# Name Phylogeny: A Generative Model of String Variation <br> <br> Nicholas Andrews, Jason Eisner, Mark Dredze 

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Department of Computer Science, CLSP, HLTCOE, Johns Hopkins University

## A Example Name Phylogeny



Edges from the root correspond to generating new name strings (entities)

Edges between strings are "mutations"
Generative model
Given a sequence of tokens, there are two options to generate the next name mention (token)

1. Pick an existing token $x$ with probability $1 /(\alpha+k)$
1.1 Copy $x$ verbatim with probability 1 - $\mu$
1.2 Mutate $\times$ with probability
2. Generate a new string with probability $\alpha /(\alpha+k)$

Generative model in action


Mutation model
We use a simple conditional PFST with latent edit and no-edit regions.

The basic character-level operations are insertions, deletions, substitions, and copies.

The parameters $\theta$ of the model determine the probabilities of transitions between latent regions and of the different character-level operations.

From token phylogenies to type phylogenies
All copy edges are collapsed (see vertices in bold below)


The first token in each collapsed vertex is a mutation; the rest are copies

Approximation: we forbid multiple tokens of the same type to be derived from mutations

## Inference via EM

Iterate until convergence:

1. E-step: Given $\theta$, compute a distribution over phylogenies (spanning trees)
2. M-step: Re-estimate transducer parameters $\theta$ given marginal edge probabilities

Inner EM loop in the M-step to sum over alignments between input/output strings

## Data

Wikipedia redirects are used as ground truth for entity name variations

The frequency of each name variant is estimated using the Google/Stanford crosswiki dataset (Spitkovsky and Chang, 2012)

For evaluation, 500 entities are sampled and their name tokens are divided into
5 training folds and 1 test fold
The training dataset contains $\sim 4000$ distinct strings
Supervision constrains the phylogeny
Subsets of the training folds restrict possible spanning trees:
(1) Edges only allowed between labeled types of the same entity and unsupervised types.
(2) No edges from unsupervised types to supervised types
(Both supervised and unsupervised types may derive from the root.)

## Experiments

## At training time

1. Estimate transducer parameters with EM
2. Find the single best phylogeny given the learned model

## At test time

1. Attach test tokens to the inferred phylogeny
2. Calculate precision and recall for the connected component the test token was attached to

Precision: fraction of name variants in the connected component of the same entity as the test token
Recall: fraction of all name variants for the test token found in the connected component

Baseline: Flat Tree

As a baseline, we limit the depth of the phylogeny to 2 so each name variant either
(1) mutates from a fixed canonical name or
(2) is generated from scratch (the root)


The most frequent name variant for each entity is selected as the canonical name.

## Results

For the baseline and the full model, we vary: - the proportion labeled data at training time - the parameter $\alpha$


