CS 6120/CS4120: Natural Language Processing

Instructor: Prof. Lu Wang
College of Computer and Information Science
Northeastern University
Webpage: www.ccs.neu.edu/home/luwang
Logistics

• Assignment 3 is extended! Due November 23rd, 6pm.

• Office hour 11/21
  • 1-2pm, Tirthraj, 462 WVH
  • 4-5pm, Vish, 462 WVH (extra)
Conversational Agents aka Dialog Agents

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

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

2. Chatbots
“Make an appointment for Tuesday”

What time is your appointment?

“215”

OK, I can create your meeting. Note that you already have an appointment at 2:15 pm. Shall I schedule it anyway?

“No”
“Find restaurants near me”

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”

Ich 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 • 800 feet • Closed now

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 • 800 feet
⭐⭐⭐⭐⭐ (69) on Yelp • $

La Ciccia
Sardinian • 0.4 miles
Architectures for Practical Dialog Systems

• Finite-State
  *Simple information: e.g., passwords or credit cards*

• Frame-Based
  *All commercial and academic system (SIRI etc.)*
How SIRI works

Figure from Jerome Bellegarda
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?
5. Do you want to go from <FROM> to <TO> on <DATE>?
6. 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.
Sytem 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 state of the art: Frame-based dialogue

• A kind of *mixed initiative*
  • The conversational initiative shifts between system and user
• The structure of the **frame** guides dialogue
Frame-based dialogue

• Invented up the hill in 1977:

GUS, A Frame-Driven Dialog System

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.

Artificial Intelligence Journal, 1977

• Still the state of the art (in real world systems)
  • SIRI based on GUS architecture
The Frame

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

<table>
<thead>
<tr>
<th>Slot</th>
<th>Question</th>
</tr>
</thead>
<tbody>
<tr>
<td>ORIGIN</td>
<td>What city are you leaving from?</td>
</tr>
<tr>
<td>DEST</td>
<td>Where are you going?</td>
</tr>
<tr>
<td>DEPT DATE</td>
<td>What day would you like to leave?</td>
</tr>
<tr>
<td>DEPT TIME</td>
<td>What time would you like to leave?</td>
</tr>
<tr>
<td>AIRLINE</td>
<td>What is your preferred airline?</td>
</tr>
</tbody>
</table>
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!
The Natural Language Understanding Component

Show me morning flights from Boston to SF on Tuesday.

SHOW:

FLIGHTS:

ORIGIN:

CITY: Boston
DATE: Tuesday
TIME: morning

DEST:

CITY: San Francisco
Often called "dialog state" detection

**Dialog state**: representation of what the user wants at any point in a dialog

- Which slots got filled in the last sentence?
- What is the current state of the frame?
  - All the values of the filled slots
- What is the user's last "dialogue act":
  - Did they ask me a **question**?
  - **Inform** me of something?
How to do Frame-based Natural Language Understanding?

Rule-based
Statistical
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 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
Statistical Natural Language Understanding

• Statistical 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
Statistical Slot filling

• Given a sentence:

  I want to go to Santa Fe

• Classifier predicts which slot the user wants to fill

  Output: (ORIGIN, DESTINATION, DEPARTURE-DATE, AIRLINE)

  Features: Words, Named Entities

• Classifier or sequence model predicts the filler:

  ORIGIN  DEST

  I want to go from Boston to SF
Evaluation

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

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 Act Detection

• Dialog Act: The dialog function of the sentence
  • Question
  • Command
  • Suggestion

• Given a user's sentence:
  • How many Italian restaurants are in walking distance?

• Was that a question?
Dialogue Act detection is hard

Can you give me a list of the flights from Atlanta to Boston?

• This looks like an QUESTION.
  • It has a question-mark, starts with "can you"
• If so, the answer is:
  • YES.
• But really it’s a COMMAND, a polite form of: 
  Please give me a list of the flights...

• What looks like a QUESTION can be a COMMAND
Dialogue Act detection

Rules
Statistical Classifiers
Another example of dialogue act detection: Correction Detection

- If system misrecognizes an utterance, and either
  - Rejects
  - Via confirmation, displays its misunderstanding
- Then user has a chance to make a

  correction
    - Repeat themselves
    - Rephrasing
    - Saying “no” to the confirmation question.
Corrections

• Unfortunately, corrections are harder to recognize than normal sentences!
  • Swerts et al (2000): corrections misrecognized twice as often as non-corrections!!!
  • Why?
    • Prosody seems to be largest factor: hyperarticulation
    • Liz Shriberg example:
      • “NO, I am DE-PAR-TING from Jacksonville”
A Labeled dialogue (Swerts et al)

<table>
<thead>
<tr>
<th>Turn</th>
<th>Corr Type</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>S</strong></td>
<td>On which day of the week do you want to leave?</td>
</tr>
<tr>
<td><strong>S</strong></td>
<td>Sorry, I didn’t hear anything from you. Please tell me your departure date</td>
</tr>
<tr>
<td><strong>U</strong></td>
<td><em>Um tomorrow</em></td>
</tr>
<tr>
<td><strong>S</strong></td>
<td>You can say the departure date. For example, you can say ‘today’ or ‘tomorrow’. Or you can say ‘Monday’, ‘Tuesday’, ‘Wednesday’...</td>
</tr>
<tr>
<td><strong>U</strong></td>
<td><em>Tomorrow</em></td>
</tr>
<tr>
<td><strong>S</strong></td>
<td>You can say the departure date. For example, you can say ‘today’ or ‘tomorrow’. Or you can say ‘Monday’, ‘Tuesday’, ‘Wednesday’...</td>
</tr>
<tr>
<td><strong>U</strong></td>
<td><em>Saturday</em></td>
</tr>
<tr>
<td><strong>S</strong></td>
<td>Do you want to leave on Saturday?</td>
</tr>
<tr>
<td><strong>U</strong></td>
<td>Yes</td>
</tr>
<tr>
<td><strong>S</strong></td>
<td>What time do you want to leave?</td>
</tr>
<tr>
<td><strong>U</strong></td>
<td><em>Eight o’clock A M</em></td>
</tr>
<tr>
<td><strong>S</strong></td>
<td>Do you want to leave around ten o’clock p.m</td>
</tr>
<tr>
<td><strong>U</strong></td>
<td><em>Eight o’clock</em></td>
</tr>
<tr>
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</table>
Machine learning to detect user corrections: features

• Lexical information (*no, correction, I don’t, swear words*)
• Prosodic indicators of hyperarticulation
  • pause duration, word duration
• Length
• LM probability
• Various dialogue features (repetition)
Deciding what to say:

Dialog act generation
   Should I do a "Confirmation" dialog act?

Natural Language Generation
   Given that I'm confirming, what exactly should I say?
Grounding

• Why do elevator buttons light up?
• 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
• What is the linguistic correlate of this?
Grounding and Confirmation

• We need to know whether an action succeeded or failed
• Talking is an action!
• I need to know if my action succeeded
  • i.e. the hearer understood my turn!
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

| C1:  | ...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. |
| C6:  | OK I’ll take the 5ish flight on the night before on the 11th. |
| C7:  | OK.                                              |
Grounding examples

Display:
  C: I need to travel in May
  A: And, what day in May did you want to travel?

Acknowledgement
  C: He wants to fly from Boston
  A: mm-hmm
  C: to Baltimore Washington International
Grounding examples (2)

• Acknowledgement + next relevant contribution
  And, what day in May did you want to travel?
  And you’re flying into what city?
  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
From 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?
Confirmation

• Errors: Speech is a pretty errorful channel
  • Humans use grounding to confirm that they heard correctly
  • Automatic speech recognition (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

S: Which city do you want to leave from?
U: Baltimore
S: Do you want to leave from Baltimore?
U: Yes
Explicit confirmation

U: I’d like to fly from Denver Colorado to New York City on September 21st in the morning on United Airlines

S: Let’s see then. I have you going from Denver Colorado to New York on September 21st. Is that correct?

U: Yes
Implicit confirmation: display

U: I’d like to travel to Berlin
S: When do you want to travel to Berlin?

U: Hi I’d like to fly to Seattle Tuesday morning
S: Traveling to Seattle on Tuesday, August eleventh in the morning. Your name?
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).
Implicit and Explicit

• Early systems: all-implicit or all-explicit
• Modern systems: adaptive
• How to decide?
  • ASR system can give confidence metric.
  • This expresses how convinced system is of its transcription of the speech
  • If high confidence, use implicit confirmation
  • If low confidence, use explicit confirmation
• Should also consider cost of an error: Explicit confirmation before moving money or booking flights
Rejection

• “I’m sorry, I didn’t understand that.”
• Reject when:
  • ASR confidence is low
  • Best interpretation is semantically ill-formed (e.g., where is sushi (Sochi))
• Might have four-tiered level of confidence:
  • Below confidence threshold, reject
  • Above threshold, explicit confirmation
  • If even higher, implicit confirmation
  • Even higher, no confirmation
Chatbots

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

• Rule-based
  • Pattern-action rules (Eliza)
  • + a mental model (Parry)
• Corpus-based (from large chat corpus)
  • Information Retrieval
  • Neural network encoder-decoder
Chatbots

• Sometimes for fun
• Originally designed as a testbed for clinical therapy
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*
Domain: Rogerian psychology interview

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”

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
• Most chatbots trying to pass Turing test choose a domain with similar properties
Eliza pattern/transform rules

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

You hate me

WHAT MAKES YOU THINK I HATE YOU

0 means kleene * (zero or more of some words)
The 3 is the constituent number in pattern
Eliza Rules

<table>
<thead>
<tr>
<th>keyword</th>
<th>pattern</th>
<th>Ranked list of transforms</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>((K, ((D_1, R_{1,1}, R_{1,2}) \cdots (R_{1,m_1}))))</td>
<td>((D_1, R_{1,1}, R_{1,2}) \cdots (R_{1,m_1})))</td>
</tr>
<tr>
<td></td>
<td>((D_2, R_{2,1}, R_{2,2}) \cdots (R_{2,m_2})))</td>
<td>((D_2, R_{2,1}, R_{2,2}) \cdots (R_{2,m_2})))</td>
</tr>
<tr>
<td></td>
<td>(\vdots)</td>
<td>(\vdots)</td>
</tr>
<tr>
<td></td>
<td>((D_n, R_{n,1}, R_{n,2}) \cdots (R_{n,m_n})))</td>
<td>((D_n, R_{n,1}, R_{n,2}) \cdots (R_{n,m_n})))</td>
</tr>
</tbody>
</table>
ELIZA architecture

• Examine each word $w$ in user sentence
  • Return the $w$ with highest keyword rank

• If $w$ exists:
  • Check each rule for $w$ in ranked order
    • Choose first one that matches sentence
    • Apply transform

• If no keyword applies, either
  • Apply the transform for the “NONE” key, or
  • Grab an action off the memory queue
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 (transformation rules)
  I (transformation rules)
PLEASE GO ON
THAT’S VERY INTERESTING
I SEE
Memory

(MEMORY MY
 (0 YOUR 0 = LETS DISCUSS FURTHER WHY YOUR 3)
 (0 YOUR 0 = EARLIER YOU SAID YOUR 3)
 (0 MY 0 = DOES THAT HAVE ANYTHING TO DO WITH THE FACT THAT 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
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
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.
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 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?  \[\rightarrow\]  (your 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)
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]
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)
Eliza Meets Parry

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.
IR-based models of chatbots

• Cleverbot
• Xiaoice

• Mine conversations of human chats or human-machine chats
• Choose a response by
  • Taking user's turn and finding a (tf-idf) similar turn in the dataset
  • Grabbing whatever the response was to that turn.
Deep-learning chatbots

• Train on:
  • movie dialogue databases
  • twitter conversations
• Train a deep neural net to map from one users turn to the other users response
Seq2seq model architecture

Sample outputs from a neural net chatbot

How are you doing?
I’m good, you?

I am off all week next week
What time you going?

I know you love me and I love you.
I love you too, baby.

Ohh I’ve never seen that! How long does it take you guys to learn the drill?
Like 2 weeks ago!!

Jiwei Li, Michel Galley, Chris Brockett, Jianfeng Gao, and Bill Dolan. 2015. A Diversity-Promoting Objective Function for Neural Conversation Models.
Sample output from a neural net chatbot

<table>
<thead>
<tr>
<th>#</th>
<th>Context</th>
<th>Message</th>
<th>Generated Response</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>noone can help this time its just i need a support system</td>
<td>well im here to support you. whenever you need me.</td>
<td>thank you, i appreciate it thank you</td>
</tr>
<tr>
<td>2</td>
<td>nothin much, and how’s the book?!</td>
<td>its good but i’m only like halfway through cuz i don’t feel like reading. i’m so bored ...</td>
<td>that’s good! i have the book but i’m bored too.</td>
</tr>
<tr>
<td>3</td>
<td>i know! what have you been up to?</td>
<td>nothing really. just enjoying summer and working. what about you?</td>
<td>that’s good! nothing much just listening to music and you?</td>
</tr>
<tr>
<td>4</td>
<td>thanks, your way better though</td>
<td>i’m just good at expressing it that’s but trust me you are freakin amazing and waay better</td>
<td>i know, i don’t trust you, i love you too! thank you</td>
</tr>
<tr>
<td>5</td>
<td>thanks. yeah i am good thanks! good, but angry!</td>
<td>sorry you’re angry, been there, i hope it dies down.</td>
<td>i hope so too, i’m not angry!!</td>
</tr>
<tr>
<td>6</td>
<td>figures. bass. never ate bass.</td>
<td>bass is extremely mild with very few bones. my favorite after catfish</td>
<td>that’s one of the bones, chuck bass is mine too.</td>
</tr>
<tr>
<td>7</td>
<td>today i will walk for miles. clear sky clear mind.</td>
<td>love walking. it’s too hot to walk this month. i haven’t moved. frozen. heat freeze.</td>
<td>it’s too hot for you to get up to honey if you have frozen yogurt to you.</td>
</tr>
<tr>
<td>8</td>
<td>arresting people over “nasty tweets”? please. i’m not condoning what the lad did but arresting him is a tad silly.</td>
<td>it appears he may have sent death threats to others while all this was kicking off too.</td>
<td>i did! he was armed with nuclear threats? that’s what’s happening to you.</td>
</tr>
</tbody>
</table>