

# Integrating Beam Search and Error States

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Group 6

# What we have done

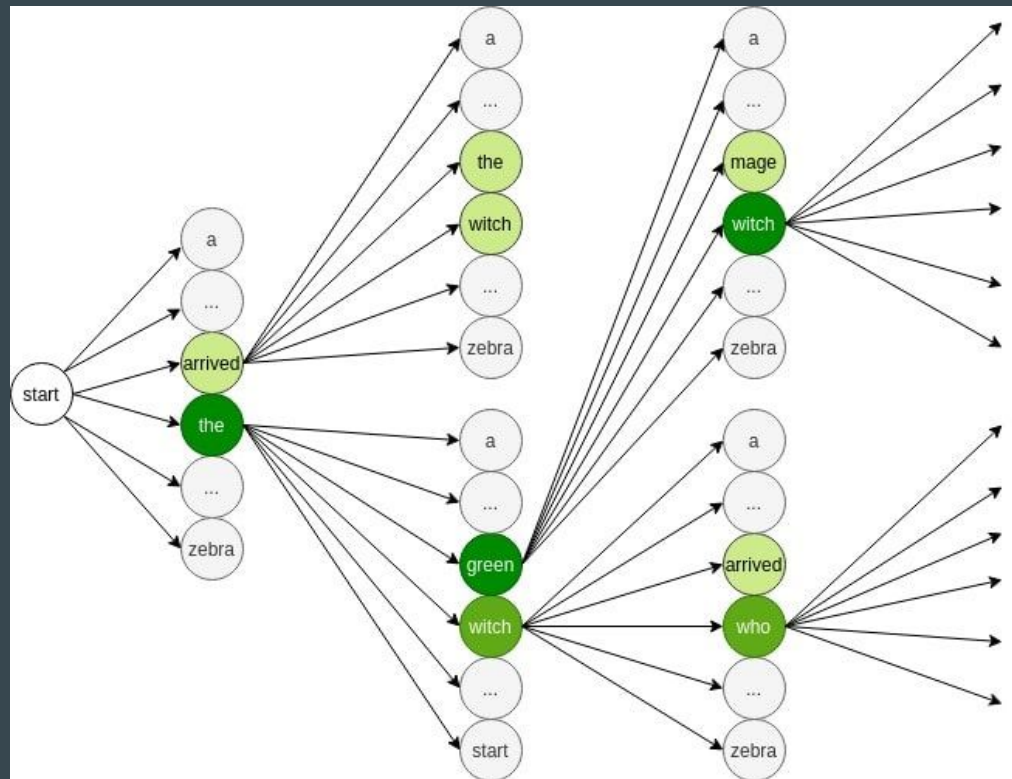
- Extension from baseline project
- Exchange the greedy search to a beam search
- Best first beam search.
- Introduce error states.

# Why is this interesting?

- Beam search will improve accuracy
- Best first implementation can efficiency without at the cost of accuracy.
- Learn from its mistakes.

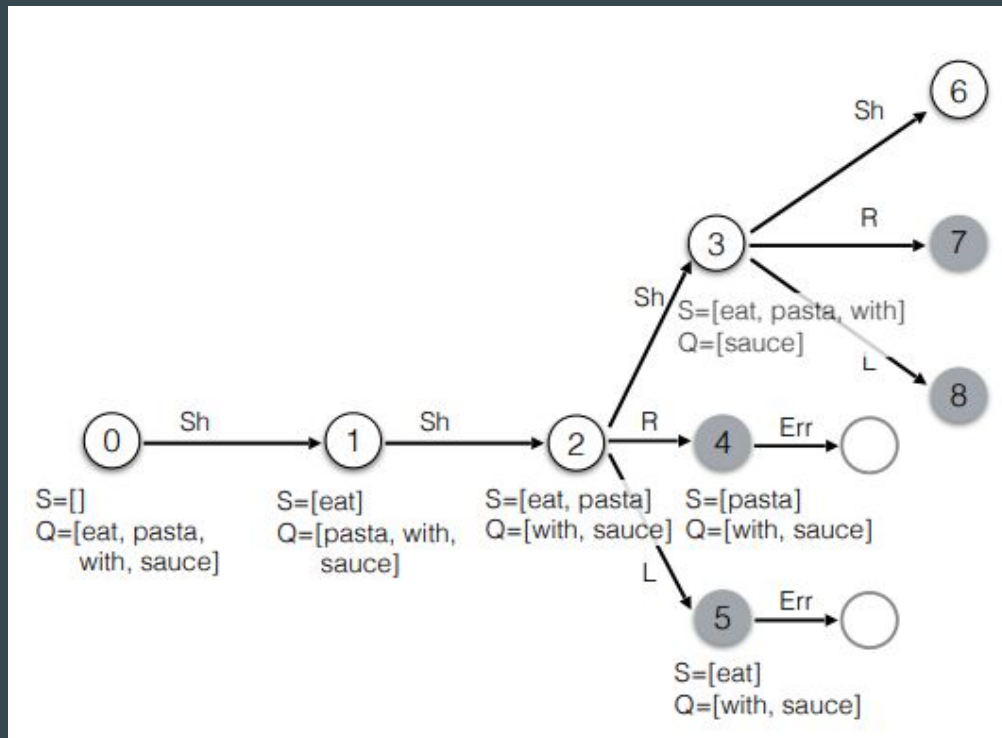
# BEAM SEARCH

- Efficient Search Algorithm
- Maintains Beam of Top Hypotheses
- Expands Promising Hypotheses
- Prunes Unlikely Hypotheses



# ERROR STATES

- Address parsing mistakes
- Improve accuracy
- Recover from mistakes



# Related work

If beam search is the answer,  
what was the question?

Best-First Beam Search

Efficient Structured Inference for  
Transition Based Parsing with Neural  
Networks and Error States

Investigated beam search

Try to answer why it is effective

Analyzed impact of beam  
widths

Proposed best-first beam search

Prioritizes the highest scoring  
paths at all times

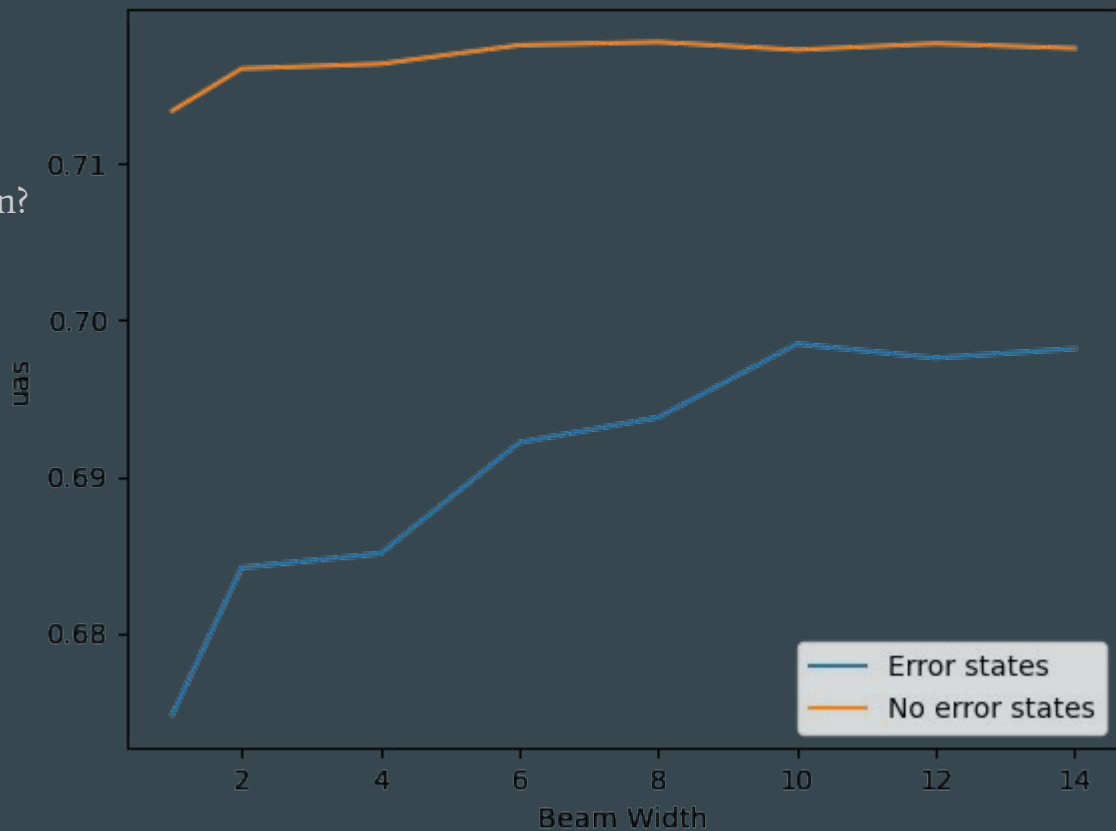
Similar accuracy but faster

Integrated error states for  
transition-based parsing

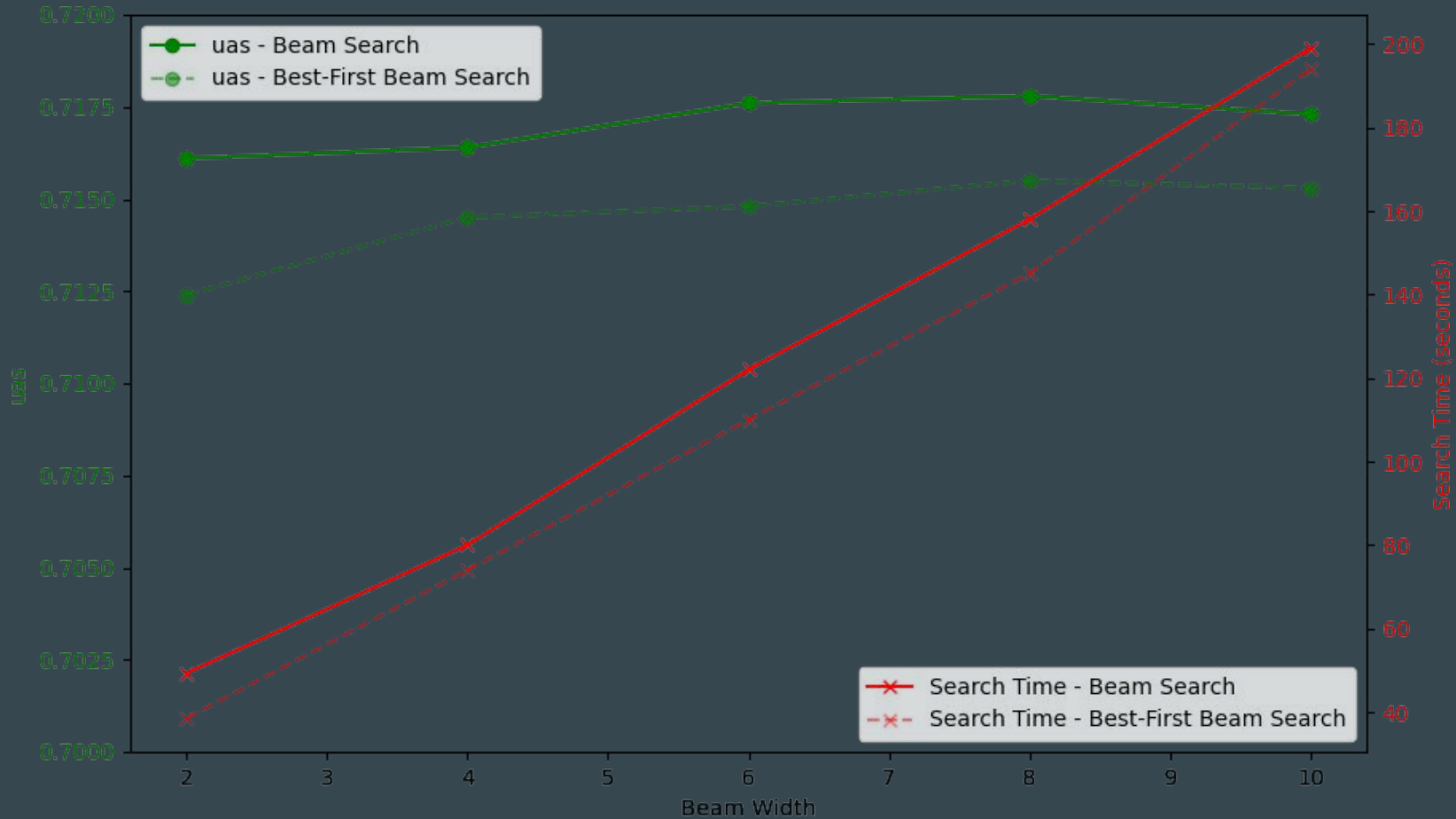
Improved accuracy on various  
datasets

# Results

- Best beam width?
- Use Error States or not?
  - Use all Error States or a fraction?
- How well does it work for different languages?
  - Size of datasets impactful?

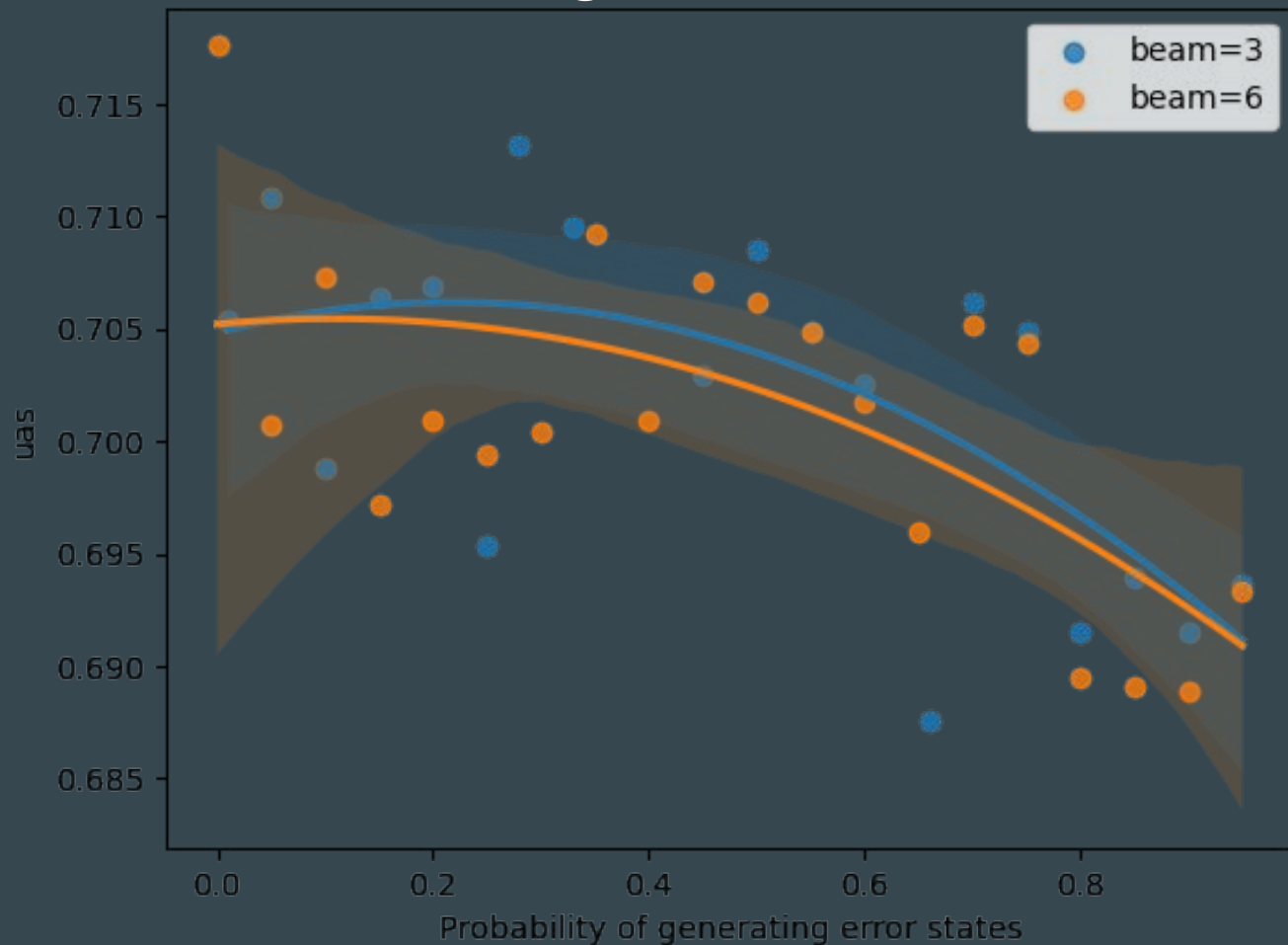


# Best-First Beam Search





# Generating less error states



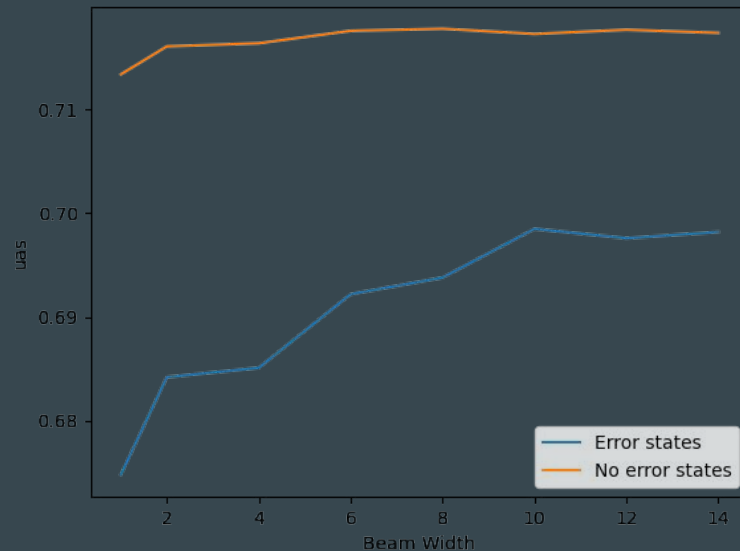
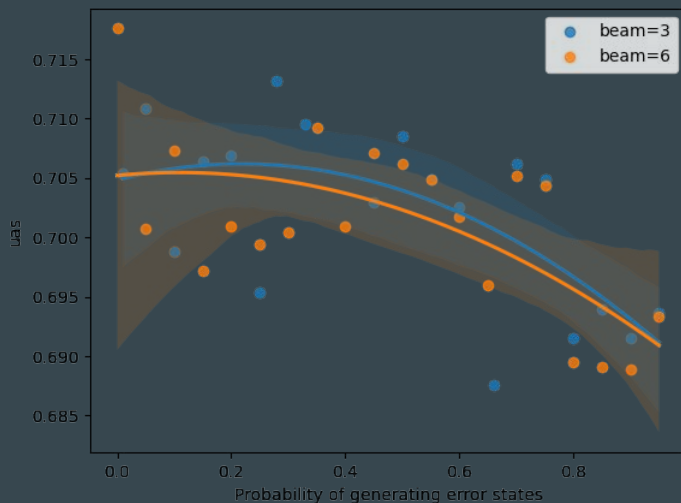
# Why did it not work?

Features:

They use 14 and 25

We use 6

Dataset



# Conclusions

Beam Search is better than greedy

Error states don't necessarily improve performance

~50% of error states might be optimal