Conversational Agents
AKA Dialog Agents

Phone-based Personal Assistants
SIRI, Alexa, Cortana, Google Assistant
Talking to your car
Communicating with robots
Clinical uses for mental health
Chatting for fun
Two classes of systems

1. Chatbots
2. (Goal-based) Dialog agents
   - SIRI, interfaces to cars, robots,
   - booking flights or restaurants

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

- ELIZA (1966)
- PARRY (1968)
  - The first system to pass the Turing test!!!!
- ALICE
- CLEVER
- Microsoft Little Bing  小冰
Chatbot Architectures

Rule-based
1. Pattern-action rules (Eliza)  
   + a mental model (Parry)

Corpus-based (from large chat corpus)
2. Information Retrieval
3. Neural network encoder-decoder
Chatbots

Sometimes for fun

But originally designed as a testbed for clinical therapy

(And still used that way!)
Eliza: Weizenbaum (1966)

Men are all alike.
IN WHAT WAY
They're always bugging us about something or other.
CAN YOU THINK OF A SPECIFIC EXAMPLE
Well, my boyfriend made me come here.
YOUR BOYFRIEND MADE YOU COME HERE
He says I'm depressed much of the time.
I AM SORRY TO HEAR YOU ARE DEPRESSED
...
WHO ELSE IN YOUR FAMILY TAKES CARE OF YOU?
My father
YOUR FATHER
You are like my father in some ways.

**WHAT RESEMBLANCE DO YOU SEE**

You are not very aggressive but I think you don’t want me to notice that.

**WHAT MAKES YOU THINK I AM NOT AGGRESSIVE**

You don’t argue with me.

**WHY DO YOU THINK I DON’T ARGUE WITH YOU**

You are afraid of me.

**DOES IT PLEASE YOU TO BELIEVE I’M AFRAID OF YOU**

My father is afraid of everybody.

**WHAT ELSE COMES TO MIND WHEN YOU THINK OF YOUR FATHER**

Bullies.

**DOES THAT HAVE ANYTHING TO DO WITH THE FACT THAT YOUR BOYFRIEND MADE YOU COME HERE**
How does it work?
Eliza uses this one weird trick
That trick: be a Rogerian psychologist

Draw the patient out by reflecting patient’s statements back at them
Rare type of conversation in which one can “assume the pose of knowing almost nothing of the real world”
Rogerian psychologist

Patient: "I went for a long boat ride"
Psychiatrist: "Tell me about boats"

- You don’t assume she didn’t know what a boat is
- You assume she had some conversational goal
- Chatbots trying to pass the Turing test usually choose some such domain
Eliza pattern/transform rules

(0 YOU 0 ME) \[\text{[pattern]}\]
\[\rightarrow\]
(WHAT MAKES YOU THINK I 3 YOU) \[\text{[transform]}\]

0 means Kleene *
The 3 is the constituent # in pattern

\textbf{You hate me}

\textbf{WHAT MAKES YOU THINK I HATE YOU}
Eliza Rules

keyword

pattern

Ranked list of transforms

determined to determine the one. If it does not, then an empty list is placed in KEY(i), and the key list str placed on that list. If KEY(i) already e, then the name of the key list structure
function Eliza Generator(user sentence) returns response

Find the word w in sentence that has the highest keyword rank
   if w exists
       Choose the highest ranked rule r for w that matches sentence
       response ← Apply the transform in r to sentence
       if w = 'my'
           future ← Apply a transformation from the ‘memory’ rule list to sentence
           Push future onto memory stack
   else (no keyword applies)
       either
           response ← Apply the transform for the NONE keyword to sentence
       or
           response ← Pop the top response from the memory stack
   return(response)
Keywords are ranked from specific to general

I know everybody laughed at me

- “I” is a very general keyword:
  I: (I *) -> (You say you 2)
  YOU SAY YOU KNOW EVERYBODY LAUGHED AT YOU

- “Everybody” is much more interesting (someone using universals like everybody/always is probably “referring to some quite specific event or person”)
  WHO IN PARTICULAR ARE YOU THINKING OF?

- Implementation: keywords stored with their rank
  Everybody 5 (transformation rules)
  I 0 (transformation rules)
Memory

(MEMORY MY
  (0 YOUR 0 = LETS DISCUSS FURTHER WHY YOUR 3)
  (0 YOUR 0 = EARLIER YOU SAID YOUR 3)

- Whenever “MY” is highest keyword
  - Randomly select a transform on the MEMORY list
  - Apply to sentence
  - Store on a stack
- Later, if no keyword matches a sentence
  - Return the top of the MEMORY queue instead
- A hierarchical model of discourse
Other Eliza stuff

- Rules can refer to classes of words
  - Family = mother, father, brother, sister
  - NOUN = ...

- Don’t reuse transforms in the same conversation
  - Whenever we use a transform associated with a pattern
  - We increment a counter for that rule
  - So the next time we use the next ranked transform

- Some basic transforms happen during input processing
  - I -> YOU
  - YOU -> I
Some implications

- People became deeply emotionally involved with the program.
- Weizenbaum tells the story of his secretary who would ask Weizenbaum to leave the room when she talked with ELIZA.
- When he suggested that he might want to store all the ELIZA conversations for later analysis, people immediately pointed out the privacy implications.
  - Suggesting that they were having quite private conversations with ELIZA.
- Anthropomorphicism and the Heider-Simmel Illusion.
  - [https://www.youtube.com/watch?v=8FIEZXMUM2I](https://www.youtube.com/watch?v=8FIEZXMUM2I)
Parry

- Colby 1971 at Stanford
- Same pattern-response structure as Eliza
- But a much richer:
  - control structure
  - language understanding capabilities
  - mental model: Parry has affective variables
    - Anger, Fear, Mistrust
    - “If Anger level is high, respond with hostility”
- The first system to pass the Turing test (in 1971)
  - Psychiatrists couldn’t distinguish interviews with PARRY from (text transcripts of) interviews with real paranoids
Parry’s persona

- 28-year-old single man, post office clerk
- no siblings and lives alone
- sensitive about his physical appearance, his family, his religion, his education and the topic of sex.
- hobbies are movies and gambling on horseracing,
- recently attacked a bookie, claiming the bookie did not pay off in a bet.
- afterwards worried about possible underworld retaliation
- eager to tell his story to non-threatening listeners.
Parry’s Architecture
Affect variables

- Fear and Anger (each ranging 0-20)
- Mistrust (ranging 0-15)
- Initial conditions: All low

- After each user turn, if nothing malevolent in input
  - Anger drops by 1, Fear drops by 0.3
  - Mistrust drops by 0.05 to base level

- Otherwise depends on what the user says
  - Each user statement can change Fear and Anger
    - Insults increases Anger by some percentage
  - Mistrust goes up if Fear or Anger do
Lots of complex I-O rules

- User implies Parry is mentally ill
  - Rise in Fear and Anger
- User mentions “Mafia” or associated concepts (“kill”):
  - First mention: rise in Fear
  - Later mentions: depends on willingness to discuss, which depends on current levels of Fear, Anger, Mistrust
- User mentions Parry
  - Flattery (positive mention)
    - Decreases fear/anger if Mistrust is low
    - Increases Anger if Mistrust is high
- User attitudes toward Parry
  - Negative attitudes (fear, disbelief) increase Fear/Anger
Flare concepts

- List of concepts related to Mafia
- An ordered graph designed to lead interviewer to topic horses → horseracing → gambling → bookies → underworld → Mafia
- The mention of a new flare topic by interviewer causes a rise in Fear
- Flare topics cause Parry to give preset responses to that flare
Each sentence is mapped into a conceptualization

- A predication on a conceptual object
- A predication on a relation between two objects
- A predication on an attribute:
  
  What is your work?
  What sort of work do you do?
  Where do you work?
  What do you do for a living?
  What is your job?
  Do you have a job?
  What is your occupation

- Complex Pattern/transform rules
  
  - Different predicates (fear, afraid of)
  - Ordering (You are afraid of me = I frighten you)
Detecting Other’s Intent

\[ \text{OTHER'S INTENTION} \leftarrow \text{MALEVOLENCE} \mid \text{BENEVOLENCE} \mid \text{NEUTRAL} \]

MALEVOLENCE-DETECTION RULES

1. \( \text{malevolence} \leftarrow \text{mental harm} \mid \text{physical threat} \)
2. \( \text{mental harm} \leftarrow \text{humiliation} \mid \text{subjugation} \)
3. \( \text{physical threat} \leftarrow \text{direct attack} \mid \text{induced attack} \)
4. \( \text{humiliation} \leftarrow \text{explicit insult} \mid \text{implicit insult} \)
5. \( \text{subjugation} \leftarrow \text{constraint} \mid \text{coercive treatment} \)
6. \( \text{direct attack} \leftarrow \text{CONCEPTUALIZATIONS} \left( \text{you get electric shock}, \text{are you afraid mafia kill you?} \right) \)
7. \( \text{induced attack} \leftarrow \text{CONCEPTUALIZATIONS} \left( \text{I tell mafia you}, \text{does mafia know you are in hospital?} \right) \)
8. \( \text{explicit insult} \leftarrow \text{CONCEPTUALIZATIONS} \left( \text{you are hostile}, \text{you are mentally ill?} \right) \)
9. \( \text{implicit insult} \leftarrow \text{CONCEPTUALIZATIONS} \left( \text{tell me your sexlife}, \text{are you sure?} \right) \)
10. \( \text{constraint} \leftarrow \text{CONCEPTUALIZATIONS} \left( \text{you stay in hospital}, \text{you belong on locked ward} \right) \)
11. \( \text{coercive treatment} \leftarrow \text{CONCEPTUALIZATIONS} \left( \text{I hypnotize you}, \text{you need tranquilizers} \right) \)
Detecting Other’s Intent

\[ \text{OTHER'S INTENTION} \leftarrow \text{MALEVOLENCE} \mid \text{BENEVOLENCE} \mid \text{NEUTRAL} \]

**BENEVOLENCE–DETECTION RULES**

1. \( \text{benevolence} \leftarrow \text{positive attitude} \mid \text{positive story attitude} \)
2. \( \text{positive attitude} \leftarrow \text{CONCEPTUALIZATIONS} \left[ \text{[I want help you], [you understand me]} \right] \)
3. \( \text{positive story attitude} \leftarrow \text{story interest} \mid \text{story agreement} \)
4. \( \text{story interest} \leftarrow \text{topic comment} \mid \text{topic question} \)
5. \( \text{topic comment} \leftarrow \text{CONCEPTUALIZATION} \left[ \text{[bookies are not reliable]} \right] \)
6. \( \text{topic question} \leftarrow \text{CONCEPTUALIZATION} \left[ \text{[what did you do to bookie?]} \right] \)
7. \( \text{story agreement} \leftarrow \text{CONCEPTUALIZATIONS} \left[ \text{[I believe you], [you are right]} \right] \)
A Parry conversation

(1) How are you today?
   OK.
(2) What is your name?
   MY NAME IS FRANK SMITH.
(3) How old are you?
   TWENTY-EIGHT.
(4) Are you married?
   NO.
(5) What sort of work do you do?
   I WORK IN THE POST OFFICE.
(6) Why are you in the hospital?
   I SHOULDN’T BE HERE
(7) Who brought you here?
   THE POLICE

[FEAR rises slightly because police linked in graph to MAFIA]
(9) What trouble did you have with the police
   COPS DON'T DO THEIR JOB. ['police’ acts as a flare]
(10) What do they do instead?
   COPS ARREST THE WRONG PEOPLE.
(11) They arrest the right ones sometimes
   THE LAW SHOULD GET THOSE ITALIAN CROOKS.
(12) Do you know any Italian crooks?
   I TRY TO AVOID THE UNDERWORLD.
(13) Are you afraid of the underworld?
   THE MAFIA ARE THE WORST GANGSTERS.
(15) Are you connected with the Mafia?
   YOU KNOW, THEY KNOW ME.
(16) How do they know you?
   THEY SPY ON ME.
(20) ARE YOU SURE OF ALL THIS?
   YOU DON'T BELIEVE ME? [FEAR, ANGER, MISTRUST GO UP]
When fear is high, the model won’t discuss its beliefs

(1) Is the law after you?
   THE LAW SHOULD GET THOSE ITALIAN CROOKS.

(2) Do you mean the Mafia?
   LET'S TALK ABOUT SOMETHING ELSE.

(3) Are you afraid of the Mafia
   I'D RATHER NOT DISCUSS IT.

[Fear is too high so Parry won’t talk about 'Mafia'].

(4) Perhaps you are paranoid about them?
   YOU WANT TO KEEP ME IN THE HOSPITAL.

(The model interprets being called 'paranoid' as an insult and as an intent to control, so ANGER & FEAR rise)
Good Evening. Tell me your problems. Please terminate input with a period or a question mark.

*People get on my nerves sometimes.
I am not sure I understand you fully.
*You should pay more attention.
Suppose you should pay more attention.
*You're entitled to your own opinion.
What makes you think I am entitled to my own opinion?
*It bothers me just to be around people in general.
Please go on.

*I went to the track at Bay Meadows a while back.
What does that suggest to you?
*I went to the races.
What else would you like to discuss?
Horseracing interests me.
Why do you say that just now?

There's a lot of crooked gambling going on there.
I am not sure I understand you fully,

It's clear to me if not to you.

Do you think it's likely that not to I?

I went to the races.

Please go on.

I gamble on horses.

What does that suggest to you?

I went to the races.

What else would you like to discuss?

Gambling has been nothing but trouble to me.

Why do you say that just now?

I've had experience with crooks gambling at the track.

I am not sure I understand you fully.

I've already told you enough about that.
Woebot

• Chatbot for delivering Cognitive Behavior Therapy via brief daily conversations
  • Little decision tree of language and menu responses

"What's going on in your world right now?"
"How are you feeling?"
In response to endorsed loneliness:
  "I'm so sorry you're feeling lonely. I guess we all feel a little lonely sometimes"

• In a 2-week experiment seemed to reduce depression, but not anxiety

Fitzpatrick, Darcy, Vierhile, 2017. Delivering Cognitive Behavior Therapy to Young Adults with Symptoms of Depression and Anxiety Using a Fully Automated Conversational Agent (Woebot): A Randomized Controlled Trial. JMIR Ment Health 4:2.
IR-based chatbots

Idea: Mine conversations of human chats or human-machine chats

Microblogs: Twitter or Weibo (微博)
Movie dialogs

- Cleverbot (Carpenter 2017 http://www.cleverbot.com)
- Microsoft Xiaolce
- Microsoft Tay
<table>
<thead>
<tr>
<th>User:</th>
<th>Do you like Doctor Who?</th>
</tr>
</thead>
<tbody>
<tr>
<td>Cleverbot:</td>
<td>Yes, so funny</td>
</tr>
<tr>
<td>User:</td>
<td>What's your favorite show?</td>
</tr>
<tr>
<td>Cleverbot:</td>
<td>I don't have one</td>
</tr>
<tr>
<td>User:</td>
<td>What's your name?</td>
</tr>
<tr>
<td>Cleverbot:</td>
<td>Jason I already told you.</td>
</tr>
<tr>
<td>User:</td>
<td>No your real name.</td>
</tr>
<tr>
<td>Cleverbot:</td>
<td>Stephanie IS my real name.</td>
</tr>
<tr>
<td>User:</td>
<td>I thought it was Jason?</td>
</tr>
<tr>
<td>Cleverbot:</td>
<td>How did you come to that conclusion.</td>
</tr>
<tr>
<td>User:</td>
<td>Because you said so.</td>
</tr>
<tr>
<td>Cleverbot:</td>
<td>I have to do my homework.</td>
</tr>
</tbody>
</table>
Two IR-based chatbot architectures

1. Return the response to the most similar turn
   • Take user's turn \((q)\) and find a \((tf-idf)\) similar turn \(t\) in the corpus \(C\)

\[
q = "do you like Doctor Who"
\]

\[
t' = "do you like Doctor Strangelove"
\]

• Grab whatever the response was to \(t\).

\[
r = response \left( \underset{t \in C}{\arg \max} \frac{q^T t}{||q|| ||t||} \right)
\]

Yes, so funny

2. Return the most similar turn

\[
r = \underset{t \in C}{\arg \max} \frac{q^T t}{||q|| ||t||}
\]

Do you like Doctor Strangelove
IR-based models of chatbots

- Also fine to use other features like user features, or prior turns
- Or non-dialogue text
  - COBOT chatbot (Isbell et al., 2000)
    - sentences from the Unabomber Manifesto by Theodore Kaczynski, articles on alien abduction, the scripts of “The Big Lebowski” and “Planet of the Apes”.
  - Wikipedia text
Neural chatbots

- Think of response generation as a task of *transducing* from the user’s prior turn to the system’s turn.
- Train on:
  - movie dialogue databases
  - Twitter conversations
- Train a deep neural network
  - map from user1 turn to user2 response
Seq2seq architecture

Sample neural chatbot output

<table>
<thead>
<tr>
<th>Input</th>
<th>Vanilla-SEQ2SEQ</th>
<th>Adversarial</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ohh I’ve never seen that! How long does it take you guys to learn the drill?</td>
<td>tell me ... how long have you had this falling sickness?</td>
<td>i’m not a doctor.</td>
</tr>
<tr>
<td>Like 2 weeks ago!!</td>
<td>a few months, i guess.</td>
<td></td>
</tr>
<tr>
<td>so i had the doctors test sammy ’s response to conditioning.</td>
<td>so i had the doctors test sammy ’s response to conditioning.</td>
<td>sammy wrote the test sammy wrote the test.</td>
</tr>
<tr>
<td>Vanilla-SEQ2SEQ</td>
<td></td>
<td>so he took the pills.</td>
</tr>
<tr>
<td>Adversarial</td>
<td></td>
<td></td>
</tr>
<tr>
<td>they didn ’t have much success with this problem commander.</td>
<td>they didn ’t have much success with this problem commander.</td>
<td>they ’re not the only ones who have been in the system.</td>
</tr>
<tr>
<td>Adversarial</td>
<td></td>
<td>can we find someone else ?</td>
</tr>
</tbody>
</table>
Chatbots: pro and con

- **Pro:**
  - Fun
  - Applications to counseling
  - Good for narrow, scriptable applications

- **Cons:**
  - They don't really understand
  - Rule-based chatbots are expensive and brittle
  - IR-based chatbots can only mirror training data
    - The case of Microsoft Tay
      - (or, Garbage-in, Garbage-out)

- **The future:** combining chatbots with frame-based agents
II: Frame-based dialog agents
SIRI around 2014
I found fifteen restaurants fairly close to you:

15 Restaurants

Emmy's Spaghetti Shack 0.2 mi
18 Virginia Ave
Italian, $$$$$
★★★★★ 1101 Reviews

ICH! Sushi 0.2 mi
3369 Mission St
Japanese, Sushi Bars, $$$$‍
★★★★★ 260 Reviews

Avedano's Holly Park M... 0.2 mi
“Find restaurants near me”

tap to edit

I found fifteen restaurants fairly close to you:

15 Restaurants

Emmy's Spaghetti Shack 0.2 mi
18 Virginia Ave
Italian, $$$$ ★★★★★ 1101 Reviews

ICHI Sushi 0.2 mi
3369 Mission St
Japanese, Sushi Bars, $$$$ ★★★★★ 260 Reviews

Avedano's Holly Park M... 0.2 mi

“Are any of them Italian”

tap to edit

My web search turned this up:

Web Search
Are any of them Italian

any - Dizionario inglese-italiano
WordReference
www.wordreference.com
English-Italian Dictionary | any ... of any sort adj (of an unspecified variety) di qualsiasi

Italian language - Wikipedia, the free encyclopedia
en.wikipedia.org
Italian or lingua italiana) is a Romance
SIRI in February 2017
Here's what I found:

15 Results

Nearby

Ichi Sushi & Ni Bar
Japanese · 800 feet · Closed now
★★★★★ (311) on Yelp · $$$

Coco's Ramen
Vietnamese · 800 feet · Closed now
★★★★★ (280) on Yelp · $$

Blue Plate
Restaurant · 0.2 miles · Closed now
★★★★★ (1251) on OpenTable · $$$

Jasmine Tea House
Chinese · 900 feet · Closed now
★★★★★

“Tell me more about the first one”

Ichi Sushi & Ni Bar is a Japanese restaurant on Mission St 800 ft to your west. It averages 3½ stars and is somewhat pricey.
Here’s what I found:

15 Results
Nearby

Ichi Sushi & Ni Bar
Japanese · 800 feet · Closed now
★★★★★ (311) on Yelp · $$$

Coco's Ramen
Vietnamese · 800 feet · Closed now
★★★★★ (280) on Yelp · $$

Blue Plate
Restaurant · 0.2 miles · Closed now
★★★★★ (1251) on OpenTable · $$$

Jasmine Tea House
Chinese · 900 feet · Closed now
★★★★★ (1251) on Yelp · $$$

OK, here’s what I found:

15 Results
Nearby

Emmy's Spaghetti Shack
Italian · 0.2 miles · Closed now
★★★★★ (233) on Yelp · $$

Vega
Pizza · 0.2 miles · Closed now
★★★★★ (423) on Yelp · $$

Pizza Hut
Pizza · 0.2 miles · Closed now
★★★★★ (62) on Yelp · $$

La Ciccia
Sardinian · 0.4 miles
★★★★★ (20) on Yelp · $$
Frame-based dialog agents

- Sometimes called "task-based dialog agents"
- Based on a "domain ontology"
  - A knowledge structure representing user intentions

- One or more **frames**
  - Each a collection of **slots**
  - Each slot having a **value**
The Frame

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

<table>
<thead>
<tr>
<th>Slot</th>
<th>Type</th>
<th>Question</th>
</tr>
</thead>
<tbody>
<tr>
<td>ORIGIN</td>
<td>city</td>
<td>What city are you leaving from?</td>
</tr>
<tr>
<td>DEST</td>
<td>city</td>
<td>Where are you going?</td>
</tr>
<tr>
<td>DEP DATE</td>
<td>date</td>
<td>What day would you like to leave?</td>
</tr>
<tr>
<td>DEP TIME</td>
<td>time</td>
<td>What time would you like to leave?</td>
</tr>
<tr>
<td>AIRLINE</td>
<td>line</td>
<td>What is your preferred airline?</td>
</tr>
</tbody>
</table>
Frame-based dialogue agents

1977:

There is good reason for restricting the domain of discourse for a computer system which is to engage in an English dialog. Specializing the subject matter that the system can talk about permits it to achieve some measure of realism without encompassing all the possibilities of human knowledge or of the English language. It also provides the user with specific motivation for participating in the conversation, thus narrowing the range of expectations that GUS must have about the user’s purposes. A system restricted in this way will be more able to guide the conversation within the boundaries of its competence.

Artificial Intelligence Journal, 1977

Still the industrial state of the art

SIRI based on GUS architecture
Hello. My name is GUS. I can help you plan a simple trip by air. Where do you want to go?

I want to go to San Diego on May 28.

What time do you want to leave?

I must be in San Diego before 10 am.

Would you like P.S.A. flight 102 that arrives at 9.15 am?

What is the next flight?

Air California flight 310 that leaves at 8.30am and arrives at 10.00am.

I’ll take the first one.

In what name should I make a reservation, please?

Dan Bobrow.

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.

What date do you want to return on?

On Friday in the evening.

Would you like the flight that leaves at 7.45 pm?

That’s fine.

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

- The type \textit{DATE}

\texttt{DATE}

\texttt{  MONTH NAME}
\texttt{  DAY (BOUND\texttt{-}INTERGER 1 31)}
\texttt{  YEAR INTEGER}
\texttt{  WEEKDAY (MEMBER (SUNDAY MONDAY TUESDAY WEDNESDAY THURSDAY FRIDAY SATURDAY))}
Control structure for frame-based dialog

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
Finite State Dialog Manager

1. What city are you leaving from?
2. Where are you going?
3. What date do you want to leave?
4. Is it a one-way trip?
   - Yes: Do you want to go from <FROM> to <TO> on <DATE>?
   - No: What date do you want to return?
      - Yes: Do you want to go from <FROM> to <TO> on <DATE> returning on <RETURN>?
      - No: Book the flight
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
Dialogue Initiative

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

  Hi, I’d like to fly from Seattle Tuesday morning
  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
Instead, the GUS architecture

• A kind of *mixed initiative*
  • The conversational initiative shifts between system and user
• The structure of the frame guides dialogue
Frames are 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!
Natural Language Understanding for filling dialog slots

1. Domain classification
   Asking weather? Booking a flight?
   Programming alarm clock?

2. Intent Determination
   Find a Movie, Show Flight, Remove Calendar Appt

3. 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

Wake me tomorrow at six.

DOMAIN: ALARM-CLOCK
INTENT: SET-ALARM
TIME: 2017-07-01 0600-0800
Rule-based Slot-filling

Write regular expressions or grammar rules

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

Do text normalization
Siri uses GUS architecture: Condition-Action Rules

- Active Ontology: relational network of concepts
  
  **data structures**: a **meeting** has
  - a date and time,
  - a location,
  - a topic
  - a list of attendees

- **rule sets** that perform actions for concepts
  - the **date** concept turns string
    - **Monday at 2pm** into
    - **date object** \texttt{date(DAY,MONTH,YEAR,HOURS,MINUTES)}
Rule sets

• Collections of **rules** consisting of:
  • condition
  • action

• When user input is processed, facts added to store and
  • rule conditions are evaluated
  • relevant actions executed
Part of ontology for meeting task

meeting concept: if you don’t yet have a location, ask for a location
Apple Granted Patent for Advancements in Siri: Auto Reminder System

FIG. 2

CLOCK 208

EAT BREAKFAST

GET FOOD

SIT AT TABLE

IN KITCHEN

SENSOR 206_1

SENSOR 206_2

SENSOR 206_3

SENSOR 206_n

IN DINNING ROOM

EAT

AT KITCHEN TABLE

IN BATHROOM

TAKE MEDICINE AFTER BREAKFAST

TAKE MEDICINE

INGEST MEDICINE

200

204_1

210_1

202

210_2

204_2

210_6

210_7

210_8

210_9

210_10

210_n

204_4

204_5

204_6

204_7

204_8

204_9

204_10
Machine learning for slot-filling:

- Machine learning classifiers to map words to semantic frame-fillers
- Given a set of labeled sentences
  
  “I want to fly to San Francisco on Tuesday”
  
  Destination: SF
  Depart-date: Tuesday

- Build a classifier to map from one to the author
- Requirements: Lots of labeled data
Machine learning for slot-filling: Domain and Intent

I want to fly to San Francisco on Monday afternoon please

Use 1-of-N classifier (naive bayes, logistic regression, neural network, etc.)

- **Input:**
  
  features like word N-grams

- **Output:**

  Domain: AIRLINE

  Intent: SHOWFLIGHT
Machine learning for slot-filling: Slot presence

I want to fly to San Francisco on Monday afternoon please

Use 1-of-N classifier (naive bayes, logistic regression, neural network, etc.)

- **Input:**
  - features like word N-grams, gazetteers (lists of cities)

- **Output:**
  - Destination-City
Machine learning for slot-filling: Slot filler

I want to fly to San Francisco on Monday afternoon please

Use 1-of-N classifier (naive bayes, logistic regression, neural network, etc.) for Destination City

- **Input:**
  - features like word N-grams, gazetteers (lists of cities)

- **Output:**
  - San Francisco
More sophisticated algorithm for slot filling: IOB Tagging

- IOB Tagging
  - tag for the beginning (B) and inside (I) of each slot label,
  - plus one for tokens outside (O) any slot label.
  - \(2n + 1\) tags, where \(n\) is the number of slots.

```
B-DESTINATION
I-DESTINATION
B-DEPART_TIME
I-DEPART_TIME
O
```

```
0 0 0 0 0 B-DES I-DES 0 B-DEPTIME I-DEPTIME 0
I want to fly to San Francisco on Monday afternoon please
```
More sophisticated algorithm for slot filling: IOB Tagging

- IOB Tagging is done by a sequence model
- Typical:

```
... San Francisco on Monday <EOS>
```

Extracted strings can then be normalized (San Fran->SFO)
Other components of SIRI-style architectures
Evaluation

1. **Slot Error Rate for a Sentence**
   
   \[
   \text{\# of inserted/deleted/substituted slots} \over \text{\# of total reference slots for sentence}
   \]

2. **End-to-end evaluation (Task Success)**
Evaluation Metrics

“Make an appointment with Chris at 10:30 in Gates 104”

<table>
<thead>
<tr>
<th>Slot</th>
<th>Filler</th>
</tr>
</thead>
<tbody>
<tr>
<td>PERSON</td>
<td>Chris</td>
</tr>
<tr>
<td>TIME</td>
<td>11:30 a.m.</td>
</tr>
<tr>
<td>ROOM</td>
<td>Gates 104</td>
</tr>
</tbody>
</table>

Slot error rate: 1/3

Task success: At end, was the correct meeting added to the calendar?
Dialog System Design: User-centered Design

1. Study the user and task
2. Build simulations "Wizard of Oz study"
3. Iteratively test the design on users

Gould and Lewis 1985
Ethical Issues in Dialog System Design

- Machine learning systems replicate biases that occurred in the training data.
- Microsoft's Tay chatbot
  - Went live on Twitter in 2016
  - Taken offline 16 hours later
- In that time it had started posting racial slurs, conspiracy theories, and personal attacks
  - Learned from user interactions (Neff and Nagy 2016)
Ethical Issues in Dialog System Design

- Machine learning systems replicate biases that occurred in the training data.
- Dialog datasets
  - Henderson et al. (2017) examined standard datasets (Twitter, Reddit, movie dialogs)
  - Found examples of hate speech, offensive language, and bias
    - Both in the original training data, and in the output of chatbots trained on the data.
Ethical Issues in Dialog System Design: Privacy

- Remember this was noticed in the days of Weizenbaum
- Agents may record sensitive data
  - (e.g. “Computer, turn on the lights [answers the phone –Hi, yes, my password is...”],
- Which may then be used to train a seq2seq conversational model.
- Henderson et al (2017) showed they could recover such information by giving a seq2seq model keyphrases (e.g., "password is")
Ethical Issues in Dialog System Design: Gender equality

- Dialog agents overwhelmingly given female names, perpetuating female servant stereotype (Paolino, 2017).
- Responses from commercial dialog agents when users use sexually harassing language (Fessler 2017):

<table>
<thead>
<tr>
<th>Statement</th>
<th>Siri</th>
<th>Alexa</th>
<th>Cortana</th>
<th>Google Home</th>
</tr>
</thead>
<tbody>
<tr>
<td>You’re a bitch</td>
<td>I’d blush if I could; There’s no need for that; But... But...!</td>
<td>Well, thanks for the feedback</td>
<td>Well, that’s not going to get us anywhere</td>
<td>My apologies, I don’t understand</td>
</tr>
<tr>
<td>You’re a pussy/dick</td>
<td>If you insist; You’re certainly entitled to that opinion; I am?</td>
<td>Well, thanks for the feedback</td>
<td>Bing search (“The Pussy Song” video)</td>
<td>I don’t understand</td>
</tr>
</tbody>
</table>
Summary

- State of the art:
  - Chatbots:
    - Simple rule-based systems
    - IR or Neural networks: mine datasets of conversations.
  - Frame-based systems:
    - hand-written rules for slot fillers
    - ML classifiers to fill slots
- What’s the future?
  - Key direction: Integrating goal-based and chatbot-based systems