# Feature Selection for Cross-Linguistic Parse Ranking

Comparison of Parse Ranking Accuracy for English and Japanese

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## Research Questions

- What are the parse selection accuracies in different languages for different feature sets?
- Do the same feature sets work for different languages?

## Languages and Features

#### Languages

- English jhpstg corpus, ERG grammar
- Japanese Tanaka corpus, JACY grammar

#### **Features**

- Grandparenting 0
- Active Edges true, false
- Constituent Weight 1, 2, 0
- N-gram 3,4
- N-gram Backoff true, false

#### **Evaluation Metric**

1-best exact match

#### Raw Results

#### Fixed dimensions

- Grandparenting 0
- Relative Tolerance  $1 \times 10^{-8}$
- Variance 1

#### Best Results

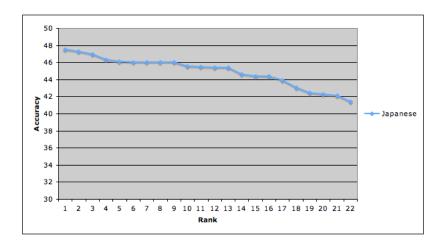
- Japanese 47.46
- CW=2, AE=false, N-gram=3, N-gram Backoff = true
- English 37.13
- CW=2, AE=false, **N-gram=4**, N-gram Backoff = true



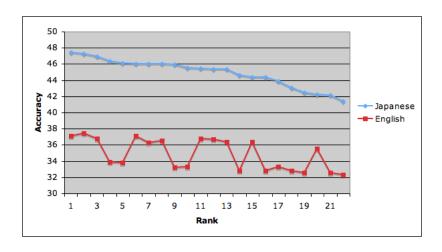
### Cross-Linguistic Feature Sets

- For a single language just pick the best feature set
- How do you quantitatively navigate the feature space for more than one language?
- Which feature subsets make two languages the most dissimilar?

# Japanese and English Accuracy



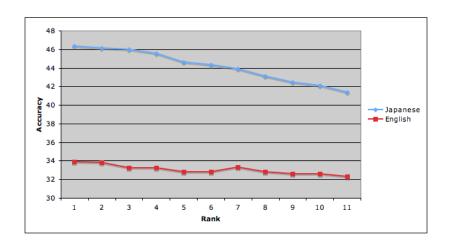
# Japanese and English Accuracy



# Relative Monotonicity

- Japanese is monotonically decreasing by construction
- Which data points can I remove from English to make it also monotonic?

# Japanese and English Accuracy, Active Edges = false



# Quantifying Correlation

Use Pearson's Rank Correlation Coefficient

$$\rho = \frac{n(\sum x_i y_i) - (\sum x_i)(\sum y_i)}{\sqrt{n(\sum x_i^2) - (\sum x_i)^2} \sqrt{n(\sum y_i^2) - (\sum y_i)^2}}$$

where  $x_i$  and  $y_i$  are corresponding rankings

- ullet ho ranges from -1 (anticorrelated) to +1 (correlated)
- $\rho = 0$  is uncorrelated

#### Discussion

Feature Exclude	Pearson's Correlation	p-value
Active Edges=false	0.89	0.0003
Active Edges=true	0.87	0.0005
N-gram backoff = true	0.85	0.0010
N-gram backoff = false	0.80	0.0032
N-gram=0	0.71	0.0003
CW=0	0.69	0.0098
None	0.63	0.0015
CW=1	0.63	0.0121
CW=2	0.62	0.0097
N-gram=4	0.62	0.0421
N-gram=3	0.53	0.0746



# Preliminary Conclusions

- Mostly the same feature set performs equally well for Japanese and English
- Methodology for extracting most discriminative features uses correlation coefficient
- Most discriminative English/Japanese feature is Active Edges

#### **Future Work**

- Finish generating grid points
- Test stability on different iterations and fold numbers
- Different accuracy metrics