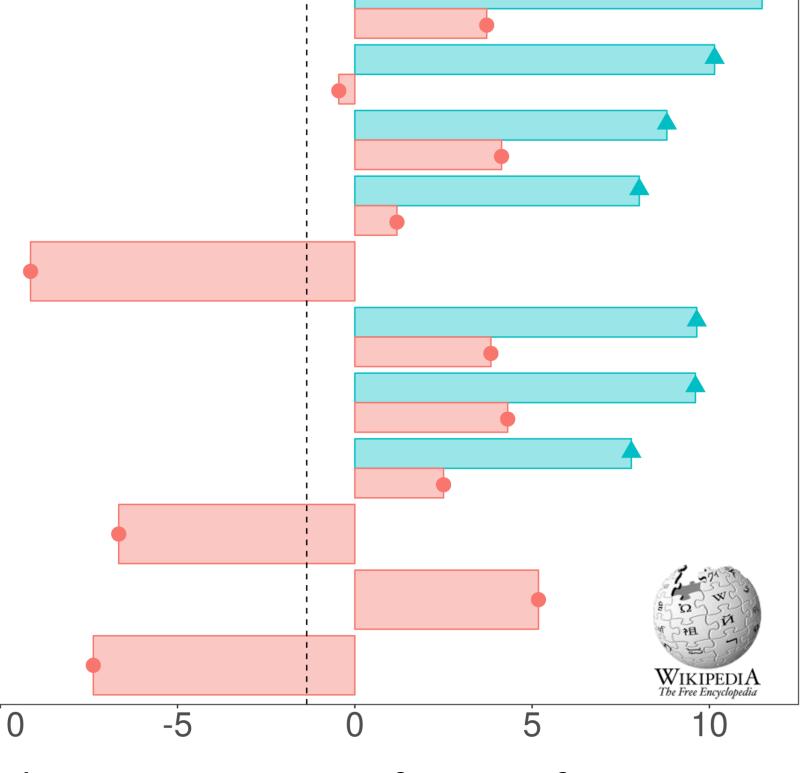
## Higher SPR Correlations in Learned Frame Trigger Embeddings



Frames rival and outperform strong lexical models

Extended context and frames are complementary

Predict role fillers for higher correlation



Percent Improvement over word2vec and tensor factorization without any semantic frame information

# Learning Paraphrases and Inflectional Relations

Filler | sep

Under three newswire models, what triggers are most similar to anticipated?

1 foresaw 2 figuring 3 alleviated 4 craved 5 jeopardized

6 pondered 7 kidded 8 constituted 9 uttering 10 forgiven

6 rubbed 1 aggravate 2 scouts 7 analyzes 3 jams 8 jailed 9 discerned 4 appealing 5 provoking

Filler | sep + fn-role

2 anticipating 3 anticipates 4 stabbing 10 halved 5 separate

6 intimidated 1 anticipate 7 separating 8 separates 9 drag 10 guarantee

PropBank | sep

#### Conclusion

Semantic frames obtained from large, disparate corpora can be used to learn enriched word vectors resulting in higher semantic proto-role based correlations.



