

A Network-based End-to-End Trainable Task-oriented Dialogue System

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- Intro
- Neural Dialogue System
- Wizard-of-Oz Data Collection
- Experiments
- Conclusion & Discussion

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The two paradigms

M: Hello, welcome to the Cambridge dialogue system, What kind of food would you like?

H: Yeah I want to find a restaurant that serves European food.

M: Hotel du Vin and Bistro is a nice place. It serves European food.

H: Uh what is the address and phone number?

M: Hotel du Vin and Bistro is a nice place. Their address is XXX and their phone is YYY.

H: Thank you goodbye.

M: Thank you for using this system.

[Su et al, 2016]

A: How old are you?

B: I'm 16. Why are you asking?

A: I thought you were 12.

B: What made you think so?

A: I don't know what you are talking about.

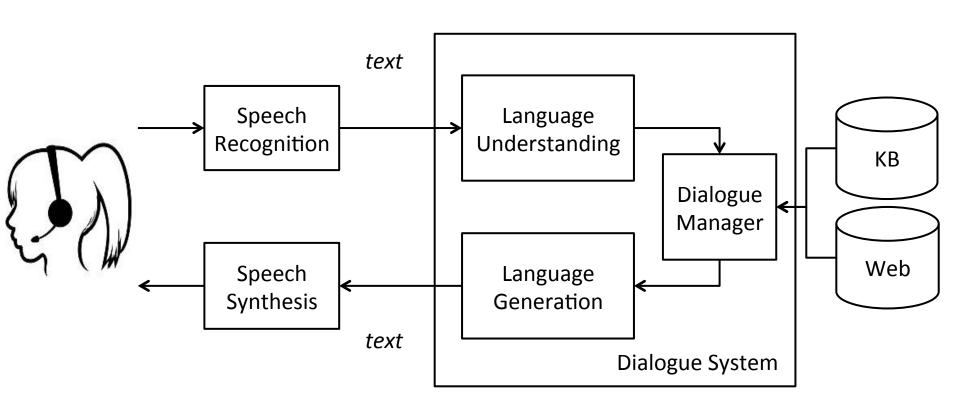
B: You don't know what you are saying.

The two paradigms

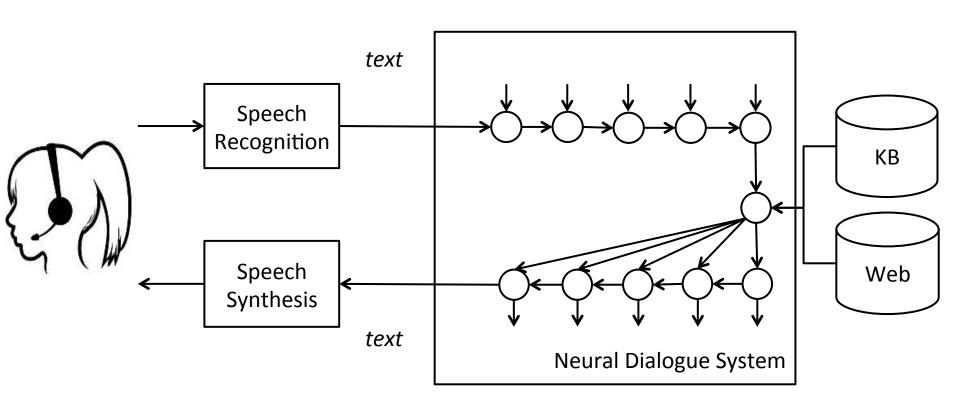
- Task-Oriented Dialogue Systems
 - Goal-oriented
 - Require precise understanding, hard to collect data.
 - Modular, highly handcrafted, restricted ability, but meaningful/ useful systems.
- Chat-based Conversational Agents
 - Chit-chat (non-goal).
 - Vast amount of data (but probably not helpful).
 - End-to-end, highly data-driven, but meaningless/inappropriate responses, unreliable systems.
- Can we train a useful (complete tasks) dialogue system directly from data?
- How can we collect the data to train this model?

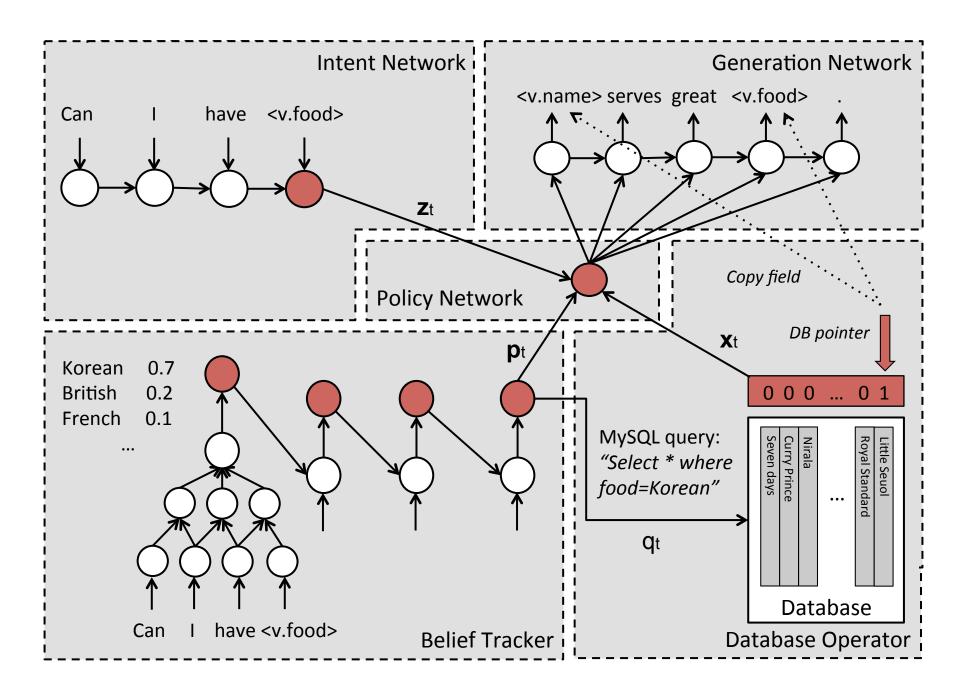
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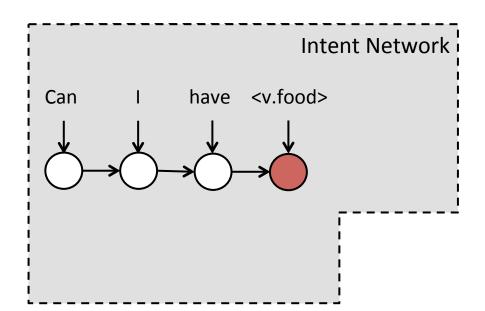
Traditional Dialogue Systems

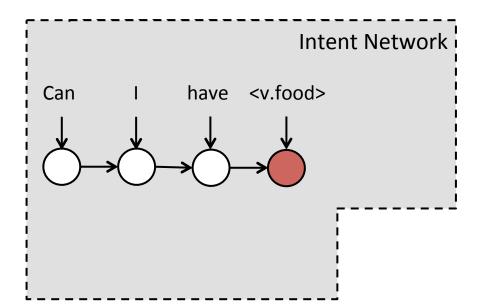


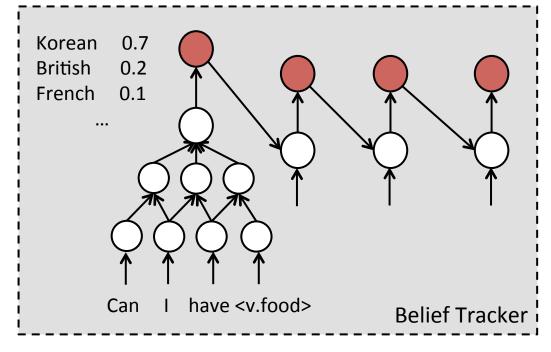
Neural Dialogue Systems



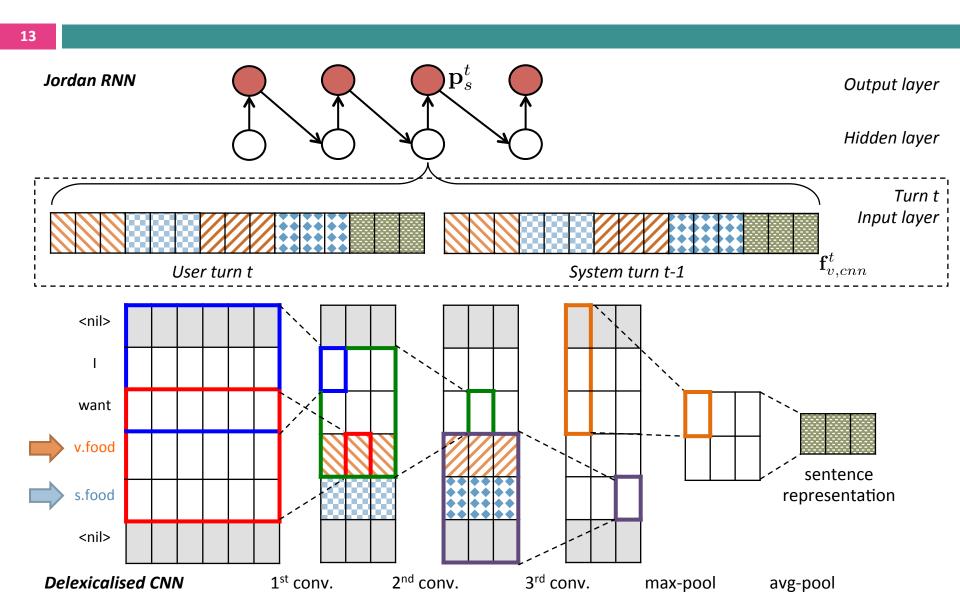


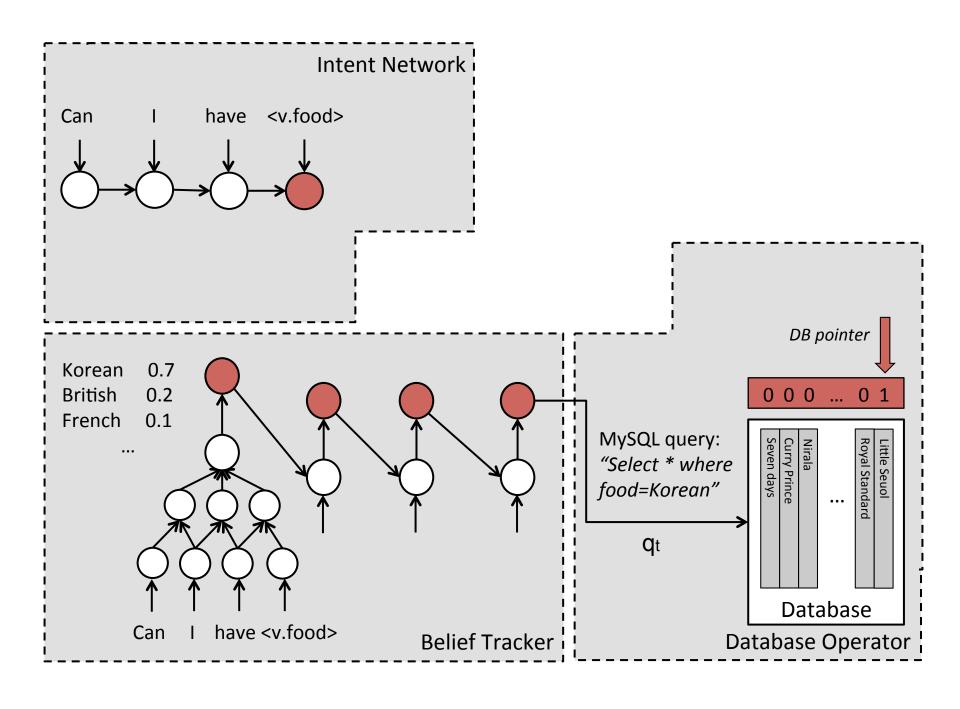


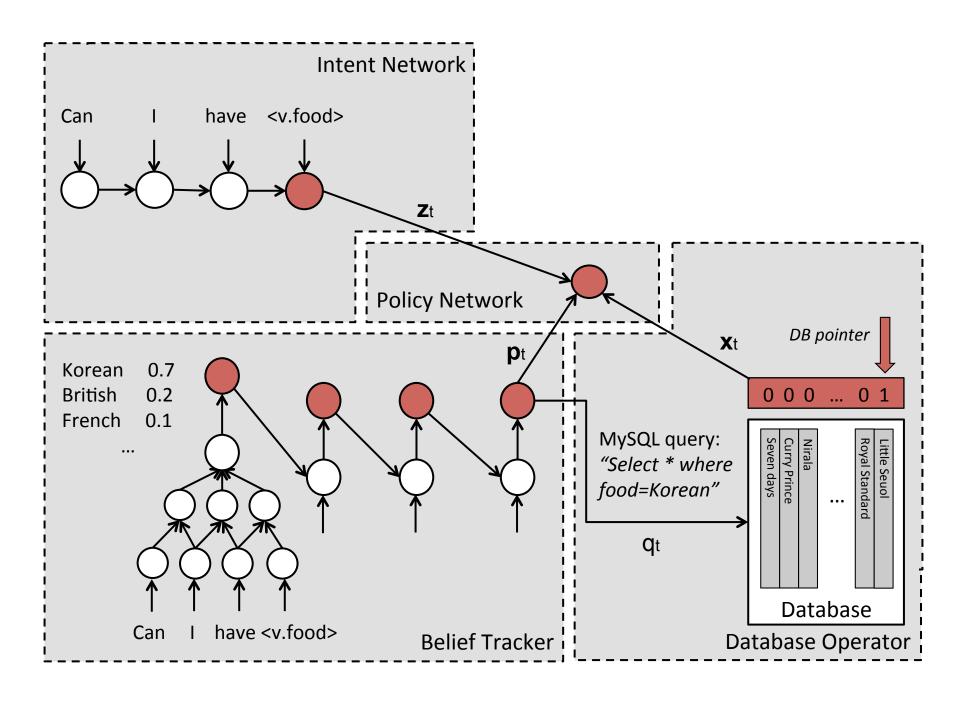


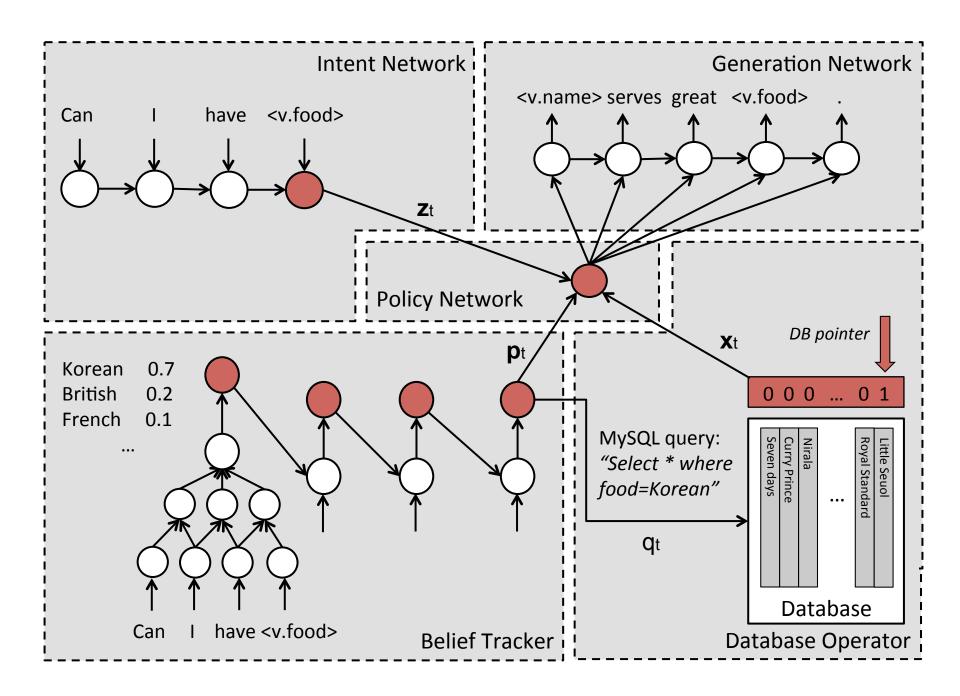


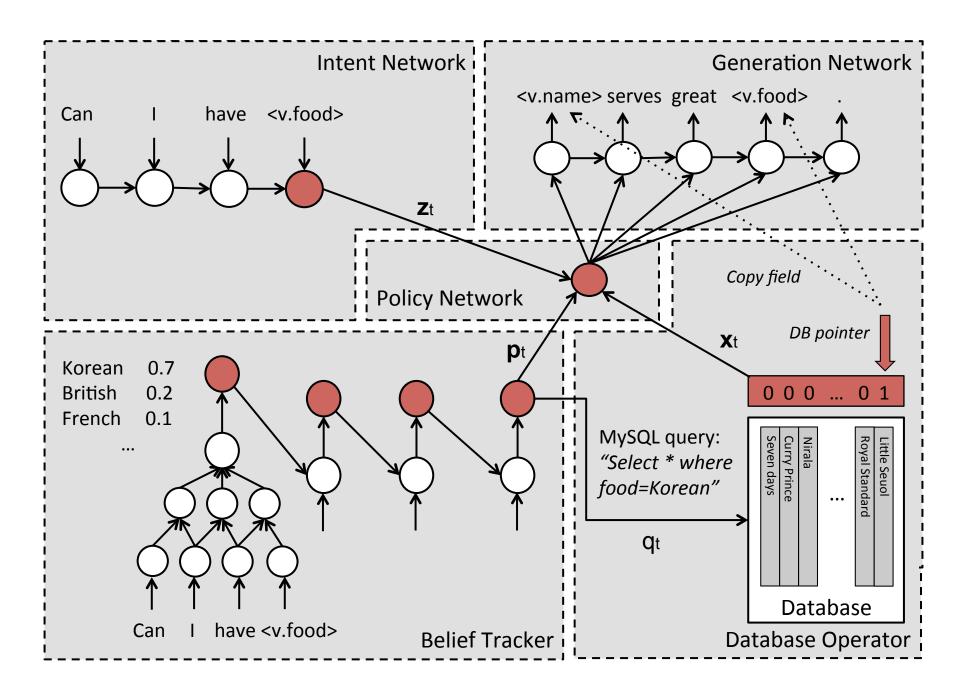
Jordan RNN-CNN belief trackers











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Task: Find a restaurant,

cheap, Korean, North Ask phone number



Hi, I want a cheap Korean restaurant.

What **area** are you looking for?

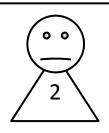


What user wants?			
Food	Korean		
Price	Cheap		
Area	N/A		

Search Table					
Little Seoul	***				
Best Korea					

Task:

Find a restaurant, cheap, Korean, North Ask phone number



Hi, I want a cheap Korean restaurant.

What **area** are you looking for?

Somewhere in the north.

Little Seoul is nice one in the north.



What user wants?				
Food	Korean			
Price	Cheap			
Area	North			

Search Table				
Little Seoul				

Task:

Find a restaurant. cheap, Korean, North Ask phone number







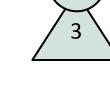
Somewhere in the **north**.

Little Seoul is nice one in the north.

What is the **phone number**?

Its phone number is **01223456789**.





What user wants?				
Food	Korean			
Price	Cheap			
Area	North			

Search Table					
Little Seoul					

Task:

Find a restaurant. cheap, Korean, North Ask phone number



Hi, I want a cheap Korean restaurant.

What **area** are you looking for?

Somewhere in the **north**.

Little Seoul is nice one in the north.

What is the **phone number**?

Its phone number is **01223456789**.

Thank you very much, good bye.

Thank you for using the system.



What user wants?

Food	Korean
Price	Cheap
Area	North

Search Table

Little Seoul

Hi, I want a cheap Korean restaurant.

What **area** are you looking for?

Somewhere in the north.

Little Seoul is nice one in the north.

What is the **phone number**?

Its phone number is **01223456789**.

Thank you very much, good bye.

Thank you for using the system.

What user wants?				
Food	Korean			
Price	Cheap			
Area	North			

- Online parallel version of WOZ on MTurk
 - Randomly hire a worker to be user/wizard.
 - Task: Enter an appropriate response for one turn.
 - Repeat the process until all dialogues are finished.
- Example user page

Task 02004: You are looking for and it should serve gastropub food. You don't care about the price range. You want to know the address.

Info Desk: Hello , welcome to the Cambridge restaurant system . You can ask for restaurants by area, price range or food type . How may I help you ? Customer: i want a gastropub food Info Desk: There are 4 restaurants serving gastropub food, what price range do you want?				
Next turnl				
Customer : (Your response)				
I dont care about the price range, just give me the address please.	Submit the HIT			
_				

Example wizard page

nfo Desk : Hello , welcome to the Cambridge restaurant system . You can ask for restaurants by area, price range or food type . How may I help you ? Customer : I want a gastropub food in Desk : There are 4 restaurants serving gastropub food, what price range do you want ? Customer : I dont care	
Next turn!	
Please modify the following answers based on the latest customer response: • What to sue suer want? What is the food type the user wants? What is the area the user wants? What is the price range the user wants? dont care • What does user ask? Is the user asking for food type of an offered venue? Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user asking for postcode of an offered venue? No \$ Is the user mentioning any restaurant names? No \$ Is the user mentioning any restaurant names?	
Info Desk: (Your response) I would recommend backstreet bistro, a great gastropub restaurant in the centre. do you want their phone number? Submit the HIT	

Name	Food	Area	Price Range	Phone	Address	Postcode
backstreet bistro	gastropub	centre	expensive	01223 306306	2 Sturton Street City Centre	C.B 1, 2 Q.A
royal standard	gastropub	east	expensive	01223 247877	290 Mill Road City Centre	C.B 1, 3 N.L
the cow pizza kitchen and bar	gastropub	centre	moderate	01223 308871	Corn Exchange Street	C.B 2, 3 Q.F
the slug and lettuce	gastropub	centre	expensive		34 - 35 Green Street	C.B 2, 3 J.U
nil	gastropub	nil	nil	nil	nil	nil
Showing 1 to 4 of 4 entries (filtered from 1)	(O total entries)					Previous 4 Next

Data Statistics

Ontology:

- Cambridge restaurant domain, 99 venues.
- 3 informable slots: area, price range, food type
- 3 requestable slots: address, phone, postcode

Dataset

- 676 dialogues, ~2750 turns
- 3000 HITS, takes 3 days, costs ~400 USD
- Data cleaning takes 2-3 days for one person

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Experiments

- Experimental details
 - Train/valid/test: 3/1/1
 - SGD, l2 regularisation, early stopping, gradient clip=1
 - Hidden size = 50, Vocab size: ~500
- Two stage training:
 - Training trackers with label cross entropy
 - Training other parts with response cross entropy
- Decoding
 - Beam search w/ beam width 10
 - Decode with average word likelihood

Response Generation Task

Model	Match (%)	Success (%)	BLEU
Seq2Seq [Sutskever et al, 2014]	-	-	0.1718
HRED [Serban et al, 2015]	-	-	0.1861
Our model w/o req. trackers	89.70	30.60	0.1799
Our full model	86.34	75.16	0.2313
Our full model + attention	90.88	80.02	0.2388

Human evaluation

Quality assessment

Metric	N2N
Success	98%
Comprehension Naturalness	4.11 4.05
# of dialogues:	245

System Comparison

Metric	N2N	Modular	Tie
Subj. Success	96.95%	95.12%	_
Avg. # of Turn	3.95	4.54	-
Comparisons(%)			
Naturalness	46.95^{*}	25.61	27.44
Comprehension	45.12^{*}	21.95	32.93
Preference	50.00^{*}	24.39	25.61
Performance	43.90*	25.61	30.49

^{*} p < 0.005, # of comparisons: 164

Human

Example dialogues

: Hi, I'm hungry for some Indonesian food

Human : Restaurant in any area that serves Vietnamese food : thanh binh serves vietnamese food type , is there anything else i can help you with? : What is the phone number and post code : the number is 01223 362456 and the area code is c.b 3, 0 a.f . is there anything else i can do for you? : Good bye : thank you for using the cambridge restaurant system .

Machine : i am sorry, there are no indonesian restaurants. would you like a different food?

Human : Oh no. How about Indian?

Machine : there are 9 restaurants serving indian types of food. do you have a preference on side of town?

Human : East side please

Machine : sitar tandoori is a indian restaurant in the east side of town would you like their information?

Human : Yes, the address and phone number please

Machine : sitar tandoori is located at 43 high street cherry hinton cherry hinton and their phone number is 01223 249955 .

Human : Okay, thank you

Machine : thank you, goodbye.

Example dialogues

Human : hello

Machine : what cuisine would you like ?

Human : afghan

Machine : i am sorry , there are no afghan restaurants . would you like something else ?

Human : korean

Machine : little seoul serves korean food, it is located at 108 regent street city centre.

Human : is it expensive?

Machine : little seoul serves korean food and is in the expensive price range.

Human : what area is that

Machine : little seoul is located in the centre area of town . would you like their number ?

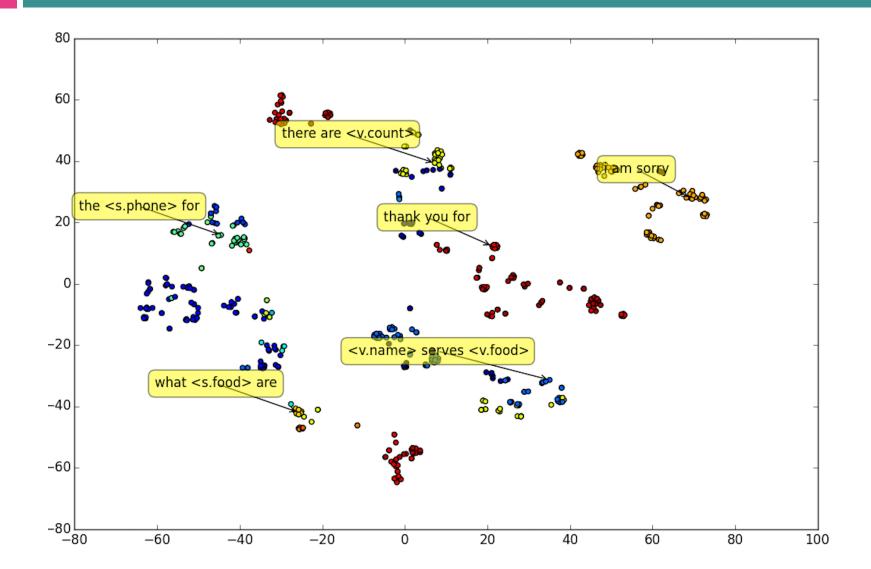
Human : yes

Machine : the telephone number is 01223 308681 . is there anything else i can help you with today ?

Human : no goodbye

Machine : thank you for using the cambridge restaurant system, goodbye.

Visualising action embedding



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Conclusion & Discussion

- An end-to-end trainable task-oriented dialogue system architecture is proposed.
- A complementary WOZ data collection is also proposed (no latency, parallel, cheap).
- Results show that it can learn from human-human conversations and help users to complete tasks.

• Explicit language grounding is crucial, but what is the best way to represent semantics?

The paper

Tsung-Hsien Wen, David Vandyke, Nikola Mrksic, Milica Gasic, Lina M.R. Barahona, Pei-Hao Su, Stefan Ultes, and Steve Young. A Network-based End-to-End Trainable Taskoriented Dialogue System. arXiv preprint: 1604.04562 2016.

References

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- Word-Based Dialog State Tracking with Recurrent Neural Networks, M. Henderson, B. Thomson and S. Young. SigDial 2014.
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Thank you! Questions?

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