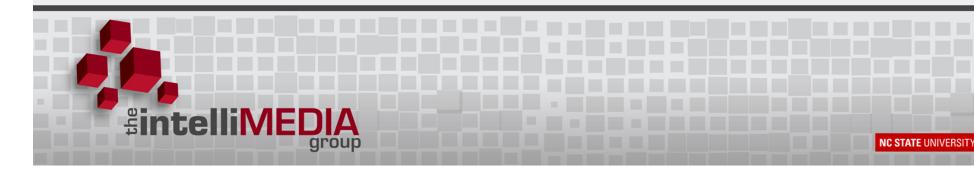
Leveraging Hidden Dialogue State to Select Tutorial Moves

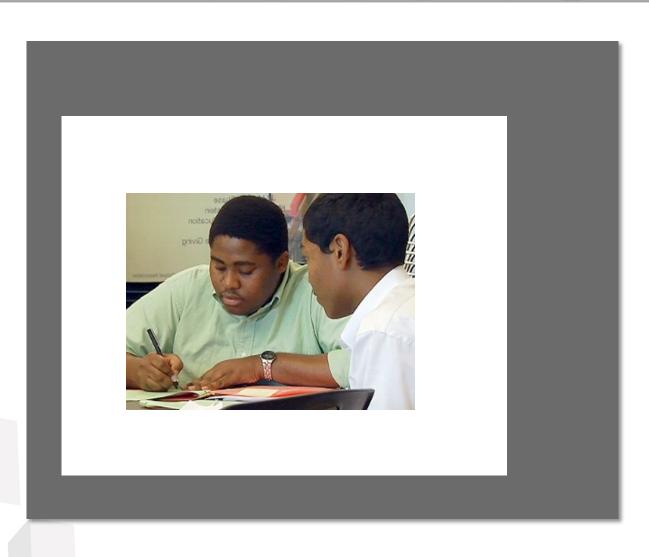
Kristy Elizabeth Boyer, Robert Phillips, Eun Young Ha, Michael Wallis, Mladen Vouk, & James Lester

Department of Computer Science
North Carolina State University



Introduction





Tutorial Dialogue Systems

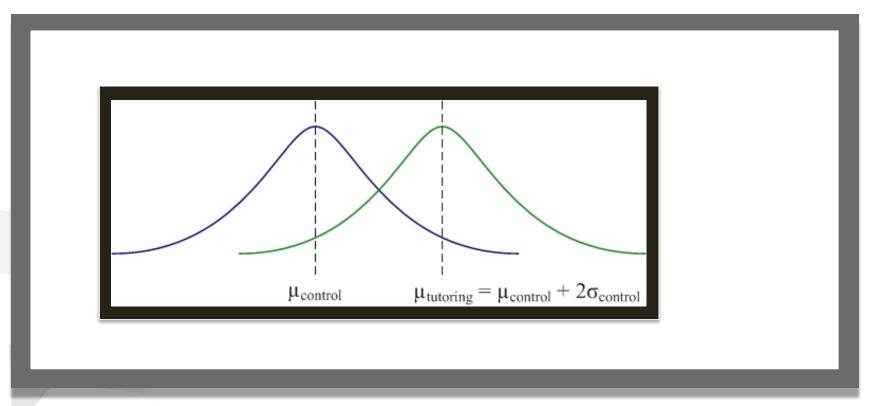


- Geometry (Aleven et al., 2004)
- Circulatory System (Evens & Michael, 2006)
- Physics (Graesser et al., 2005; Jordan et al., 2006)
- Data Structures (Fossati et al., 2008)
- And many more...

Challenge



Never achieved effectiveness equal to the most effective human tutors



Reasons?



 Need more sophisticated natural language dialogue (Graesser et al., 1994)

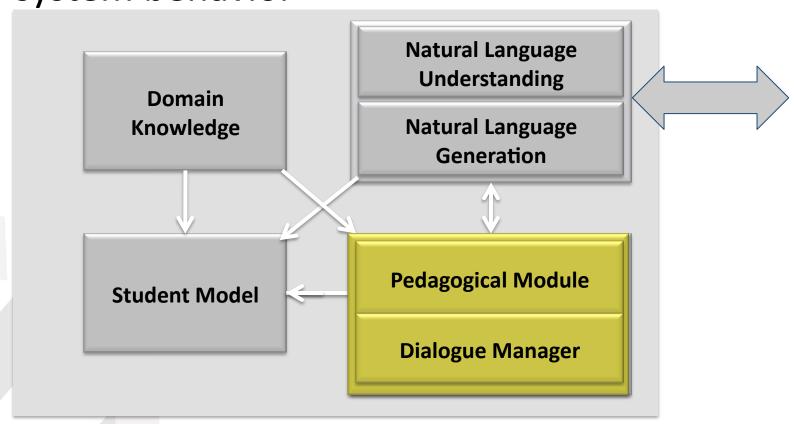
AND/OR

 Need to facilitate mastery learning (Van Lehn, 2008)

Possible Solution in Either Case



Data-driven authoring of tutorial dialogue system behavior



Hypothesized Benefits



Data-driven tutorial dialogue systems:

- Reduced development time
- Increased number of hours of instruction for students
- Flexible dialogue strategies
- Reflect approaches of effective human tutors

Goal of This Work



Learn

a data-driven tutorial dialogue strategy

as evidenced by

predicting human tutors' dialogue acts within a corpus

Approach



Perform

sequence prediction

with models that leverage

hidden dialogue state

Related Work



Data-driven dialogue policy creation through:

- Reinforcement learning
- Direct corpus-based extraction

Reinforcement Learning



- Markov assumption (Levin et al. 2000)
- Reinforcement learning (Frampton & Lemon, 2009)
- Challenges include sparse data and large state spaces (Ai et al., 2007; Tetreault & Litman, 2008; Henderson et al., 2008; Heeman, 2007; Young et al., 2009)
- Comparing specific tutorial dialogue tactic choices (Chi et al., 2008)

Corpus-Based Extraction



Assumes that a good dialogue policy is realized in successful human-human dialogues

- Financial domain (Hardy et al., 2006)
- Catalogue ordering (Bangalore et al., 2008)
- Maptask conversational game (Poesio & Mikheev, 1998)

Strategies in Tutorial Dialogue



Data-driven exploration

- CIRCSIM-TUTOR (Evens & Michael, 2006)
- ITSPOKE (Forbes-Riley, Rotaru, Litman, & Tetreault, 2007; Forbes-Riley & Litman, 2009)
- KSC-PAL (Kersey, Di Eugenio, Jordan, & Katz, 2009)

Hidden Dialogue State



- Learned by hidden Markov models (Boyer et al., 2009a)
- Qualitative resemblance to tutorial dialogue strategies (Boyer et al., 2009b)
- Hypothesized to boost prediction of tutor moves

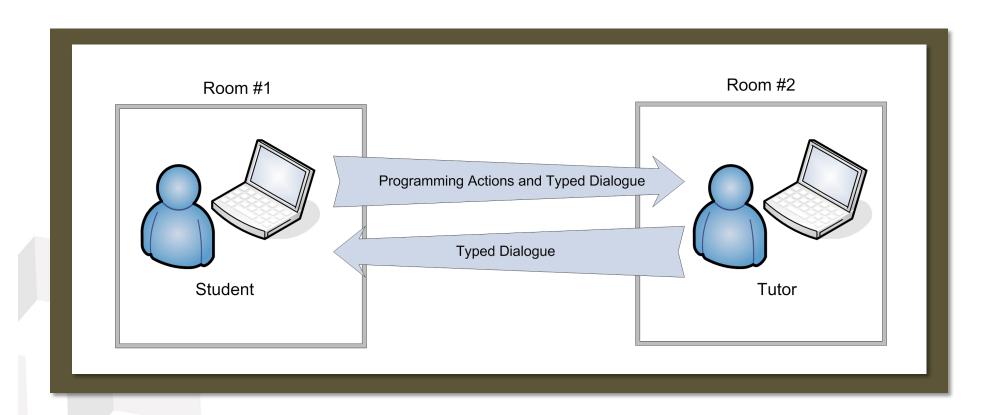
Corpus Study



- 48 Human-human tutoring dialogues
- Domain: introductory computer programming
- Task-oriented, separate, parallel task event stream

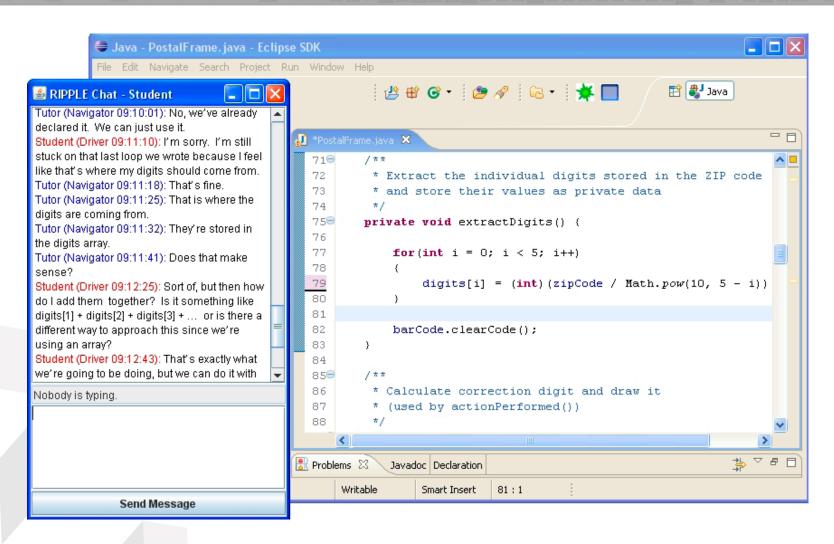
Corpus Collection





Remote Tutoring Environment





Corpus



- 1,468 student utterances and 3,338 tutor utterances
- 3,793 semantic student task actions
- Significant learning gain from pretest to posttest (7% average, p<0.0001)
- Annotated with dialogue act tags and task/ subtask structure

Dialogue Act Annotation



- Scheme inspired by tags for
 - Conversational speech (Stolcke et al., 2000)
 - Task-oriented dialogue (Core & Allen, 1997)
 - Tutoring (Litman & Forbes-Riley, 2006)
- Inter-rater reliability on 10% of corpus was κ =0.80.

Dialogue Act Tags



Description
Request for feedback on task or conceptual utterance.
Asides not relevant to the tutoring task.
Acknowledgement/thanks.
Negative assessment with explanation.
Lukewarm assessment of task action or conceptual utterance.
Negative assessment with explanation.
Negative assessment of task action or conceptual utterance.
Positive assessment with explanation.
Positive assessment of task action or conceptual utterance.
Task or conceptual question.
Task or conceptual assertion.

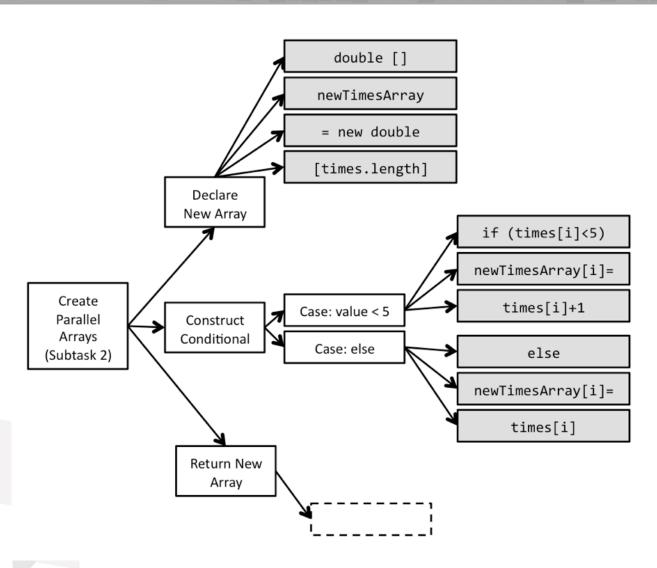
Task Annotation



- Corpus includes 97,509 keystroke-level student task events
- Manually aggregated into 3,793 subtask actions
- Annotated with task/subtask structure
- Marked for correctness

Subtask Structure Annotation





Task Annotation Results



- Subtask annotation (66 leaves)
 - Inter-rater reliability study on 20% of corpus
 - Simple kappa = 0.58
 - Weighted kappa = 0.86
- Correctness annotation (4 tags)
 - Correct, Buggy, Incomplete, and Dispreferred
 - Simple kappa = 0.80

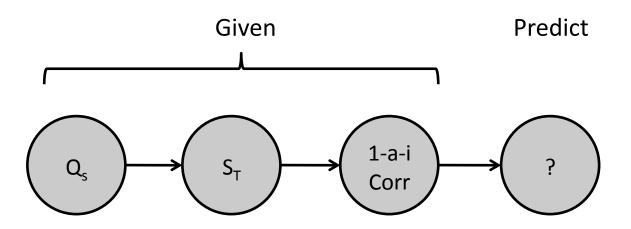
Tagged Excerpt



Time Stamp		Dialogue Stream	Task Stream	
		so do i have to manipulate the array this		
2008-04-11 18:23:45	Student:	time? [Q]		
2008-04-11 18:23:53	Tutor:	this time, we need to do two things [S]		
		first, we need to create a new array to hold		
2008-04-11 18:24:02	Tutor:	the changed values [S]		
2008-04-11 18:24:28			i]
2008-04-11 18:24:28			n	1-a-l
2008-04-11 18:24:28			t	Buggy
2008-04-11 18:24:28			\sp	
2008-04-11 18:24:35			\del	_
2008-04-11 18:24:36			\sp	
2008-04-11 18:24:36			d	1
2008-04-11 18:24:36			0	
2008-04-11 18:24:36			u	
2008-04-11 18:24:36			b	1-a-i
2008-04-11 18:24:37			1	Correc
2008-04-11 18:24:37			е	
2008-04-11 18:24:37			\sp	
2008-04-11 18:24:39				

Sequence Modeling Task





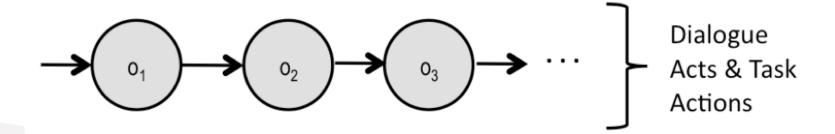
Three models



- First-order Markov models (MMs)
- Hidden Markov models (HMMs)
- Hierarchical Hidden Markov models (HHMMs)

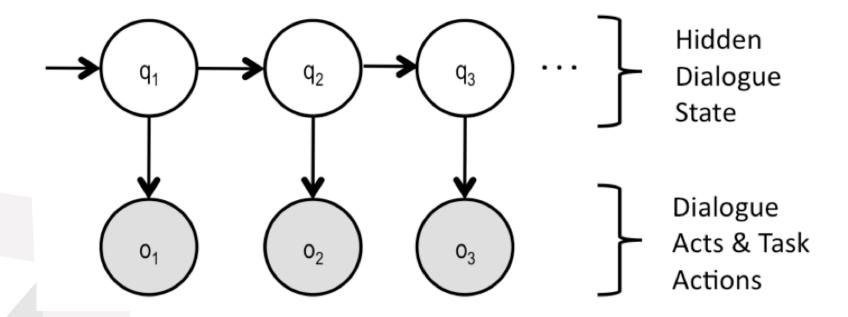
Markov model





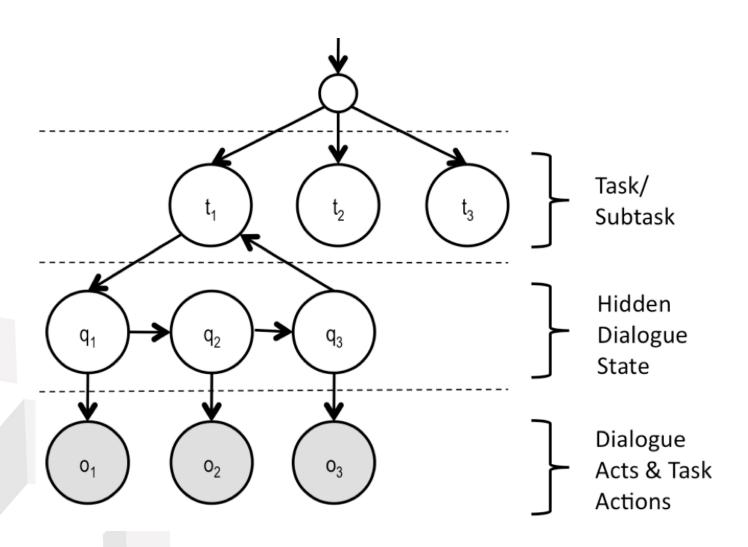
HMM





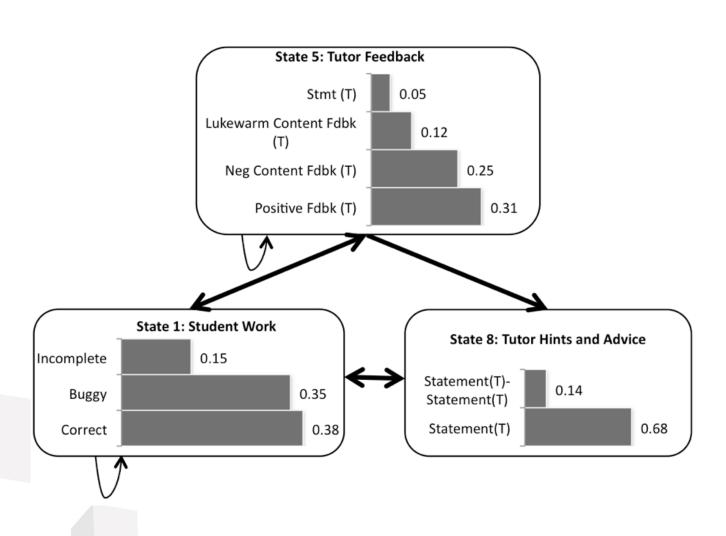
HHMM





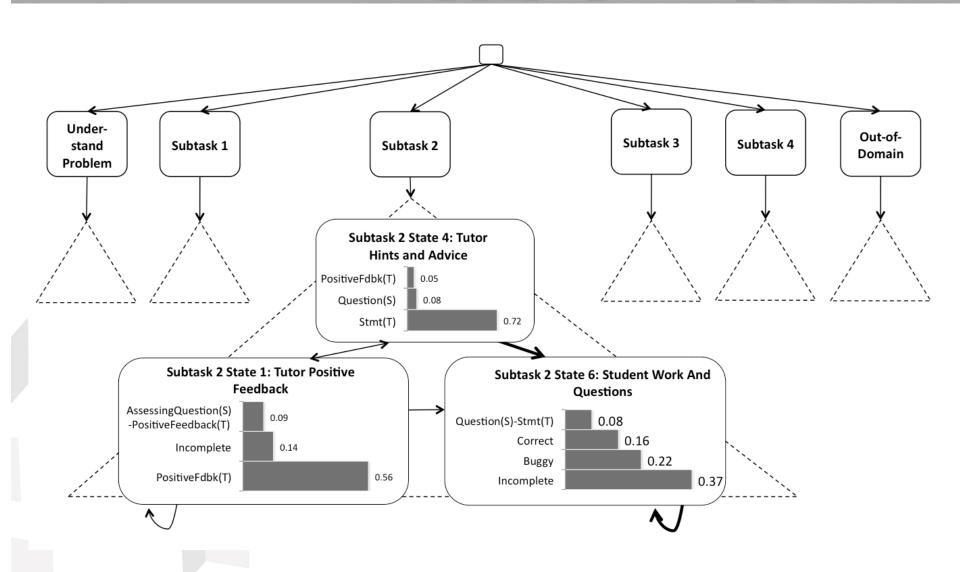
Portion of Learned HMM





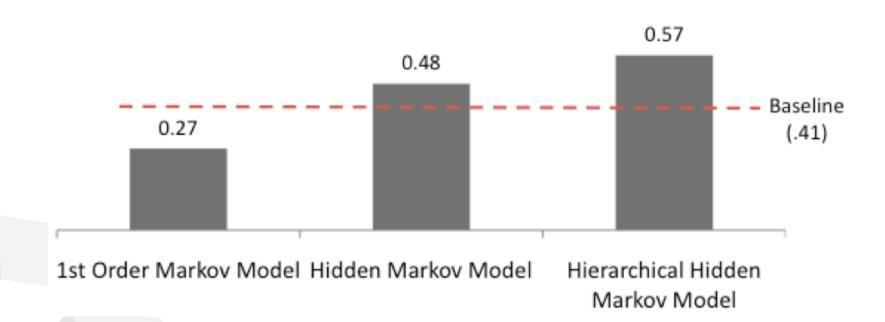
Portion of Learned HHMM





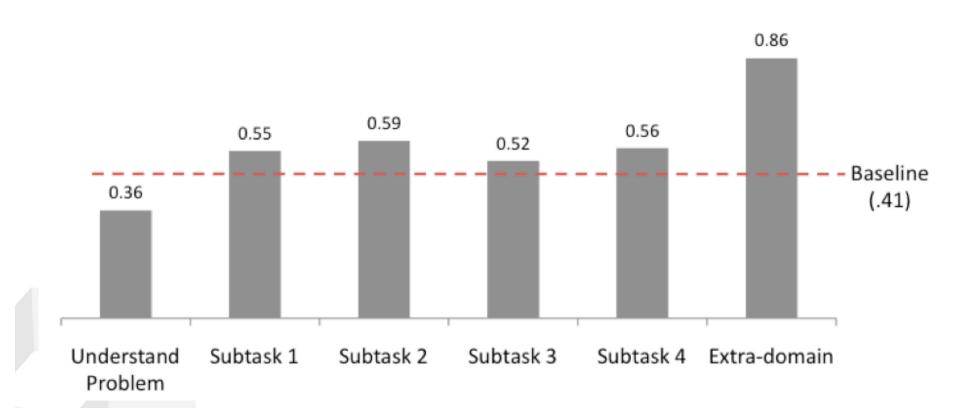
Prediction Results by Model





HHMM Results by Subtask





Results in Context



- Catalogue ordering domain
 - Flat model 55% accuracy (Bangalore et al., 2008)
 - Hierarchical model 35.6% accuracy (Bangalore & Stent, 2009)
- Our domain, tutoring for introductory computer programming
 - Flat model 48% accuracy
 - Hierarchical model 57% accuracy

Conclusions



- Hidden dialogue state useful; HMMs outperform MMs
- Task/subtask structure useful; HHMMs outperform HMMs
- Understand Problem phase very challenging to model
- Extra-domain conversation most straightforward to predict

Future Work



- Maintain multiple hypotheses for tutorial moves
- Leverage learner characteristics
- Develop fully unsupervised dialogue models
- Ultimate goal: create highly effective datadriven tutorial dialogue systems

A Shameless Plug



The Third Workshop on Question Generation

QG 2010



www.questiongeneration.org/QG2010

June 18, 2010, Pittsburgh

Acknowledgments















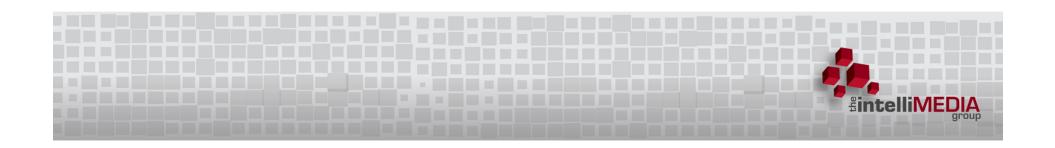








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