

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

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# Overview

- Introduction
- Model
- Wizard-of-Oz Data Collection
- Empirical Experiments
- Conclusions

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# Introduction

- Treat as a POMDP and use RL to train dialogue policies
- Build end-to-end trainable, non-task-oriented conversational systems using seq2seq model
- The authors propose a model by balancing the strengths and the weaknesses of these two



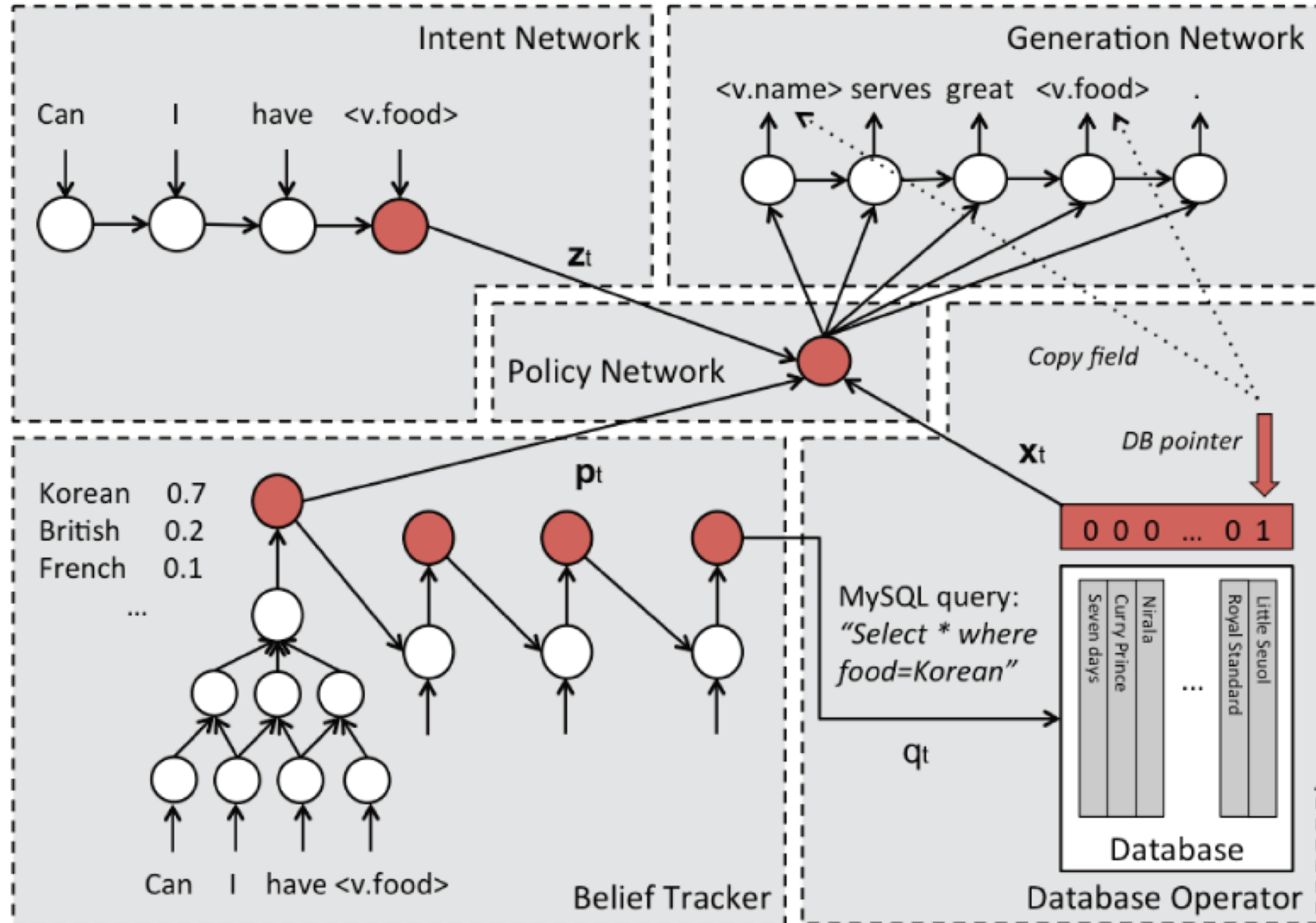
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- **Model**
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# Model

- Intent Network
- Belief Trackers
- Database Operator
- Policy Network
- Generation Network

# Model



# Intent Network

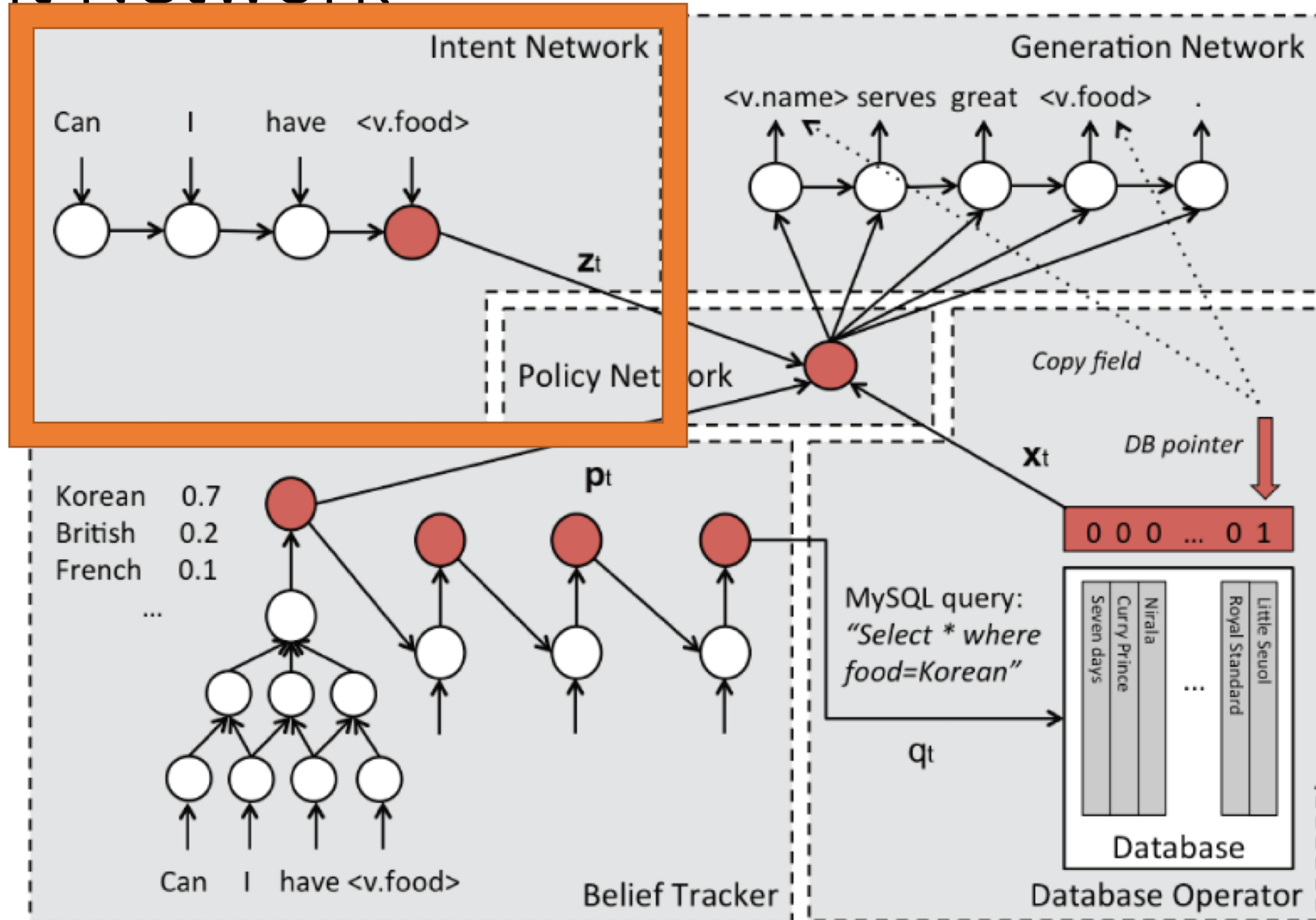
- Encoder in the sequence-to-sequence framework
- Typically, an LSTM network is used

$$\mathbf{z}_t = \mathbf{z}_t^N = \text{LSTM}(w_0^t, w_1^t, \dots w_N^t)$$

- Alternatively, a CNN can be used

$$\mathbf{z}_t = \text{CNN}(w_0^t, w_1^t, \dots w_N^t)$$

# Intent Network



# Belief Trackers

- Core component of the model
- Every slot has its belief tracker
- Each tracker is a Jordan type RNN with a CNN feature extractor

# Belief Trackers

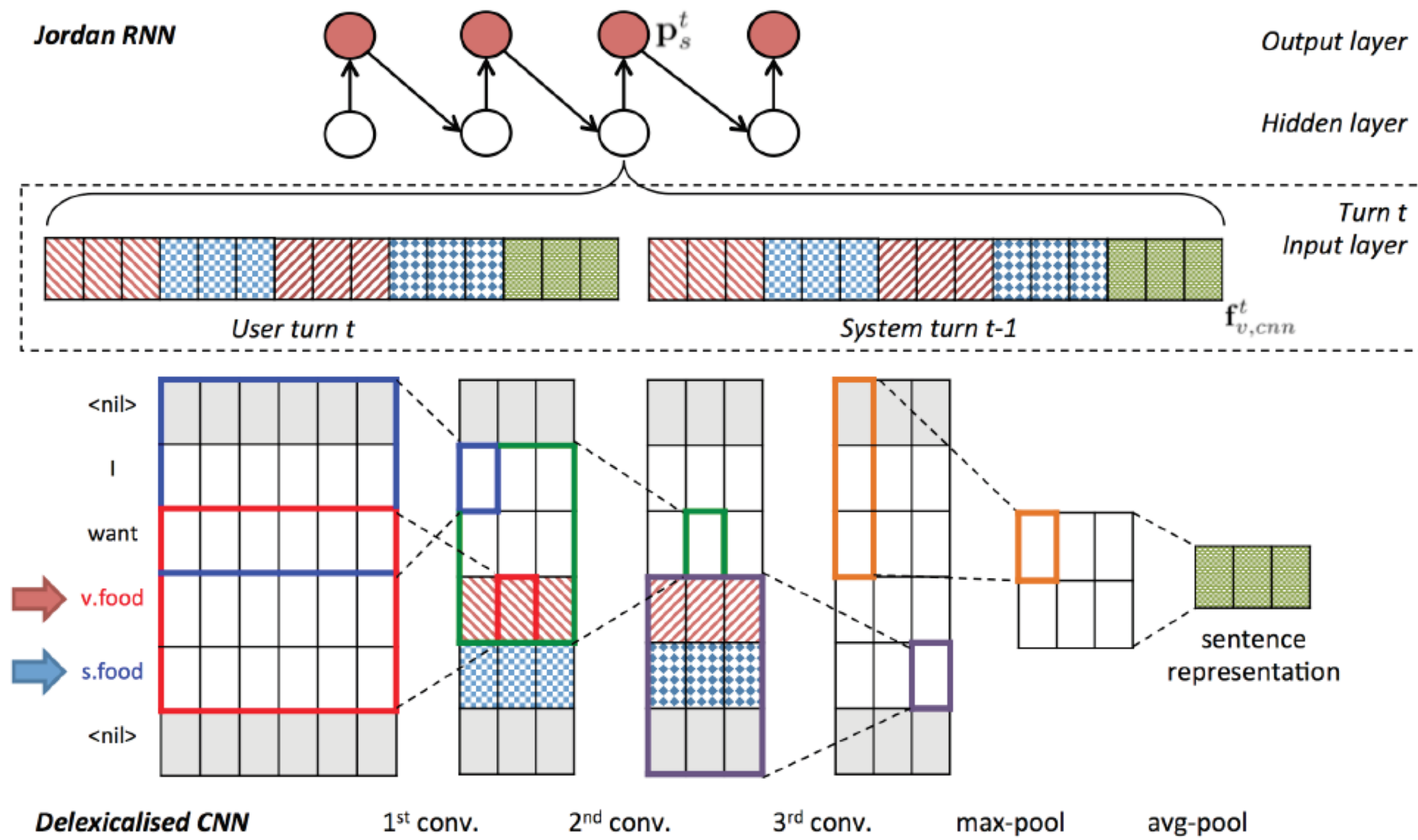
$$\mathbf{f}_{v,cnn}^t = \text{CNN}_{s,v}^{(u)}(u_t) \oplus \text{CNN}_{s,v}^{(m)}(m_{t-1})$$

$$\mathbf{f}_v^t = \mathbf{f}_{v,cnn}^t \oplus p_v^{t-1} \oplus p_{\emptyset}^{t-1}$$

$$g_v^t = \mathbf{w}_s \cdot \text{sigmoid}(\mathbf{W}_s \mathbf{f}_v^t + \mathbf{b}_s) + b'_s$$

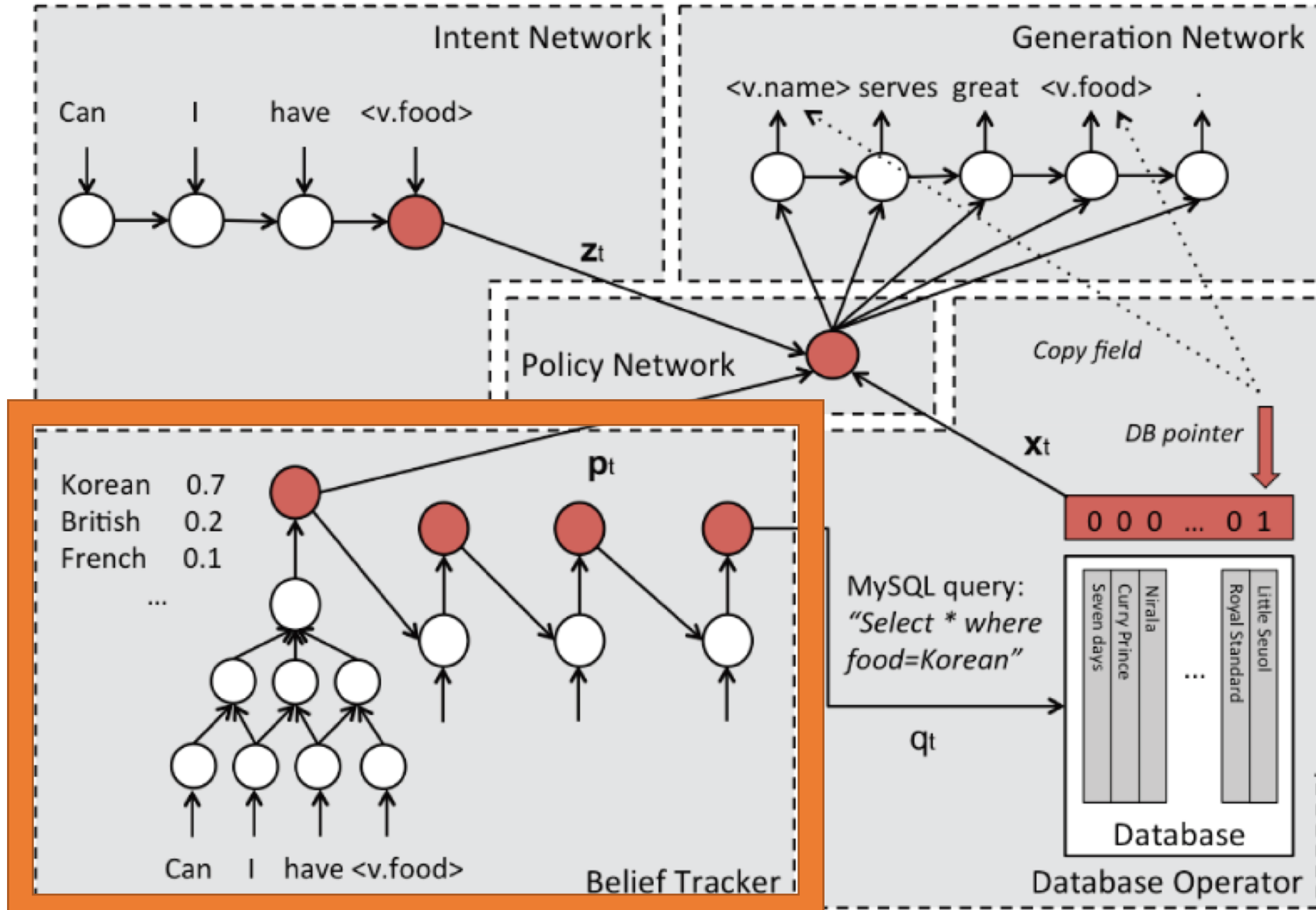
$$p_v^t = \frac{\exp(g_v^t)}{\exp(g_{\emptyset,s}) + \sum_{v' \in V_s} \exp(g_{v'}^t)}$$

# Belief Trackers





# Belief Trackers



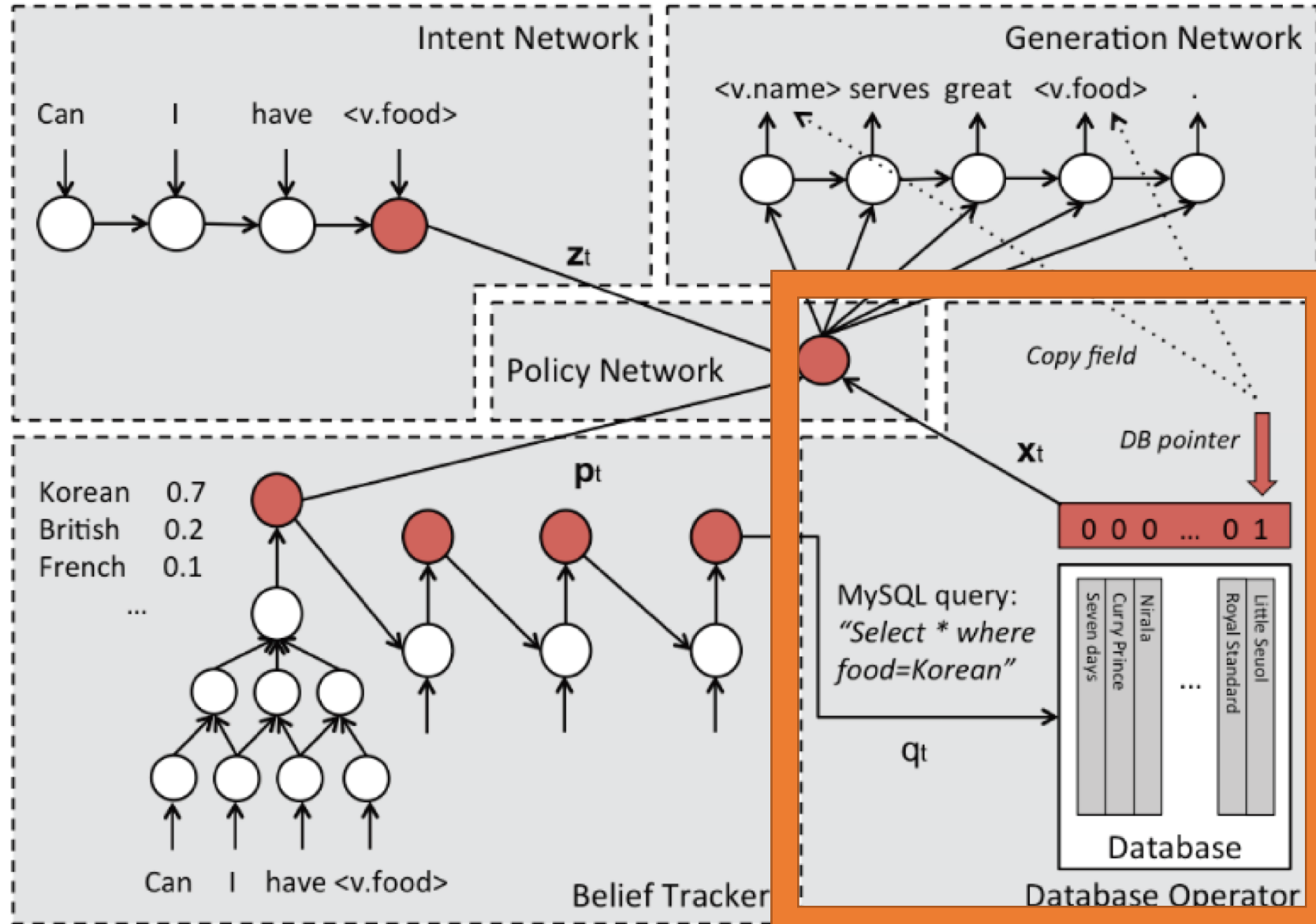
# Database Operator

- The DB query  $q_t$  is formed by

$$q_t = \bigcup_{s' \in S_I} \{\operatorname{argmax}_v \mathbf{p}_{s'}^t\}$$

- Then query is applied to the DB to create a binary truth value vector  $x_t$  over DB entities
- The entity referenced by the entity pointer is used to form the final system response

# Database Operator

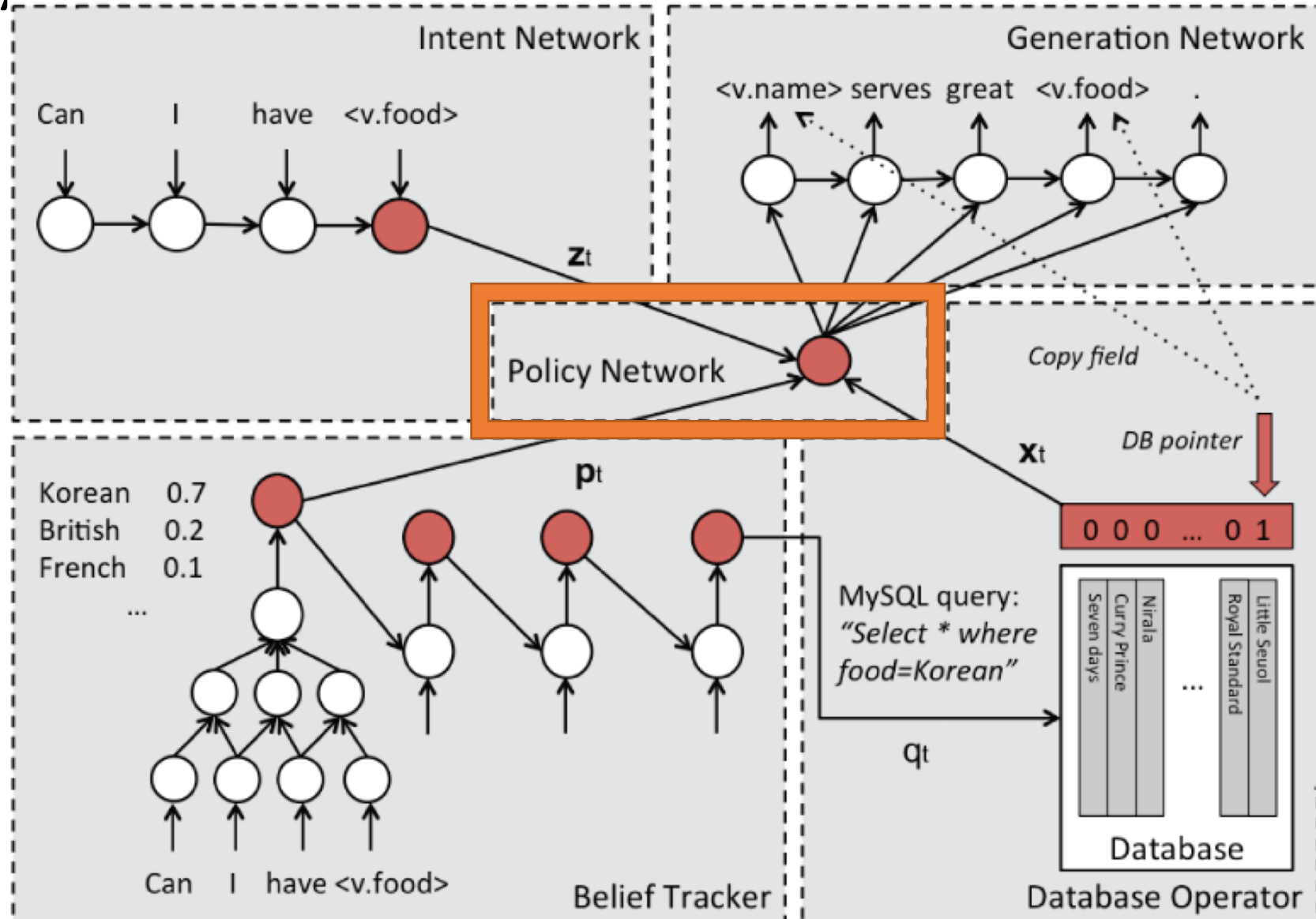


# Policy Network

- Can be viewed as the glue binding other modules together

$$\mathbf{o}_t = \tanh(\mathbf{W}_{zo}\mathbf{z}_t + \mathbf{W}_{po}\hat{\mathbf{p}}_t + \mathbf{W}_{xo}\hat{\mathbf{x}}_t)$$

# Policy Network



# Generation Network

$$P(w_{j+1}^t | w_j^t, \mathbf{h}_{j-1}^t, \mathbf{o}_t) = \text{LSTM}_j(w_j^t, \mathbf{h}_{j-1}^t, \mathbf{o}_t)$$

- Once the output token sequence has been generated, the generic tokens are replaced by their actual values

# Generation Network

- Attentive Generation Network

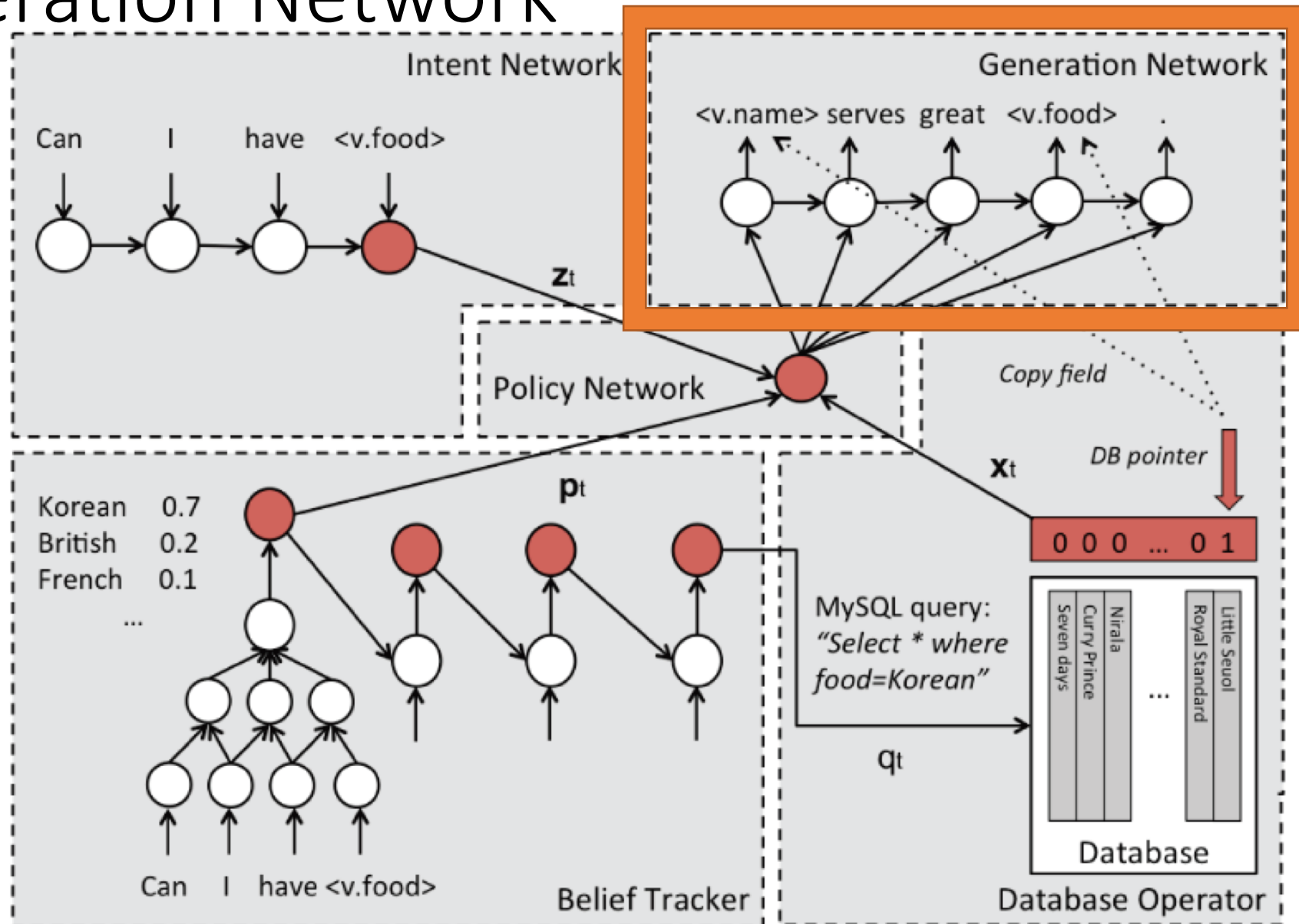
$$\mathbf{o}_t^{(j)} = \tanh(\mathbf{W}_{zo}\mathbf{z}_t + \hat{\mathbf{p}}_t^{(j)} + \mathbf{W}_{xo}\hat{\mathbf{x}}_t)$$

$$\hat{\mathbf{p}}_t^{(j)} = \sum \alpha_s^{(j)} \tanh(\mathbf{W}_{po}^s \cdot \hat{\mathbf{p}}_s^t)$$

$$\alpha_s^{(j)} = \text{softmax}(\mathbf{r}^\top \tanh(\mathbf{W}_r \cdot \mathbf{u}_t))$$

$$\mathbf{u}_t = \mathbf{z}_t \oplus \hat{\mathbf{x}}_t \oplus \hat{\mathbf{p}}_s^t \oplus \mathbf{w}_j^t \oplus \mathbf{h}_{j-1}^t$$

# Generation Network





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# Wizard-of-Oz Data Collection

- This paper proposed a novel crowdsourcing version of the Wizard-of-Oz paradigm
- Designed two webpages on Amazon Mechanical Turk, one for wizards and the other for users

# Wizard-of-Oz Data Collection

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 turn!

Customer : (Your response)

I don't care about the price range, just give me the address please.

Submit the HIT

# Wizard-of-Oz Data Collection

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 ?  
Customer : I dont care

Next turn

Please **modify** the following answers based on the latest customer response:

- What does user want?  
What is the **food type** the user wants?   
What is the **area** the user wants?   
What is the **price range** the user wants?
- What does user ask?  
Is the user asking for **food type** of an offered venue?   
Is the user asking for **price range** of an offered venue?   
Is the user asking for **area** of an offered venue?   
Is the user asking for **postcode** of an offered venue?   
Is the user asking for **phone number** of an offered venue?   
Is the user asking for **address** of an offered venue?   
Is the user mentioning any **restaurant names**?

Finish

Info Desk : (Your response)

I would recommend backstreet bistro, a great gastropub restaurant in the centre. do you want their phone number ?

Name	Food	Area	Price Range	Phone	Address	Postcode
backstreet bistro	gastropub	centre	expensive	01223 306306	2 Sturton Street City Centre	CB 1, 2 QA
royal standard	gastropub	east	expensive	01223 247677	280 Mill Road City Centre	CB 1, 3 NL
the cow pizza kitchen and bar	gastropub	centre	moderate	01223 308871	Com Exchange Street	CB 2, 3 QF
the slug and lettuce	gastropub	centre	expensive	-	34 - 35 Green Street	CB 2, 3 JU
<input type="text" value="nil"/>	<input type="text" value="gastropub"/>	<input type="text" value="nil"/>	<input type="text" value="nil"/>	<input type="text" value="nil"/>	<input type="text" value="nil"/>	<input type="text" value="nil"/>

Showing 1 to 4 of 4 entries (Filtered from 110 total entries)

Previous  Next

# Wizard-of-Oz Data Collection

- 99 restaurants in the DB
- 3000 HITs (Human Intelligence Tasks) in total
- 680 dialogues after data cleaning
- Cost ~ 400 USD

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# Empirical Experiments

Tracker type	Informable			Requestable		
	Prec.	Recall	F-1	Prec.	Recall	F-1
cnn	99.77%	96.09%	97.89%	98.66%	93.79%	96.16%
ngram	99.34%	94.42%	96.82%	98.56%	90.14%	94.16%

# Empirical Experiments

Encoder	Tracker	Decoder	Match(%)	Success(%)	T5-BLEU	T1-BLEU
<b>Baseline</b>						
lstm	-	lstm	-	-	0.1650	0.1718
lstm	turn recurrence	lstm	-	-	0.1813	0.1861
<b>Variant</b>						
lstm	rnn-cnn, w/o req.	lstm	89.70	30.60	0.1769	0.1799
cnn	rnn-cnn	lstm	88.82	58.52	0.2354	0.2429
<b>Full model w/ different decoding strategy</b>						
lstm	rnn-cnn	lstm	86.34	75.16	0.2184	0.2313
lstm	rnn-cnn	+ weighted	86.04	78.40	0.2222	0.2280
lstm	rnn-cnn	+ att.	90.88	80.02	0.2286	0.2388
lstm	rnn-cnn	+ att. + weighted	90.88	83.82	0.2304	0.2369



# Empirical Experiments

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

Metric	NN	HDC	Tie
Subj. Success	96.95%	95.12%	-
Avg. # of Turn	3.95	4.54	-
<b>Comparisons(%)</b>			
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

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# Conclusions

- Combines modularly connected model and end-to-end trainable model
- First end-to-end NN-based model that can conduct meaningful dialogues in a task-oriented application

Thank you