

# Translation of Unknown Words in Low Resource Languages

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31 October 2016

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It is based on this paper:

<http://www.cs.jhu.edu/~huda/papers/gujral2016AMTA.pdf>

bib:

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# Out of Vocabulary Words (OOVs)

- Hindi → English:
  - It वूवन पैंट्स, graphic टीज, Polo T शर्टे, शर्टे, शॉर्ट्स, स्कर्टे and bright embroidered jackets etc are included.
- Uzbek → English:
  - Quvayt o'yinga how ko'ryapmiz with the preparation.

# Goals

- Generate candidates for each OOV
- Select the best one

# How big is this problem?

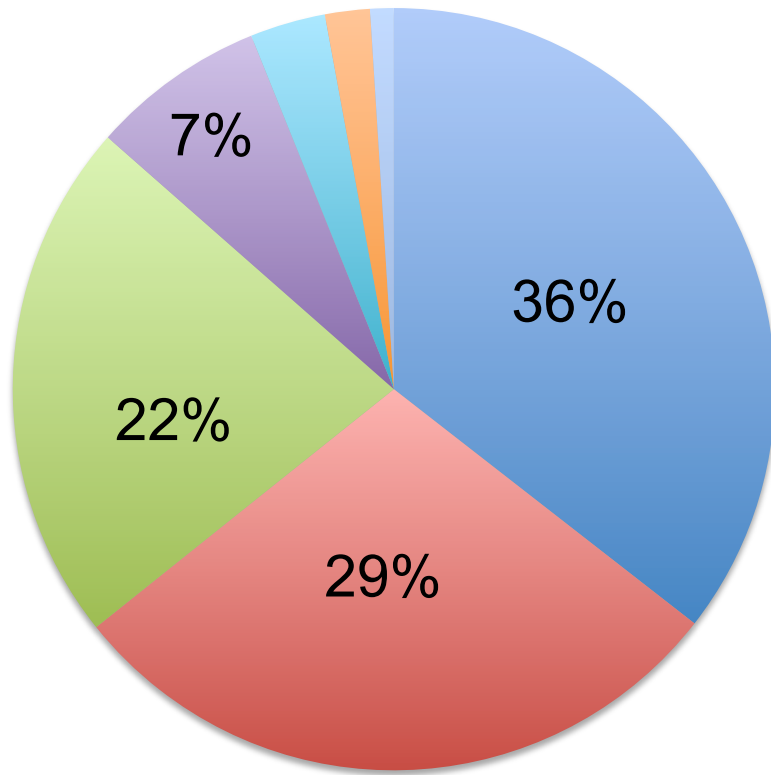
# Data

- Hindi → English
  - WMT14
  - News
  - Training
    - 274k sentences
  - Test
    - 2.5k sentences
    - ~1 OOV/sentence
    - ~5% OOVs
- Uzbek → English
  - LORELEI
  - News, Wikipedia, social media
  - Training
    - 55k sentences
  - Test
    - 1k sentences
    - ~4 OOVs/sentence
    - ~20% OOVs

# OOV Examples

- Names
  - هدى → Huda
- Misspellings
  - grammer/grammar
- Inflections
  - play/plays/playing
- Borrowed words
  - हैलोवीन → Halloween
- Reinflected Borrowings
  - स्कर्ट → skirts
  - *Googlear* → to Google
- Content words
  - अटकलें → speculation

# Distribution of OOVs



- Named Entities
- Borrowed Words
- Source Content Words
- Misspellings & Typos
- Acronyms
- Reinflected Borrowings
- Numbers & Punctuation



# MT System

- **Moses** (Koehn et al. 2007)
  - Phrase Based
- **Large English language model**
  - WMT English '07-'12

# Methods

# Methods

- Transliteration
- Levenshtein distance
- Word Embeddings

# Transliteration

- هدى → Huda
- हैलोवीन → Halloween
- Unsupervised Moses mode (Durrani et al. 2014)
  - Character translation model
  - Incorporate larger English language model
- Uzbek is already written in Latin script, keep original spelling
- Generate 1 candidate

# Levenshtein distance

- grammer/grammar
- play/plays
- Minimum number of:
  - insertions
  - deletions
  - substitutions

# Levenshtein distance

- qilyapmiz → doing
- qilyapsiz → doing Levenshtein distance = 1
- Find source words with distance  $\leq 2$  from OOV
  - Use their English translation as translation candidate
- Generate 18 candidates on average

# Word Embeddings

rumors

अफवाह    doubts    अटकलें

rumours

suspicious

misgivings

worry

speculation

worried

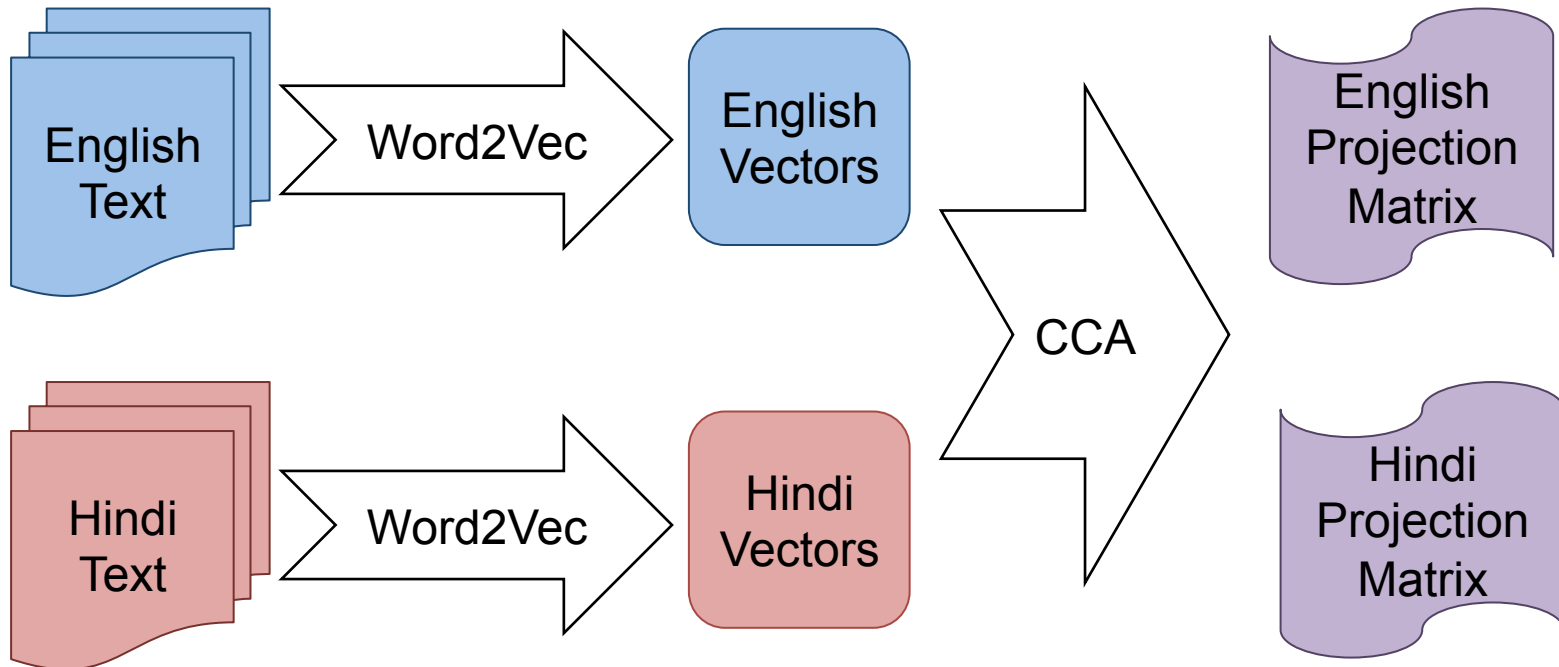
करोड

# Word Embeddings

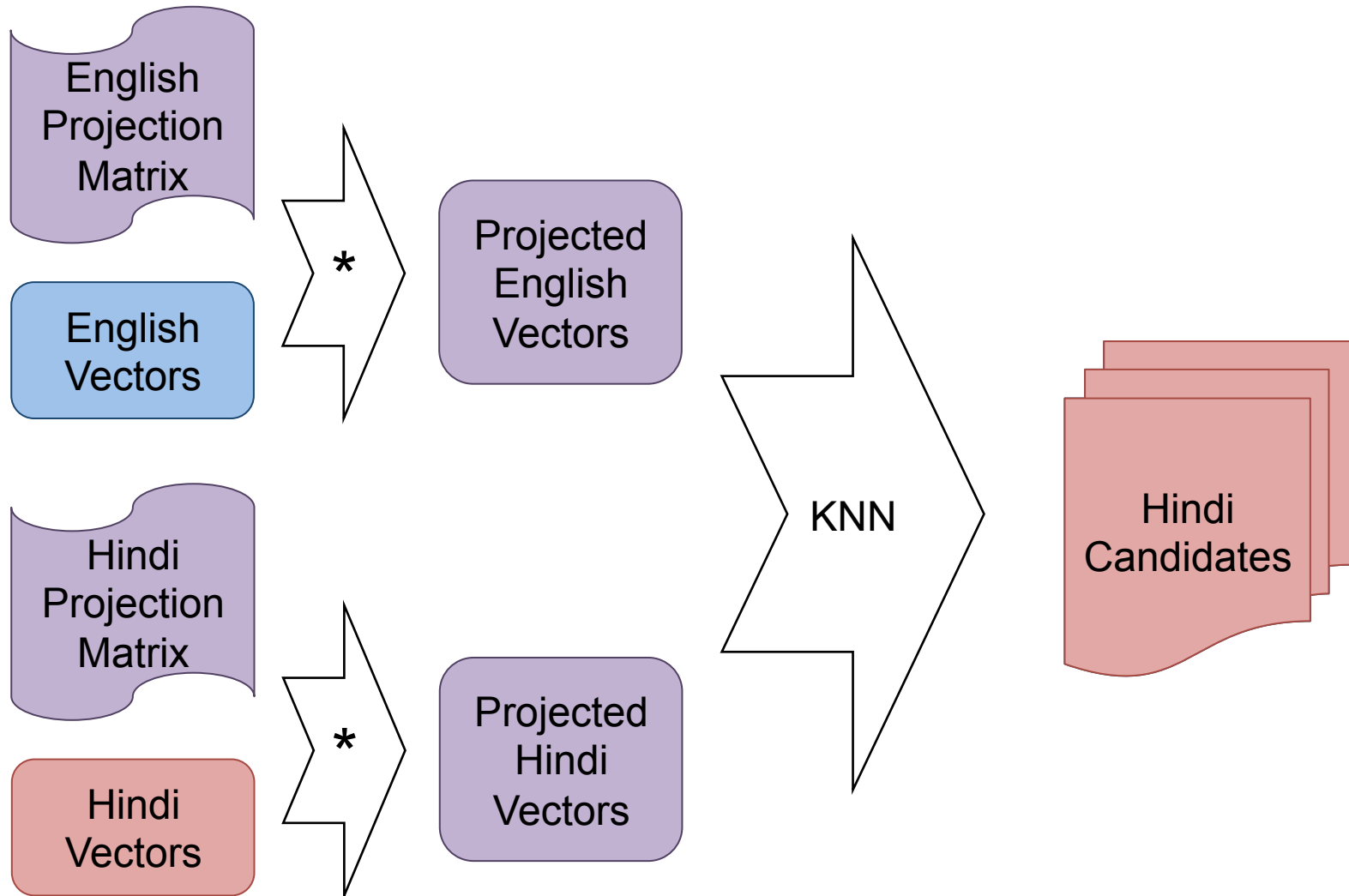
- **Word2vec** (Mikolov et al. 2013)
  - monolingual corpora
- **Multilingual word vectors** (Faruqui & Dyer 2014)
  - monolingual vectors
  - alignments
  - Canonical Correlation Analysis (CCA)
- **Generates 20 candidates**



# Word Embeddings



# Word Embeddings



# Word Embeddings

rumors

अफवाह

doubts

अटकलें

rumours

suspicious

misgivings

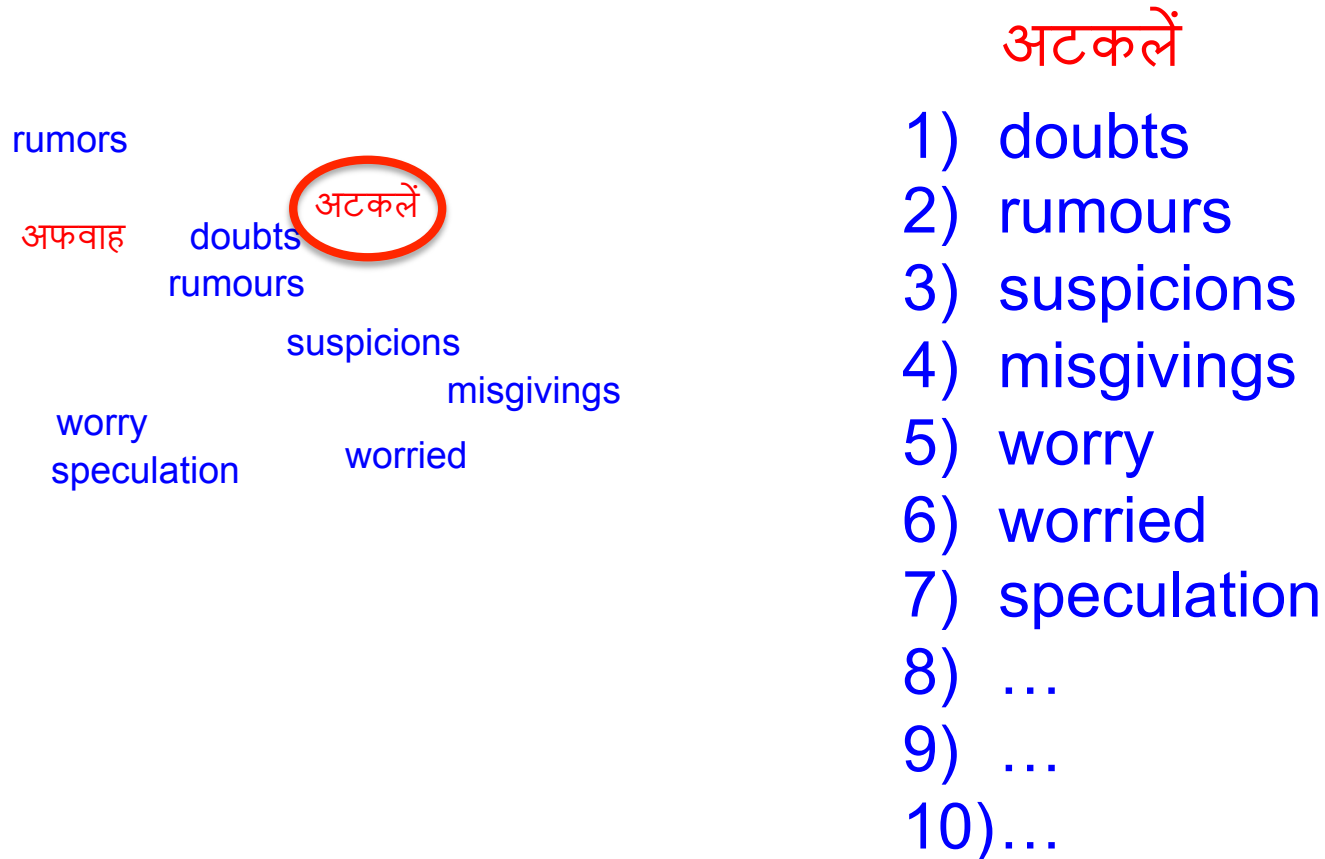
worry

speculation

worried

करोड

# Word Embeddings



# Word Embeddings



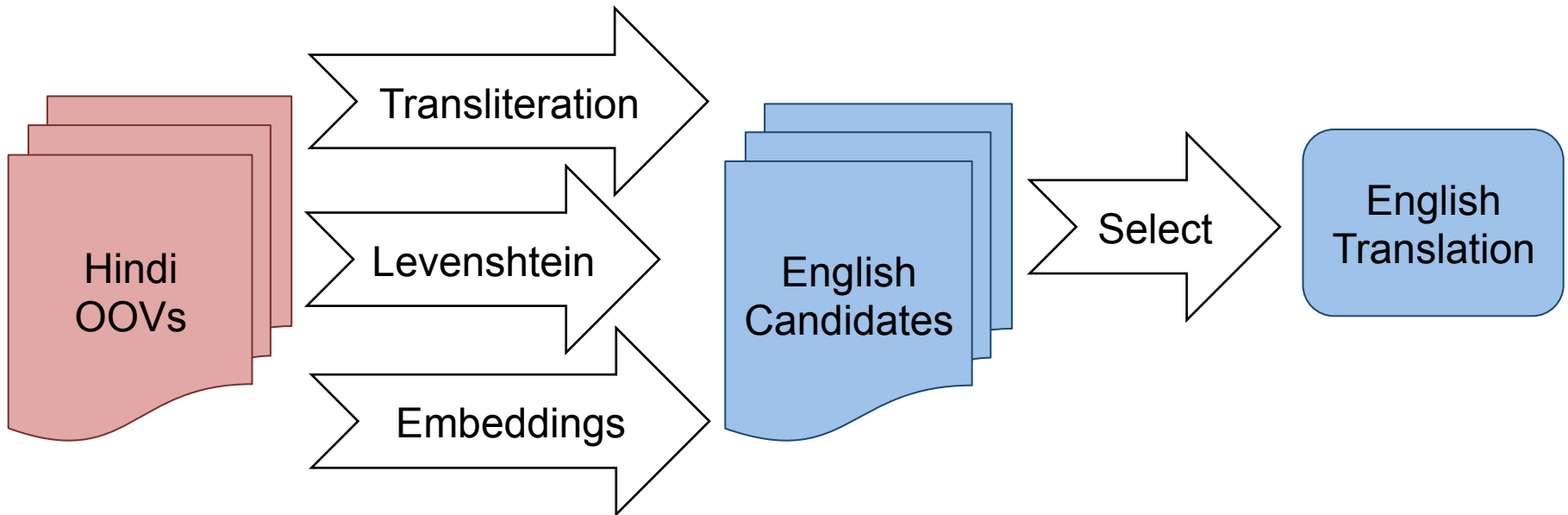
# Word Embeddings

अटकलें

- |                |          |
|----------------|----------|
| 1) doubts      | 1) rumor |
| 2) rumours     | 2) crore |
| 3) suspicions  | 3) ...   |
| 4) misgivings  | 4) ...   |
| 5) worry       | 5) ...   |
| 6) worried     | 6) ...   |
| 7) speculation | 7) ...   |
| 8) ...         | 8) ...   |
| 9) ...         | 9) ...   |
| 10)...         | 10)...   |

# Integration

# Integration





# Integration

- Language Model
- Phrase table

# Language Model

- Large English language model
- XML markup in Moses (Koehn & Haddow, 2009)
- Selection occurs during decoding

# Phrase Table

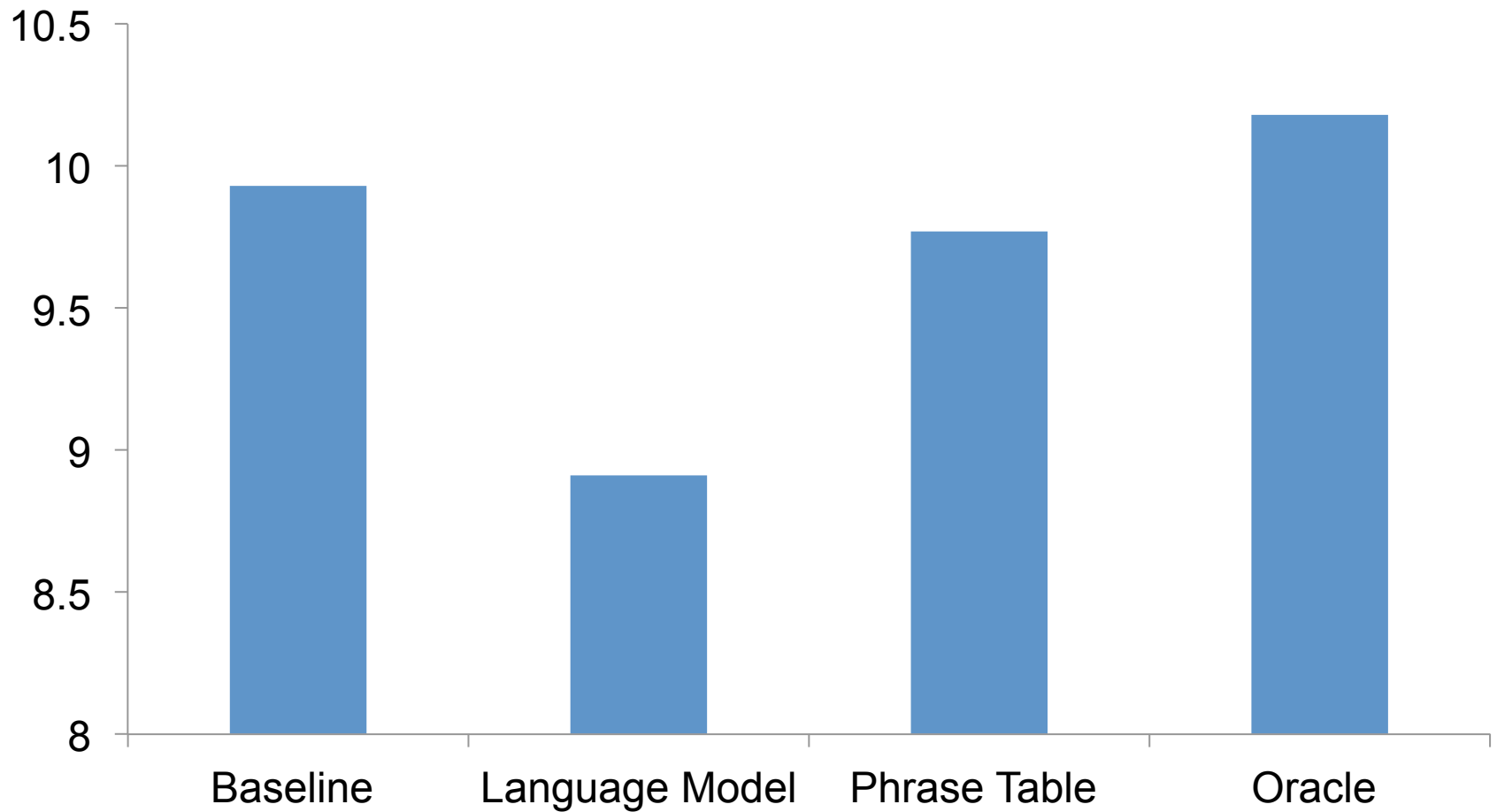
- Secondary Phrase Table only includes OOVs
- Features:
  - Method
  - Word Vector Distance
  - Levenshtein distance
  - Inverse frequency in Monolingual corpus

# Results

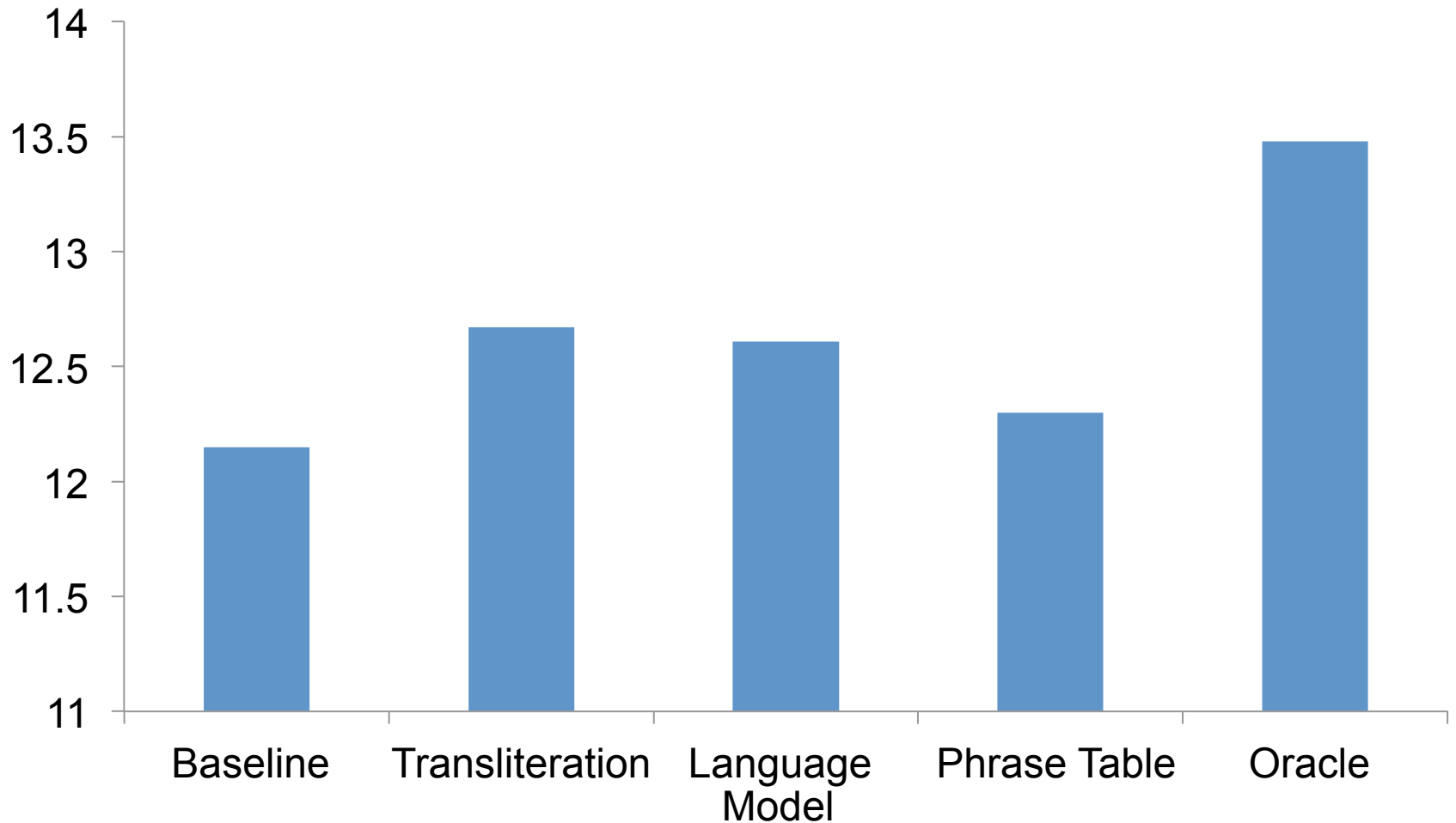
# Oracle

- Upper bound on how well a selection method can do given current generation methods
  - Select word from list of candidates that is in the reference

# BLEU - Uzbek



# BLEU - Hindi



# Beyond BLEU

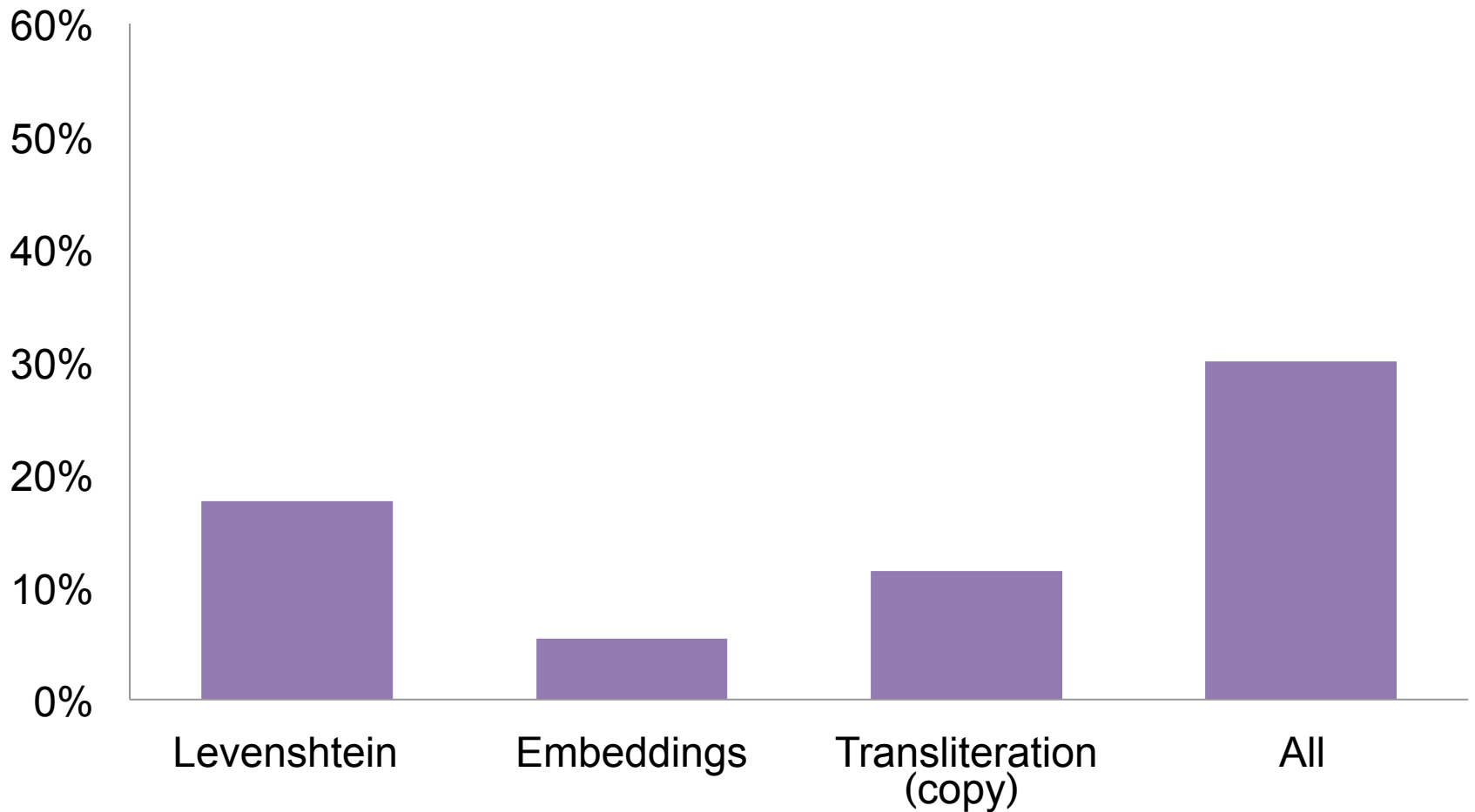
- Goals:
  - generate candidates for each OOV
    - How well can we generate translation candidates?
  - select the best one
    - How well can we select from the translation candidates?



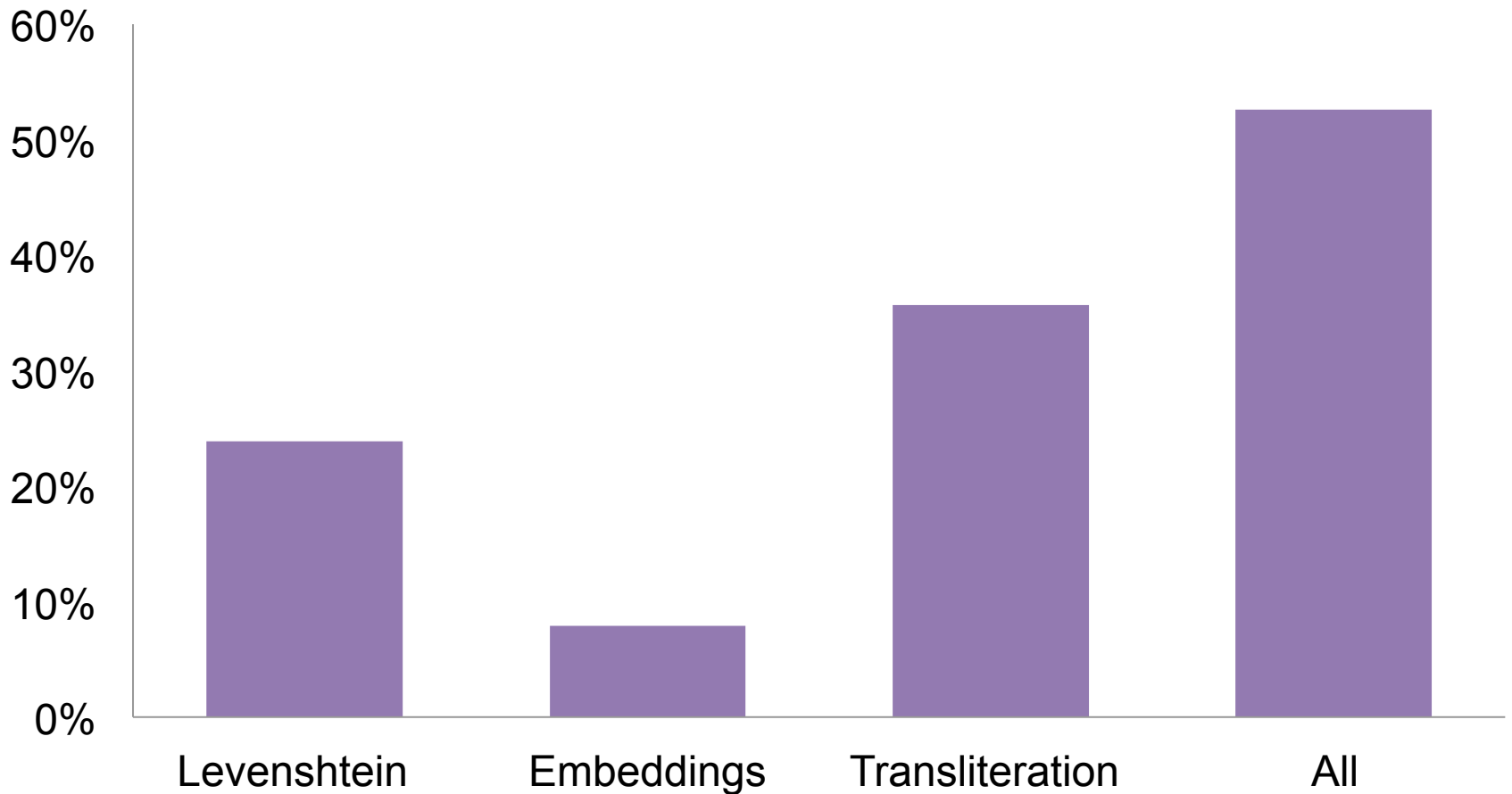
# Coverage

- How well can we generate translation candidates?
  - Was one of the candidates **generated** by this method in the reference?

# Coverage - Uzbek



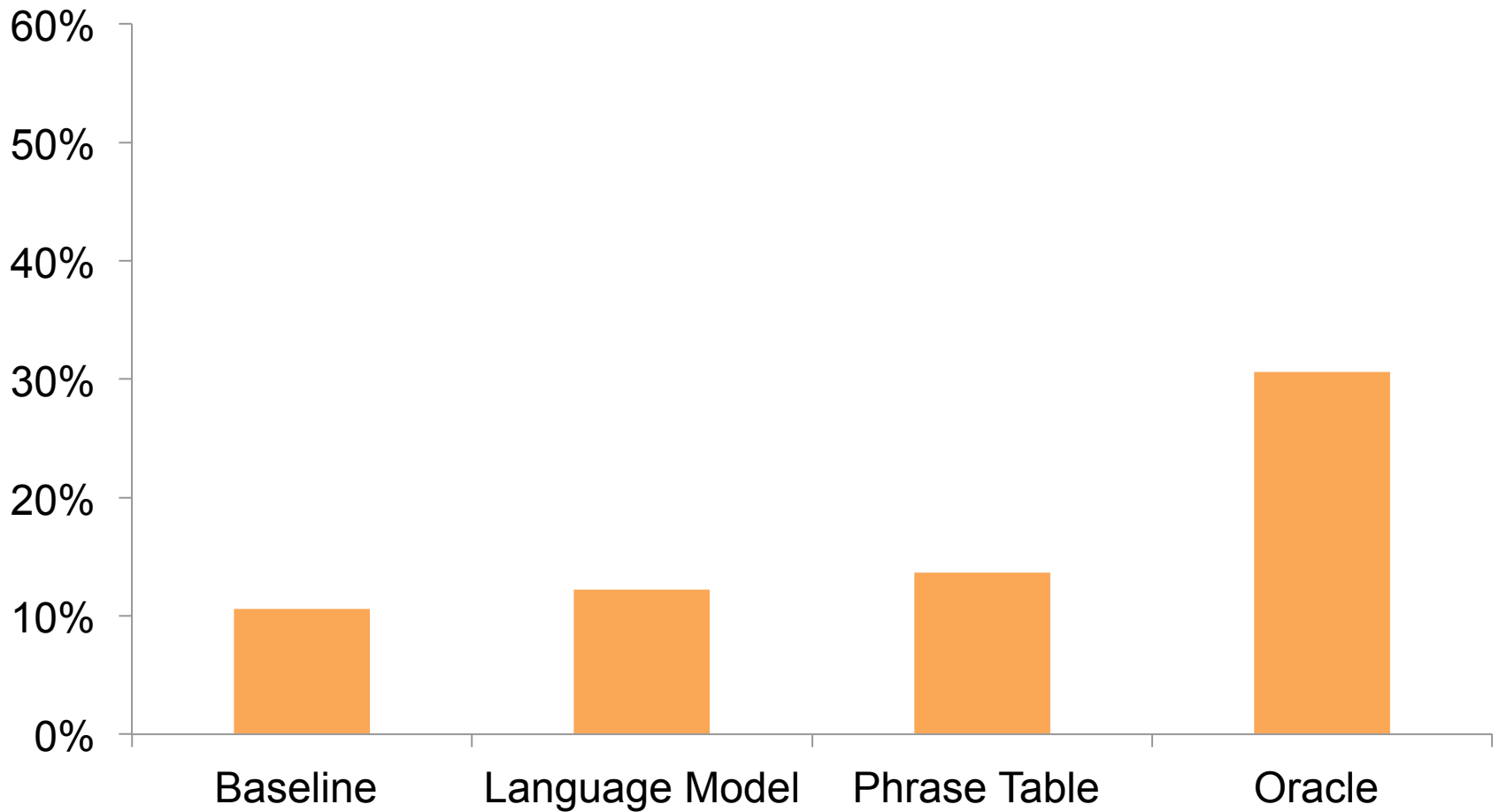
# Coverage - Hindi



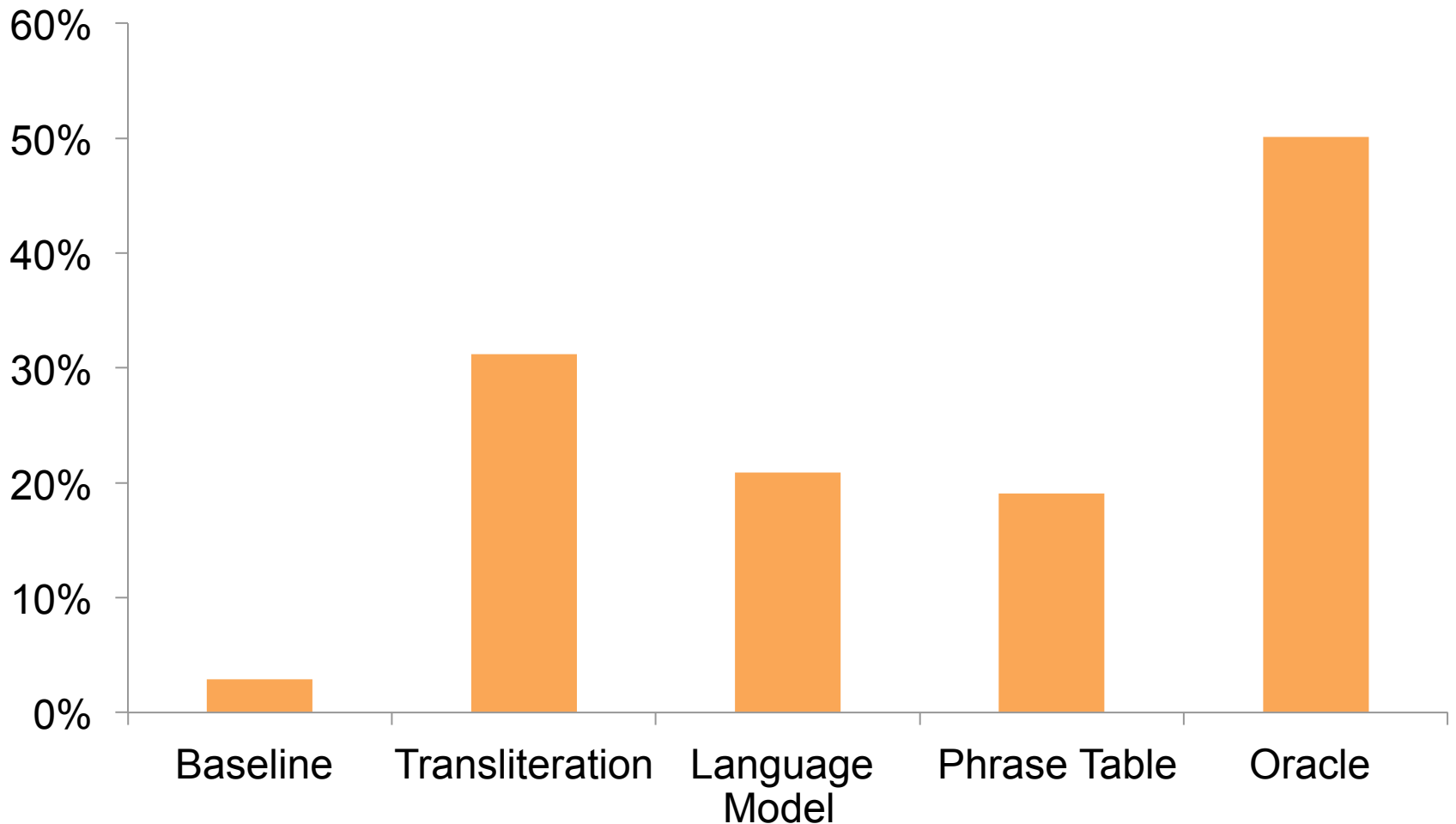
# Accuracy

- How well can we select from the translation candidates?
  - Is the word we **selected** in the reference?

# Accuracy - Uzbek



# Accuracy - Hindi



# Conclusion & Future Work

- Generate Quality translations
  - Selection does not perform as well
- Improved selection methods
- More sophisticated embedding projection
- Analysis of what methods work on which types of OOVs

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