# Dirichlet Process Mixture Language Model

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#### Abstract

We extend the work in Mielke and Eisner (2018) with a Dirichlet Process mixture language model.

#### 1 Problem Statement

Open vocabulary LM with open class latent lexemes. Maybe explicitly model some other stuff with local RVs?

Let's see what we can discover...

### 2 Model

Generative story. Let  $\alpha$  be a hyperparameter and  $G_0 \sim N(0, I_d)$ .

- 1. Draw  $\eta_i^* \mid G_0 \sim G_0, i = \{1, 2, \ldots\}$ , the embeddings for the latent lexemes.
- 2. For each index  $t \in 1, \ldots, T$ 
  - (a) Draw  $V_{it} \mid \alpha, x_{< t} \sim \text{Beta}(1, \alpha), i = \{1, 2, ...\}$ , the parameters of the stick breaking process.
  - (b) Draw lexeme  $Z_t \mid \{v_{1t}, v_{2t}, \ldots\} \sim \operatorname{Cat}(\pi(\mathbf{v}_t))$
  - (c) Draw word  $X_t \mid z_t \sim p(x_t \mid \eta_{z_t}^*)$

asdf

- 2.1 Parameterization
- 3 Training and Inference
- 4 Anticipated Failure Cases

1.

REFERENCES Justin Chiu

## References

Sebastian J. Mielke and Jason Eisner. 2018. Spell once, summon anywhere: A two-level open-vocabulary language model. CoRR, abs/1804.08205.