

# Mapping Unseen Words to Task-Trained Embedding Spaces

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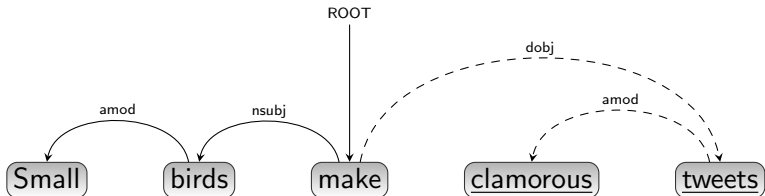
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## Consider an example for syntactic parsing:

- ▶ Setting: Supervised neural network based model → uses and updates word embeddings.
- ▶ '*clamorous*' and '*tweets*' are rare or unseen words.



- ▶ Problem: What embeddings to use for '*clamorous*', '*tweets*'? Their default embeddings don't match the task-trained embedding space!

## Proposal:

- ▶ A neural network based mapper that maps initial embeddings to the task-trained embeddings space.
- ▶ We propose a weighted multiple-loss criterion.
- ▶ Our architecture allows to have some control over different domains by tuning hyperparameters.

## Mapper Architecture

- ▶ Basic Architecture:

$$G(e_i^o) = W_2(\text{hard tanh}(W_1 e_i^o + b_1)) + b_2$$

- Multi-loss Criterion:

$$\text{loss}(y, \hat{y}) = \alpha \sum_{j=1}^n |y_j - \hat{y}_j| + (1 - \alpha) \sum_{j=1}^n |y_j - \hat{y}_j|^2$$

- ▶ Regularization

$$F(\theta) = L(\theta) + \lambda_1 \|\theta\|_1 + \frac{\lambda_2}{2} \|\theta\|_2^2$$

# The Pipeline

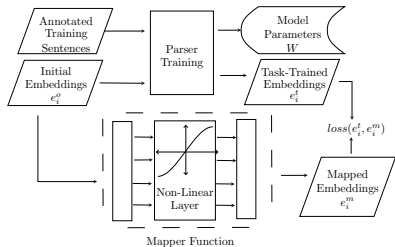


Figure: Mapper training

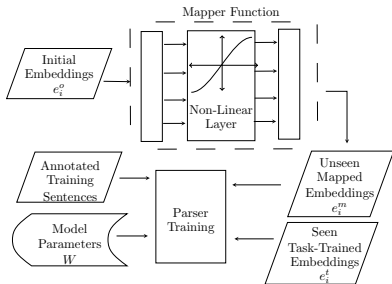
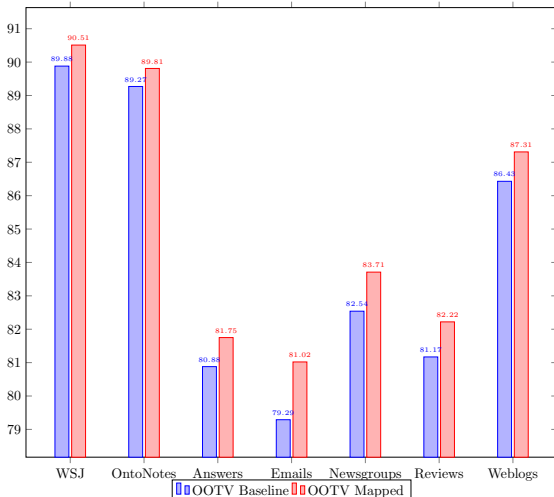


Figure: Parsing with Mapped Embeddings

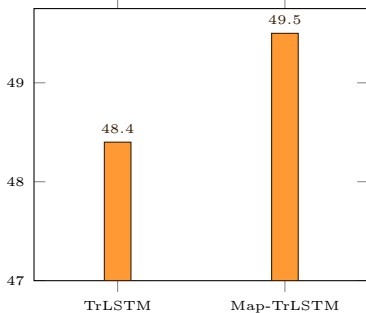
# Results on Dependency Parsing

Dependency Parsing experiments with WSJ and Web Treebank

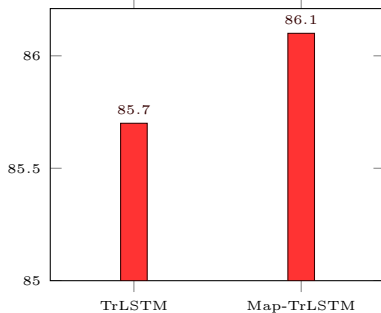


## Results on Downstream Task

Experiments on Downstream Task - Sentiment Analysis Task using Dependency TreeLSTM

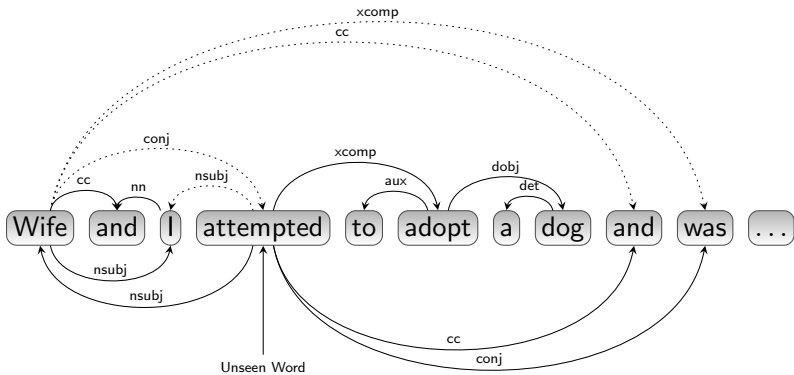


 Fine-grained Classification



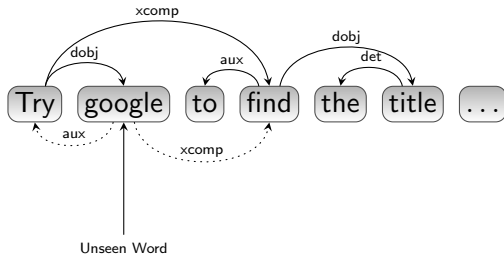
 Binary Classification

## What Works





## What Doesn't



## Observations

- ▶ We achieve significant improvements in dependency parsing accuracy across several domains.
- ▶ We also observe improvements over a downstream task that uses dependency output.
- ▶ Our approach is simple, effective, and applicable to various other settings.

Thanks