

HELPR: A Framework to Break the Barrier across Domains in Spoken Dialog Systems

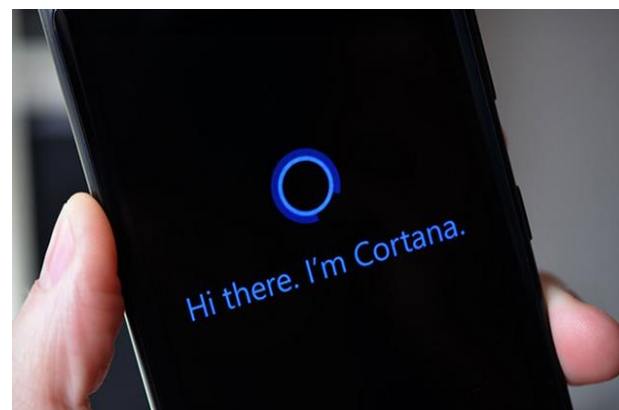
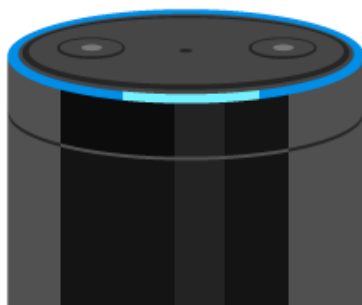
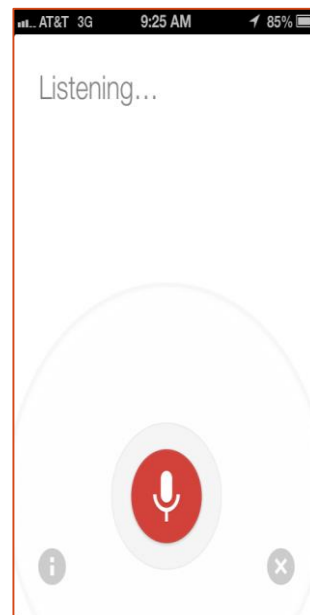
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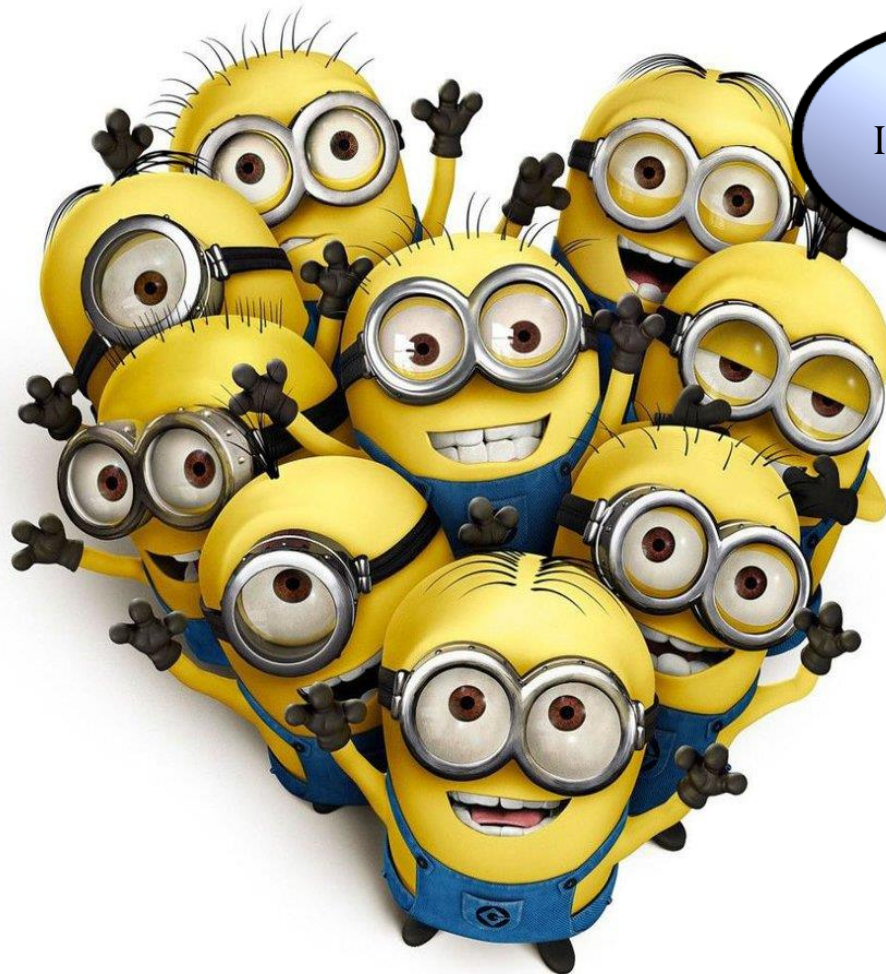
Carnegie Mellon University



Book a restaurant!



I can do Yelp!



Show me bus to China Palace.



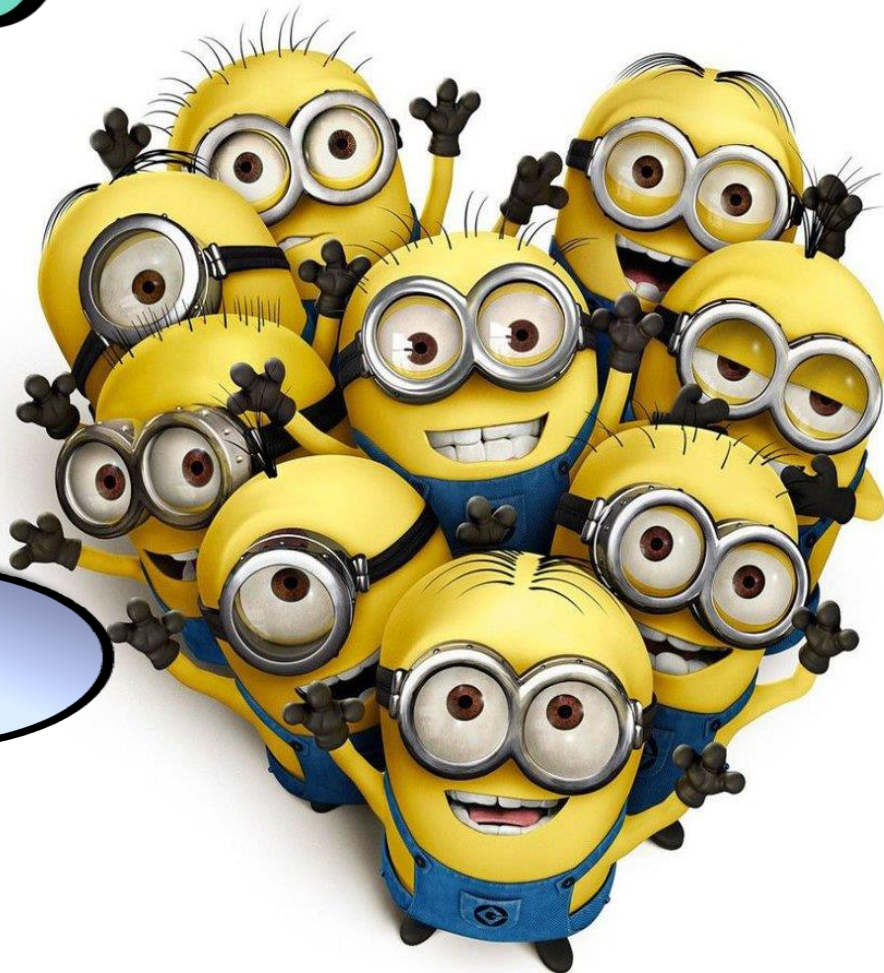
I know Maps!



Tell kids we will have Chinese tonight.



I will Text!



Problems (potential improvement)

- **Cannot handle complex intention**
- **Passively support cross-domain dialog**
- **No shared context**

Problem 1 --- Complex Intention

Plan a dinner for me and my kids!

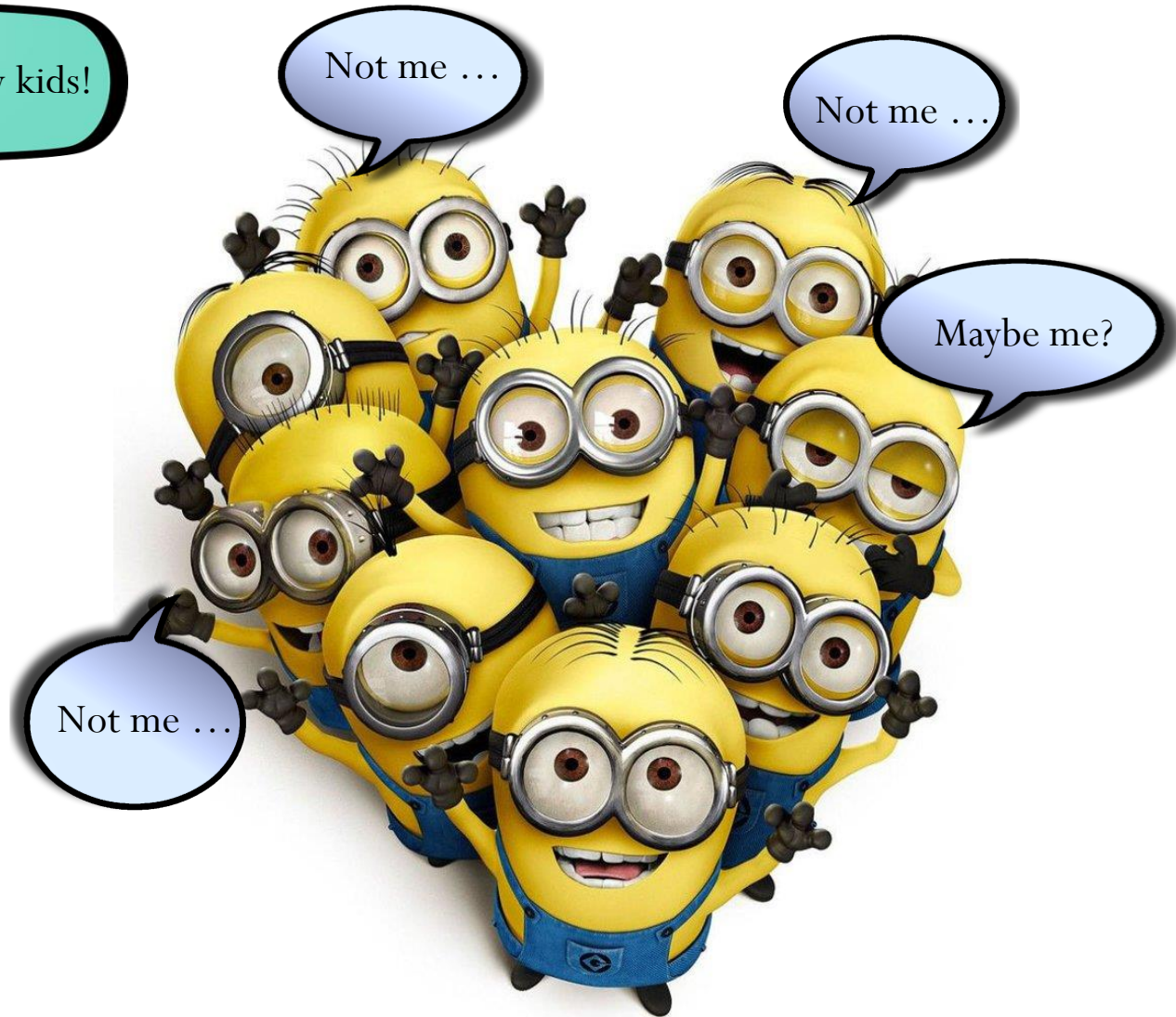


Not me ...

Not me ...

Maybe me?

Not me ...



Problem 1 --- Complex Intention

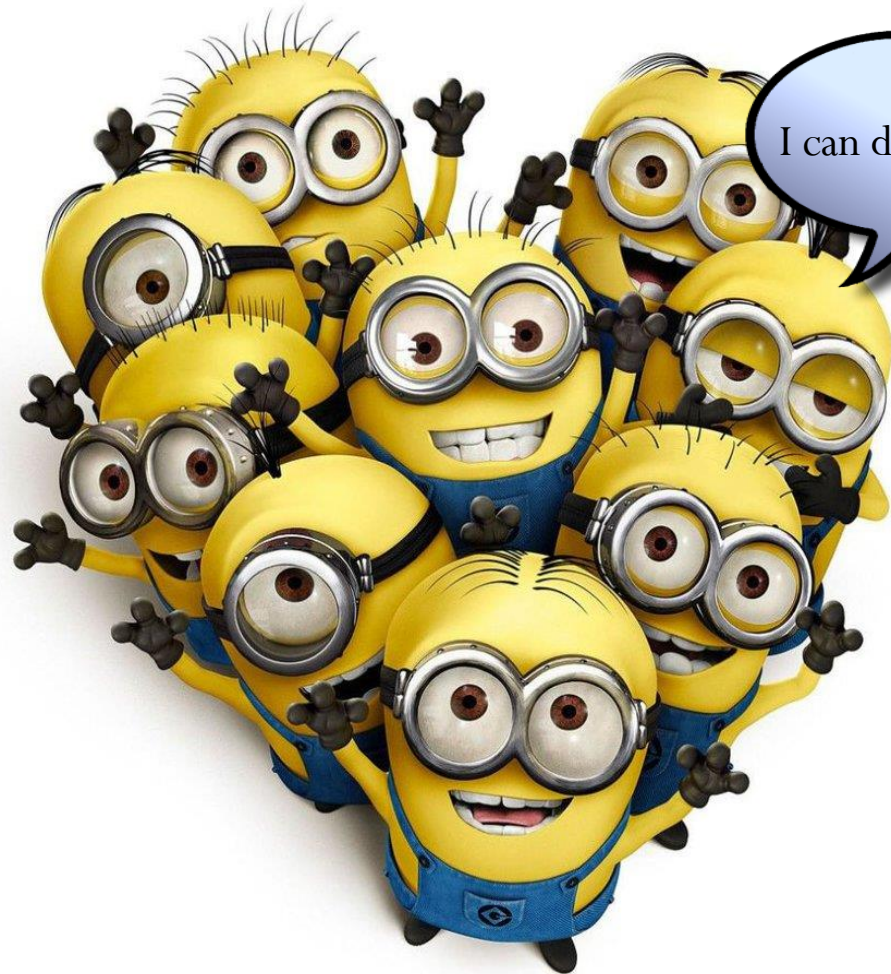
- Given a high-level/complex input, **ONE** closest domain is selected

Problem 2 -- Passive Cross-Domain Support

Book a restaurant!



I can do Yelp!



Problem 2 -- Passive Cross-Domain Support



Problem 2 -- Passive Cross-Domain Support



Problem 2 -- Passive Cross-Domain Support

- Given the current context, not aware of the next step/domain.

Problem 3 -- No Shared Context



Book a restaurant!



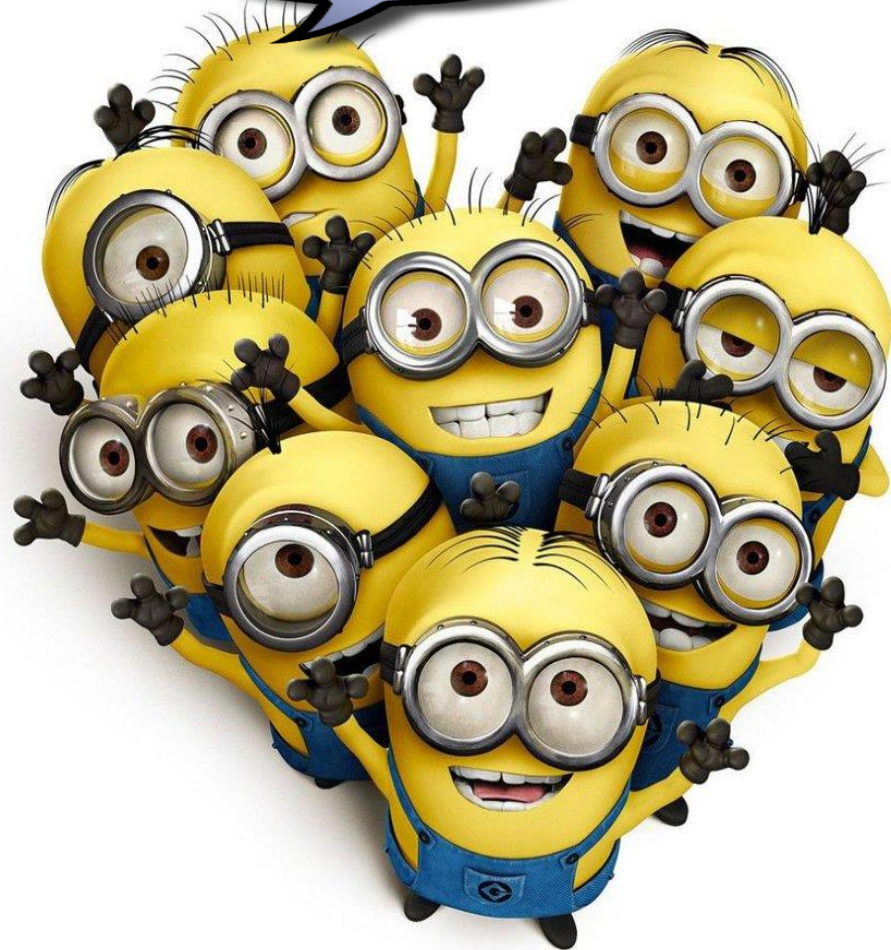
OK. China Palace has good review.

Problem 3 -- No Shared Context

Show me bus route **there**.



I know **Maps!** Where is your **destination?**



Problem 3 --- No Shared Context

- Not aware of the relationship with previous domains.

Solutions

- Cannot handle complex intention
 - **Sol**: Intention Recognition and Understanding (IWSDS'16, IUI'16)
- Passively support cross-domain dialog
 - **Sol**: Context-aware domain prediction (NIPS-SLU'15)
- No shared context
 - **Sol**: co-reference resolution ('there' -> 'China Palace')

Intention Understanding

Plan a dinner!

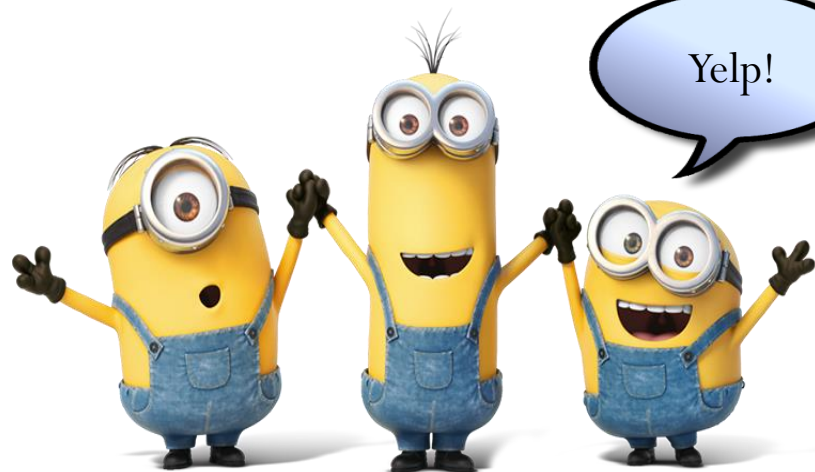


Yes Dad!
Special Team 1!



Intention Understanding

Find a restaurant!



Intention Understanding

Find a bus route!



Maps!



Intention Understanding

Message Agnes & Edith!



Messenger!



Intention Understanding

Set up meeting!



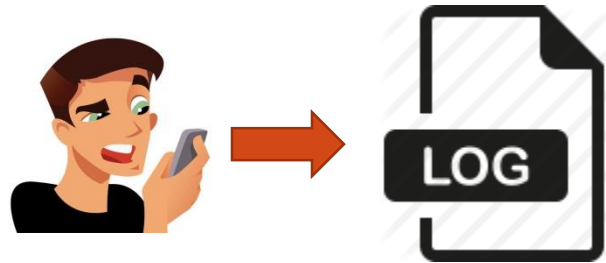
Yes Dad!
Special Team 2!



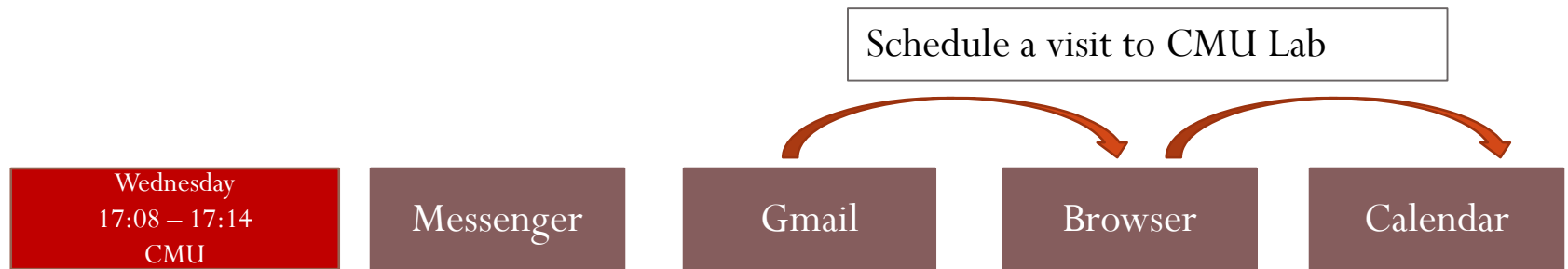
Approaches

- Step 1: Observe human user perform multi-domain tasks
 - Data collection
- Step 2: Learn to assist at task level
 - Map an input command to a set of domains
 - Talk at task level

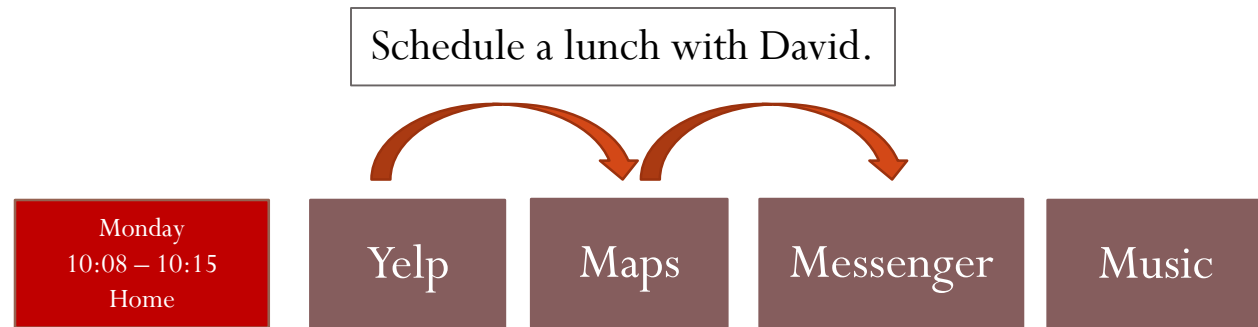
Data Collection 1 - Smart Phone



- Log app invocation + time/date/location
- Separate log into episodes if there is 3 minute inactivity



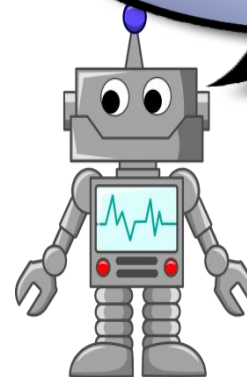
Data Collection 2 – Wizard-of-Oz



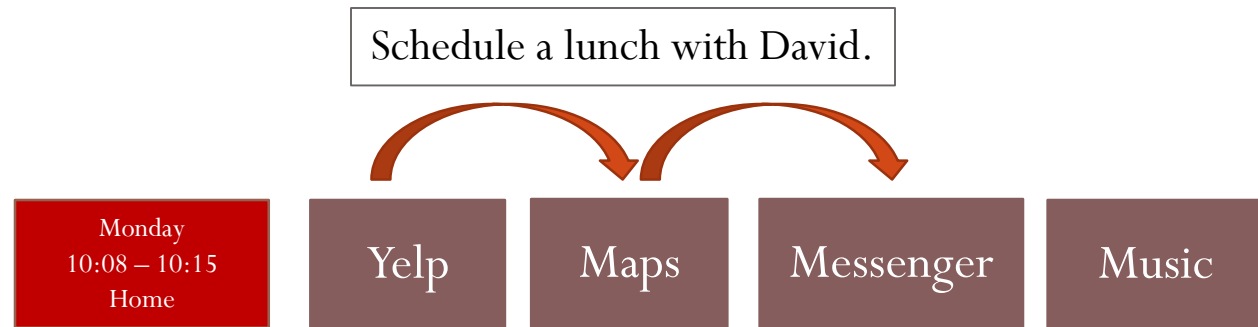
Find me an Indian place near CMU.



Yuva India is nearby.



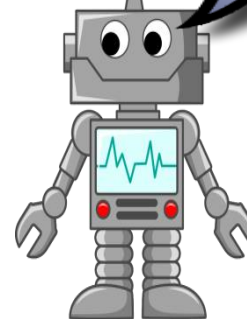
Data Collection 2 – Wizard-of-Oz



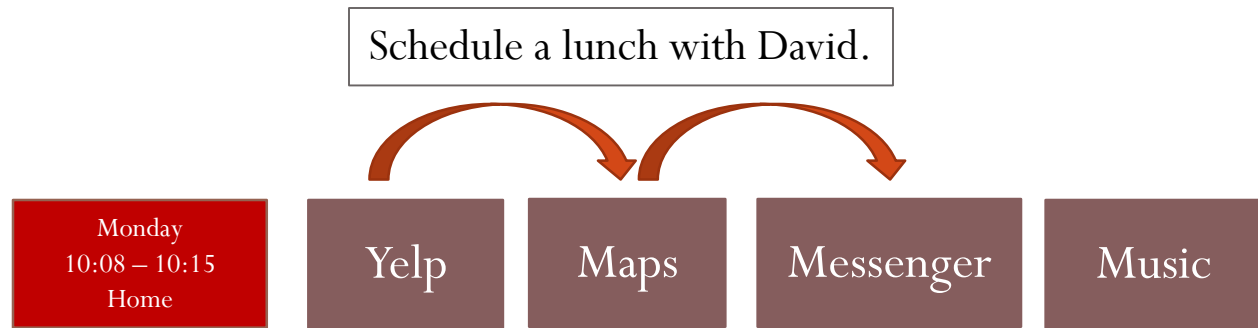
When is the next bus to school?



In 10 min, 61C.



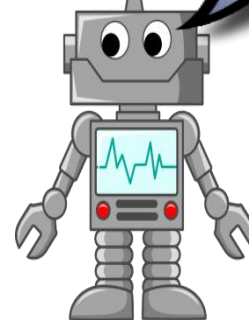
Data Collection 2 – Wizard-of-Oz



Tell David to meet me there in 15 min.




Message sent.

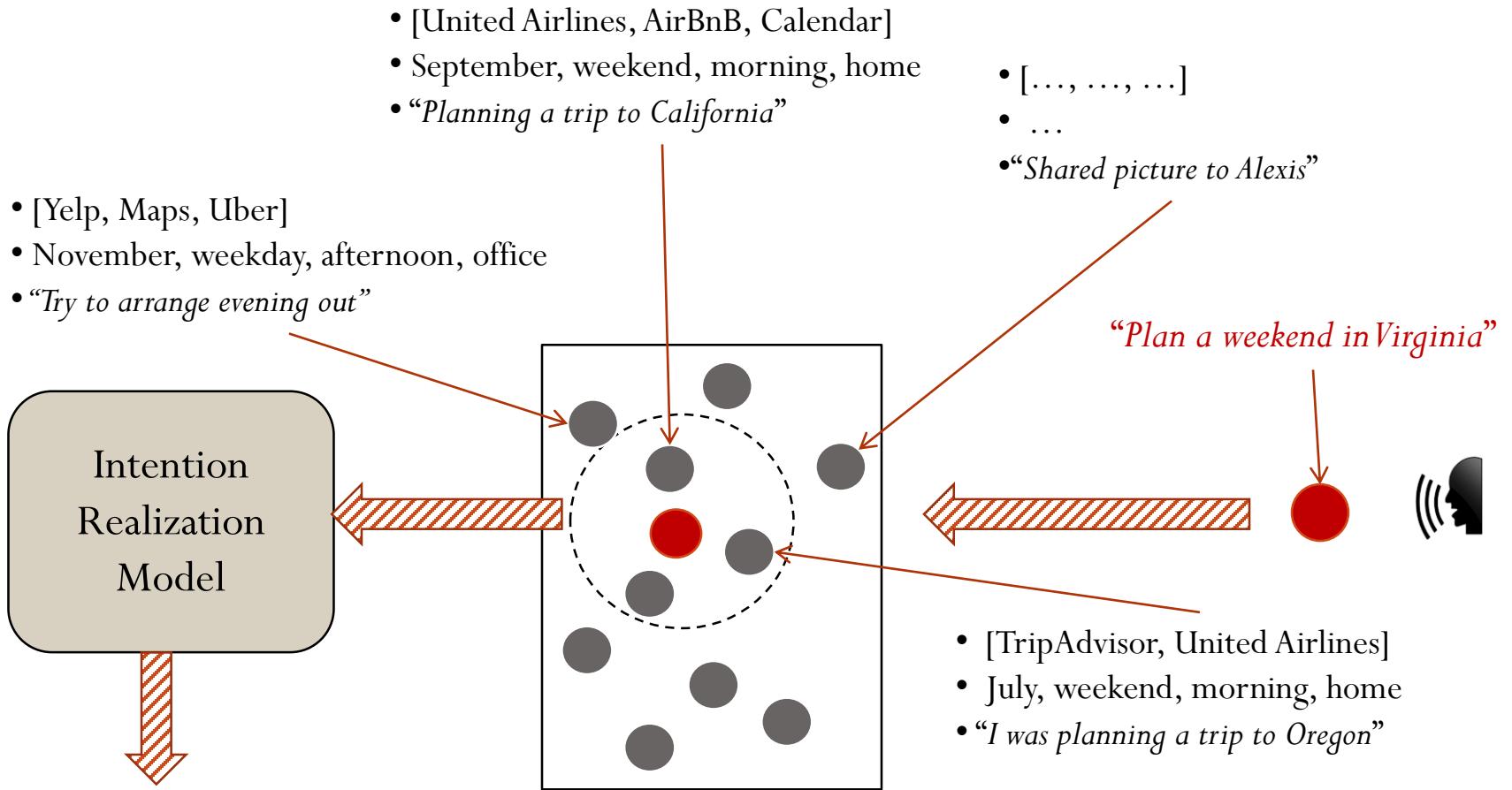


Data Collection Summary

- 533 real-life multi-domain interactions from 14 real users
 - 12 native English speakers
 - 4 males & 10 females
 - Mean age: 31
 - Total # unique apps: 130 (Mean = 19/user)



Resources	Examples	Usage
App sequences	Yelp->Maps->Messenger	structure/arrangement
Task descriptions	“Schedule a lunch with David”	nature of the intention, language reference
User utterances	“Find me an Indian place near CMU.”	language reference
Meta data	Monday, 10:08 – 10:15, Home	contexts of the tasks

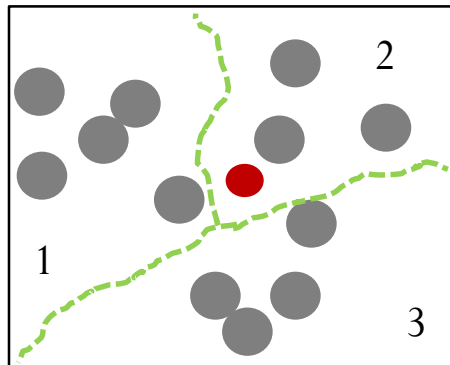


Infer:

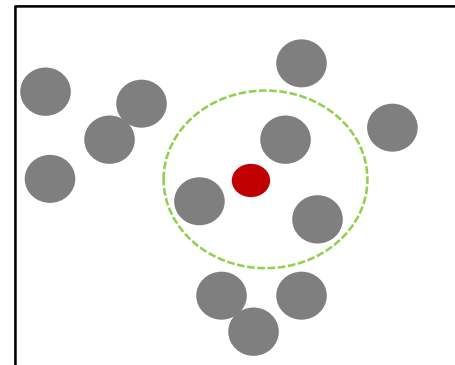
- 1) Supportive Domains: *United Airlines, TripAdvisor, AirBnB*
- 2) Summarization: “*plan trip*”

Find similar past experience

- Cluster-based:
 - K-means clustering on user generated language
- Neighbor-based:
 - KNN



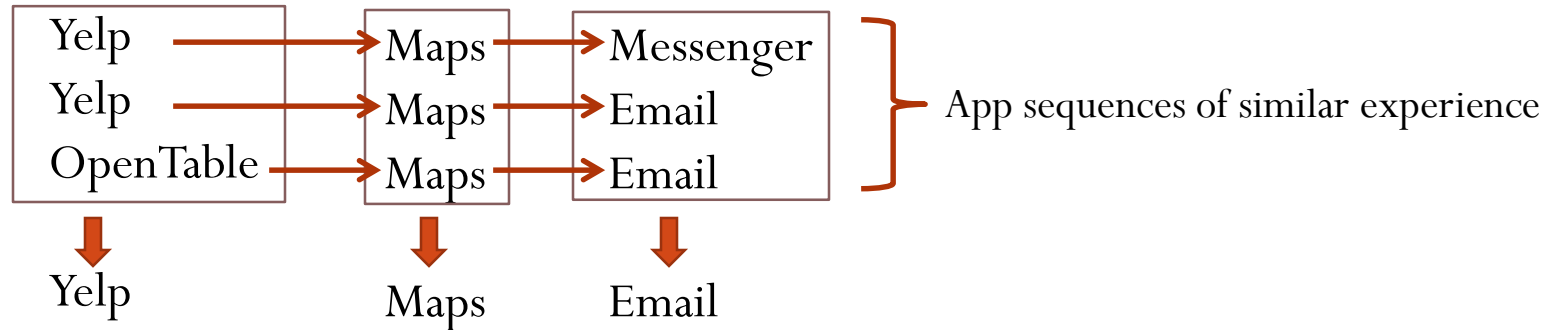
Cluster-based



Neighbor-based

Realize domains from past experience

- Representative Sequence



- Multi-label Classification

Obstacles

- Language-mismatch
 - **Sol:** Query Enrichment (QryEnr)
 - [“shoot”, “photo”] -> [“shoot”, “take”, “photo”, “picture”]
- App-mismatch
 - **Sol:** Ideally, use functionality (extracted from app descriptions) to compute distance between two apps

Realize domains from language

	Personalized			Generic		
	Precision	Recall	F	Precision	Recall	F
Cluster-based	71.4	47.4	55.7	48.4	16.5	23.8
+QryEnr	63.5	51.6	56.3	31.1	27.6	27.9
+AppSim	71.4	47.4	55.7	56.5	19.1	27.8
+QryEnr+AppSim	63.5	51.6	56.3	40.4	35.1	36.1
Neighbor-based	72.2	41.6	51.3	45.8	12.3	19.1
+QryEnr	68.1	49.9	57.0	35.2	17.4	22.9
+AppSim	72.2	41.6	51.3	55.6	14.6	22.7
+QryEnr+AppSim	68.1	49.9	57.0	42.2	21.5	28.0

Realize domains from language

	Personalized			Generic		
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Cluster-based	71.4	47.4	55.7	48.4	16.5	23.8
+QryEnr	-	+	+	-	+	+
+AppSim	N/A	N/A	N/A	+	+	+
+QryEnr+AppSim	-	+	+	-	+	+
Neighbor-based	72.2	41.6	51.3	45.8	12.3	19.1
+QryEnr	-	+	+	-	+	+
+AppSim	N/A	N/A	N/A	+	+	+
+QryEnr+AppSim	-	+	+	-	+	+

Talk at task level

- Techniques:
 - (Extractive/abstractive) summarization
 - Key phrase extraction
- User study:
 - Key phrase extraction + user generated language
 - Ranked list of key phrases + user's binary judgment

Looking up math problems. (Desc)
Now open a browser. (Utt)
Go to slader.com. (Utt)
Doing physics homework. (Desc)
...
Go to my Google drive. (Utt)
Look up kinematic equations. (Utt)
Now open my calculator so I can plug in numbers. (Utt)



1. solutions online
2. project file
3. Google drive
4. math problems
5. physics homework
6. answers online
...

Talk at task level

- Metrics
 - Mean Reciprocal Rank (MRR)
- Result:
 - $MRR > 0.6$
 - understandable verbal reference among the top 2 of the list

Looking up math problems. (Desc)
Now open a browser. (Utt)
Go to slader.com. (Utt)
Doing physics homework. (Desc)
...
Go to my Google drive. (Utt)
Look up kinematic equations. (Utt)
Now open my calculator so I can plug in numbers. (Utt)



1. solutions online
2. project file

3. Google drive
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...

Conclusion

- Collected real-life cross-domain interactions from real users
- HELPR framework to learn to assist at task level
 - Suggest a set of supportive domains to accomplish the task
 - Personalized model > Generic model
 - The gap can be reduced by QryEnr + AppSim
 - Generate language reference to communicate verbally at task level

Future Work

- From multiple domains to a **unified** dialog
 - Dynamically and automatically configure the dialog flow
 - Actively maintain a shared context
 - Reduce redundancy
 - Improve naturalness

Thanks!

- References:

- M. Sun, YN. Chen and A. I. Rudnicky, "An Intelligent Assistant for High-Level Task Understanding".(to appear) ACM Conference on Intelligent User Interfaces (IUI), 2016.
- M. Sun, YN. Chen and A. I. Rudnicky, "HELPR: A Framework to Break the Barrier across Domains in Spoken Dialog Systems". (to appear) International Workshop on Spoken Dialogue Systems (IWSDS) 2016.
- M. Sun, YN. Chen and A. I. Rudnicky, "Understanding User's Cross-Domain Intentions in Spoken Dialog Systems". NIPS Workshop on Machine Learning for SLU and Interaction 2015.