



EMNLP
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Infinite SCAN: An Infinite Model of Diachronic Semantic Change

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Sense Modeling with a Topic Model

1

Creating Inputs

target word: "coach"



Diachronic Corpus



Usage with Time Stamp

```
1853 The driver made room for the trunk on the top of thecoach.
1900 The chair passed the coach, the horses proceeding at a walk.
1949 Tell him if I start coaching, it'll be as a head coach at a top school.
2003 Football coach and other top school officials have been interviewed.
```

⋮



Snippets of Contextual Information

```
1853 [driver, make, room, trunk]
1900 [chair, pass, horse, proceed, walk]
1949 [tell, start, coach, head, top, school]
2003 [football, top, school, official, interview]
```

⋮

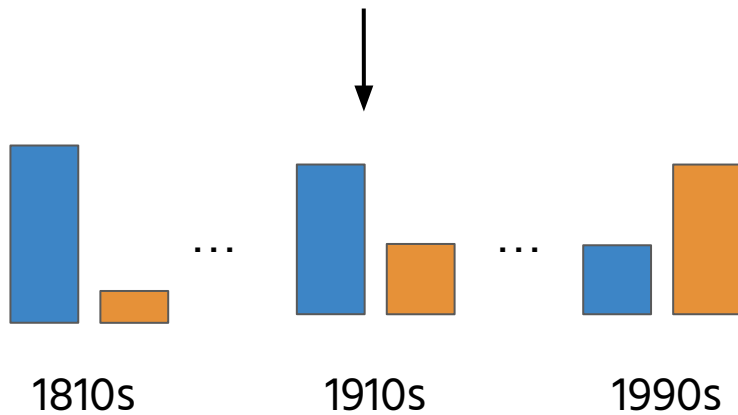
Sense Modeling with a Topic Model

2

Sense Modeling

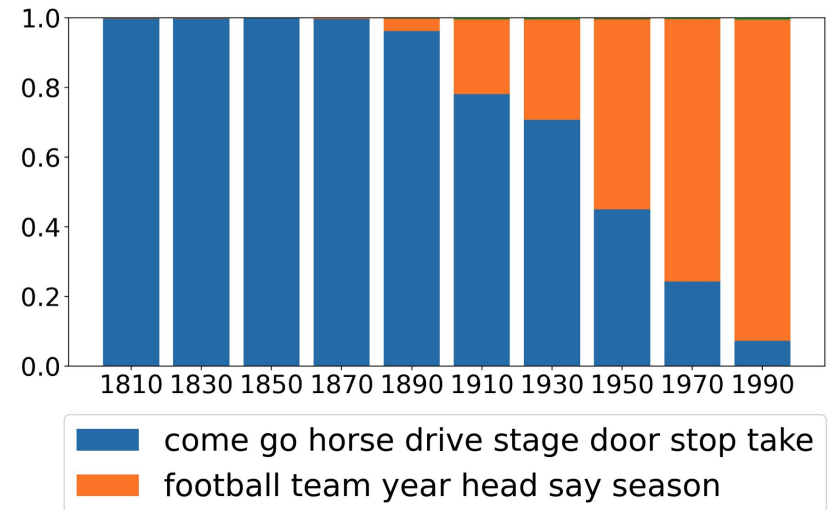
Snippets (inputs)

```
1853 [driver, make, room, trunk]
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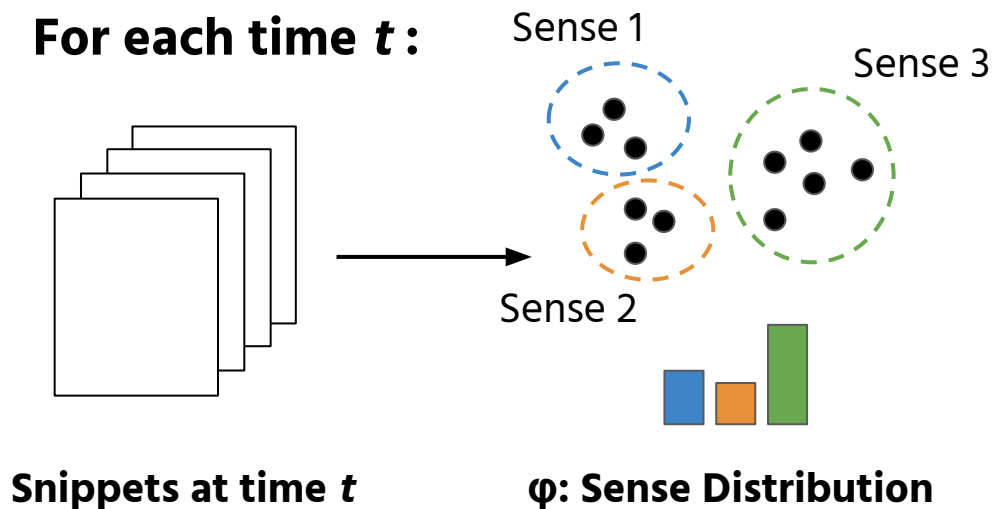
3

Analysis

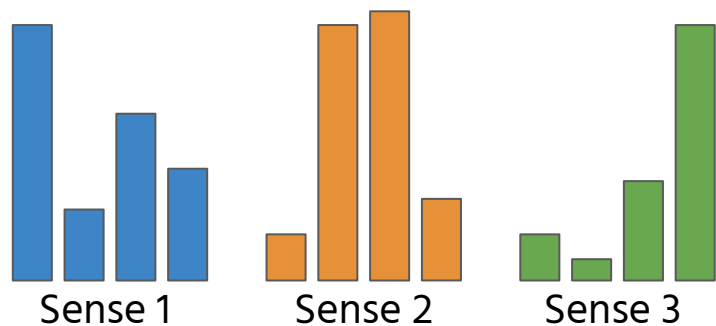


Preliminaries: SCAN

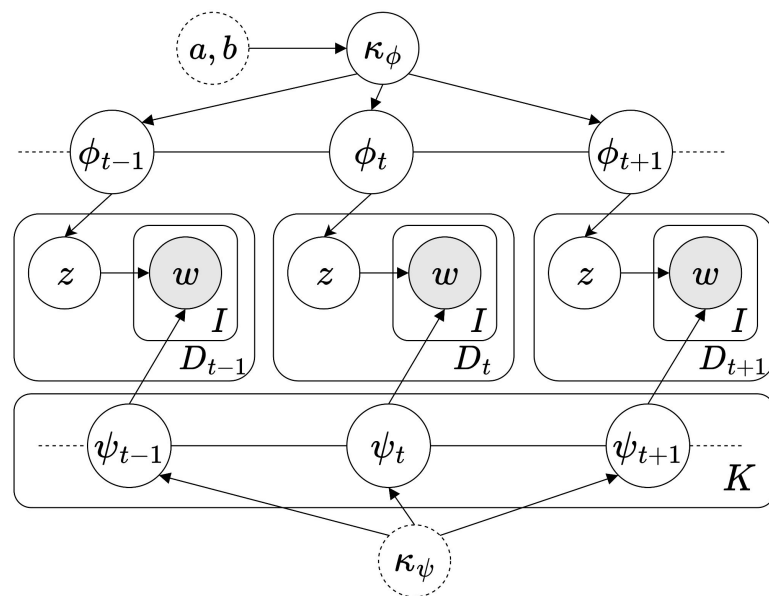
SCAN: Bayesian Model of Sense Change [Frermann+, 2016]



Snippets at time t

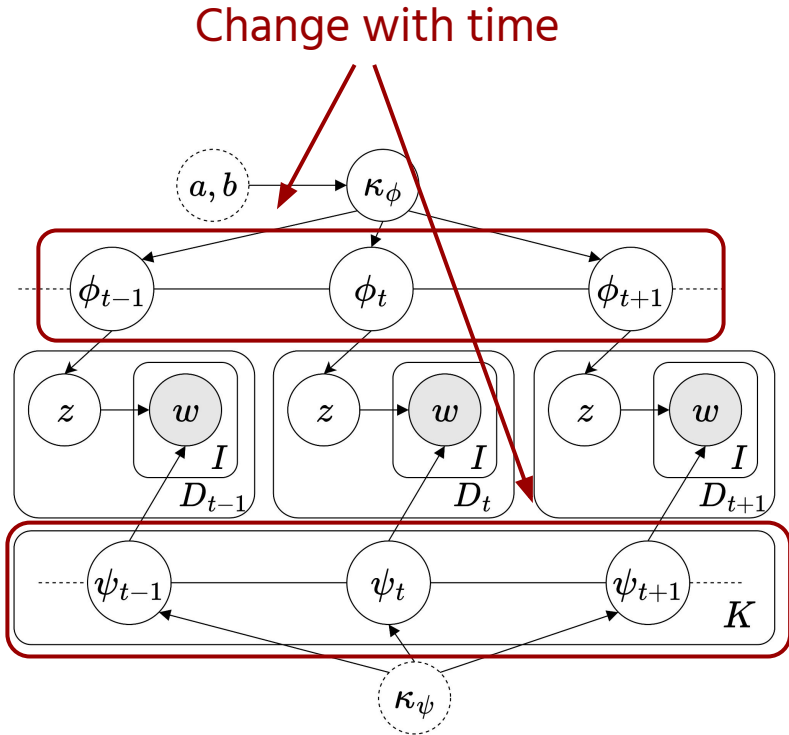
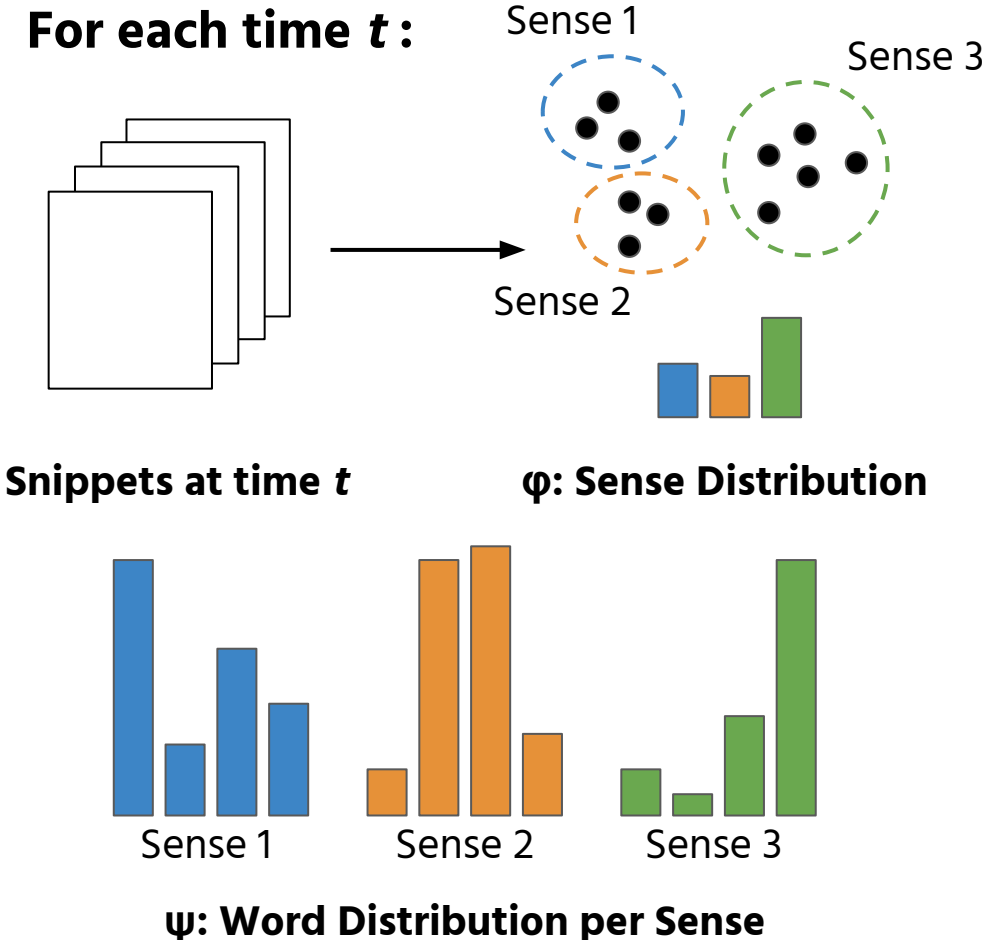


ψ : Word Distribution per Sense



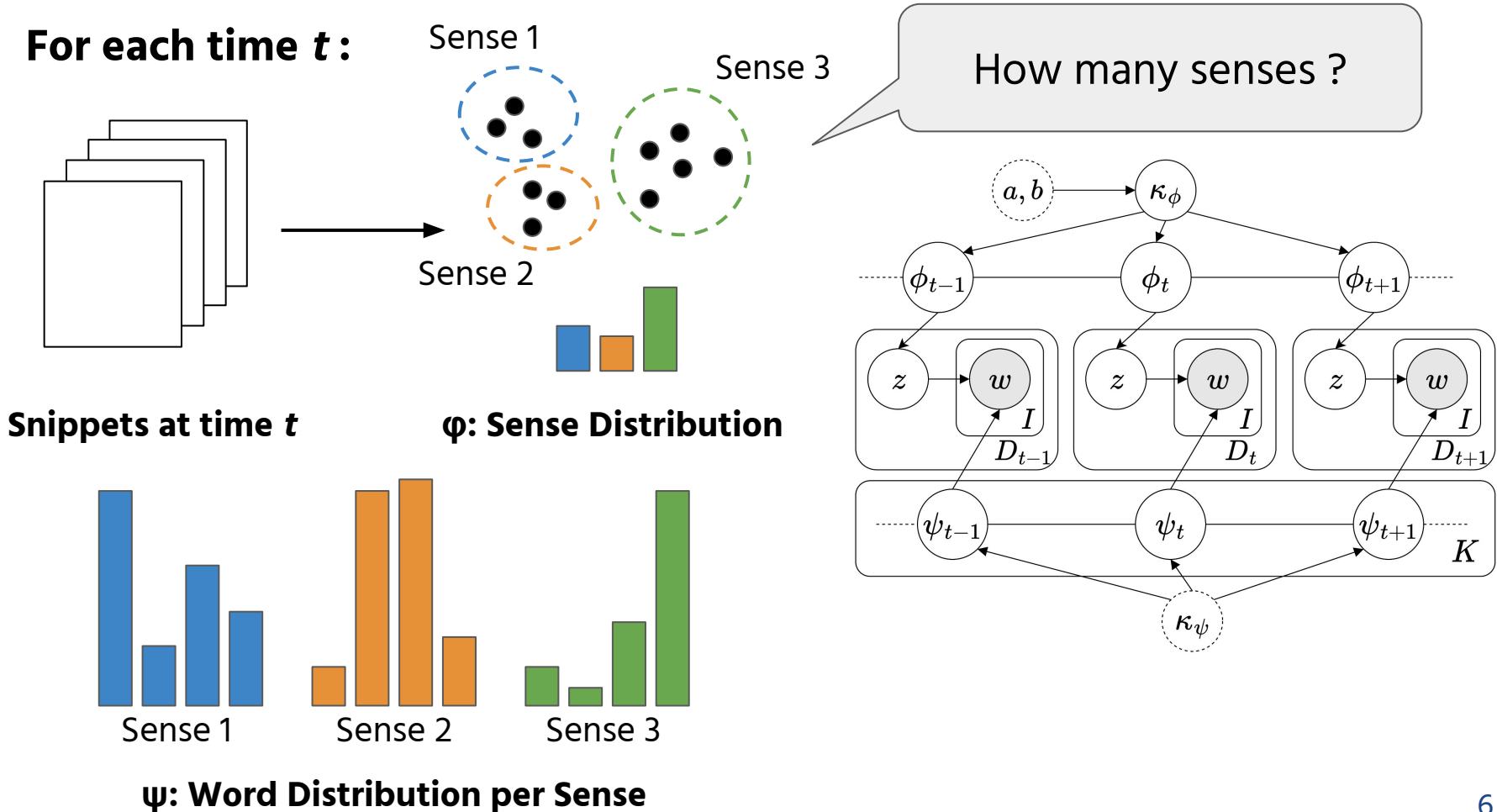
Preliminaries: SCAN

SCAN: Bayesian Model of Sense Change [Frermann+, 2016]



Preliminaries: SCAN

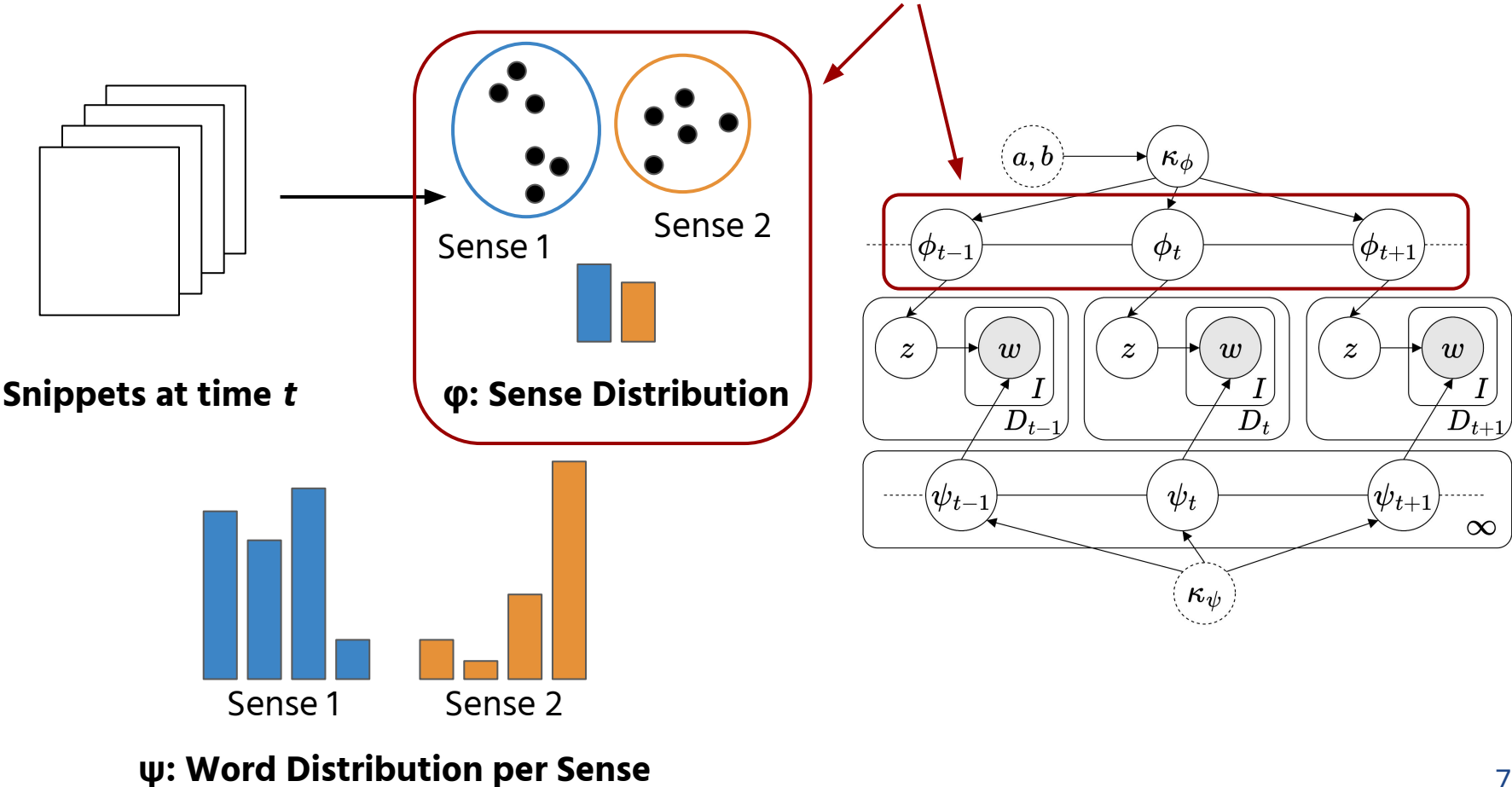
SCAN: Bayesian Model of Sense Change [Frermann+, 2016]



Proposal: Infinite SCAN

Infinite SCAN: Infinite Model of Diachronic Semantic Change

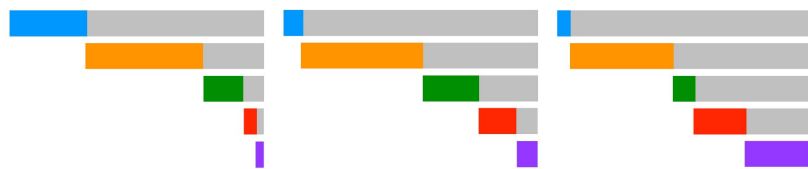
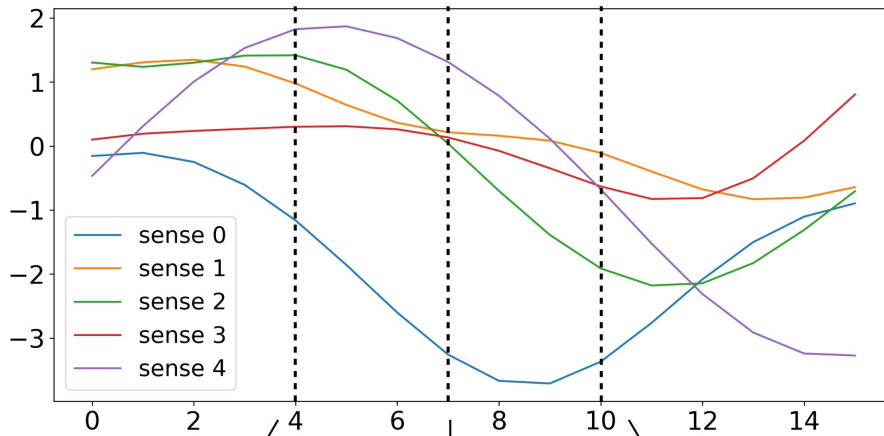
Automatically estimate the number of senses using Dirichlet process (Stick-breaking process)



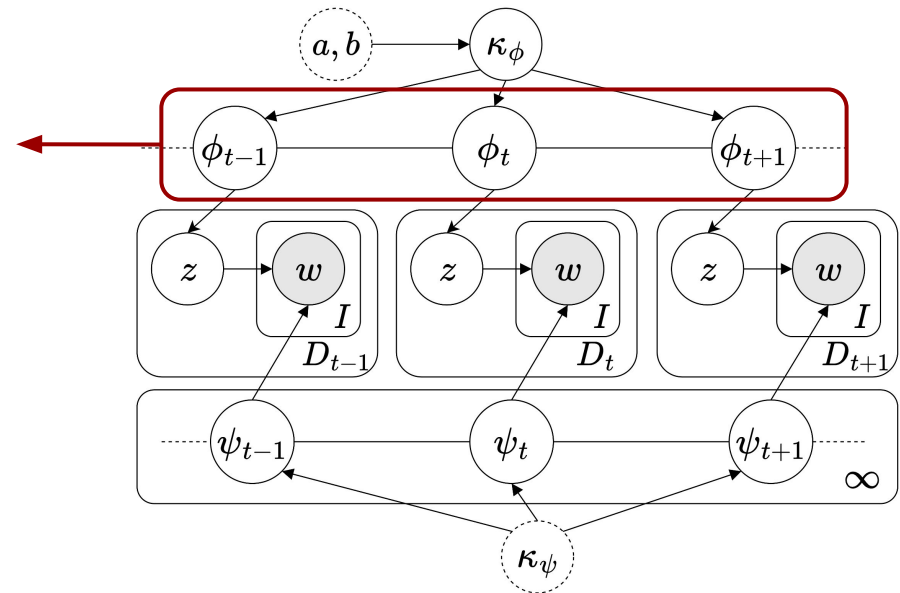
Proposal: Infinite SCAN

Stick-breaking transformation of Gaussian sense distribution

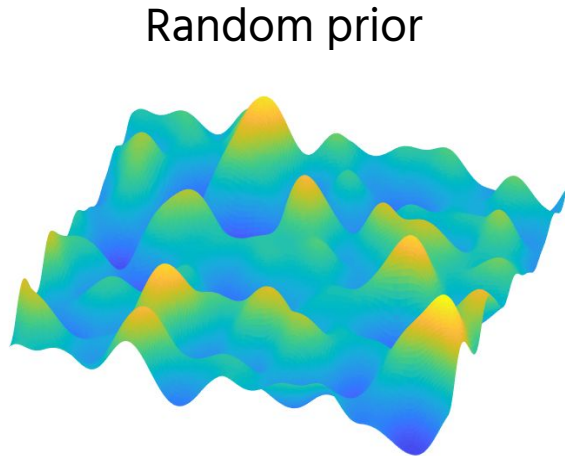
Logistic normal priors on
Gaussian Markov Random Field



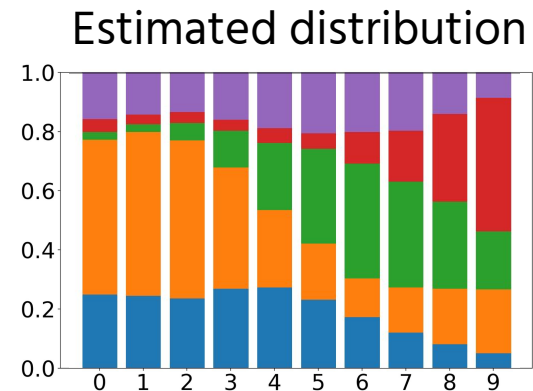
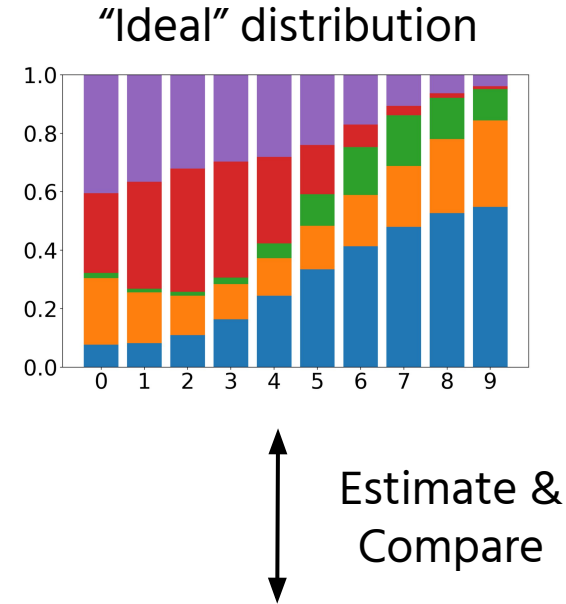
Projecting to K-1 simplex using Logistic
stick-breaking process (not softmax)



Setup: Experiments on Artificial Data



Draw



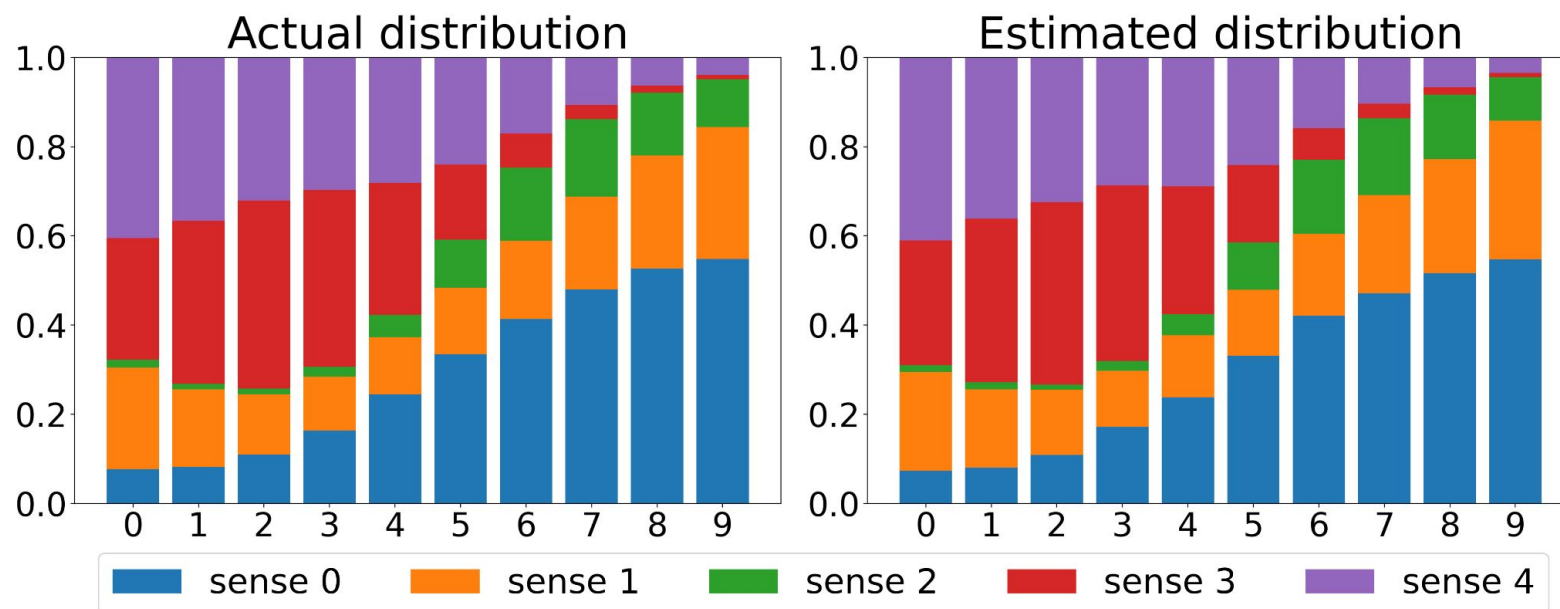
- Sense distribution: Draw from Gaussian process
- Sense-word distribution: Draw from Zipfian distribution

Experiments on Artificial Data

For artificial data with ranging the number of senses 1 to 5:

- Correctly estimate the number of senses except for $S = 1$
- Properly estimate the actual distribution: $KL(\text{Actual} \parallel \text{Estimated}) \approx 0.0$

Example for the number of senses: $S = 5$



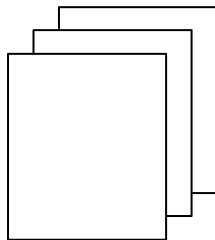
*Kullback-Leibler divergence (Actual \parallel Estimated) = **0.004***

Setup: Experiments on Real Data

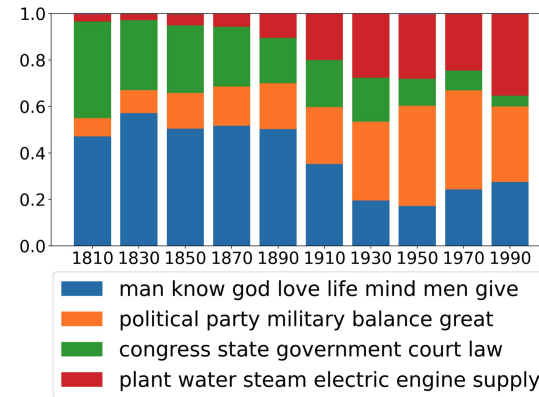


Diachronic Corpus
CCOHA [Alatrash+, 2020]

Extract



Snippets



Estimated distribution

Evaluate

Gold number of
senses

Sense inventory of
OntoNotes [Hovy+, 2006]

Known
Sense change

Previous work
[Giulianelli+, 2020;
Frermann+, 2016]

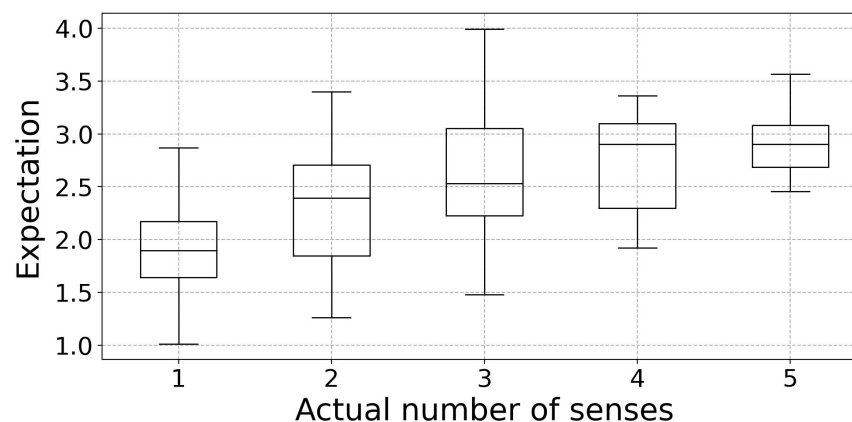
Experiments on Real Data

Can the model correctly estimate the number of senses?

For 120 target words randomly selected from OntoNotes:

Model	Accuracy	PCC
HDP-LDA	0.258	0.019
BERT + K-means	0.217	0.026
BERT + DBSCAN	0.125	-0.070
SCAN ($K = 5$)	0.158	0.141
SCAN ($K = 8$)	0.000	0.087
Infinite SCAN	0.358	0.474

Prediction results for the number of senses.

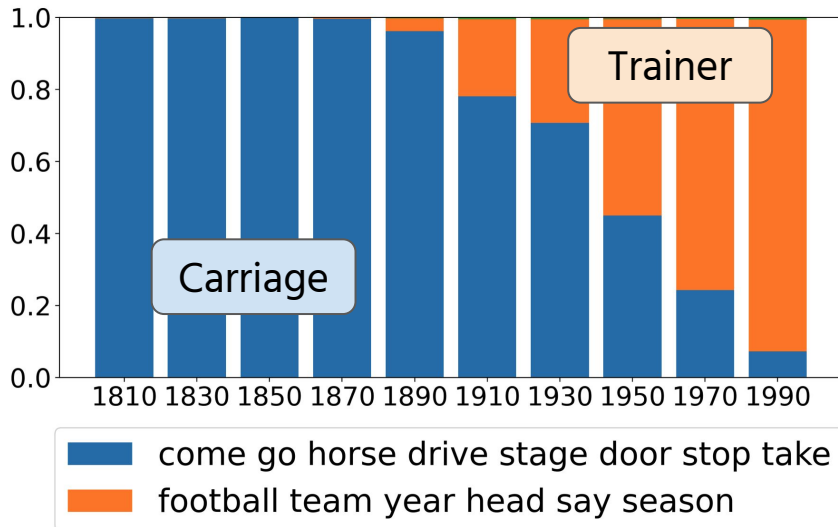


Correlation between actual and estimated number of senses by Infinite SCAN.

Experiments on Real Data: “coach”

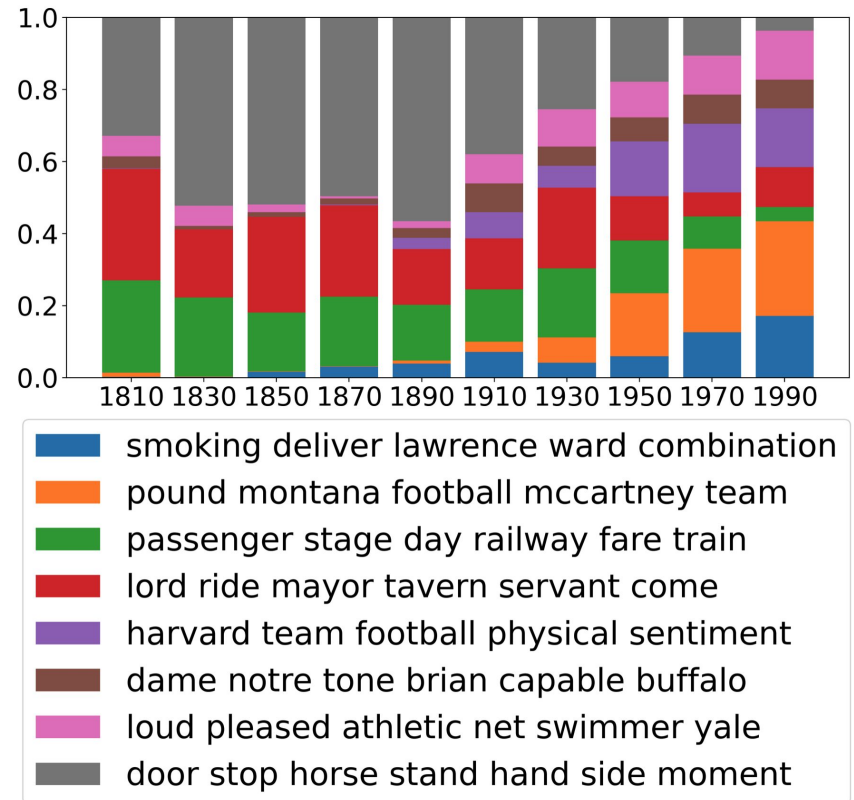
Can the model properly describe the semantic change?

Infinite SCAN



Semantic shift: “carriage” → “trainer”

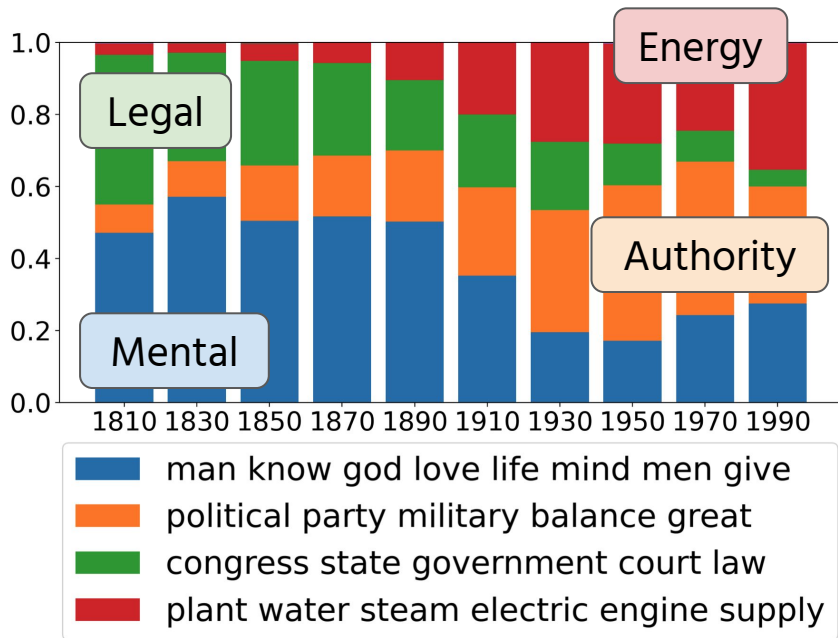
SCAN



Experiments on Real Data: “power”

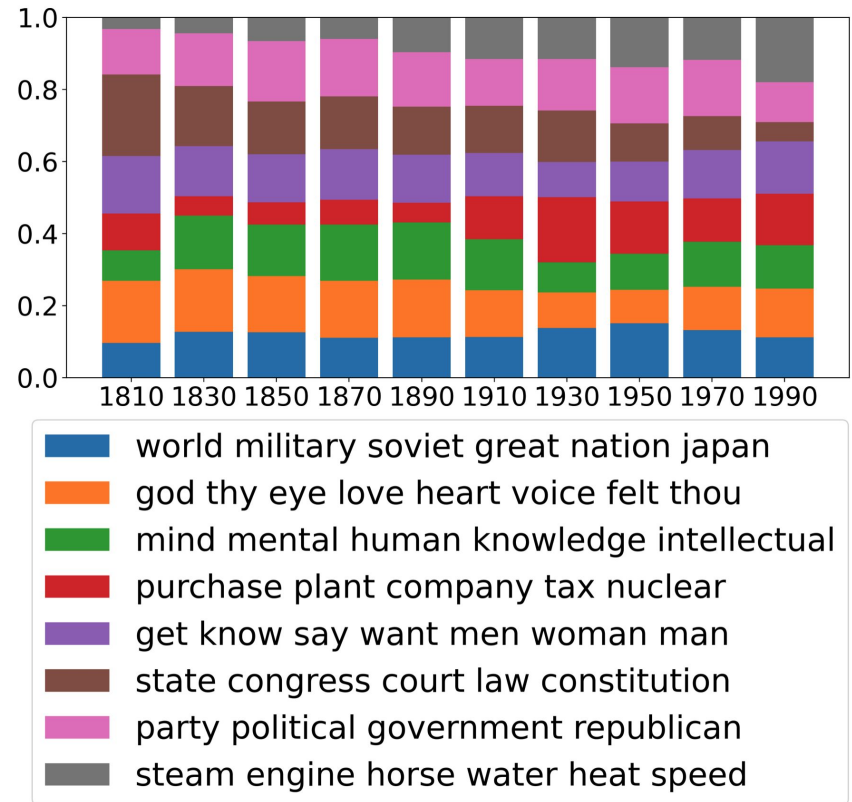
Can the model properly describe the semantic change?

Infinite SCAN



Sense birth: “energy”

SCAN



Conclusions / Future Work

- Infinite SCAN: An infinite model of diachronic semantic change
 - based on dynamic topic model
 - combining Gaussian Markov random fields with a Dirichlet process
- Outperformed baseline models
 - interpretability of model output
 - estimation of the number of senses
- Captured semantic changes in line with previous studies
- In future, working on classification of change patterns

Appendices: Setup of experiments on Artificial Data

Generation of artificial data

- Sense distribution: Draw from Gaussian process with the number of senses K
- Sense-word distribution: Draw from Zipfian distribution with the vocabulary size V

Parameter settings

- Number of time points: $T = 10$
- Original vocabulary size: $V = 5,000$
- Snippet length: $l = 10$
- Number of senses: $S = 1 \sim 5$
- Sample size: $N = 10,000$

Metric

- Kullback-Leibler distance between actual and estimated distribution

#Senses	SCAN		Infinite SCAN
	$K = 5$	$K = 8$	
1	1.523	1.997	0.468
2	0.335	0.578	0.039
3	0.216	0.735	0.030
4	0.212	0.150	0.061
5	0.004	0.017	0.004

Table 1: Kullback-Leibler divergence (lower is better) between actual sense distribution and sense distribution estimated by each model for the artificial data.

Appendices: Setup of experiments on Real Data

Historical corpus

Clean Corpus of Historical American English (CCOHA) [Alatrash+, 2020]

Target words with gold number of senses

randomly selected 120 words (noun, verbs) with $S \leq 5$ from OntoNotes [Hovy+, 2006]

Target words with semantic change description

- “*coach*” [Giulianelli+, 2020; Aida+, 2021]
- “*power*” [Frermann+, 2016]

Baseline models

- HDP-LDA [Teh+, 2010]
- BERT (base-uncased) [Devlin+, 2019] + K-means, DBSCAN
- SCAN [Frermann+, 2016]