

# Cascaded Text Generation with Markov Transformers

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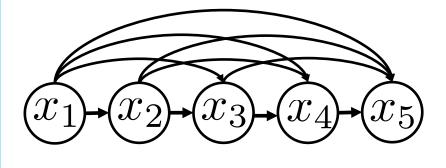
github.com/harvardnlp/cascaded-generation

We present a fast and fluent text generation method using bounded-order Markov models. To decode from high-order models efficiently, we propose a cascaded decoding approach that prunes the search space using lower-order models, and we introduce a Markov transformer that can parameterize the entire cascade.

#### 1/ Motivation

Fully Autoregressive (AR)

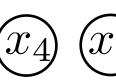
Nonautoregressive (NAR)











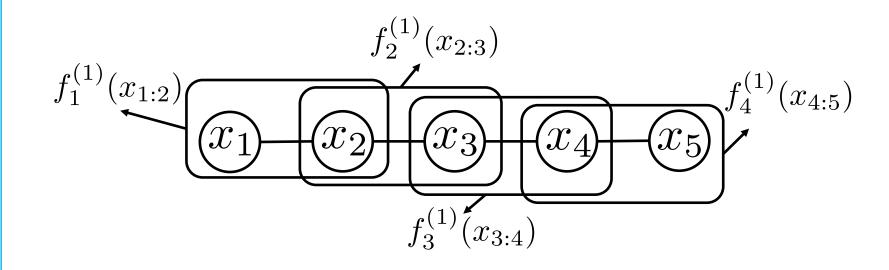
- Decoding: beam search
- Fluent but serial
- Decoding: argmax
- Parallel but disfluent
- We propose a method that's both fast and fluent

### 2/ Model: Bounded-Order MRF

• An *m*-th order Markov Random Field (MRF)

$$P^{(m)}(x_{1:L};\theta) \propto \exp \sum_{l=1}^{L-m} f_l^{(m)}(x_{l:l+m};\theta)$$

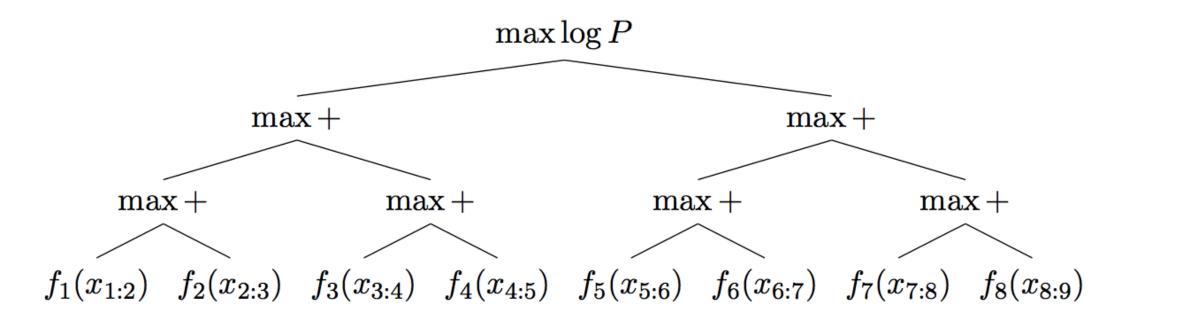
- Each  $f_i^{(m)}$  models dependencies among adjacent (m+1) words
- Example: m = 1, L = 5



- **Special Cases:** 
  - $\circ m = 0$ : nonautoregressive
  - $\circ m = L 1$ : fully autoregressive
  - 0 < m < L 1: bounded-order models (this work)

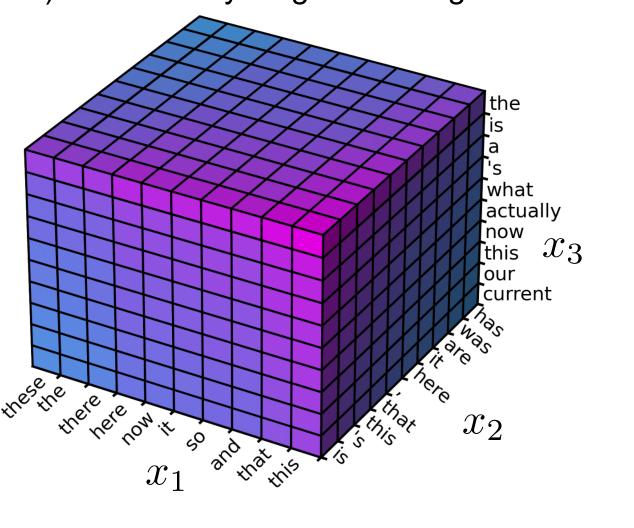
# 3/ Parallel Decoding

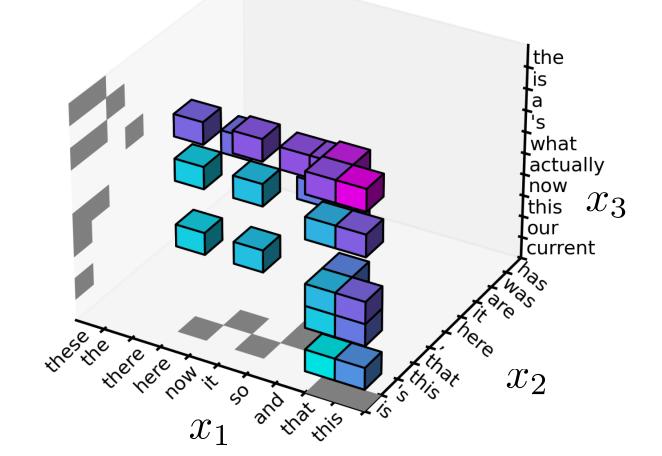
- Bounded-order models can be decoded in parallel
- In practice, takes <1% total time



# 4/ Cascaded Decoding

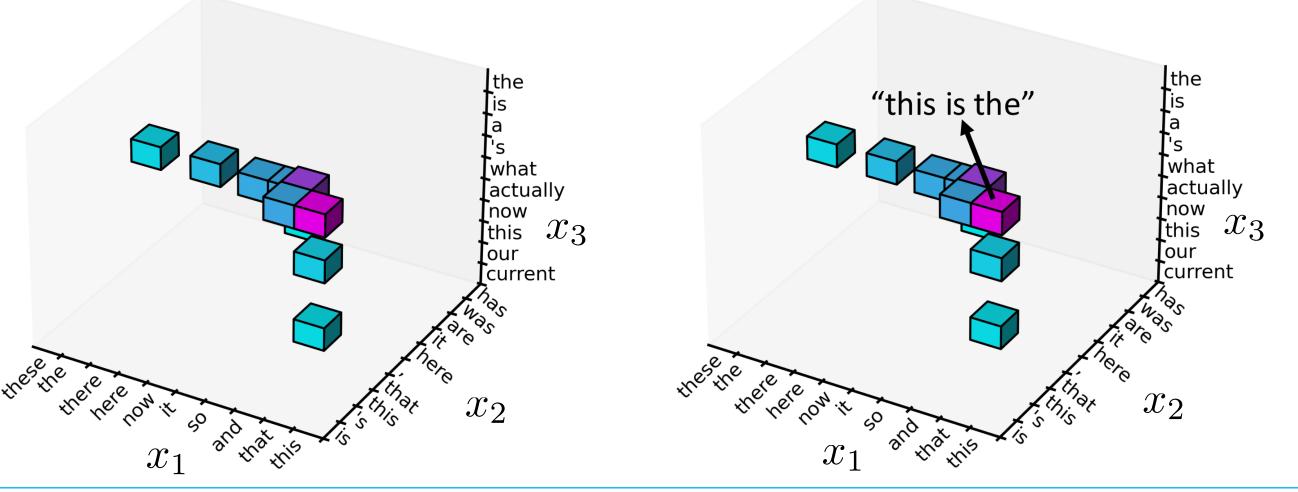
- Parallel decoding for an M-th order model takes  $O(V^{M+1} \log L)$  time
- Impractical even when *M*=1
- Cascaded decoding filters unlikely n-grams using lower-order models
- Example: M= 3
- a) filter unlikely unigrams using m = 0 b) filter unlikely bigrams using m = 1





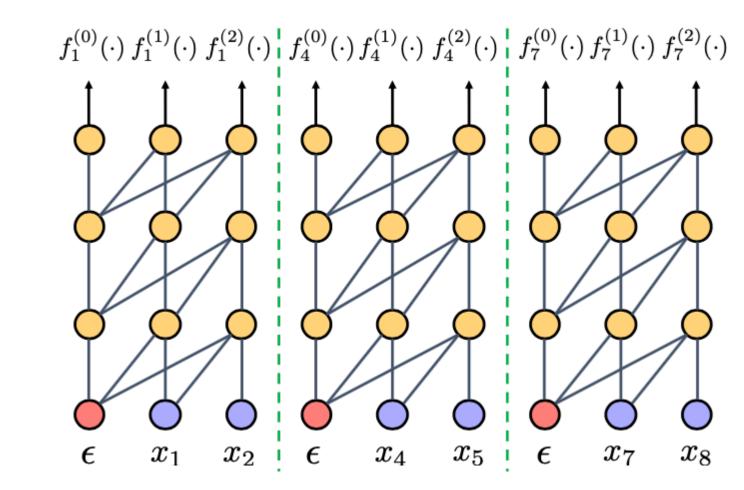






#### 5/ Parameterization: Markov Transformer

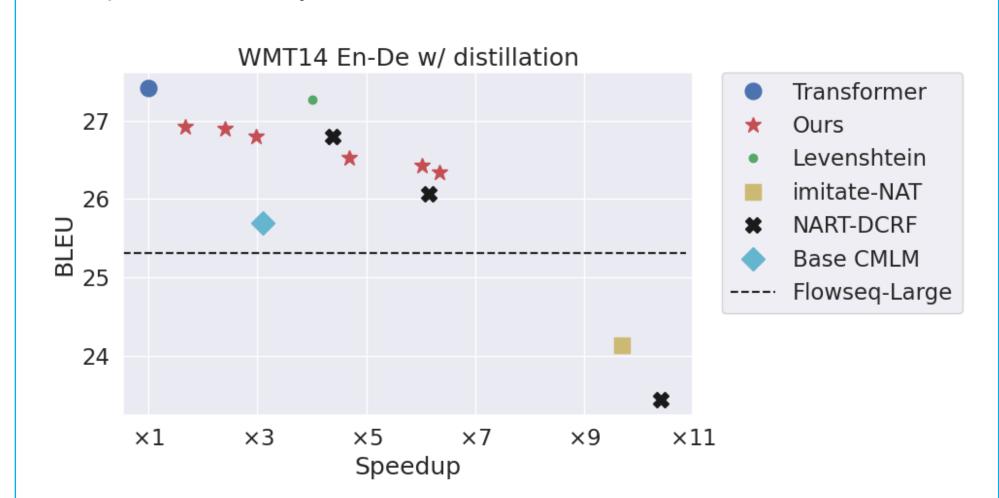
- Parameterize the entire cascade with a single transformer
- Training: insert M-spaced barriers with random offset



• Test: can be applied as any Markov model with m < M

## 6/ Results & Analysis

Speed/accuracy tradeoff



Fewer repetitions



