

# **Robust Entity Clustering** via Phylogenetic Inference

#### Nicholas Andrews with Jason Eisner and Mark Dredze



human language technology center of excellence



Did **Taylor swift** just dis harry sytles on the #grammys ? Lmao



Did **Taylor swift** just dis harry sytles on the #grammys ? Lmao

Lets see how bad **T Swift** will be. #grammys



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Watching the Grammy's - it's clear that **T-Swizzle** is on drugs. Lots of drugs.



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Taylor swift is apart of the Illuminati #grammys



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LL Cool J is looking reallIII chizzled!

## Entity clustering





Did **Taylor swift(1)** just dis harry sytles on the(1) #grammys ? Lmao

Lets see how bad **T Swift(1)** will be. #grammys

Watching the Grammy's - it's clear that **T-Swizzle** (1) is on drugs. Lots of drugs.



**Taylor swift(1)** is apart of the Illuminati #grammys

Ladies STILL love LL Cool James(2).



LL Cool J(2) is looking realll chizzled!



2

### Key idea: "Directed" name variation



Did **Taylor swift** just dis harry sytles on the #grammys ? mao **Abbreviation** 

Lets see how bad **T Swift** will be. #grammys

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**Taylor swift** is apart of the Illuminati #grammys

Ladies STILL love **LL Cool James**.



**LL Cool J** is looking realll chizzled!

#### "Directed" name variation



Did **Taylor swift** just dis harry sytles on the #grammys ? mao

Lets see how bad **T Swift** will be. #grammys Copy

Watching the Grammy's - it's clear that **T-Swizzle** is on drugs. Lots of drugs.

**Taylor swift** is apart of the Illuminati #grammys

Ladies STILL love **LL Cool James**.



**LL Cool J** is looking realll chizzled!

#### "Directed" name variation



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Lets see how bad **T Swift** will be. #grammys

Watching the Grammy's - it's clear that **T-Swizzle** is on drugs. Lots of drugs.

-izzle

**Taylor swift** is apart of the Illuminati #grammys

Ladies STILL love **LL Cool James**.



**LL Cool J** is looking realll chizzled!

#### "Directed" name variation



Did **Taylor swift** just dis harry sytles on the #grammys ? I mao

Lets see how bad **T Swift** will be. #grammys

Watching the Grammy's - it's clear that **T-Swizzle** is on drugs. Lots of drugs.

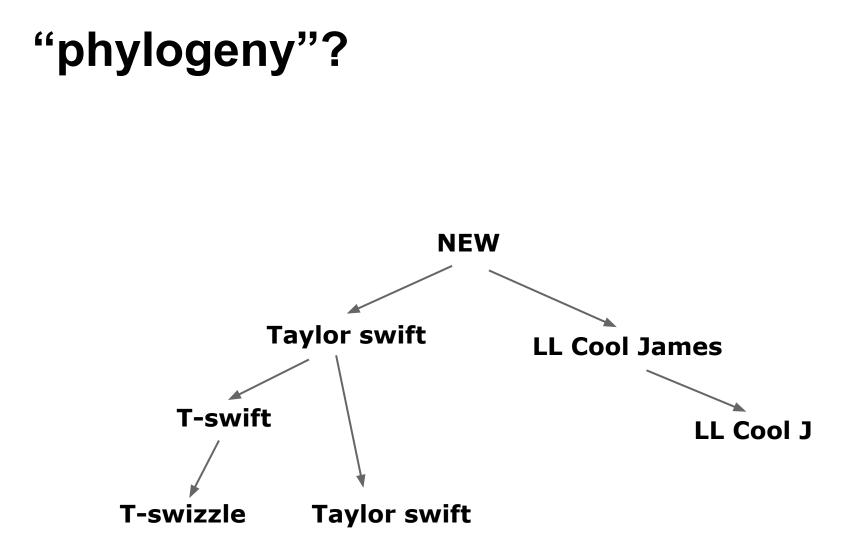
Taylor swift is apart of the Illuminati #grammys

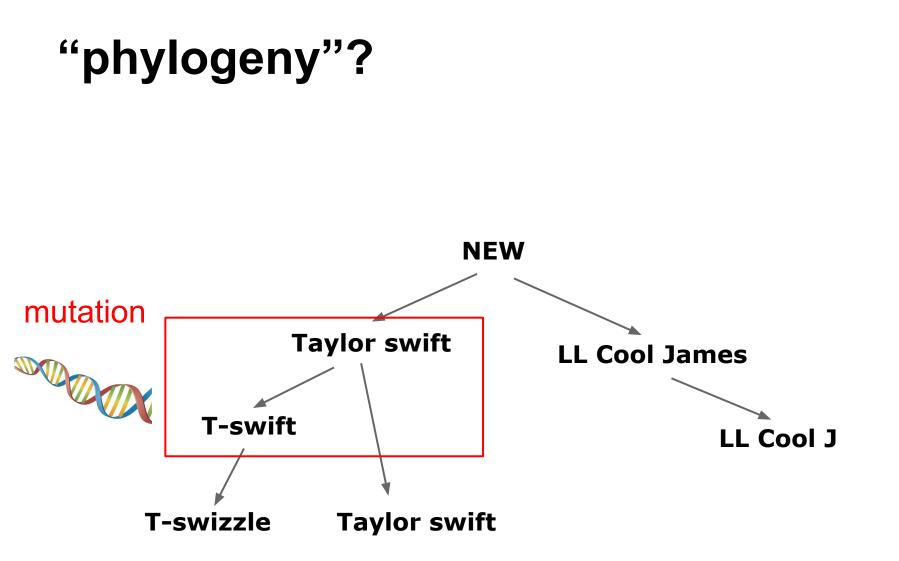
Ladies STILL love **LL Cool James**.



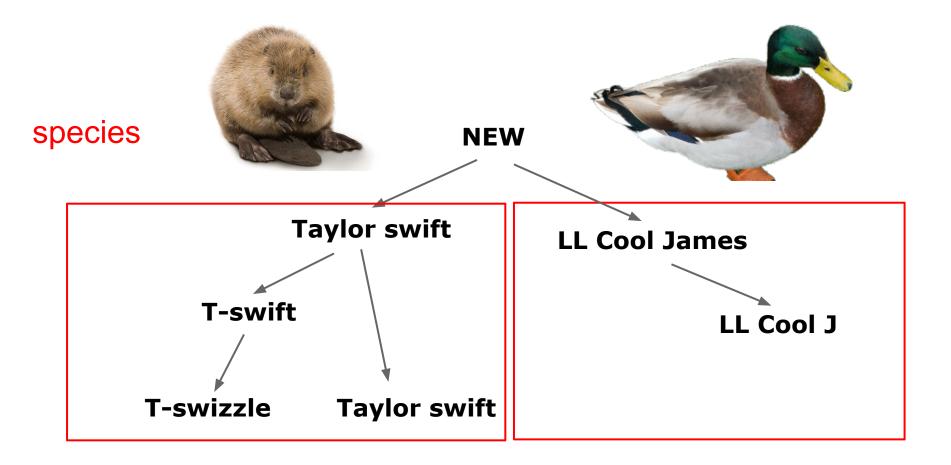
Abbreviation

LL Cool J is looking reallIII chizzled!









## A Generative Story



Did [Taylor swift] just dis harry sytles

Lets see how bad **[T Swift]** will be. #grammys

it's clear that **[T-Swizzle]** is on drugs

[Taylor swift] is apart of the Illuminati

Ladies STILL love [LL Cool James].

[LL Cool J] is looking realIIII chizzled!

## A Generative Story



Did **Taylor swift** just dis harry sytles

Lets see how bad **T Swift** will be. #grammys

it's clear that **T-Swizz** 

Taylor swift is apart of the Illuminati

Ladies STILL love LL Cool James

LL Cool J is looking reallIII chizzled!

## **Pick topics**



# 152123DidTaylor swiftjust dis harry sytles

 3
 4
 1
 2
 5
 1
 2
 7

 Lets see how bad
 T Swift
 will be. #grammys

 3
 1
 2
 5
 1
 2
 4

 it's clear that
 T-Swizzle
 is on drugs

5 1 3 2 3 8 Taylor swift is apart of the Illuminat

12310Ladies STILL loveLL Cool James.

102517LL CoolJ is looking reallill chizzled!

#### [ ] is a placeholder for an entity mention

7

## **Pick words | topics**





Did [Taylo? swift] just dis harry sytles

Lets see how bad [T Swift] will be. #grammys

5

5 it's clear that [T-Swizzle] is on drugs

5 [Taylor swift] is apart of the Illuminati

**10** Ladies STILL love [LL Cool James].

**10** [LL Cool J] is looking realIIII chizzled!

5

NEW



Did [Taylor swift] just dis harry sytles 5 Lets see how bad [T Swift] will be. #grammys 5 it's clear that [T-Swizzle] is on drugs 5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James] 10 [LL Cool J] is looking realIII chizzled!

NEW



5 Did [Taylor swift] just dis harry sytles Lets see how bad [T S? [ft] will be. #grammys 5 it's clear that [T-Swizzle] is on drugs 5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James] 10 [LL Cool J] is looking realll chizzled!



NEW 5 Did [Taylor swift] just dis harry sytles 5 Lets see how bad [T Swift] will be. #grammys 5 it's clear that [T-Swizzle] is on drugs 5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James] 10 [LL Cool J] is looking realIII chizzled!



Did [Taylor swift] just dis harry sytles

5

Lets see how bad [T Swift] will be. #grammys

it's clear that [T-Sv?zzle] is on drugs

5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James].

**10** [LL Cool J] is looking realIIII chizzled!

NEW



5 Did [Taylor swift] just dis harry sytles 5 Lets see how bad [T Swift] will be. #grammys 5 it's clear that [T-Swizzle] is on drugs 5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James] 10 [LL Cool J] is looking realll chizzled!



Did [Taylor swift] just dis harry sytles

5

5

Lets see how bad [T Swift] will be. #grammys

EW

it's clear that [F-Swizzle] is on drugs

[Taylo?swift] is apart of the Illuminati

#### 10

Ladies STILL love [LL Cool James].

**10** [LL Cool J] is looking realIIII chizzled!

NEW



5 Did [Taylor swift] just dis harry sytles Lets see how bad [T Swift] will be. #grammys it's clear that [T-Swizzle] is on drugs 5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James] 10 [LL Cool J] is looking realll chizzled!

5

$$p(\text{parent}) = \frac{\exp\left(\mathbf{f}(\text{parent}, \text{child}) \cdot \phi\right)}{Z}$$



[Taylor swift] is apart of the Illuminati 10 NEW

Ladies STILL love [LL Cool James]

**10** [LL Cool J] is looking realIIII chizzled!

#### Unknown parameters

$$p(\text{parent}) = \frac{\exp\left(\phi_{10,10} + \phi_{\text{equal}}\right)}{Z}$$

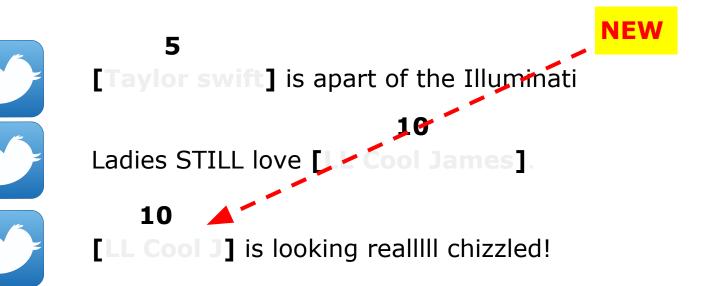
5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James]. 10 [LL Cool ]] is looking reallIII chizzled!

$$p(\text{parent}) = \frac{\exp(\phi_{5,10} + \phi_{\text{diff}})}{Z}$$

Solution in the second seco

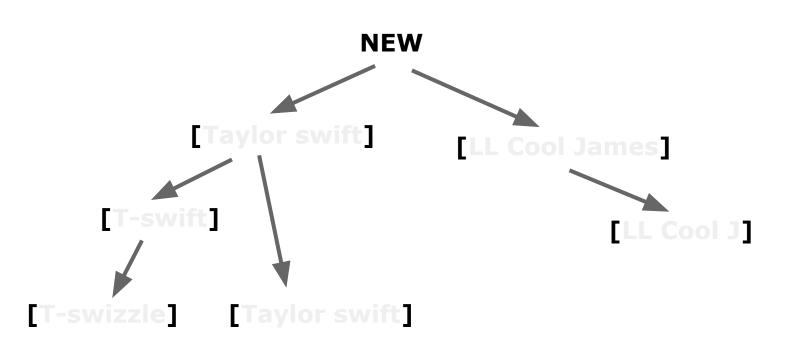
#### Hyperparameter: controls # of entities

$$p(\text{parent}) = \frac{\exp(\phi_{\text{new}})}{Z}$$



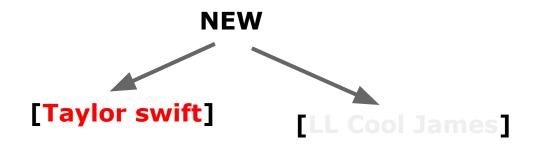
#### **Another view**

## Note: No names yet, that's next...



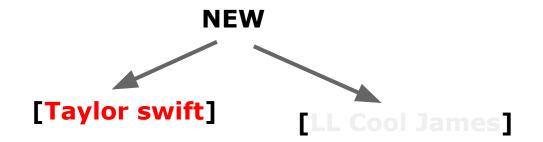
#### Generating a name

#### If parent = NEW: name new entity



#### **Generating a name**

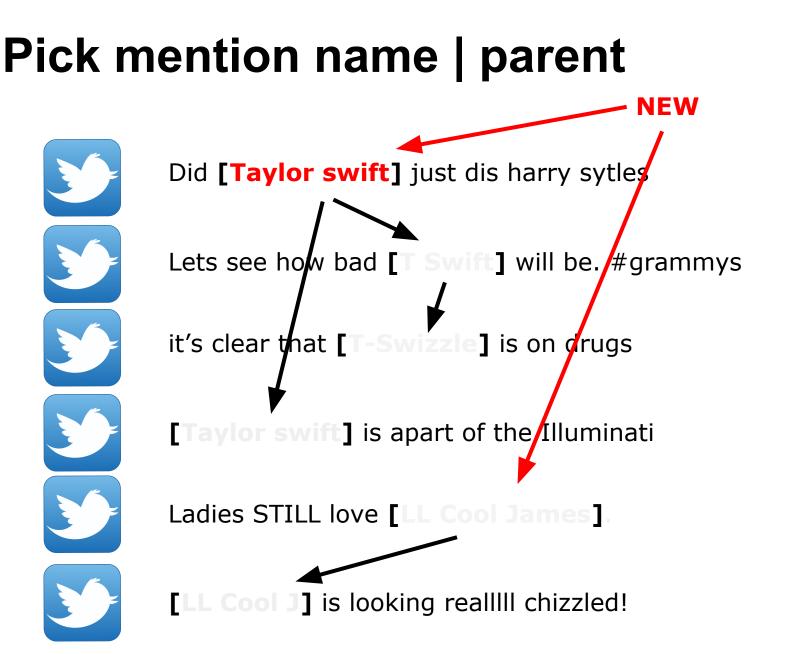
#### If parent = NEW: name new entity

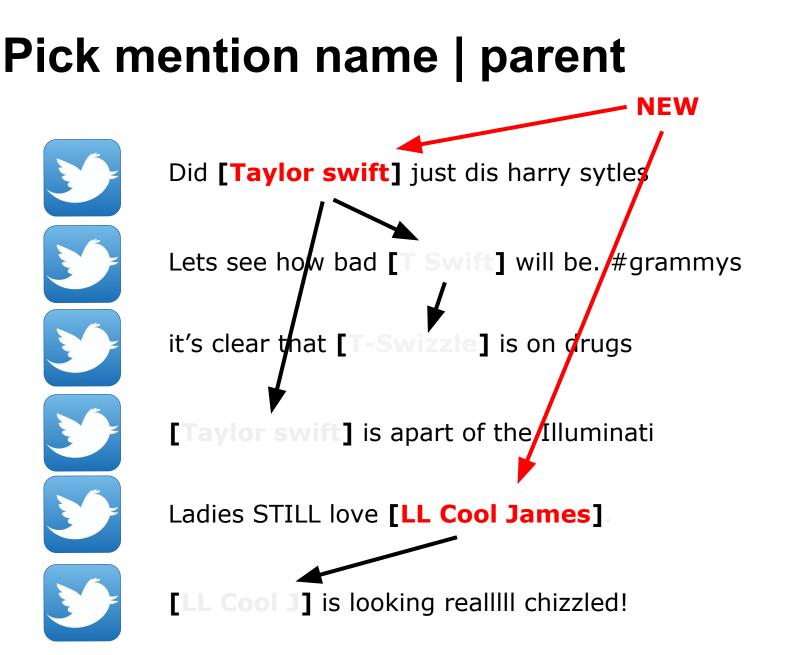


We draw a name from a simple character LM with trainable parameters  $\boldsymbol{\theta}$ 

(... room for fancier distributions over  $\sum^*$  ...)

#### **Pick mention name | parent** FW 5 Did [Taylor swift] just dis harry sytles Lets see how bad [T Swift] will be. #grammys it's clear that [T-Swizzle] is on drugs 5 [Taylor swift] is apart of the Illuminati 10 Ladies STILL love [LL Cool James] 10 [LL Cool J] is looking reallll chizzled!





# **Generating a name**

If parent != NEW: copy (maybe mutate) the parent's name



We draw the name from a stochastic contextual edit model (see Cotterell et al., 2014) with trainable parameters  $\theta$ 

Taylor swift Teeee-Swift Copy Del Subst Copy Stop

# **Generating a name**

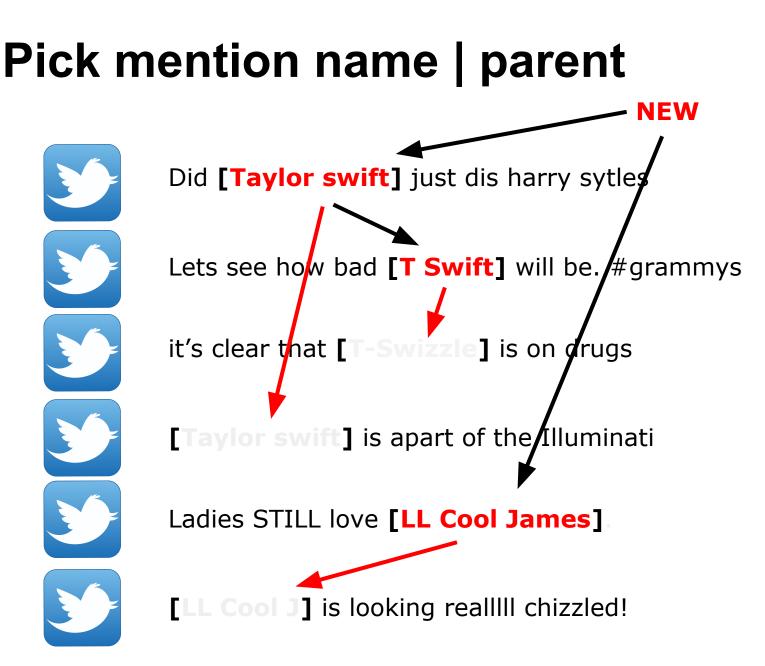
If parent != NEW: copy (maybe mutate) the parent's name

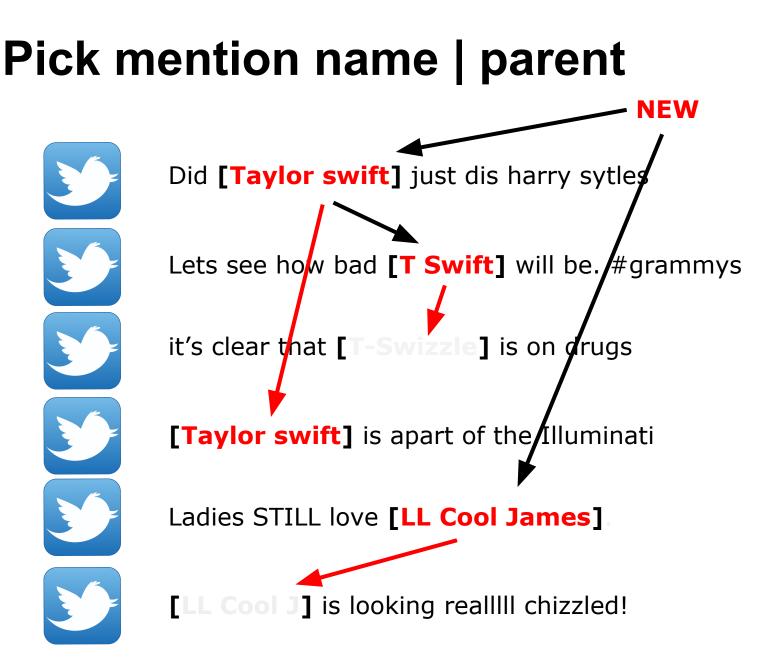
[Taylor swift] [T-Swift]

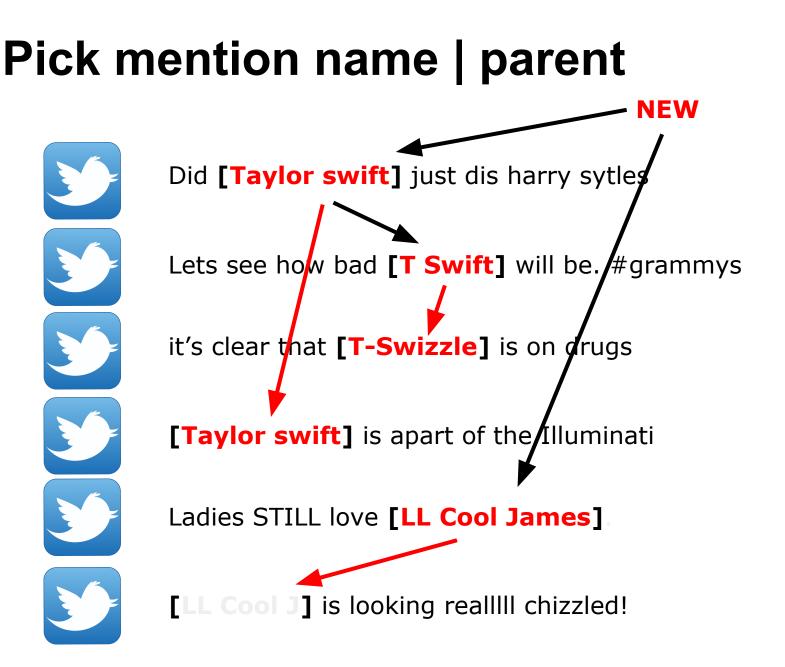
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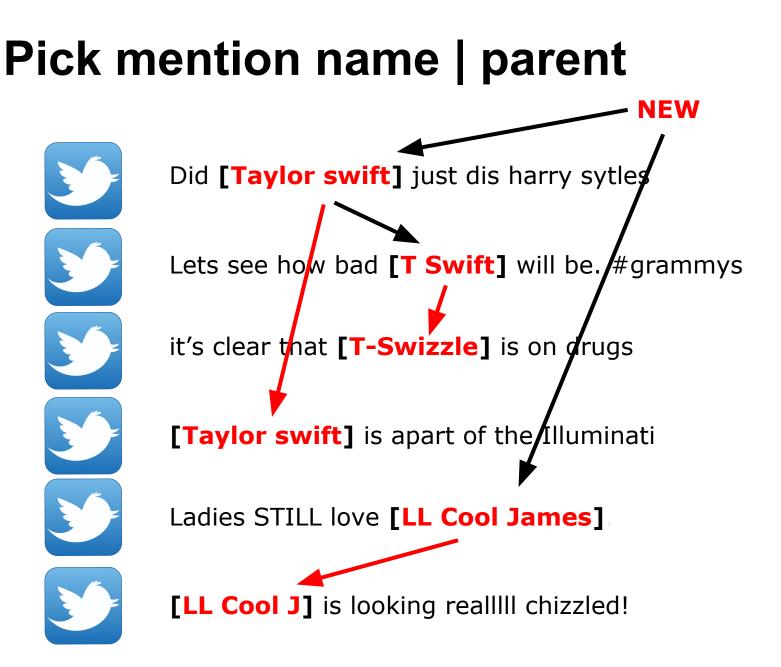
 $\theta$  gives the contextual probability of different character edits

- copy, delete, substitute *c*, insert *c* 

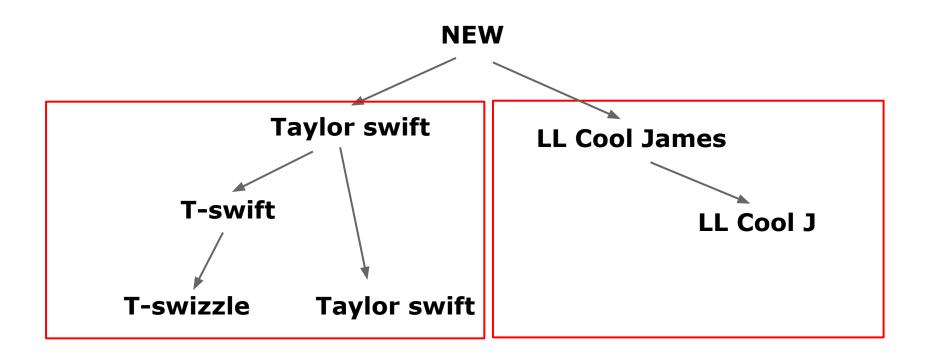




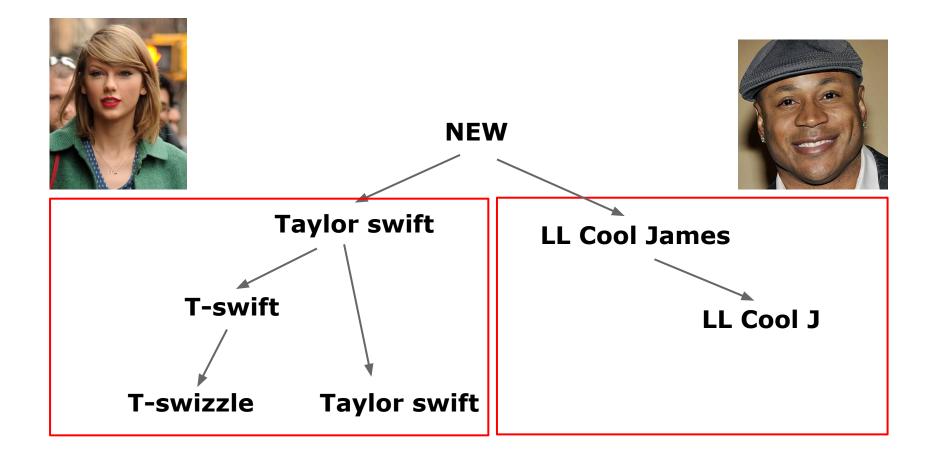




## **End result**



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## **Inference** (fixed parameters)



Did **[Taylor swift]** just dis harry sytles

Lets see how bad **[T Swift]** will be. #grammys

it's clear that **[T-Swizzle]** is on drugs

[Taylor swift] is apart of the Illuminati

Ladies STILL love **[LL Cool James]**.



Did [Taylor swift] just dis harry sytles

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IEW







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Did [Taylor swift] just dis harry sytles

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EW

# Many possible topic assignments

5

Did [Taylor swift] just dis harry sytles



**5** Lets see how bad **[T Swift]** will be. #grammys

#### 5

it's clear that [T-Swizzle] is on drugs

5 [Taylor swift] is apart of the Illuminati

#### 10

Ladies STILL love [LL Cool James].

#### 10

# Many possible topic assignments



Did [Taylor swift] just dis harry sytles



2 it's clear that **[T-Swizzle]** is on drugs

3 [Taylor swift] is apart of the Illuminati

#### 5

Lets see how bad **[T Swift]** will be. #grammys

Ladies STILL love [LL Cool James].

# Many possible topic assignments

3

Did [Taylor swift] just dis harry sytles



Lets see how bad [T Swift] will be. #grammys 3

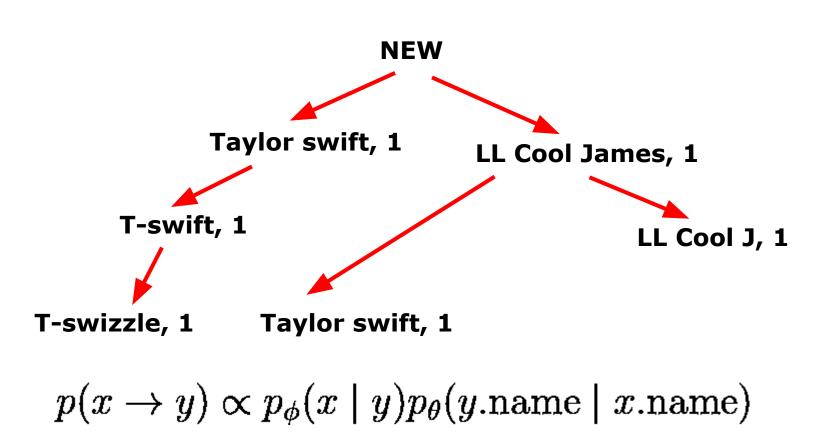
it's clear that [T-Swizzle] is on drugs

3 [Taylor swift] is apart of the Illuminati

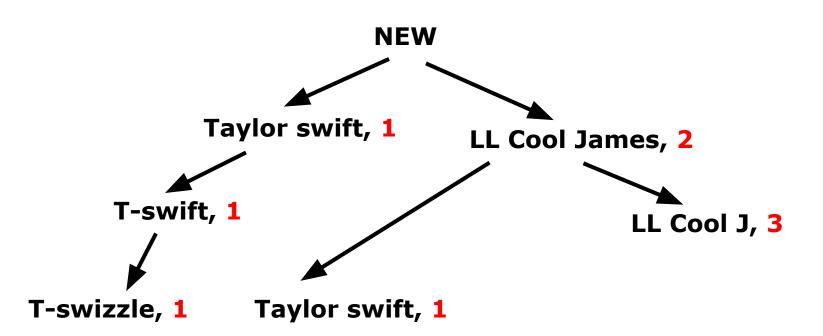
#### 3

Ladies STILL love [LL Cool James].

Sample phylogeny | topics

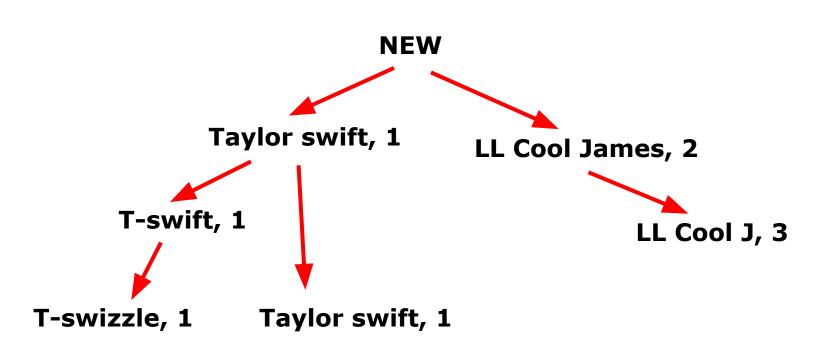


Sample topics | phylogeny

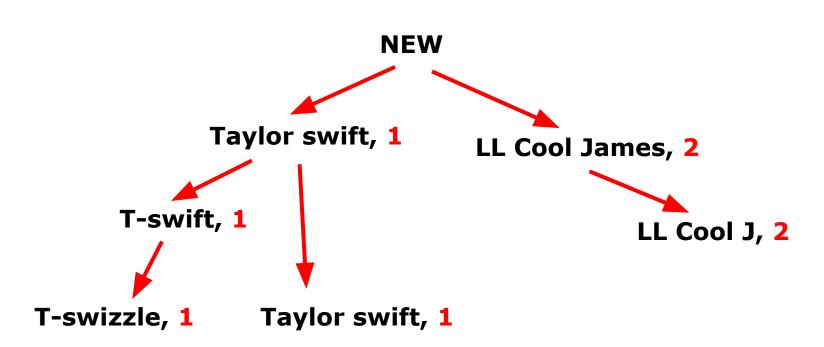


(We use tree-structured BP to construct proposals; see paper for details.)

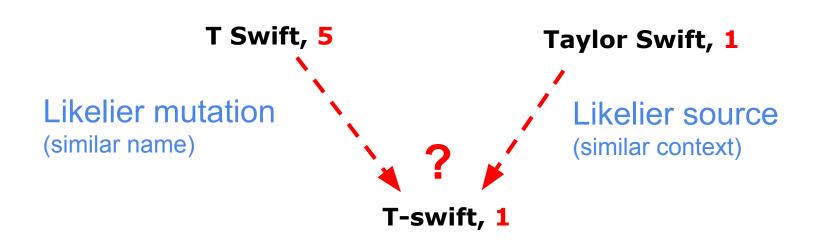
Sample phylogeny | topics



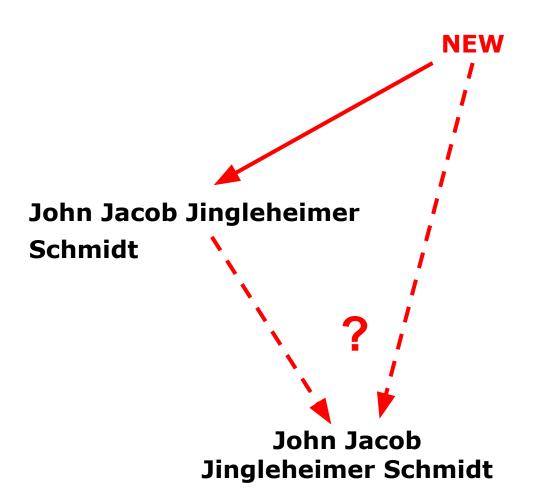
Sample topics | phylogeny



## What is the sampler thinking about?



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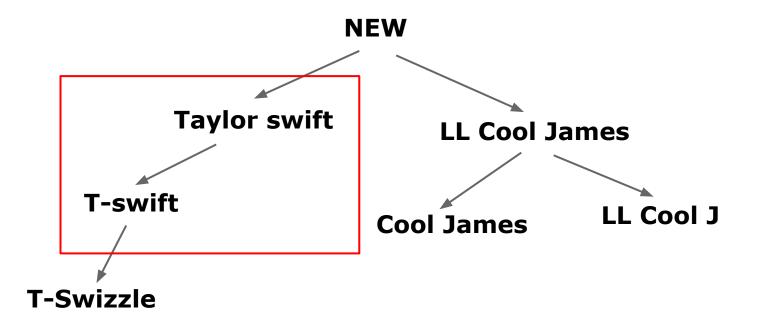
Prefer to **copy** rather than generate from scratch

(short or common names *would* be plausible to generate twice)

(see paper for an improved mutation model that considers pragmatics)

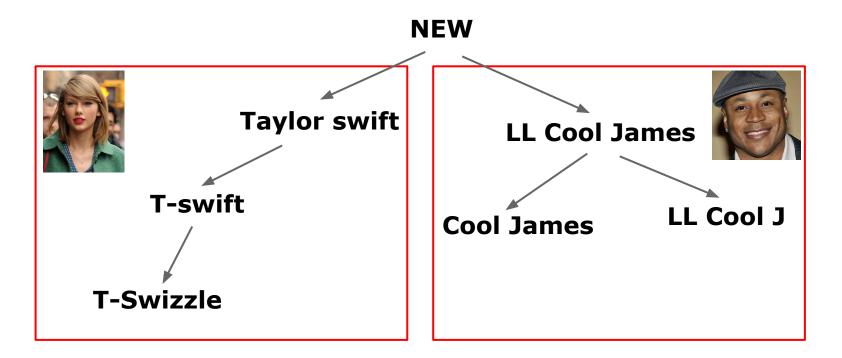
#### What do samples tell us?

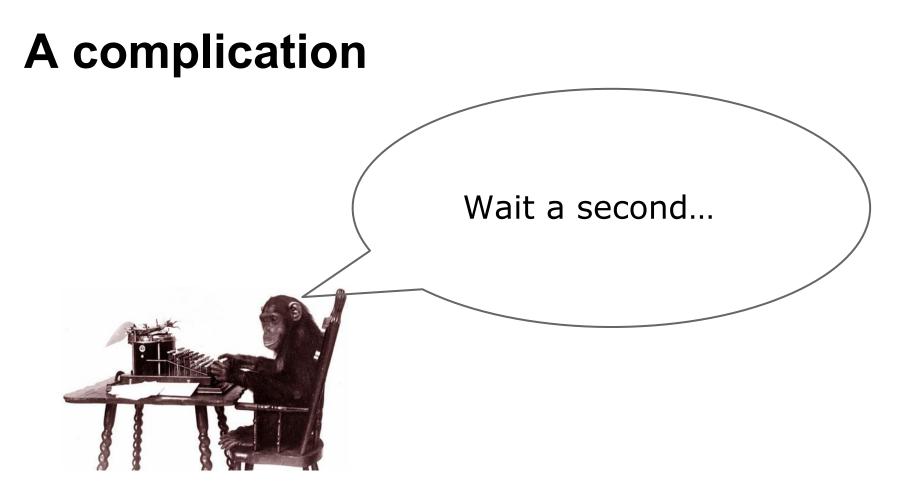
1. Which names are copies of other names (used for EM training of all parameters)



## What do samples tell us?

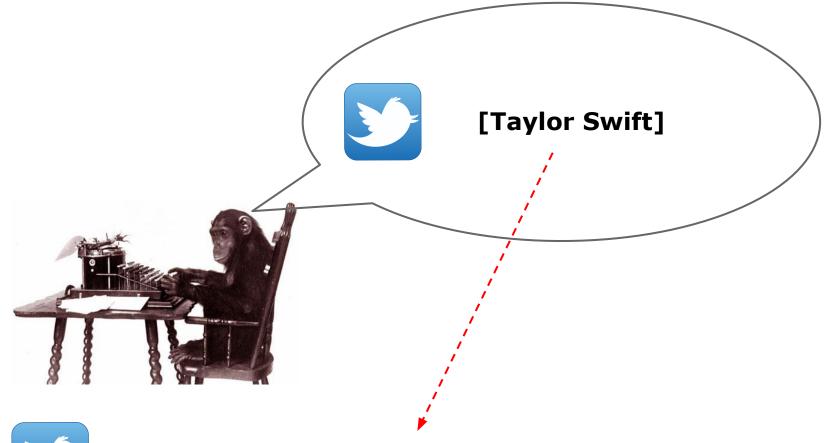
# 2. Which names corefer (used for your IE task)



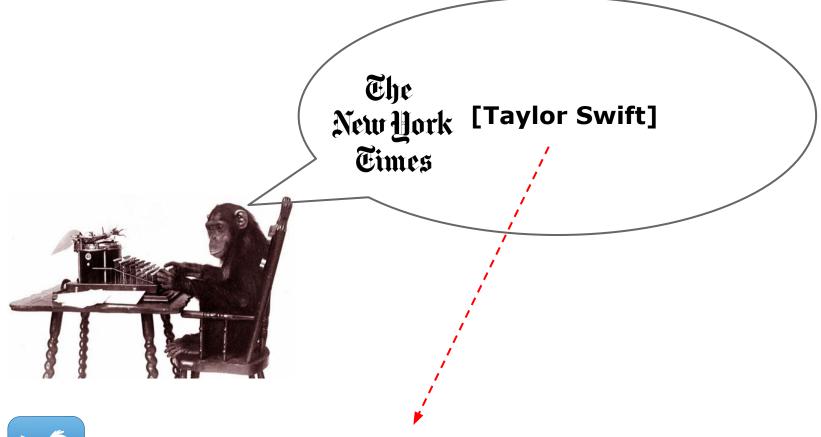




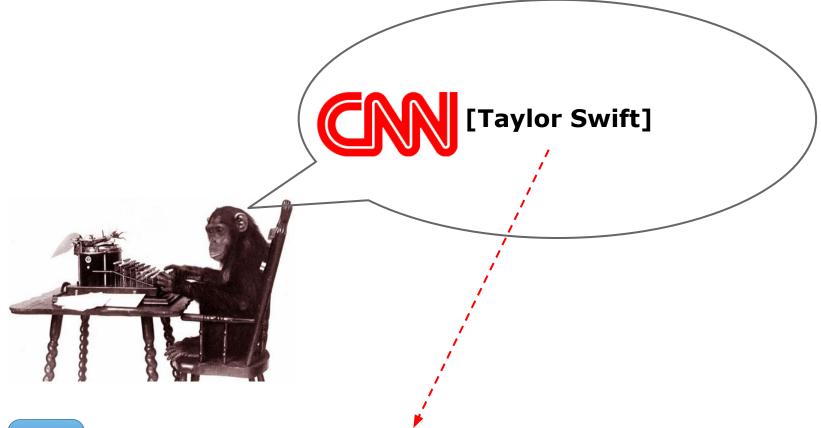
#### Authors copy from multiple sources



#### Authors copy from multiple sources

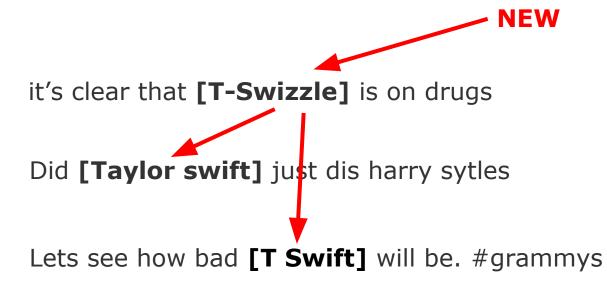


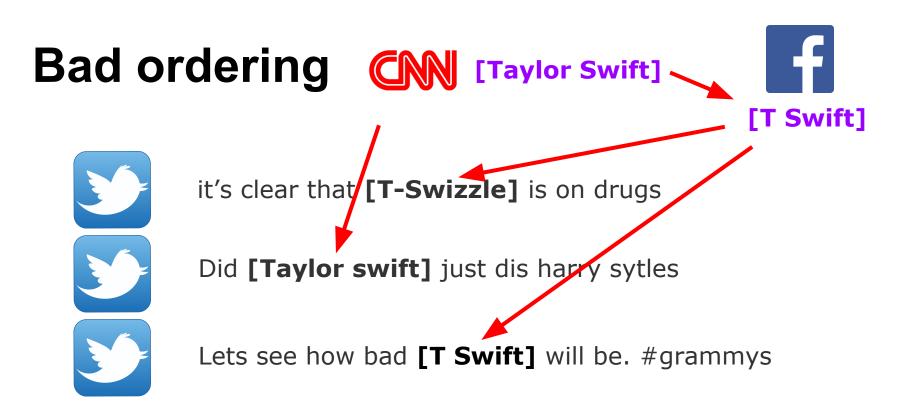
#### Authors copy from multiple sources



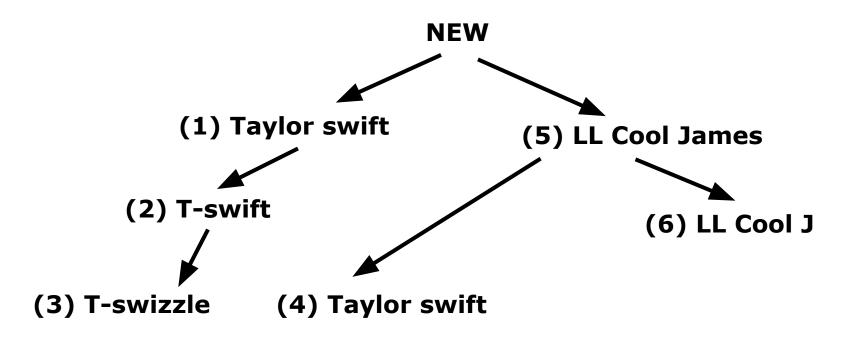
# **Bad ordering**





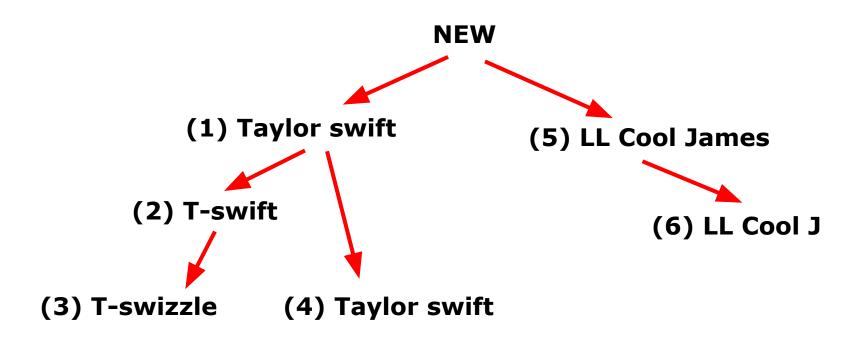


#### Solution: Treat order as unknown



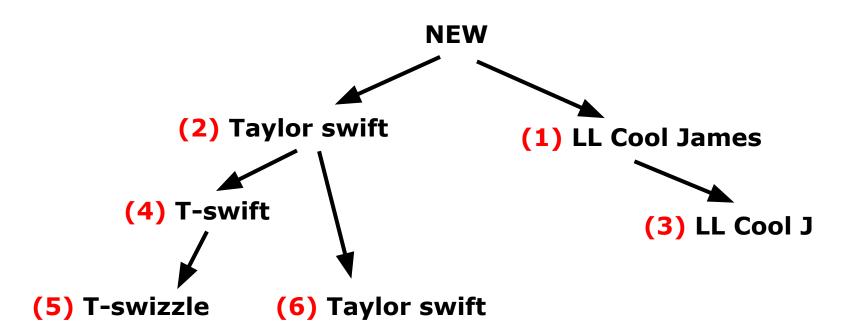
#### **Updated Block Gibbs (fixed params)**

Sample phylogeny | ordering, topics



#### **Updated Block Gibbs (fixed params)**

Sample ordering | phylogeny, topics



(we use a proposal distribution and correct with MH; see paper for details)

# Summary

So far we've seen:

- The generative story
- A sampler for posterior inference

Wrap up with:

- Parameter estimation
- MBR decoding
- Experiments

### **Parameter Estimation**

#### Monte Carlo EM

Repeat:

- E-Step: Count edges in sample
- **M-Step:** Take stochastic gradient step
  - Update\* parent model parameters
  - Update\* mutation model parameters

# **Consensus Clustering**

To get a single "hard" clustering C for evaluation, we use minimum Bayes risk:

#### $\operatorname{argmin}_{C} \mathbf{E}_{C'} [ \operatorname{loss}(C, C') ]$

- Minimize expected loss of C
  - With respect to C' drawn from the model posterior
  - Estimate this using our samples of C'

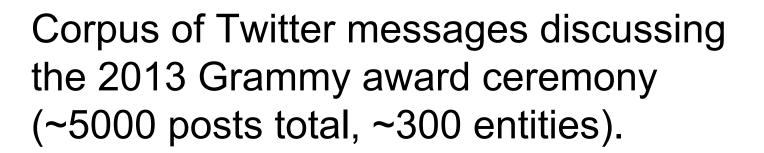
(we approximate argmin using spectral clustering; see paper for details)

# Summary

#### **Unsupervised clustering procedure:**

- 1. Train model using Monte Carlo EM
- 2. Sample from the posterior
- 3. Pick the MBR clustering given posterior

# Evaluation: Twitter



- Procedure: 4-fold cross validation
  - **Train:** Tune weight of picking NEW as the parent to control precision / recall trade-off
  - **Test:** Run clustering procedure with this weight fixed

System	B <sup>3</sup> F1
Exact-Match	69.8

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Exact-Match	69.8
Green et al. (2012)	79.3

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Phylo (no context)	88.7

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Phylo + Topic	91.8

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Phylo + Topic	91.8
Phylo + Topic + MBR	91.9

# Results: ACE 2008

Corpus of news articles, mostly politics ~4000 mentions, ~2000 entities

#### • Procedure

**Train:** Tune weight of picking NEW as the parent to control precision / recall trade-off

Test: Run clustering procedure with this weight fixed

#### **Evaluation: ACE 2008**

System	PER B3 F1	ORG B3 F1
Exact-Match	88.8	87.1

#### **Evaluation: ACE 2008**

System	PER B3 F1	ORG B3 F1
Exact-Match	88.8	87.1
Green et al. (2012)	91.9	90.3

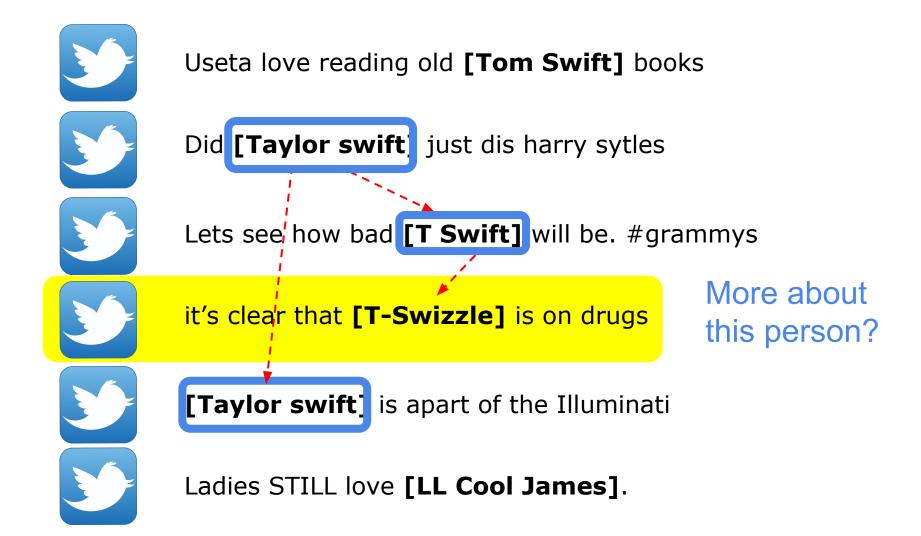
#### **Evaluation: ACE 2008**

System	PER B3 F1	ORG B3 F1
Exact-Match	88.8	87.1
Green et al. (2012)	91.9	90.3
Phylo+Topic+MBR	92.7	87.6

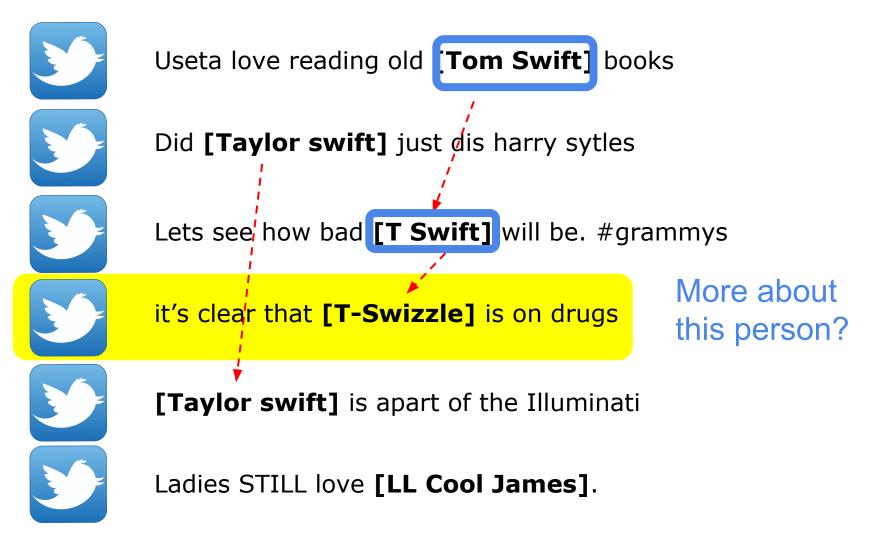
#### Thanks!

- More experiments in the paper
   Name canonicalization
- Code will be released soon
  - <u>https://bitbucket.org/noandrews/phyloinf</u>
- Future uses of model and code?
  - Track diffusion of memes through social media
  - Derivational morphology

### Samples tell us which entities corefer



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### **Coref probabilities from many samples**

