### Discovering Latent Structure in Task-Oriented Dialogues

Ke Zhai and Jason D. Williams

June 22, 2014

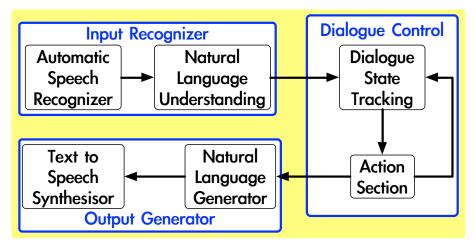
Dept. of Computer Science University of Maryland College Park zhaike@cs.umd.edu



Microsoft Research Redmond Jason.Williams@microsoft.com

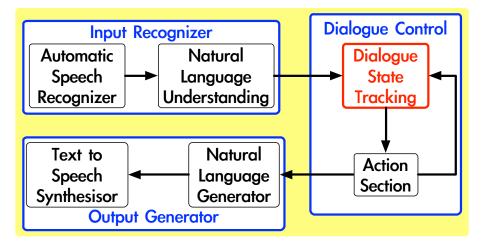
## Microsoft<sup>®</sup> Research

<sup>1</sup>Internship work at Microsoft research, Redmond



- Input Recognizer: speech-to-text and sematic analysis
- Dialogue Control: flow control and decision making
- Output Generator: natural language and speech generator

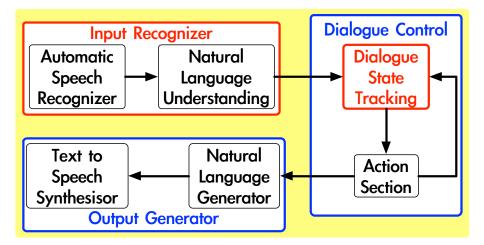
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#### Key Challenge

tracking dialogue states and understanding conversation flows

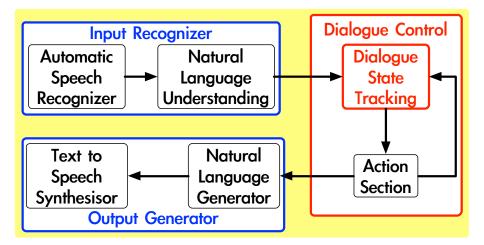
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#### Input Recognition System

feedback with hypothesis to improve recognition accuracy

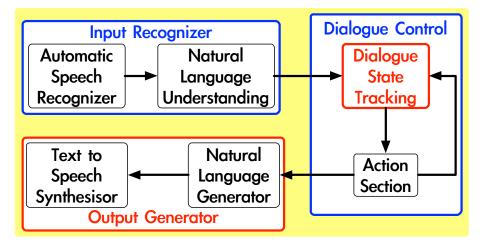
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#### **Dialogue Control System**

control dialogue flow, for example, reinforcement learning

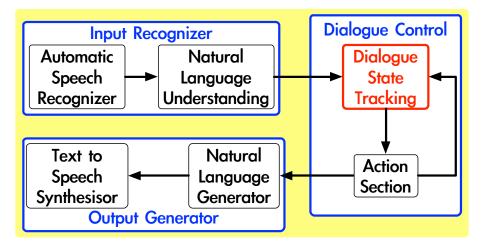
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#### **Output Generator System**

model languages in conversation and generate meaningful output

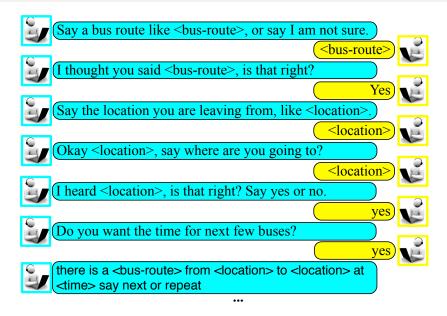
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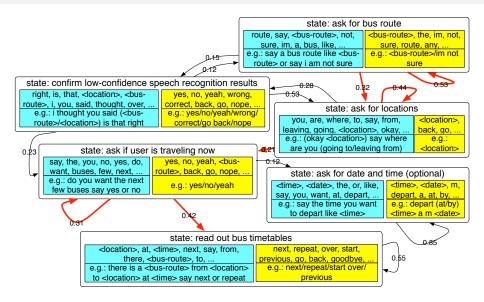


#### Key Challenge: Dialogue State Tracking

Automatically! Unsupervised! Annotation-free!

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### RoadMap



- Existing Models: LM-HMM and LM-HMMS
- Proposed Models: ТМ-НММ and ТМ-НММSS
- 3 Experiments: qualitative and quantitative evaluations
  - Conclusion: what's exciting in Zhai & Williams (2014)...

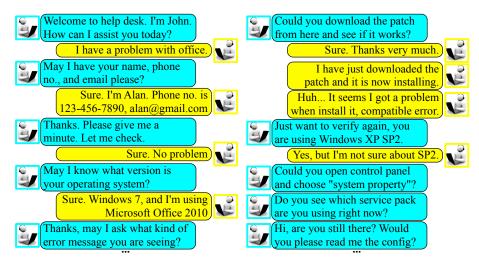
### Outline



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- Proposed Models: TM-HMM and TM-HMMSS
- 3 Experiments: qualitative and quantitative evaluations
  - 4 Conclusion: what's exciting in Zhai & Williams (2014)...

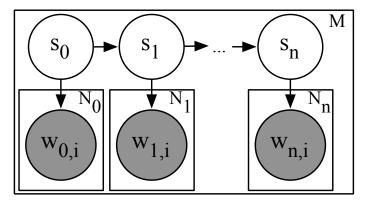
### Motivating Dialogue Example



TechSupport: a set of human-human online technical support chats

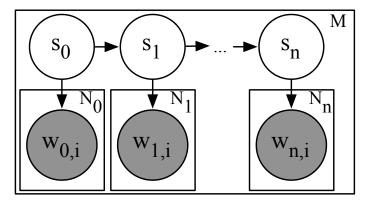
### Conversation Model (Barzilay & Lee, 2004)

LM-HMM: HMM with state language model



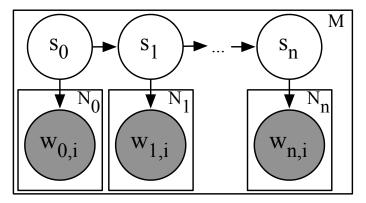
### Conversation Model (Barzilay & Lee, 2004)

LM-HMM: HMM with state language model



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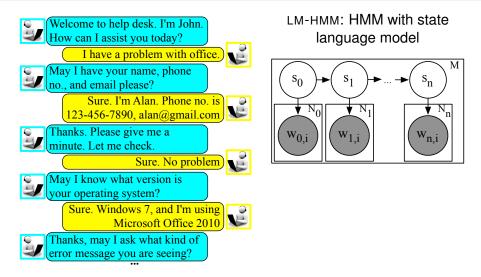
LM-HMM: HMM with state language model



#### Problem

- not ALL words are state-dependent.
- ignores commonalities across different states.

### **Revisit Motivating Dialogue Example**



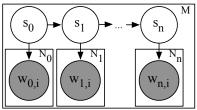
Similar dialogue flow patterns, but different problems/domains/topics.

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### **Revisit Motivating Dialogue Example**



# LM-HMM: HMM with state language model

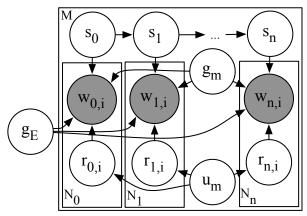


- not all words are state-dependent
- ignores commonalities across different states/sessions

Similar dialogue flow patterns, but different problems/domains/topics.

### Conversation + Topic Model (Ritter et al., 2010)

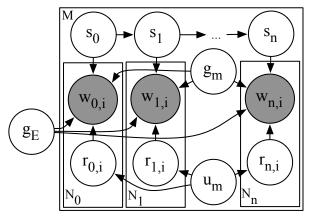
LM-HMMS: HMM with state language model and source generator



According to source indicator  $r_{0,i}$ , generate a word from: 1) *state language* model; 2) *session topic*  $g_m$ ; or 3) background topic  $g_E$ .

### Conversation + Topic Model (Ritter et al., 2010)

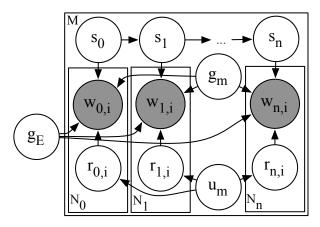
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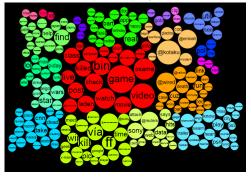


but still, does not share topics across different sessions fine for social media dialogues, e.g. twitter chats but not so well for task-oriented dialogues

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### Free-style Dialogues vs. Task-oriented Dialogues

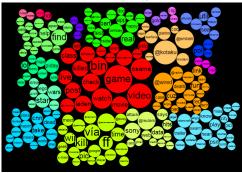
Free-style Diffused Topics

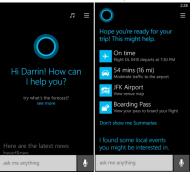


### Free-style Dialogues vs. Task-oriented Dialogues

#### Free-style Diffused Topics

#### Task-oriented In-domain Topics





### Free-style Dialogues vs. Task-oriented Dialogues

#### Free-style Diffused Topics

#### Task-oriented In-domain Topics



#### **Task-oriented Dialogues**

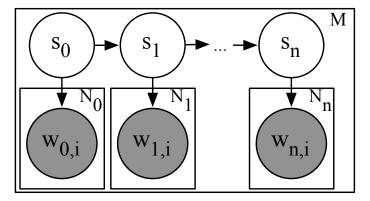
Topics are shared across different dialogues, or maybe even across different utterances in a dialogue session.

### Outline

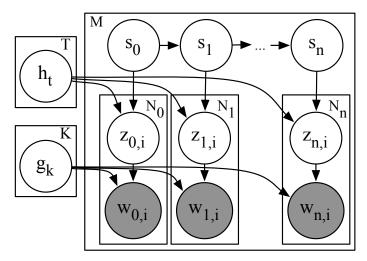


- Existing Models: LM-HMM and LM-HMMS
- Proposed Models: ТМ-НММ and ТМ-НММSS
- 3) Experiments: qualitative and quantitative evaluations
- Conclusion: what's exciting in Zhai & Williams (2014)...

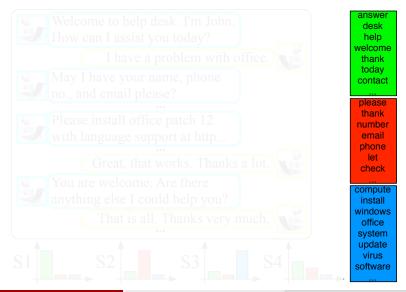
#### LM-HMM: HMM with state language model



<u>⊢мтм</u>-нмм: HMM with state-language topic model

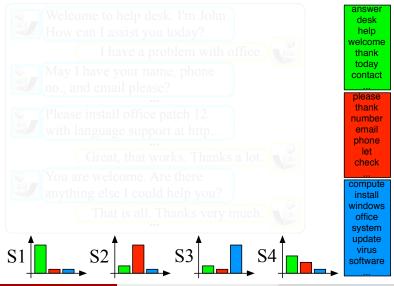


#### тм-нмм: HMM with state topic model



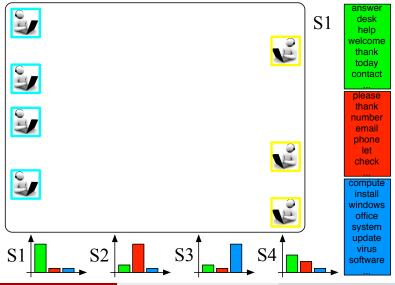
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#### тм-нмм: HMM with state topic model



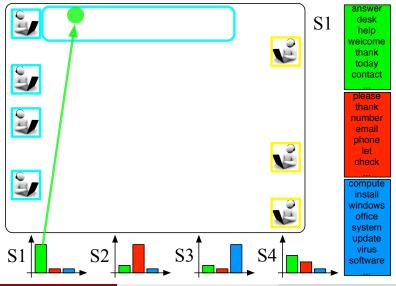
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**Dialogue Models** 



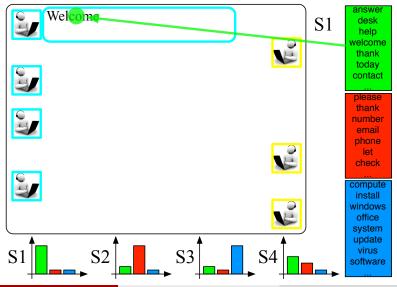
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**Dialogue Models** 



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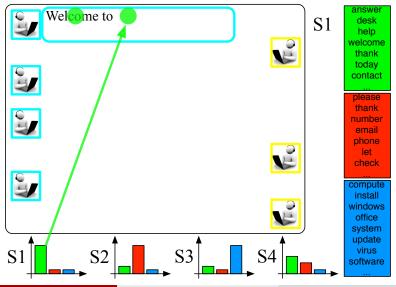
**Dialogue Models** 



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**Dialogue Models** 

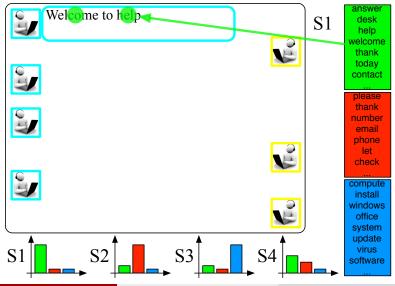
тм-нмм: HMM with state topic model



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**Dialogue Models** 

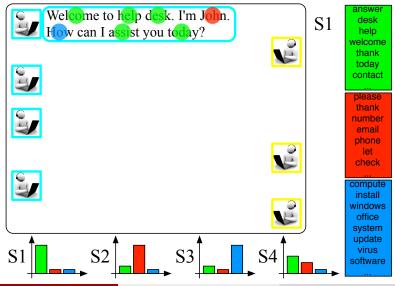
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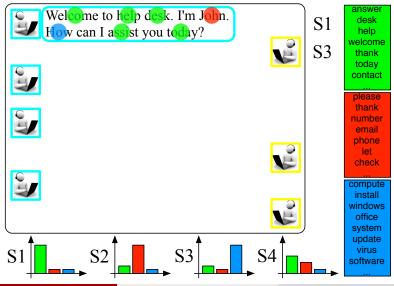
**Dialogue Models** 

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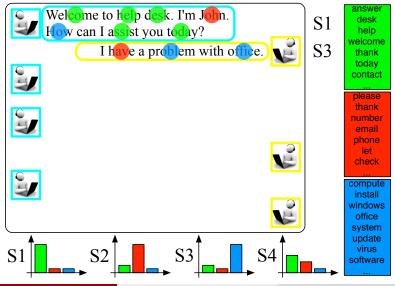
**Dialogue Models** 



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**Dialogue Models** 

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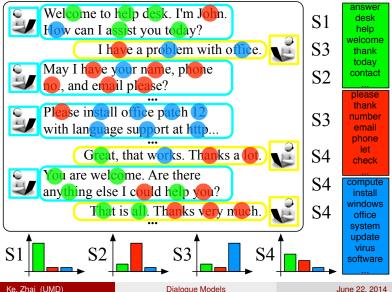


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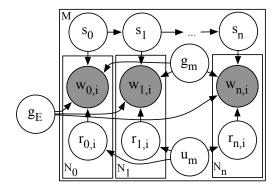
**Dialogue Models** 

#### тм-нмм: Share Topics across States

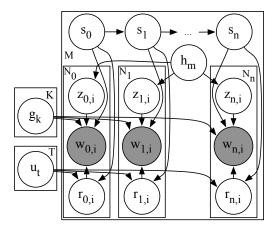
#### тм-нмм: HMM with state topic model



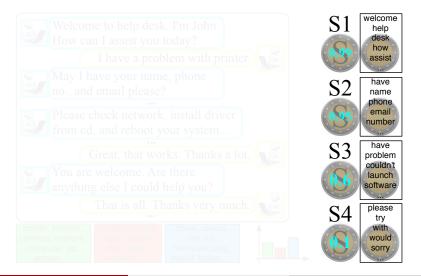
#### LM-HMMS: HMM with state language model and source generator



LMTM-HMMSS: HMM with state language model and state source generator and session topic model

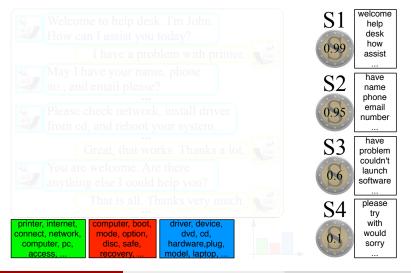


#### TM-HMMSS: HMM with state language model and state source generator and session topic model



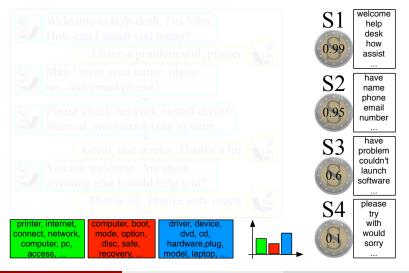
#### Ke, Zhai (UMD)

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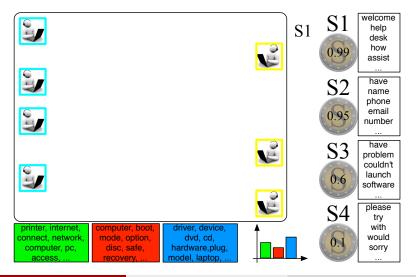
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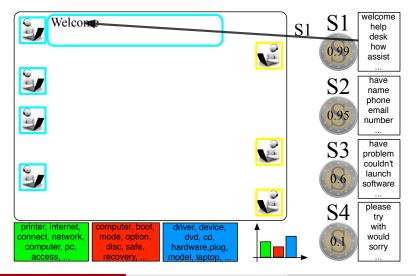
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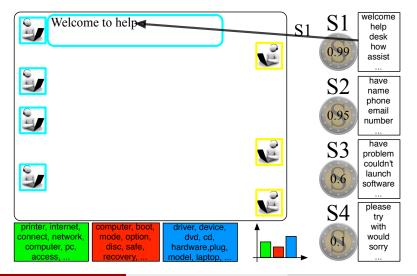
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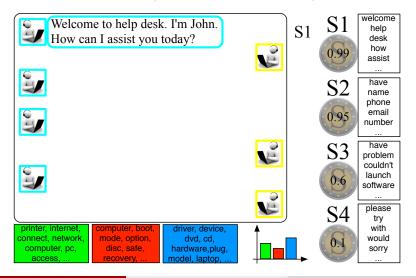
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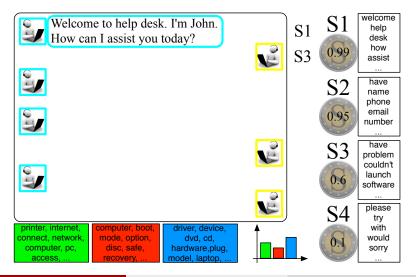
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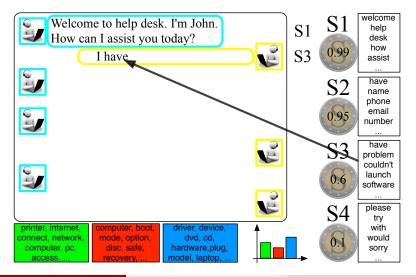
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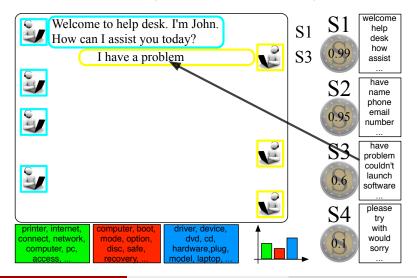
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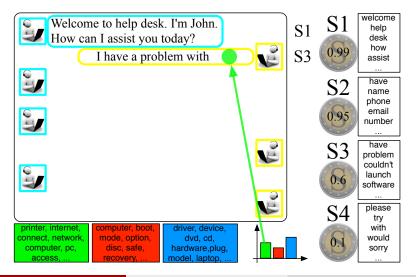
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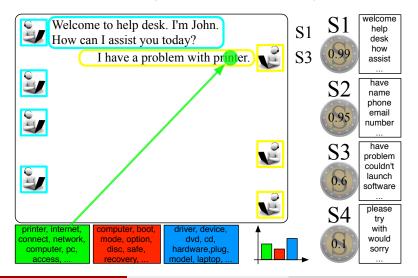
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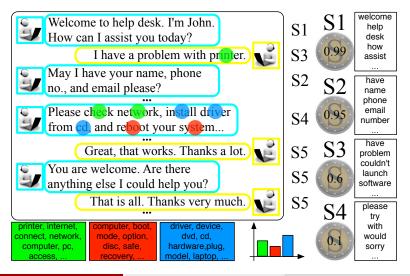
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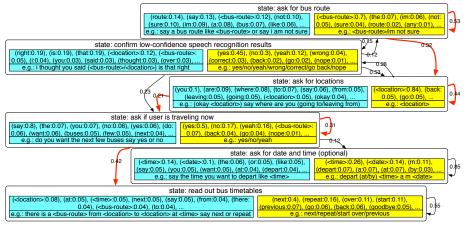
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#### Outline



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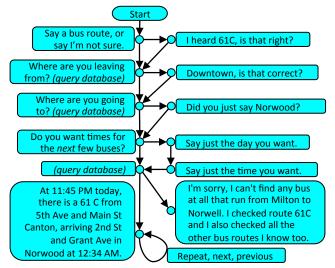
### Inferred Flowchart on BusTime Dataset



# TM-HMM with 10 states and 10 topics (transition cutoff threshold is 0.1)

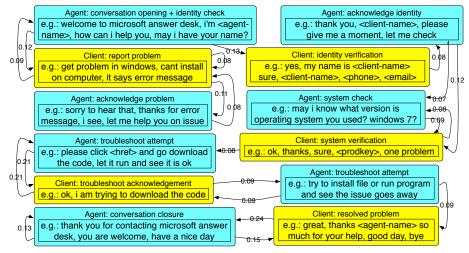
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### Inferred Flowchart on BusTime Dataset



Hand-crafted reference flowchart for BusTime (Williams, 2012)

# Inferred Flowchart on TechSupport Dataset



TM-HMMSS with 20 states and 20 topics (transition cutoff threshold is 0.05)

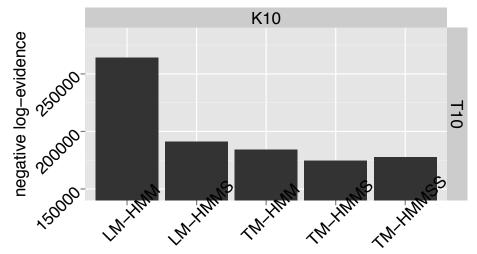
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# **Inferred Topics**

Topic	Top Ranked Words
browser	internet, explor, browser, ie, open, websit, googl, download, click,
0.	chrome,
backup	file, restor, system, comput, back, folder, creat, option, dont, delet,
boot	comput, boot, mode, option, disc, safe, recoveri, repair, back, clean,
	cd, disk,
update	updat, window, servic, instal, pack, run, comput, download, check,
•	restart, inform, system, error, fix,
network	connect, internet, printer, comput, network, pc, print, access, wireless,
	hp, cable, adapt, router, speed,
anti-virus	viru, scan, comput, remov, secur, run, system, anti, essenti, infect,
	defend, softwar, program, protect, antiviru, malwar,
hardware	driver, devic, drive, dvd, cd, hardwar, issu, model, laptop, plug, soft-
110	ware, usb,
windows	window, upgrad, 8, download, 7, instal, bit, vista, pro, system,
office	offic, 2010, word, microsoft, home, excel, version, 2007, student,
	document, trial, 2013,
TM-HMMSS with 20 states and 20 topics	

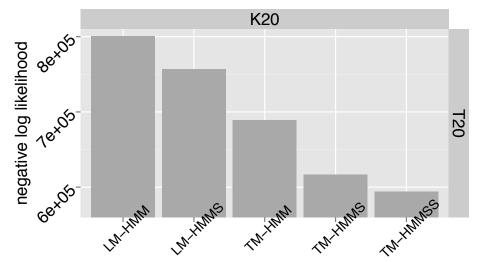
(transition cutoff threshold is 0.05)

#### Log-likelihood Evaluation on BusTime Dataset



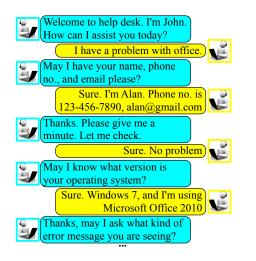
Negative log-likelihood (lower the better) on held-out test set with 10 states and 10 topics

# Log-likelihood Evaluation on TechSupport Dataset

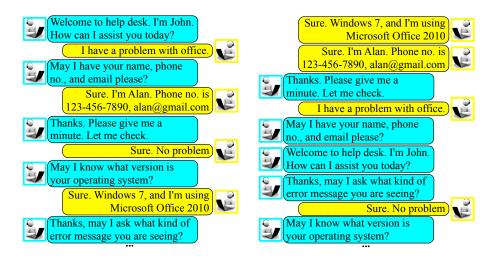


Negative log-likelihood (lower the better) on held-out test set with 20 states and 20 topics

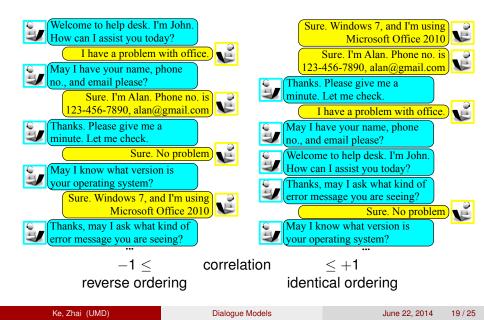
# Ordering Task



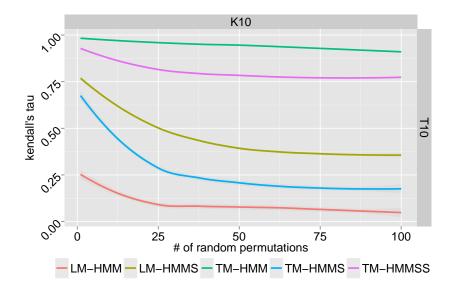
# **Ordering Task**



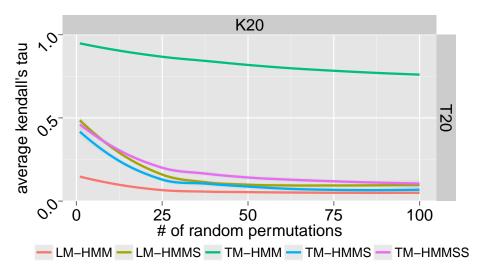
# **Ordering Task**



#### Ordering Task on BusTime Dataset



#### Ordering Task on TechSupport Dataset



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# Conclusion

#### In this talk

- synthesize hidden Markov models and topic models
- propose unsupervised generative models: TM-HMM and TM-HMMSS
- discover two horizons of a dialogue system automatically
  1) a generic flow-chart and 2) a domain of topics

#### What's exciting in Zhai & Williams (2014)...

- Plate notations, generative stories and MCMC inference
- Experiments on BusTime and TechSupport datasets
  - BusTime: transcripts from human-computer interactive system
  - TechSupport: a set of human-computer online chat dialogues
  - Qualitative: meaningful dialogue flows, aligned with reference design
  - Quantitative: log-likelihood and ordering task on held-out test set
- Sensitivity to different parameter settings

# Thank you! Questions?

Special Acknowledgement Alan Ritter, Bill Dolan Jordan Boyd-Graber Kai (Anthony) Lui

- Barzilay, Regina and Lee, Lillian. Catching the drift: Probabilistic content models, with applications to generation and summarization. In Conference of the North American Chapter of the Association for Computational Linguistics, pp. 113–120, 2004.
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- Zhai, Ke and Williams, Jason D. Discovering latent structure in task-oriented dialogues. In Proceedings of the Association for Computational Linguistics, 2014.