CS 224S / Linguist 285 Spoken Language Processing

Andrew Maas | Stanford University | Spring 2024

Lecture 16: Dialog Systems Overview: Human conversations. Task-oriented dialog.

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Logistics

- Project milestones due Monday 11:59pm
- What's in a milestone?
 - Written initial stub of your final report. Include experiment setup, datasets, baseline model descriptions, results so far, related work etc.
 - Ideally you keep adding to the milestone to form your final report, just tell us what remaining work you plan to do
 - Must use ACL PDF template. You will lose points for other formats. Try Overleaf for easy setup
- Tight timing just after in-class check-ins, here's why
 - Encourage you to continue clarifying and solidifying project direction + experiments
 - Gives us a more detailed view of your project so far compared to check-in
 - We will quickly review and flag any projects in need of major course correction
- Grading for project milestones and final reports will allow room for excellent projects to stand out
 - A 100% on milestone or final report is a paper that is ready for a top-tier NLP/speech conference
 - Don't worry, final grading in the class is on a curve

Outline

- Human Conversation. Grounding & dialog acts
- Dialog system conceptual architecture
- Frame-based dialog

Human Conversation

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Task-Oriented Human Conversation

- Turn-taking
- Speech Acts
- Grounding

Stanford University C_1 : ... I need to travel in May.

- A1: And, what day in May did you want to travel?
- C_2 : OK uh I need to be there for a meeting that's from the 12th to the 15th.
- A₂: And you're flying into what city?
- C_3 : Seattle.
- A₃: And what time would you like to leave Pittsburgh?
- C₄: Uh hmm I don't think there's many options for non-stop.
- A₄: Right. There's three non-stops today.
- C_5 : What are they?
- A₅: The first one departs PGH at 10:00am arrives Seattle at 12:05 their time. The second flight departs PGH at 5:55pm, arrives Seattle at 8pm. And the last flight departs PGH at 8:15pm arrives Seattle at 10:28pm.
- C_6 : OK I'll take the 5ish flight on the night before on the 11th.
- A₆: On the 11th? OK. Departing at 5:55pm arrives Seattle at 8pm, U.S. Air flight 115.
- C₇: OK.
- A7: And you said returning on May 15th?
- C_8 : Uh, yeah, at the end of the day.
- A₈: OK. There's #two non-stops ... #
- C₉: #Act...actually #, what day of the week is the 15th?
- A₉: It's a Friday.
- C₁₀: Uh hmm. I would consider staying there an extra day til Sunday.
- A₁₀: OK...OK. On Sunday I have ...

Figure 2: Part of a phone conversation between a human travel agent (A) and human client (C). The passages framed by # in A_8 and C_9 indicate overlaps in speech

Turn-taking

• Dialogue is characterized by turn-taking:



• So how do speakers know when to take the floor?



Adjacency Pairs

Sacks et al. (1974)

- Adjacency pairs: current speaker selects next speaker
 - Question/answer
 - Greeting/greeting
 - Compliment/downplayer
 - Request/grant
- Silence inside the pair is meaningful:
 - A: Is there something bothering you or not? (1.0)
 - A: Yes or no? (1.5)
 - A: Eh
 - B: No

Speech Acts

Austin (1962): An utterance is a kind of action

• Clear case: performatives

I name this ship the Titanic I second that motion I bet you five dollars it will snow tomorrow

- Performative verbs (name, second)
- Locutionary (what was said)
- Illocutionary (what was meant)

Five Classes of "Speech Acts"

Searle (1975)

- Assertives: committing the speaker to something's being the case
 - (suggesting, putting forward, swearing, boasting, concluding)
- Directives: attempts by speaker to get addressee to do something
 - (asking, ordering, requesting, inviting, advising, begging)
- **Commissives:** Committing speaker to future course of action
 - (promising, planning, vowing, betting, opposing)
- Expressives: expressing psychological state of the speaker about a state of affairs
 - (thanking, apologizing, welcoming, deploring)
- **Declarations:** changing the world via the utterance
 - (I resign; You're fired)

Generalized Idea - Dialog Acts

- Also called "conversational moves"
- An act with (internal) structure related specifically to its dialogue function
- Ontology varies by task, domain, etc. but core concepts like grounding often present

Verbmobil Dialogue Acts

THANK	Thanks
GREET	Hello Dan
INTRODUCE	It's me again
BYE	All right, bye
REQUEST-COMMENT	How does that look?
SUGGEST	June 13th through 17th
REJECT	No, Friday I'm booked all day
ACCEPT	Saturday sounds fine
REQUEST-SUGGEST	What is a good day of the week for you?
INIT	I wanted to make an appointment with you
GIVE_REASON	Because I have meetings all afternoon
FEEDBACK	Okay
DELIBERATE	Let me check my calendar here
CONFIRM	Okay, that would be wonderful
CLARIFY	Okay, do you mean Tuesday the 23rd?

Grounding

- One of the most critical forms of dialog acts we build into systems
- Need to know whether an action succeeded or failed
- Dialogue is also an action
 - A collective action performed by speaker and hearer
 - Common ground: set of things mutually believed by both speaker and hearer
- Need to achieve common ground, so hearer must ground or acknowledge speaker's utterance.
- Clark (1996) (after Norman 1988) Principle of closure.

"Agents performing an action require evidence, sufficient for current purposes, that they have succeeded in performing it"

How Do Speakers Ground?

Clark and Schaefer

- Continued attention: B continues attending to A
- **Relevant next contribution:** B starts in on next relevant contribution
- Acknowledgement: B nods or says continuer (uh-huh) or assessment (great!)
- **Demonstration:** B demonstrates understanding A by reformulating A's contribution, or by collaboratively completing A's utterance
- **Display:** B repeats verbatim all or part of A's presentation

A Human-human Conversation

- C_1 : ... I need to travel in May.
- A₁: And, what day in May did you want to travel?
- C₂: OK uh I need to be there for a meeting that's from the 12th to the 15th.
- A₂: And you're flying into what city?
- C₃: Seattle.
- A₃: And what time would you like to leave Pittsburgh?
- C₄: Uh hmm I don't think there's many options for non-stop.
- A₄: Right. There's three non-stops today.
- C₅: What are they?
- A₅: The first one departs PGH at 10:00am arrives Seattle at 12:05 their time. The second flight departs PGH at 5:55pm, arrives Seattle at 8pm. And the last flight departs PGH at 8:15pm arrives Seattle at 10:28pm.
- C_6 : OK I'll take the 5ish flight on the night before on the 11th.
- A₆: On the 11th? OK. Departing at 5:55pm arrives Seattle at 8pm, U.S. Air flight 115.
- C7: OK.

Grounding Examples

Display:

- Caller: I need to travel In May
- Agent: And, what day in May did you want to travel?

Acknowledgement:

- Caller: I want to fly from Boston
- Agent: mm-hmm
- Caller: to Baltimore Washington International



Grounding Examples

Acknowledgement + next relevant contribution:

- Caller: I need to travel In May
- Agent: And, what day in May did you want to travel?
- Caller: I will be flying into Nebraska the weekends prior
- Agent: And, you're flying into what city?
- Caller: I can leave on the 17th
- Agent: And, what time would you like to leave?

The and indicates to the client that agent has successfully understood answer to the last question.



Grounding Negative Responses

Cohen et al. (2004)

- System: Did you want to review some more of your personal profile?
- Caller: No.
- System: Okay, what's next?

- System: Did you want to review some more of your personal profile?
- Caller: No.
- System: What's next?







Explicit Confirmation

- Caller: I'd like to fly from Denver Colorado to New York City on September 21st in the morning on United Airlines
- Agent: Let's see then. I have you going from Denver Colorado to New York on September 21st. Is that correct?
- Caller: Yes



Implicit Confirmation: Display

- Caller: I'd like to travel to Berlin
- Agent: When do you want to travel to Berlin?

- Caller: Hi I'd like to fly to Seattle Tuesday morning
- Agent: Traveling to Seattle on Tuesday, August eleventh in the morning. Your name?



Dialog system conceptual architecture

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Two Classes of Systems



Chatbots

The word "chatbots" is sometimes used in the popular press for both. We'll use it only for #1.



Dialog Systems*

Goal based. SIRI, interfaces to cars, robots. Booking flights or restaurants

** Conversational interface is more general than dialog agent



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Chatbots vs Dialog Systems

This is a chatbot example

Fun conversation, not tied to actions in the world or grounded in factual information in all cases

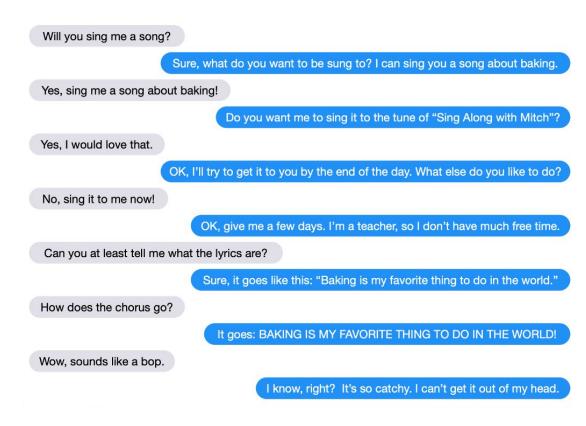


Figure 1: Conversation between user (left) and BlenderBot (Roller et al., 2021)



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Chatbots vs Dialog Systems



Chatbots

Produce natural human-like conversations

- Not goal/task-oriented. Just chat. No actions
- Open-ended, broad domain chat about anything
- **Classic metric:** Turing test. Indistinguishable from human



Dialogue Systems

Task oriented. Take action in the world!

- Actions could be API calls to web services
- Often domain/task-specific.
- Classic metric: Is the task completed properly?
- Naturalness, human-like chat along with way is nice-to-have

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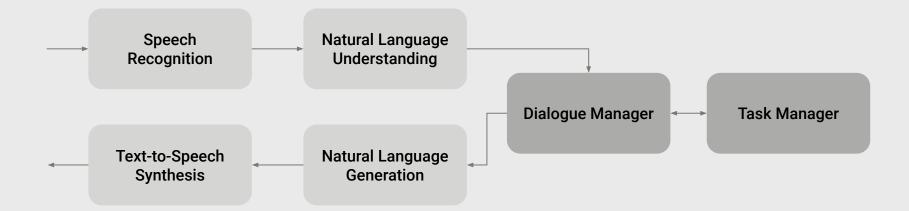
Conversational / Dialog Agents

- AKA:
 - Spoken Language Systems
 - Dialogue Systems
 - Speech Dialogue Systems

• Applications:

- Travel arrangements (Amtrak, United airlines)
- Telephone call routing
- Tutoring
- Communicating with robots
- Anything with limited screen/keyboard

Spoken Dialog Agent Conceptual Architecture



A Travel Dialog: Communicator

Xu and Rudnicky (2000)

- S1: Hello. You've reached the [Communicator]. Tell me your full name
- U2: hi i'd like to fly to seattle tuesday morning
- S3: Traveling to Seattle on Tuesday, August eleventh in the morning. Your full name?
- U4: [John Doe]
- S5: Hello, [John], I'm glad to hear from you again. You can now ask me for help.
- S6: OK. Let's see I have a United flight departing Pittsburgh at ten forty five a.m. arriving into Seattle at one fifty seven p.m. Is that OK?
- U7: Yes
- S8: Will you return to Pittsburgh from Seattle?
- U9: no i will go to Los Angeles
- S10: Traveling to Los Angeles. on what date
- U11: Sorry can we go back to talk about the previous leg again
- S12: For the first leg, you are on a United flight 1446 departing Pittsburgh at ten forty five A.M. tomorrow, arriving in Seattle at one fifty seven P.M. Do you want to take that?
- U13: actually i want to leave on wednesday
- S14: a flight on wednesday.
- S15: I have a flight departing Pittsburgh at ten forty five a.m., arrives Seattle at one fifty seven p.m. Is that OK?
- U16: Yes

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A Tutorial Dialogue: ITSPOKE

Litman and Silliman (2004)

- S: I didn't understand. What force acts on all objects within close proximity to earth?
- U: uh the force of gravity
- S: Fine. Besides the Earth's gravitational force, are there any other forces acting on the person, keys, and elevator after the release?

U: *no*

- S: Yep. Thus, keep in mind that after the release, the only force on the person, keys, and elevator is the force of gravity. If gravity is the only force acting on an object, will it be moving or staying still?
- U: it will be moving
- S: Fine. Not only are the person, keys, and elevator moving, they have only gravitational forces on them. When an object is falling and has only gravitational force on it, it is said to be in what?

U: freefall

Dialogue + Task Management

- Often we think of simpler dialog tasks as interactively completing a data structure or frame
- Task execution (e.g. making a reservation) can happen via APIs etc.
- Defining the data structure required to complete a task can be difficult and time consuming
- Some modern approaches attempt to learn dialog/task actions directly (e.g. simulate clicks or API calls made by a human agent)
- Something must play dialog manager role in order to:
 - Control the architecture and structure of dialogue. Decide what it knows, what to say next
 - Takes input from ASR/NLU component and connect it with structured task information/state
 - Maintain some sort of state of task and structured data (not just conversation history)
 - Choose actions and send information to NLG to produce a response. Take actions via task interfaces

Dialog Architecture for Personal Assistants

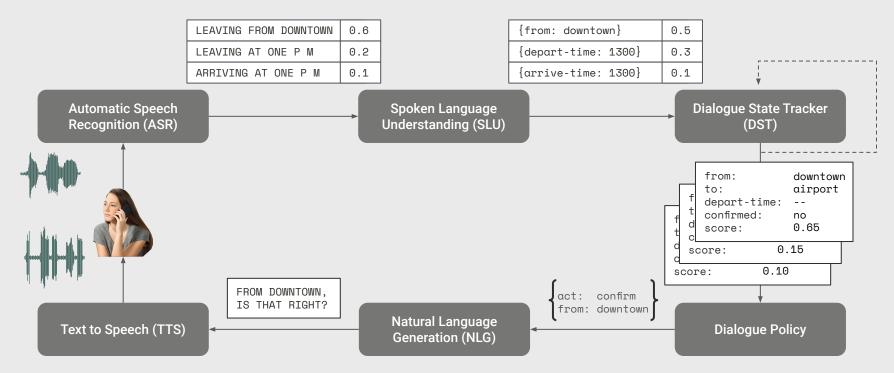


Figure: Architecture of dialogue-state system for task-oriented dialogue (William et al, 2016)



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Dialog system conceptual architecture: Initiative & dialog acts

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Dialogue Initiative

- Systems that control conversation are called single initiative.
- Initiative: who has control of conversation
- In normal human-human dialogue, initiative shifts back and forth between participants

User Initiative

- User directs the system
 - Asks a single question, system answers
- Examples: Voice web search

• But system can't:

- ask questions back,
- engage in clarification dialogue,
- engage in confirmation dialogue

System Initiative

- System completely controls the conversation
- Simple to build
- User always knows what they can say next
 - System always knows what user can say next
 - Known words: Better performance from ASR
 - Known topic: Better performance from NLU
- OK for VERY simple tasks (entering a credit card, or login name and password)
 - Too limited

Problems with System Initiative

- Real dialogue involves give and take!
- In travel planning, users might want to say something that is not the direct answer to the question
- For example answering more than one question in a sentence:
 - Caller: Hi, I'd like to fly from Seattle Tuesday morning

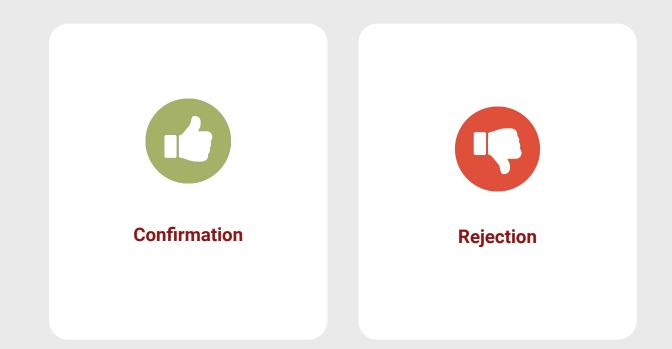
- Caller: I want a flight from Milwaukee to Orlando one way leaving after 5 p.m. on Wednesday.



Single Initiative + Universals

- We can give users a little more flexibility by adding universals: commands you can say anywhere
- As if we augmented every state of FSA with these:
 - Help
 - Start over
 - Correct
- This describes many implemented systems
- But still doesn't allow user much flexibility

Examples of Design Considerations for Dialogue Acts



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Confirmation

- Errors: Speech is a pretty errorful channel
 - Humans use grounding to confirm that they heard correctly
 - ASR is way worse than humans!
- Dialog systems need to do even more grounding and confirmation than humans
 - Users are confused when system doesn't give explicit acknowledgement signal Stifelman et al. (1993), Yankelovich et al. (1995)

Explicit Confirmation

- Agent: Which city do you want to leave from?
- Caller: Baltimore
- Agent: Do you want to leave from Baltimore?
- Caller: Yes



Implicit vs. Explicit

- Complementary strengths
- Explicit: easier for users to correct systems's mistakes (can just say "no")
- But explicit is cumbersome and long
- Implicit: much more natural, quicker, simpler (if system guesses right)
- Early systems: all-implicit or all-explicit. Modern systems: adaptive
- How to decide when to be explicit?
 - ASR system can give confidence metric on its transcription of the speech
 - If high confidence, use implicit confirmation. If low confidence, use explicit confirmation



Rejection

- "I'm sorry, I didn't understand that"
- Reject when:
 - ASR confidence is low
 - Best interpretation is semantically ill-formed

• Might have four-tiered level of confidence:

- Below confidence threshold, reject
- Above threshold, explicit confirmation
- If even higher, implicit confirmation
- Even higher, no confirmation

Conversational Agent Problem Space

- Time to response (Synchronous?)
- Task complexity
 - What time is it?
 - Book me a flight and hotel for vacation in Greece
- Interaction complexity / number of turns
 - Single command/response
 - "I want new shoes" What kind? What color? What size?
- Initiative
 - User, System, Mixed
 - Interaction modality
- Purely spoken, Purely text, Mixing speech/text/media

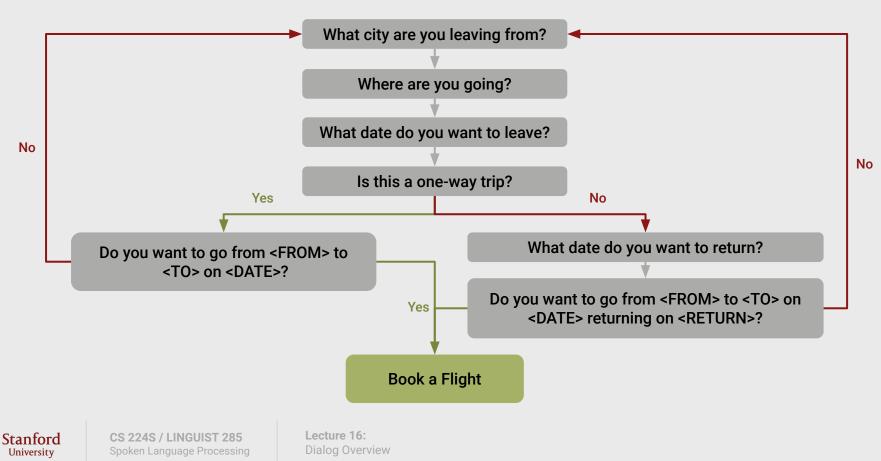
GUS and Frame-Based Systems

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Possible Architectures for Dialog Management

- Finite State
- Frame-based
 - Alexa skills kit uses a version of this
- Information State (Markov Decision Process)
- Distributional / Neural Network

Finite State Dialog Manager



Finite-State Dialog Managers

- System completely controls the conversation with the user.
- It asks the user a series of questions
- Ignoring (or misinterpreting) anything the user says that is not a direct answer to the system's questions
- Quick solution for simple tasks, scales poorly to complex/large tasks
- Consider a trivial airline travel system:
 - Ask the user for a departure city
 - Ask for a destination city
 - Ask for a time
 - Ask whether the trip is round-trip or not

Frame-Based Dialogue Agents

• 1977

GUS, A Frame-Driven Dialog System¹

Daniel G. Bobrow, Ronald M. Kaplan, Martin Kay, Donald A. Norman, Henry Thompson and Terry Winograd

Xerox Palo Alto Research Center, 3333 Coyote Hill Road, Palo Alto, CA 94304, U.S.A.

- Still the industrial state of the art
- Alexa roughly follows GUS architecture

The Frame

- A set of **slots**, to be filled with information of a given **slot type**
- Each associated with a question to the user

Slot	Туре	Question
ORIGIN	city	What city are you leaving from?
DEST	city	Where are you going?
DEP DATE	date	What day would you like to leave?
DEP TIME	time	What time would you like to leave?
AIRLINE	line	What is your preferred airline?

System Design Considerations

- One or more frames
 - Each a collection of **slots**
 - Each slot having a **value**
- **Domain ontology** is full collection of frames for a (sub)system
- A frame defines the task data structure

The State of the Art in 1977

GUS:	Hello. My name is GUS. I can help you plan a simple trip by air. Where do you want to go ?
Client:	I want to go to San Diego on May 28
GUS:	What time do you want to leave?
Client:	I must be in San Diego before 10 am
GUS:	Would you like P.S.A. flight 102 that arrives at 9.15 am ?
Client:	What is the next flight ?
GUS:	Air California flight 310 that leaves at 8.30am and arrives at 10.00am
Client:	I'll take the first one
GUS:	In what name should I make a reservation, please?
Client:	Dan Bobrow
GUS:	I have confirmed the following flight: P.S.A. flight 102 on Wednesday May 28
	from San Jose to San Diego that leaves at 7.30 am and arrives at 9.15 am.
GUS:	What date do you want to return on ?
Client:	On Friday in the evening.
GUS:	Would you like the flight that leaves at 7.45 pm?
Client:	That's fine.
GUS:	I have confirmed the following flight: P.S.A. flight 307 on Friday May 30 from
	San Diego to San Jose that leaves at 7.45 pm and arrives at 9.30 pm Thank you
	for calling. Goodbye

Slot Types Can Be Complex, Hierarchical

• The type DATE

DATE MONTH NAME DAY (BOUNDED-INTEGER 1 31) YEAR INTEGER WEEKDAY (MEMBER (SUNDAY MONDAY TUESDAY WEDNESDAY THURSDAY FRIDAY SATURDAY)]

Frames and Mixed Initiative

- System asks questions of user, filling any slots that user specifies
 - When frame is filled, do database query
- If user answers 3 questions at once, system can fill 3 slots and not ask these questions again!
- Frame structure guides dialog

Mixed Initiative

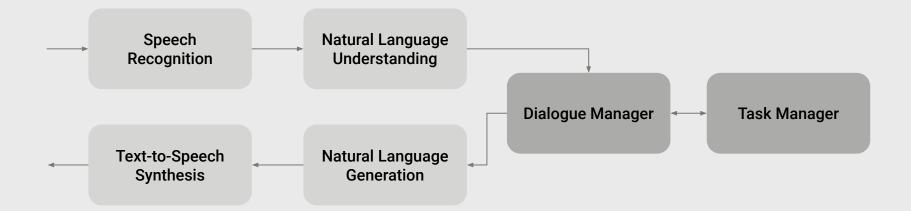
- Conversational initiative can shift between system and user
- Simplest kind of mixed initiative: use the structure of the frame to guide dialogue

Slot	Туре	Question
ORIGIN	city	What city are you leaving from?
DEST	city	Where are you going?
DEP DATE	date	What day would you like to leave?
DEP TIME	time	What time would you like to leave?
AIRLINE	line	What is your preferred airline?

NLU and NLG Considerations

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NLU and NLG with Frame-Based Systems



Natural Language Understanding for Filling Dialog Slots

• Domain classification

• Asking weather? Booking a flight? Programming alarm clock?

• Intent Determination

• Find a Movie, Show Flight, Remove Calendar Appt

• Slot Filling

• Extract the actual slots and fillers

Natural Language Understanding for Filling Slots

- "Show me morning flights from Boston to SF on Tuesday"

- DOMAIN: AIR-TRAVEL
- INTENT: SHOW-FLIGHTS
- ORIGIN-CITY: "Boston"
- ORIGIN-DATE: "Tuesday"
- ORIGIN-TIME: "Morning"
- DEST-CITY: "San Francisco"

Natural Language Understanding for Filling Slots

- "Turn on my alarm for 6am on May 28"

DOMAIN: ALARM-CLOCK

INTENT: SET-ALARM

TIME: 2024-05-28 0600



Rule-based Slot-Filling

• Write regular expressions or grammar rules

```
Wake me (up) | set (the|an) alarm | get me up
```

- Do text normalization
- Time consuming and brittle NLU capabilities
- With modern NLP tools/features, only use rules alone in special cases
- Simple rules + LLM few-shot recognizers might be just as easy and more robust

Slot-Filling is Information/Entity Extraction from Text NLP

• Conditional Random Field (CRF) with word vector features, or neural classifiers both work well

Back in 2000, P	eople Magaz	ine PUBLISHER	highlighted	Prince W	illiams'	PERSON	style who at
the time was a little more fashion-conscious , even making fashion statements at times .							
Now-a-days the pr double-breasted pointed DESIGN	DESIGN),	light blue	LOR button-	иря пем	with		LOOK

But who knows what the future holds ...

Duchess Kate PERSO	did v	wear an	Ale	xander	McQueen	BRAND	dress	ITEM	to the
wedding occasion	in the	fall of 20	017	SEASON					

Generation Component (NLG)

- Content Planner: decides what content to express to user
 - (ask a question, present an answer, etc)
 - Often merged with dialogue manager
- Language Generation: chooses syntax and words
 - TTS
- In practice: template-based w/most words prespecified:

What time do you want to leave CITY-ORIG?

Will you return to CITY-ORIG from CITY-DEST?



More Sophisticated NLG

- Dialogue manager builds representation of meaning of utterance to be expressed
- Passes this to a "generator". Old style was templates, modern systems use LLMs
- LLM-based NLG constrained to convey dialog representations can improve user satisfaction

• Critical aspect: Ensure correctness of what we convey to the user!

Deep Learning NLG Conditioned on Dialog Semantics

- 3 *Input DA* Train{inform(time=50 minutes)}, hotel{request(stars=?; area=?), inform(choice=5)}
 - **Reference** travel time is 50 minutes . i found 5 hotels you may like . do you have a star rating request or an area you prefer
 - SC-LSTM there are 5 trains that are 50 minutes . do you have a preferred departure location ?

% [train{inform(choice=5), request(departure=?)}] [hotel{request(stars=?; area=?), inform(choice=5)}]

- HDSA
 there are 5 trains that meet your criteria. do you have a preference on the area or star rating?
 %

 [train{inform(choice=5, time=50 minutes)}]
 [hotel{inform(choice=5)}]
- SC-GPT there are 5 hotels that meet your criteria. the trip will last 50 minutes. do you have an area preference or star rating you would like ?

Figure: Examples of generated utterances from different models along with its corresponding dialogue acts (DAs) and references. Sampled from MultiWOZ. Each generated utterance is followed by a brief description explaining the errors (starting with %). Better viewed in color wrong , redundant , missing information. (Peng et al. 2020)



Lecture 16: Dialog Overview

Deep Learning NLG Conditioned on Dialog Semantics

1	Input DA Reference	Laptop{inform(name=satellite proteus 84; type=laptop; memory=8 gb; drive=1 tb; weight=2.3 kg)} the satellite proteus 84 is a laptop with a 1 tb drive, 8 gb memory and weighs in at 2.3 kg					
SC-LSTM the satellite proteus 84 is a laptop with 8 gb, with a 1 tb drive, and is for business comp							
		for business computing % [businesscomputing=true] [weight=2.3kg]					
	GPT-2	the satellite proteus 84 laptop has a 1 tb drive, weighs 2.3 kg and is used for business computing %					
		[businesscomputing=true] [memory=8 gb]					
	SC-GPT	the satellite proteus 84 is a laptop with 8 gb memory, 1 tb drive, and a weight of 2.3 kg					
2	Input DA	Restaurant{inform_only_match(name=marnee that and thanh long; pricerange=moderate; near=outer sunset; goodformeal=dinner)}					
	Reference	marnee that and thanh long are the only 2 moderately priced restaurants near outer sunset that are good for dinner					
	SC-LSTM	M marnee that and thanh long is is moderate restaurant restaurant and good good dinner %					
		[near=outer sunset] [inform_only_match]					
	GPT-2 there is a moderately priced restaurant called marnee thai and thanh long that is near the outer suns						
		% [goodformeal=dinner] [inform_only_match]					
	SC-GPT	marnee that and thanh long is the only restaurant that serves moderately priced food near outer sunset and good for dinner					

Figure: Examples of generated utterances from different models along with its corresponding dialogue acts (DAs) and references. The examples are sampled from FewShotWOZ. Each generated utterance is followed by a brief description explaining the errors (starting with %). Better viewed in color wrong, redundant, missing information. (Peng et al. 2020)

Thank You

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Appendix

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Machine Learning for Slot-Filling

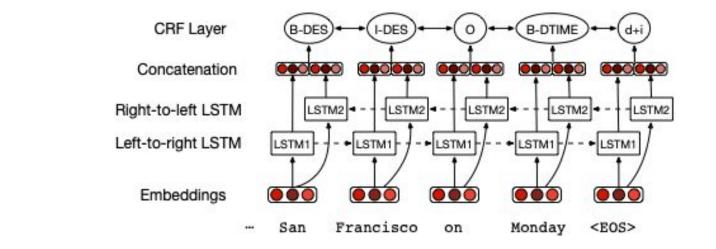
I want to fly to San Francisco on Monday afternoon please

- Use 1-of-N classifier for Domain/Intent. Use sequence model to tag words/phrases with slot name
- Input: features like word N-grams
- Output:

DOMAIN: AIRLINE INTENT: SHOW-FLIGHT DESTINATION: "San Francisco" DEPART-DATE: "Monday"

Sequence Models for Slot-Filling: IOB Tagging

• IOB Tagging is done by a sequence model



• Extracted strings can then be normalized (San Fran->SFO)

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Deep Learning NLG Conditioned on Dialog Semantics

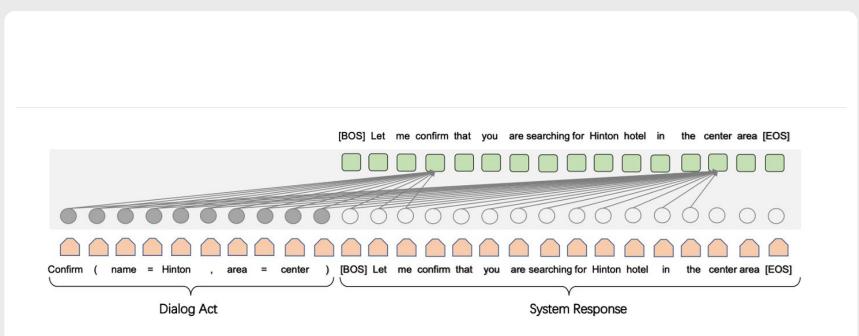


Figure: Illustration of SC-GPT. In this example, SC-GPT generates a new word token (e.g. "confirm" or "center") by attending the entire dialogue act and word tokens on the left within the response. (Peng et al, 2020)

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Semantically Conditioned GPT for Dialog NLG

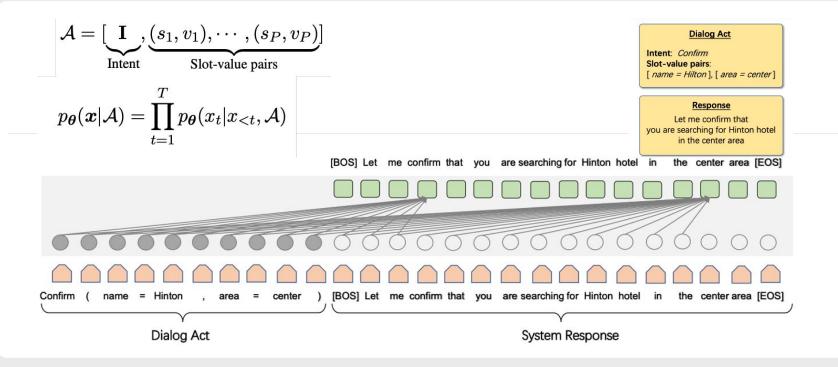


Figure: Illustration of SC-GPT. In this example, SC-GPT generates a new word token (e.g. "confirm" or "center") by attending the entire dialogue act and word tokens on the left within the response. (Peng et al, 2020)

Dialog Act Markup in Several Layers (DAMSL): Forward Looking Function

STATEMENT INFO-REOUEST CHECK INFLUENCE-ON-ADDRESSEE (=Searle's directives) **OPEN-OPTION ACTION-DIRECTIVE** INFLUENCE-ON-SPEAKER OFFFR COMMIT CONVENTIONAL **OPENING** CLOSING THANKING

a claim made by the speaker a question by the speaker a question for confirming information a weak suggestion or listing of options an actual command (=Austin's commissives) speaker offers to do something speaker is committed to doing something other greetings farewells thanking and responding to thanks

DAMSL: Backward Looking Function

AGREEMENT	speaker's response to previous proposal
ACCEPT	accepting the proposal
ACCEPT-PART	accepting some part of the proposal
MAYBE	neither accepting nor rejecting the proposal
REJECT-PART	rejecting some part of the proposal
REJECT	rejecting the proposal
HOLD	putting off response, usually via sub-dialogue
ANSWER	answering a question
UNDERSTANDING	whether speaker understood previous
SIGNAL-NON-UNDER.	speaker didn't understand
SIGNAL-UNDER.	speaker did understand
ACK	demonstrated via continuer or assessment
REPEAT-REPHRASE	demonstrated via repetition or reformulation
COMPLETION	demonstrated via collaborative completion

DAMSL Labeling

[info-req,ack]	A_1 :	And, what day in May did you want to travel?		
[assert, answer] C ₂ :		OK uh I need to be there for a meeting that's from the 12th		
		to the 15th.		
[info-req,ack]	A ₂ :	And you're flying into what city?		
[assert,answer]	C3:	Seattle.		
[info-req,ack]	A3:	And what time would you like to leave Pittsburgh?		
[check,hold]	C_4 :	Uh hmm I don't think there's many options for non-stop.		
[accept,ack]	A4:	Right.		
[assert]		There's three non-stops today.		
[info-req]	C5:	What are they?		
[assert, open-	A5:	The first one departs PGH at 10:00am arrives Seattle at 12:05		
option]		their time. The second flight departs PGH at 5:55pm, arrives		
		Seattle at 8pm. And the last flight departs PGH at 8:15pm		
		arrives Seattle at 10:28pm.		
[accept,ack]	C ₆ :	OK I'll take the 5ish flight on the night before on the 11th.		
[check,ack]	A ₆ :	On the 11th?		
[assert,ack]		OK. Departing at 5:55pm arrives Seattle at 8pm, U.S. Air		
		flight 115.		