

Predicting User Engagement Status for Online Evaluation of Intelligent Assistants

Rui Meng, Zhen Yue, Alyssa Glass



Introduction

- Intelligent Assistants (IA)
 - New portal to Internet services
 - Support a wide variety of functions
- Challenges in Online Evaluation for IA
 - Large volume of daily user requests
 - Real-time performance monitoring





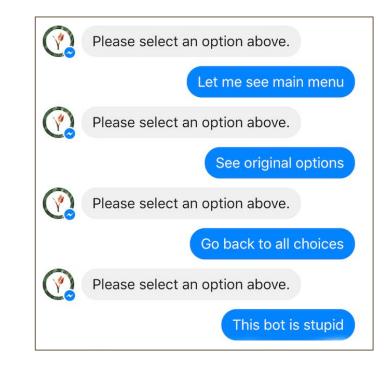
Introduction

• Intelligent Assistants (IA)

- New portal to Internet services
- Support a wide variety of functions

• Challenges in Online Evaluation for IA

- Large volume of daily user requests
- Real-time performance monitoring
- Understanding user utterances



Online IA Evaluation

• Desiderata

- Reflect <u>real-time</u> system performance
- Can <u>scale</u> to billions of requests
- Indicative of <u>end-user engagement/satisfaction</u>
- <u>Interpretable</u>, easy for engineers to locate system errors

Previous Methods for IA Evaluation

1. Evaluation on individual components

- E.g. dialogue state tracking, dialogue act classification
- Does not reflect the overall system performance and user engagement

2. Evaluation by comparing system outputs with ground-truth responses

- Commonly used in dialogue generation systems, using metrics like BLEU, METEOR
- Correlated poorly with human rating
- 3. Human-based evaluation
 - e.g. using Amazon Mechanical Turk
 - Cannot scale up; Does not reflect real-time performance

Motivation

- Lessons from Online Evaluation for IR Systems
 - User behavioral signals can be effective and reliable indicators
 - E.g. long-dwell click, query reformulation, query abandonment

• Lessons from PARADISE framework

- A classic framework for evaluating spoken dialogue agents
- A successful dialogue system should <u>maximize task success</u> and <u>minimize cost</u>
 - Requires good ways to segment tasks and measure their success/cost
- Can we find good user behavioral signals for evaluating intelligent assistants?

Session from Yahoo Captain (Family Assistant)

Side	Content	Quality
User	"Chic peas, diced tomatoes"	
Bot	["I'm learning. Can you clarify?"]	Bad
User	"Add Chic peas, diced tomatoes"	
Bot	["I'm learning. Can you clarify?"]	Bad
User	"Add to list Chic peas, diced tomatoes"	
Bot	["Great! I've added your items, Sarah.\n\nHere's your list:\n1. potatoes\n2. sweet potatoes\n3. kale\n4. can crushed tomatoes\n5. almond butter\n6. broccoli\n7. beer\n8. onions\n9. peas\n10. hamburger\n11. cauliflower\n12. chic peas\n13. diced tomatoes"]	Good

- Task success ≠ satisfactory user experience
 - Both utterance-level and task-level engagement matter

Session from DSTC-2

Side	Content	Quality
Bot	Thank you for calling the Cambridge Information system. Your call will be recorded for research purposes. You may ask for information about a place to eat, such as a restaurant, a pub, or a cafe. How may I help you?	
User	im looking for a turkish restaurant	
Bot	What kind of price range are you looking for? For example, cheap, or moderate.	G or B?
User	uh moderate	
Bot	Are you looking for somewhere near a particular venue? For example, the Museum Of Technology, Churchill College, or Clare Hall.	G or B?
User	no	
Bot	196 Meze Bar is an outstanding restaurant and it serves turkish food. It's in the moderate price range.	Good

- A successful task may need several necessary turns
 - They should not be treated as negative experience, nor positive
 - Measuring cost by number of turns only is not reasonable

Our Method

- Utterance Classification by User Engagement Status
 - Define a four-class utterance classification scheme
 - Each class represents the engagement status of a user after seeing system response
 - **Fulfillment (F):** current user request is understood and fulfilled by the system.
 - Continuation (C): current user request is understood by the system but more interactions are required to complete this request.
 - *Reformulation (R)*: current user request is wrongly or incompletely understood by the system, and user repeats/rephrases this request in the next turn.
 - Abandonment (A): current user request is not understood or fulfilled by the system, and the user abandons this request by closing the conversation or starting a new request.

	# Turn		Utterance		
		Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?		
	1	User:	Moderately priced		
		Bot:	What part of town do you have in mind?		
Request a	2	User:	North		
restaurant		Bot:	Would you like something in the cheap, moderate, or expensive price range?		
	3	User:	Moderate		
		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
		User:	A ddmass		
	4	User:	Address		
Ask address	4	Bot:	Sure , golden wok is on 191 Histon Road Chesterton		
Ask address	5		Sure, golden wok is on 191 Histon Road		
Ask address Ask number		Bot:	Sure , golden wok is on 191 Histon Road Chesterton		
		Bot: User:	Sure , golden wok is on 191 Histon Road Chesterton Phone number The phone number of golden wok is		
Ask number	5	Bot: User: Bot:	Sure , golden wok is on 191 Histon Road Chesterton Phone number The phone number of golden wok is 01223 350688		
	5	Bot: User: Bot: User:	Sure , golden wok is on 191 Histon Road Chesterton Phone number The phone number of golden wok is 01223 350688 Postcode Golden wok is a nice restaurant in the		
Ask number	6	Bot: User: Bot: User: Bot:	Sure , golden wok is on 191 Histon Road Chesterton Phone number The phone number of golden wok is 01223 350688 Postcode Golden wok is a nice restaurant in the north of town in the moderate price range		

II THE I III III IIII			
# Turn		Utterance	Label
	Bot:	Hello, welcome to the Cambridge	
1	User:	Moderately priced	R
	Bot:	What part of town do you have in mind?	
2	User:	North	С
	Bot:	Would you like something in the cheap, moderate, or expensive price range?	
3	User:	Moderate	F
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range	
4	User:	Address	F
	Bot:	Sure , golden wok is on 191 Histon Road Chesterton	
5	User:	Phone number	F
	Bot:	The phone number of golden wok is 01223 350688	
6	User:	Postcode	R
6	User: Bot:	Postcode Golden wok is a nice restaurant in the north of town in the moderate price range	R
6		Golden wok is a nice restaurant in the	R
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range	
	3	Bot:1User:Bot:2User:Bot:3User:Bot:4User:Bot:5User:	Bot: Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you? 1 User: Moderately priced Bot: Bot: What part of town do you have in mind? 2 User: North Bot: Bot: Would you like something in the cheap, moderate, or expensive price range? 3 User: Moderate Bot: Golden wok is a nice restaurant in the north of town in the moderate price range 4 User: Address Bot: Sure , golden wok is on 191 Histon Road Chesterton 5 User: Phone number The phone number of golden wok is

	# Turn		Utterance	Label
		Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?	
	1	User:	Moderately priced	R
-		Bot:	What part of town do you have in mind?	
Request a	2	User:	North	С
restaurant		Bot:	Would you like something in the cheap, moderate, or expensive price range?	
	3	User:	Moderate	F
		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range	
	4	User:	Address	F
Ask address		Bot:	Sure, golden wok is on 191 Histon Road Chesterton	
	5	User:	Phone number	F
Ask number	5	User: Bot:	Phone number The phone number of golden wok is 01223 350688	F
Ask number	5		The phone number of golden wok is	F
-		Bot:	The phone number of golden wok is 01223 350688	
Ask number - Ask postcode -		Bot: User:	The phone number of golden wok is 01223 350688 Postcode Golden wok is a nice restaurant in the	
-	6	Bot: User: Bot:	The phone number of golden wok is 01223 350688 Postcode Golden wok is a nice restaurant in the north of town in the moderate price range	R

-	# Turn		Utterance	Label	Task
		Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?		
	1	User:	Moderately priced	R	
		Bot:	What part of town do you have in mind?		
Request a	2	User:	North	С	Success
restaurant		Bot:	Would you like something in the cheap, moderate, or expensive price range?		
	3	User:	Moderate	F	
		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
	4	User:	Address	F	Success
Ask address		Bot:	Sure , golden wok is on 191 Histon Road Chesterton		Duccess
	5	User:	Phone number	F	Success
Ask number		Bot:	The phone number of golden wok is 01223 350688		Success
-	6	User:	Postcode	R	
Ask postcode		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
Ask posicoue	7	User:	Postcode	А	Failure
		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
	8	User:	Thank you good bye		

	# Turn		Utterance	Label
		Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?	
	1	User:	Moderately priced	R
		Bot:	What part of town do you have in mind?	\sim
Request a	2	User:	North	С
restaurant		Bot:	Would you like something in the cheap, moderate, or expensive price range?	
	3	User:	Moderate	F
		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range	
	4	User:	Address	F
Ask address		Bot:	Sure, golden wok is on 191 Histon Road Chesterton	
	5	User:	Phone number	F
Ask number		Bot:	The phone number of golden wok is	
			01223 350688	
	6	User:		R
Ask nostrode	6	User: Bot:	01223 350688	R
Ask postcode	6		01223 350688 Postcode Golden wok is a nice restaurant in the	R
Ask postcode		Bot:	01223 350688 Postcode Golden wok is a nice restaurant in the north of town in the moderate price range	

	# Turn		Utterance	Label
		Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?	
	1	User:	Moderately priced	R
Request a		Bot:	What part of town do you have in mind?	
	2	User:	North	C
restaurant		Bot:	Would you like something in the cheap, moderate, or expensive price range?	
-	3	User:	Moderate	F
		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range	
	4	User:	Address	F
Ask address		Bot:	Sure, golden wok is on 191 Histon Road Chesterton	
	5	User:	Phone number	F
Ask number		Bot:	The phone number of golden wok is 01223 350688	
	6	User:	Postcode	R
Ask postcodo	6	User: Bot:	Postcode Golden wok is a nice restaurant in the north of town in the moderate price range	R
Ask postcode	6		Golden wok is a nice restaurant in the	R
Ask postcode		Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range	

Our Method

- The classification implies two binary classification tasks
 - 1. Response Success
 - \circ $\;$ whether user's current utterance is reasonably responded or not

	Ongoing	Ending
Correctly responded	Continuation	Fulfillment
Wrongly responded	Reformulation	Abandonment

Our Method

- The classification implies two binary classification tasks
 - 1. Response Success
 - \circ $\;$ whether user's current utterance is reasonably responded or not
 - 2. Task completion (task boundary detection)
 - \circ $\;$ whether the current task has ended or not

	Ongoing	Ending
Correctly responded	Continuation	Fulfillment
Wrongly responded	Reformulation	Abandonment
····		

Online Metrics w/ User Engagement Status

• We define an online metric **User Engagement Score** (*UE_{score}*) taking into account both <u>task success</u> and <u>dialogue efficiency</u>.

$$UE_{SCORE} = \frac{SuccessRate + EfficiencyRate}{2}$$

- Success Rate
 - The ratio of tasks that were successfully accomplished.
- Efficiency Rate
 - Reflected by the <u>number of reformulated utterances</u> and <u>average task length</u>.

# Turn		Utterance	Label	Task
	Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?		
1	User:	Moderately priced	R	
	Bot:	What part of town do you have in mind?		
2	User:	North	С	Success
	Bot:	Would you like something in the cheap, moderate, or expensive price range?		
3	User:	Moderate	F	
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
4	User:	Address	F	Success
	Bot:	Sure , golden wok is on 191 Histon Road Chesterton		Success
5	User:	Phone number	F	Success
	Bot:	The phone number of golden wok is 01223 350688		Success
6	User:	Postcode	R	
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
7	User:	Postcode	Α	Failure
/	r – – –	Golden wok is a nice restaurant in the		
1	Bot:	north of town in the moderate price range		

$$SuccessRate = \frac{\#(TASK_{success \in S})}{\#(TASK_{\in S})}$$
$$= \frac{3}{4} = 0.75$$

# Turn		Utterance	Label	Task
	Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?		
1	User:	Moderately priced	R	
	Bot:	What part of town do you have in mind?		
2	User:	North	С	Success
	Bot:	Would you like something in the cheap, moderate, or expensive price range?		
3	User:	Moderate	F	
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
4	User:	Address	F	Success
	Bot:	Sure, golden wok is on 191 Histon Road Chesterton		Success
5	User:	Phone number	F	Success
	Bot:	The phone number of golden wok is 01223 350688		Success
6	User:	Postcode	R	
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		17.11
7	User:	Postcode	Α	Failure
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		
8	User:	Thank you good bye		

• The more reformulated utterances, the more cost

$$ReformRate = \sum_{T \in \mathbb{S}} \frac{\#(UTT_{reform \in T})}{\#(UTT_{\in T})}$$
$$= \frac{2}{7} = 0.286$$

The longer task length, the more user fatigue
α is a hyperparameter (=1), acceptable number of

turns

$$FatigueValue = \frac{\sum_{T \in \mathbb{S}} max(1, \#(UTT_{\in T}) - \alpha)}{\#(TASK_{\in \mathbb{S}})}$$
$$= \frac{2 + 1 + 1 + 1}{4} = 1.25$$

 $EfficiencyRate = \frac{1 - ReformRate}{FatigueValue} = \frac{1 - 0.286}{1.25} = 0.5712$

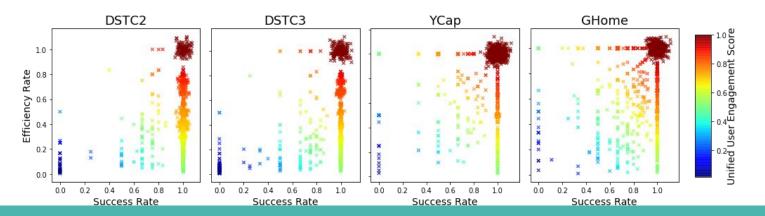
# Turn		Utterance	Label	Task	
	Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?			
1	User:	Moderately priced	R		
	Bot:	What part of town do you have in mind?			
2	User:	North	С	Success	
	Bot:	Would you like something in the cheap, moderate, or expensive price range?			
3	User:	Moderate	F		
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range			
4	User:	Address	F	Success	
	Bot:	Sure, golden wok is on 191 Histon Road Chesterton		Success	
5	User:	Ser: Phone number		Success	
	Bot: The phone number of golden wok is 01223 350688			Success	
6	User:	Postcode	R		
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		17.11	
7	User:	Postcode	Α	Failure	
	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range			
	User:	Thank you good bye			

• Average of task success and dialogue efficiency

$$UE_{SCORE} = \frac{SuccessRate + EfficiencyRate}{2}$$
$$= \frac{0.75 + 0.5712}{2} = 0.661$$

Case Study on Four Annotated Datasets

- Datasets
 - 4 IA systems: **DSTC2**, **DSTC3**, Yahoo Captain (**YCap**), Google Home (**GHome**)
 - 4,000 dialogue sessions, labelled by professional annotators, kappa=0.790.
 - Identified 35/63/9/13 low-quality sessions whose UE score is less than 0.2
 - DSTC2/DSTC3: poor ASR and language understanding ability
 - YCap: only takes user commands matching particular templates
 - **GHome**: Problems are more diverse, e.g. failed to answer open-domain questions



Automatic Engagement Status Prediction

• Goal

- Preliminary exploration of automatic prediction using machine learning methods
- Identify potential technical challenges in this task

• Experiment setting

- An utterance-level four-way classification
- Models
 - Traditional model (Logistic Regression, SVM, Random Forest) + feature engineering
 - Convolutional Neural Networks
- Extract 7 groups of features from utterances
 - 1. Basic Feature (utterance length)
 - 2. User Action Feature
 - 3. Time Interval Feature
 - 4. Lexical Feature
 - 5. Phrasal Feature
 - 6. Syntactic Feature
 - 7. Semantic/Topic Feature

• Models

• CNNs perform well without any manual feature.

Model	Accuracy	F1-score
Majority	0.6020	0.1858
Random	0.2503	0.2029
SVM	0.8410	0.6440
LR	0.8398	0.6413
RF	0.8415	0.6192
CNN.Rand	0.8287	0.6549
CNN.MultiCh	0.8367	0.6674

• Models

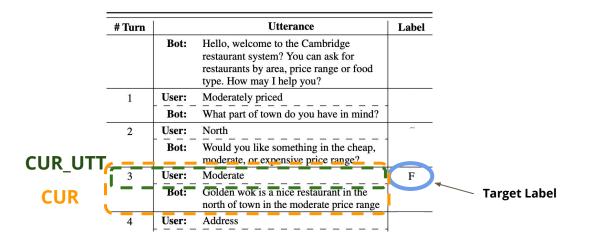
- CNNs perform well without any manual feature design.
- But classic models can improve significantly with careful feature engineering, e.g.

combining multiple groups and feature selection. =

Model	Accuracy	F1-score
Majority	0.6020	0.1858
Random	0.2503	0.2029
SVM	0.8410	0.6440
LR	0.8398	0.6413
RF	0.8415	0.6192
CNN.Rand	0.8287	0.6549
CNN.MultiCh	0.8367	0.6674

Model	w\o FeatSelect	w\Fea	tSelect
Widden	w\Sim	w\o Sim	w\Sim
(a) Basic	0.4105	-	0.4105
(b) Phrasal	0.6316	-	0.6318
(c) Syntactic	0.6280	-	0.6402†‡
(d) N-grams	0.6573	-	0.6770†‡
(e) Topic model	0.6346	-	0.6358
(f) Word2Vec	0.6521	-	0.6523
(g) Doc2Vec	0.5968	-	0.5969
(h) Skip-thought	0.6216	-	0.6216
(i) (a) + (b) + (c) + (d)	0.6694	0.6511	0.7085†‡
(j) + Topic Model	0.6720	0.6617	0.7152 †‡
(k) + Word2Vec	0.6790	0.6617	0.7135†‡
(1) $+$ Doc2Vec	0.6713	0.6631	0.7100 †‡
(m) + Skip-thought	0.6747	0.6666	0.7124†±
(n) All	0.6825	0.6589	0.7140 †‡
(o) CNN.Rand	0.6647	-	-
(p) CNN.MultiCh	0.6685	-	-

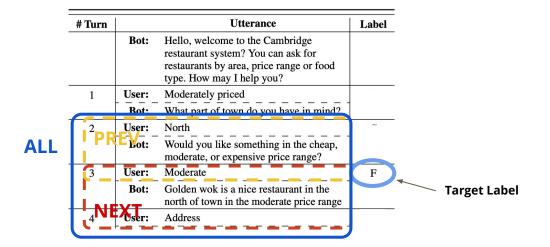
- Context Range
 - Five settings of context window
 - CUR_UTT = {user_utt_i}
 - CUR = {user_utt_i, bot_utt_i}



- Context Range
 - Five settings of context window
 - PREV = {user_utt_{i-1}, bot_utt_{i-1}, user_utt_i}
 - NEXT = {user_utt_i, bot_utt_i, user_utt_{i+1}}

# Turn	Utterance		Label	
	Bot:	Hello, welcome to the Cambridge restaurant system? You can ask for restaurants by area, price range or food type. How may I help you?		-
1	User:	Moderately priced		
	Bot:	What part of town do you have in mind?		
² PR	User: EV- Bot:	North Would you like something in the cheap, moderate, or expensive price range?	~	-
3	User:	Moderate	F	
Г — 1 ми	Bot:	Golden wok is a nice restaurant in the north of town in the moderate price range		Target La
_4NE	User:	Address		

- Context Range
 - Five settings of context window
 - ALL = {user_utt_{i-1}, bot_utt_{i-1}, user_utt_i, bot_utt_i, user_utt_{i+1}}



• Context Range

• Including the user's next utterance is critical to predict user engagement (CUR -> NEXT)

Model	CUR_UTT	CUR	NEXT	PREV	ALL
Basic	0.3425	0.3503	0.3836	0.3501†‡	0.3963†‡
Phrasal	0.3679	0.5521	0.5913	0.3709	0.5661†‡
Syntactic	0.3485	0.5530	0.6078	0.3671†‡	0.5867†‡
N-grams	0.3839	0.5694	0.6113	0.3788	0.5984†‡
Topic Model	0.2982	0.5255	0.5803	0.3464†‡	0.5829
Word2Vec	0.3704	0.5723	0.6162	0.3827†‡	0.6032†‡
Doc2Vec	0.3427	0.5379	0.5858	0.3722†‡	0.5740†‡
Skip-thought	0.3648	0.5545	0.6063	0.3692	0.6008†
CNN.Rand	0.4252	<u>0.5862</u>	0.6647	0.4153	0.6549†
CNN.MultiCh	0.4207	0.5829	<u>0.6685</u>	<u>0.4288</u>	<u>0.6674</u>

• Context Range

- Including the user's next utterance is critical to predict user engagement (CUR -> NEXT)
- Previous history is not helpful, can be even detrimental (**NEXT -> ALL**)

Model	CUR_UTT	CUR	NEXT	PREV	ALL
Basic	0.3425	0.3503	0.3836	0.3501†‡	0.3963†‡
Phrasal	0.3679	0.5521	0.5913	0.3709	0.5661†‡
Syntactic	0.3485	0.5530	0.6078	0.3671†‡	0.5867†‡
N-grams	0.3839	0.5694	0.6113	0.3788	0.5984†‡
Topic Model	0.2982	0.5255	0.5803	0.3464†‡	0.5829
Word2Vec	0.3704	0.5723	0.6162	0.3827†‡	0.6032†‡
Doc2Vec	0.3427	0.5379	0.5858	0.3722†‡	0.5740†‡
Skip-thought	0.3648	0.5545	0.6063	0.3692	0.6008†
CNN.Rand	0.4252	0.5862	0.6647	0.4153	0.6549†
CNN.MultiCh	0.4207	0.5829	<u>0.6685</u>	0.4288	0.6674

• Similarity Features

- Similarities (cosine, jaccard distance etc.) between two adjacent user utterances (user_utt_i, user_utt_{i+1})
- Huge boost for detecting reformulations.

Model	w\o Similarity	w\Similarity
Basic	0.3836	0.4105 (+2.69%)
Phrasal	0.5913	0.6316 (+4.03%)†‡
Syntactic	0.6078	0.6280 (+2.02%)†‡
N-grams	0.6113	0.6573 (+4.60%)†‡
Topic Model	0.5803	0.6346 (+5.43%)†‡
Word2Vec	0.6162	0.6521 (+3.59%)†‡
Doc2Vec	0.5858	0.5968 (+1.10%)
Skip-thought	0.6063	0.6216 (+1.53%)

Failure Cases

- Manually examine 50 random wrongly-predicted examples from GHome dataset
 - **R**eformulation / **A**bandonment / **F**ulfillment / Continuation=44% / 42% / 8% / 6%
 - Common failure reasons:
 - i. (32%) model fails to understand the relevance of a system response to a user's request
 - ii. (30%) needs long-term contextual information
 - iii. (18%) model fails to understand reformulated utterances
 - "I want the stair lights" vs. "turn on the stair lights"

Future Work

- Verify the effectiveness of proposed metrics
 - User study (offline)
 - A/B testing (online, large-scale)
- Improve automatic prediction
 - Increase NLU with pretrained language models e.g. BERT
 - Pretraining with large amounts of un-annotated dialogues



rui.meng@pitt.edu memray.me

Deep Keyphrase Generation

- Use neural networks to summarize keywords/keyphrases of a long text.
 - Deep Keyphrase Generation, ACL 2017
 - One Size Does Not Fit All: Generating and Evaluating Variable Number of Keyphrases, ACL 2020
 - An Empirical Study on Neural Keyphrase Generation, NAACL 2021
- Studies have shown its effectiveness for improving IR performance
 - By Florian Boudin & Ygor Gallina:
 - Keyphrase Generation for Scientific Document Retrieval
 - Redefining Absent Keyphrases and their Effect on Retrieval Effectiveness, arXiv 2103.12440

