Linguistic & social interaction Methods in AI research

Dong Nguyen 26 September 2019



Practicalities

Literature for today:

- (optional) *Speech & Language processing*, Section 25.5 (25.5 Natural language generation in the dialog-state model)
- (optional) Computational Sociolinguistics: A Survey, Nguyen et al. 2016 https://www.mitpressjournals.org/doi/full/10.1162/COLI_a_00258#. V-jwiqJ95cR

Language

Making sense of language is more than recognising linguistic content

- Making sense of language involves discovering the goals of different bits of content
- and the way different bits of content connect
- and the broader (social) context in which it was uttered



Speakers do not act in isolation!

(Historic view) Language allows us to *describe* the world

But...



I hereby bet you 10 euros it won't rain tomorrow



Order!



Speech Act Theory

- By Austin (*How to Do Things with Words*, 1962) and developed further by Searle.
- So far, focus had been on utterances that describe the state of affairs, but there are utterances for which it doesn't makes sense to say whether they are *true or false*
- Austin distinguished between:
 - **Constatives**: true/false statements
 - **Performatives**: action (do)

abjure, abolish, accept, acknowledge, acquit, admit, admonish, advise, affirm, agree to, announce, answer, apologize, ascribe, ask, assent, assert, assess, assume, authorize, baptize, beg, bet, bid, call upon, caution, charge, christen, claim, classify, command, commiserate, compliment, concur, congratulate, conjecture, convict, counsel, declare, declare out, delegate, demand, demur, deny, describe, diagnose, disagree, dispute, donate, dub, excuse, exempt, fire, forbid, give notice, grant, guarantee, guess, hire, hypothesize, identify, implore, inform, instruct, license, name, notify, offer, order, pardon, permit, plead, pray, predict, prohibit, promise, proscribe, query, question, rank, recommend, refuse, reject, renounce, report, request, require, rescind, resign, sanction, say, sentence, state, submit, suggest, summon, suppose, swear, tell, testify, thank, urge, volunteer, vouch for, warn, withdraw

from: Keith Allan, "Meaning and Speech Acts"

http://users.monash.edu.au/~kallan/papers/Speechacts.html

Types of acts

Austin identifies three types of acts that are performed simultaneously:

- locutionary act: basic act of uttering a linguistic expression ('what is said')
- illocutionary act: the kind of action the speaker intends to accomplish ('what is meant'), e.g. thanking, ...
 - often this is what referred to when people talk about speech acts
- *perlocutionary act*: effects produced on the audience

Relations between Acts

The same *locutionary* act can have different illocutionary forces in different contexts:

The gun is loaded (threatening, explaining)

The same *illocutionary* act can be realised by different locutionary acts:

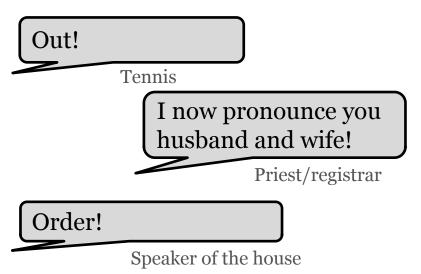
- Could you close the door
- Close the door, please

Speech Act taxonomy (Searle)

category	description	example
assertives	committing S to something being the case; causing H to believe something	I make the best cake, She is a great leader
commissives	committing S to do something	I promise that I will, I will bring the book
directives	getting H to do something	Could you close the door, Turn off the music
expressives	expressing how S feels about a situation	I'm sorry that, I'm thrilled you will
declarations	they immediately change the state of the world	You are fired

Convention

- speech acts can be linguistically marked
- large variety of how this happens



- specific socially recognised situations
- involving agents with specific socially recognised authority

Convention

- speech acts can be linguistically marked
- large variety of how this happens

The performative formula: *I (hereby) VERB X*

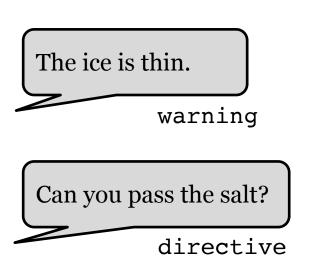
I thank you for helping me

I withdraw from the competition

I admit I was wrong

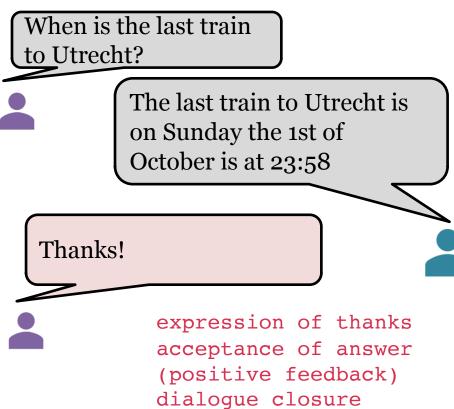
But

- very often, a speech act is not marked as such
- speech acts can be complex



But

- very often, a speech act is not marked as such
- speech acts can be complex



Dialog acts

- Speech act theory focusses on the intentions of the speaker. But a dialogue is not simply a sequence of actions each performed by individual speakers.
- Dialogue requires coordination amongst participants.
 - Many actions in dialogue serve to manage the interaction itself (e.g. grounding). These are overlooked by speech act theory
- Dialog acts extends the notion of speech acts for conversational phenomena.

Conversational structure

- Adjacency pairs: Pairing of two dialog acts
 - Questions and answers, greeting and greeting, proposal and acceptance (or rejection)

```
Use: I want to fly from UNKNOWN to London

System: Let's see, where do you want

to fly from?

User: Barcelona

System: Ok, here are some flights from

Barcelona to London
```

Identifying dialog acts

Words:

• *Please..., would you...* (request)

Conversational structure:

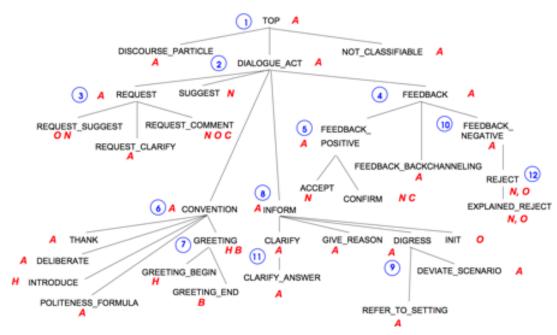
• *yeah* after a proposal vs. *yeah* following an inform

Prosody:

Final rising pitch (question)

VerbMobil-2

Acts specific to meeting scheduling domain



Dialogue Acts in VERBMOBIL-2, Alexanderssony et al., 1997

Switchboard Dialog Acts



Hi, Wanet <last name>

Conventional-opening



How are you?

Conventional-opening



I'm doing fine

Conventional-opening



Where you from?

Wh-Question



I'm from New England

Statement-non-opinion

http://compprag.christopherpotts.net/swda.html

Applications

- Dialog systems!
- E-mail assistant (Cohen et al. 2004)
- Conversation analysis (Twitter, e-mails, dating)

Bill,

Do you have any sample scheduling-related email we could use as data? -Steve

Sure, I'll put some together shortly. -Bill

Fred, can you collect the msgs from the CSPACE corpora tagged w/ the "meeting" noun, ASAP? -Bill

Yes, I can get to that in the next few days. Is next Monday ok? -Fred Assistant announces: "new email **request**, priority unknown."

Assistant: "should I add this new **commitment** to your todo list?"

Assistant: notices outgoing request, may take action if no answer is received promptly.

Assistant: notices incoming commitment. "Should I send Fred a reminder on Monday?"

(Cohen et al. 2004)

Applications

- Dialog systems!
- E-mail assistant (Cohen et al. 2004)
- Conversation analysis (Twitter, emails)

Prabhakaran & Rambow 2017 analyzed over 90k e-mails from the Enron corpus to study power structures, by making use of dialog acts:

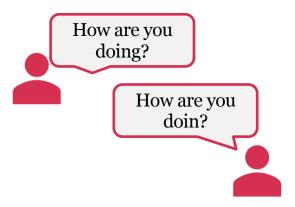
- Request Action
- Request information
- Inform
- Conventional (e.g. greetings)

Language and social interaction

Social meaning

How are you doing?

Social meaning



Social meaning



Social media

The phrase "never won an Oscar" can no longer apply to Leonardo DiCaprio. The actor, who was famously snubbed by the Academy five times, was awarded best actor for his performance in "The Revenant." DiCaprio was heavily favored to win, with many a fan and critic declaring he "deserves" it now more than ever.

(The Washington Post, Feb. 29, 2016)



LLCOOLJ. @llcoolj · Feb 29

I'm truly happy for this guy...
leonardodicaprio congrats man.. great
message.. #oscars...
instagram.com/p/BCXBk4XI6lp/



★ 184 ♥ 715
 Oprah Winfrey @Oprah · Feb 29

It Happppppened!! bravo Leo! standing O in my p.j's. #Oscars 1



Language variation carries social meaning.

Normalization throws away meaningful signals (Eisenstein 2013)





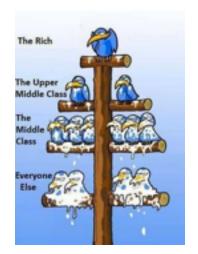


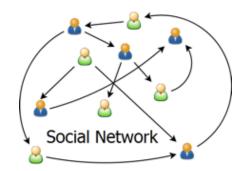
Sociolinguistics

Sociolinguistics is the descriptive study of the effect of any and all aspects of society, including cultural norms, expectations, and context, on the way language is used, and the effects of language use on society.

(Wikipedia)







Computational Sociolinguistics

Computational methods to study how language and society relate

 Combining ideas and methods from both sociolinguistics and computer science

• See Nguyen et al., Computational Linguistics, 2016

Motivation

Better NLP tools

 Testing and refining theories using large-scale naturalistic text data

- New analysis tools for sociolinguists and social scientists
- More fine-grained analyses of online behavior (e.g., user profiling, participation in campaigns)

```
Gezellig bij Emily en Charlotte.
Translation: Having fun with Emily
and Charlotte.
Hiiiiii schatjesss!
Translation: Hiiiiii cutiesss!
   @USER
Goodmorning
Saaie middag.
Translation: Boring afternoon
```

How old?

Gezellig bij Emily en Charlotte. Translation: Having fun with Emily and Charlotte.

Hiiiiii schatjesss!
Translation: Hiiiiii cutiesss!



Goodmorning

Saaie middag.
Translation: Boring afternoon

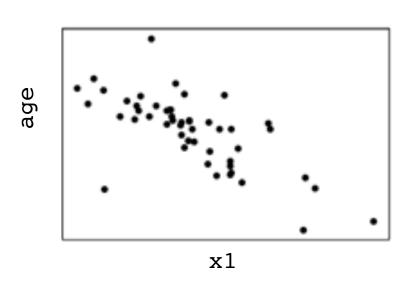
15 year old

We asked 220 people: average guess 16.7 years

How old?



Regression



How do we get the training data?

features target
$$\{ <\mathbf{x^{(1)}}, \ \mathbf{y^{(1)}}>, ..., <\mathbf{x^{(N)}}, \ \mathbf{y^{(N)}}> \}$$

Goal: Predict the target using the features

Regression task:

Output is a continuous value ($y \in \mathbb{R}$)

Notation:

Each instance $x^{(i)}$ has d features:

$$[x_1, ..., x_d]$$

 $x_i^{(i)}$: the j^{th} feature of instance i

Data collection: self-reported



user-provided information

disadvantages: biases? availability?

advantage: high accuracy



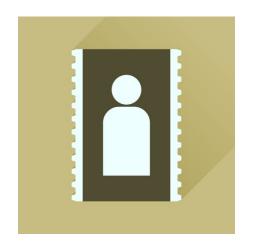
Data collection: annotation

disadvantages:

time-consuming annotator biases

advantage:

random set



text

picture

name

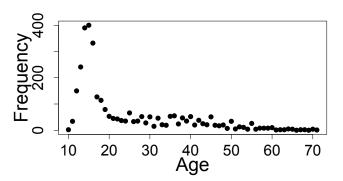
linked profiles (e.g., LinkedIn)

Data collection: derived

```
Location:
                                         Gender:
Tweets with GPS
coordinates
                                         Names
                                                    biases?
u'geo': {
                                   Popular Names in 1980 (SSA)
     u'type': u'Point',
     u'coordinates':
                                  Ran
                                   k
                                                       Female name
                                         Male name
                                              Michael
                                                             Jennifer
     [51.452131,
                                           Christopher
                                                            Amanda
          6.05643]
                                                Jason
                                                              Jessica
                                  https://www.ssa.gov/oact/babynames/
```

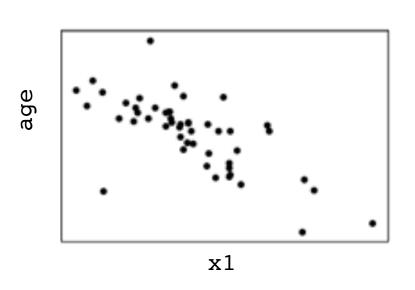
Nguyen et al (2013): data collection

- Goal: Sample a 'representative' set of Dutch Twitter users
- How: Sample based on 'het' (definite article, pronoun)
- Two annotators annotated demographic information age, based on social media profiles (Twitter, Facebook, LinkedIn), tweets etc





Regression



features target
$$\{ \langle \mathbf{x}^{(1)}, \ \mathbf{y}^{(1)} \rangle, ..., \langle \mathbf{x}^{(N)}, \ \mathbf{y}^{(N)} \rangle \}$$

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Each instance $x^{(i)}$ has d features:

$$[X_1, ..., X_d]$$

 $x_j^{(i)}$: the j^{th} feature of instance i

Age prediction: features

Younger

- first (I) and second (you) singular pronouns
- alphabetical lengthening
- capitalization of words
- slang words
- Internet acronyms

Older

- first (*we*) plural pronouns
- prepositions
- determiners
- articles
- longer words
- longer sentences
- links
- hash tags

Features: just unigrams (we also tried more advanced ones, did not help performance much)

See Nguyen et al. (2016) for a more detailed overview

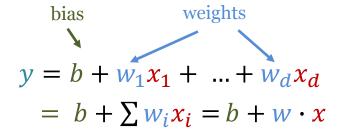
adolescents use the **most non-standard** forms





Linear regression

For each feature $\mathbf{x_j}$ we learn a weight w_j , so $w \in \mathbb{R}^d$ and $b \in \mathbb{R}$. Given an instance, map it to a real number:



This is a **linear model**.

features target
$$\{ <\mathbf{x^{(1)}}, \ \mathbf{y^{(1)}}>, ..., <\mathbf{x^{(N)}}, \ \mathbf{y^{(N)}}> \}$$

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Regression task:

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$$[x_1, ..., x_d]$$

 $x_i^{(i)}$: the j^{th} feature of instance i

Prediction experiments

We used linear regression with **L2 regularization**

How well does the model perform?

- r = 0.8845
- Mean absolute error:3.8812

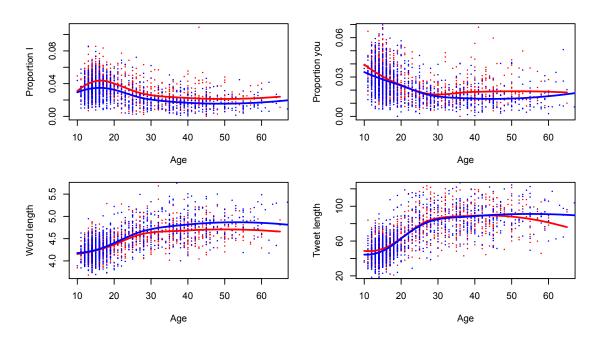
Dutch	English
school	school
ik	I
:)	:)
werkgroep	work group
stages	internships
oke	okay
xd	xd
ben	am
haha	haha
als	if

Dutch	English
verdomd	damn
dochter	daughter
wens	wish
zoon	son
mooie	beautiful
geniet	enjoy
dank	thanks
goedemorgen	good morning
evalueren	evaluate
sterkte	take care

Younger users

Older users

Language & Age



D. Nguyen, R. Gravel, D. Trieschnigg and T. Meder: "How Old Do You Think I Am?": A Study of Language and Age in Twitter. ICWSM 2013

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New analysis tools for sociolinguists and social scientists

 More fine-grained analyses of online behavior (e.g., user profiling, participation in campaigns)



In 20 odd years of having a mobile I must dropped a mobile maybe 2/3 times but luckily the screen never cracked.this year alone I dropped my mobile 6 times in 2 days and the only thing that stopped the screen from cracking was this protecting tempered screen which cracked.

I can't recommed this product enough



How can this product be 'amazons choice'? I found it to be a waste of money. Says it fits iPhone 7 but doesn't even cover the width of the active part of the screen. The packaging (which says iPhone 7/8 on a sticker on one side and iPhone S on the other) had previously been opened and both protectors had feint scratches the whole way from top to bottom. Alignment stickers not sticky enough so came off when trying to fit the protector so ended up crooked. Not happy with this purchase at all.

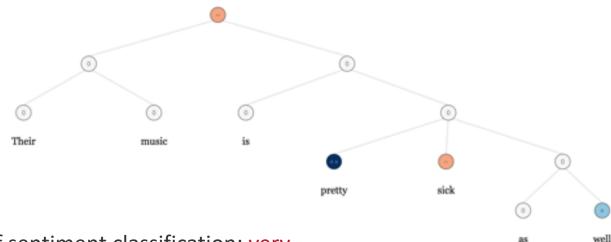
Every single shot is masterfully created and there are some genuinely scary scenes which will make you sit on the edge of your seat. (IMDB review)



What it's like to drive?
A bit scary at times but handles better in the dry

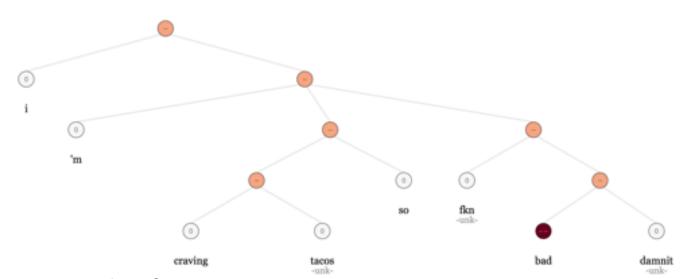
Whether 'scary' is positive or negative depends on the domain!

Their music is pretty sick as well

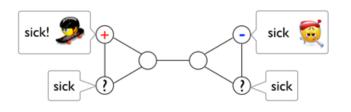


There are 5 classes of sentiment classification: very negative, negative, neutral, positive, and very positive.

i'm craving tacos so fkn bad damnit



There are 5 classes of sentiment classification: very negative, negative, neutral, positive, and very positive.



Linguistic homophily:

socially connected individuals tend to use language in a similar way.

Yang and Eisenstein (TACL 2017) exploit social network structures to improve sentiment analysis.

Motivation

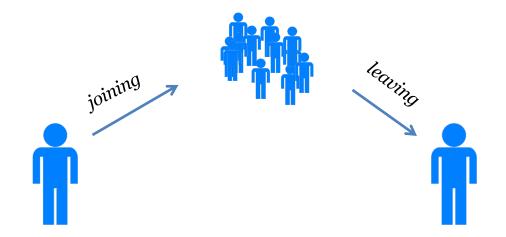
Better NLP tools

 Testing and refining theories using large-scale naturalistic text data

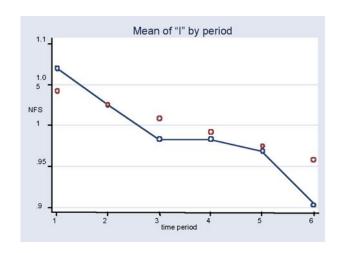
New analysis tools for sociolinguists and social scientists

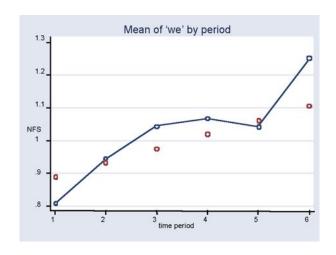
 More fine-grained analyses of online behavior (e.g., user profiling, participation in campaigns)

Linguistic change & online communities



Adoption of of community norms





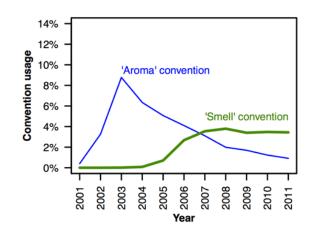
The Language of Online Intercultural Community Formation, Cassell and Tversky, Journal of Computer-Mediated Communication Volume 10, Issue 2, 2006

Online beer communities

10 years of data (2001 – 2011) **Beeradvocate**°



• 1,586,614 posts. 33,387 users



Linguistic change in Beeradvocate: Comparing language models

Community language model (q)

$$P("I am") = 0.09$$

 $P("You are") = 0.05$

User language model (p)

$$P("I am") = 0.15$$

 $P("You are") = 0.02$

How to compare two probability distributions? → Cross entropy

$$H(p,q) = -\sum p(x)\log(q(x))$$

No Country for Old Members: User Lifecycle and Linguistic Change in Online Communities, Danescu-Nculescu-Mizil et al, WWW 2013

Aside: cross-entropy

x	p(x)	q(x)	s(x)
Α	0.1	0.2	0.6
В	0.8	0.6	0.1
С	0.1	0.2	0.3

$$H(p,q) = -0.1 * ln(0.2) - 0.8 * ln(0.6) - 0.1 * ln(0.2) = 0.731$$
 $H(s,q) = 1.50$

$$H(p,q) = -\sum p(x)\log(q(x))$$

Aside: cross-entropy

Cross entropy loss! (recall: logistic regression and neural networks)

x	p(x)	q(x)	
Α	0	0.1	
В	1	8.0	
С	0	0.1	
	1	1	
true	label	clas	sifier output

$$H(p,q) = -1 * ln(0.8) = 0.223$$

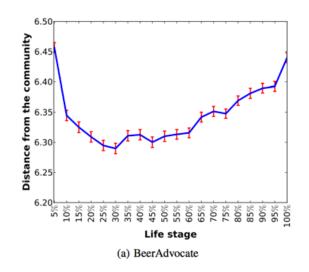
$$H(p,q) = -\sum p(x)\log(q(x))$$

Linguistic change in Beeradvocate: Comparing language models

Compare the language models of a user at different 'life stages' to the community language model at that time

Users stop adapting to language in the community.

Turns out to be useful signal for predicting whether a user will leave the community.



Life stage: %posts written out of total number

No Country for Old Members: User Lifecycle and Linguistic Change in Online Communities, Danescu-Nculescu-Mizil et al, WWW 2013

Text Generation

Natural Language Generation

Computer algorithms/systems that produce useful texts in human language

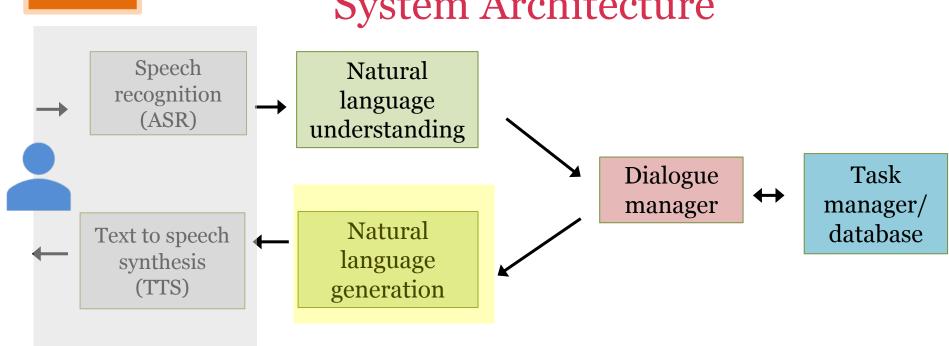
- Input: data (sensor data, logic formulas, database tables,...)
- Output: answers to queries; help messages, documents, reports, explanations, etc.

Goals can differ:

- Building models of human language use
- Building practically useful systems

RECAP!

Typical Frame-based Dialogue System Architecture



Two components

- Content planning (what to say)
 - In dialog systems: Dialog policy
- Sentence realization (how to say it)

Two components

- Content planning (what to say)
 - In dialog systems: Dialog policy
- Sentence realization (how to say it)

Given a dialog act to generate and some additional information (slots and values), how to generate the text?

NLG using sentence templates

Create a set of templates with gaps.

There is a restaurant in the X area of town called Y in the Z price range

- Fill X with a name of an area
- Fill Y with a restaurant name
- Fill Z with low / medium / high

Use syntactic rules

More variation

```
There is_{sg}/are_{plur} a restaurant<sub>sg</sub> / restaurants<sub>plur</sub> in the X area
```

Number Agreement rule:

- 1st gap = sg iff 2nd gap = sg
- 1st gap = plur iff 2nd gap = plur

Referring expressions

Adding conditions on template filling

There is a restaurant in X

Fill X with 'in the same area' iff the area where the restaurant is located is the only area mentioned in the previous utterance.

How old are you? Consistent persona message 16 and you? response What's your age? message 18. response A Persona-Based Neural What is your major? message Conversation Model, Li et al. 2016 I'm majoring in psychology response What did you study in college? message EOS england English lit. Target response where EOS ! Rob england Rob do live in Rob Rob you D Gomes25 Jinnmeow3 eaker embeddings (70k) u.s. london skinnyoflynny2 england TheCharlieZ great Rob_712 Word embeddings Dreamswalls good Tomcoatez Bob Kelly2 Kush 322 kierongillen5 monday This Is Artful tuesday The Football Bar DigitalDan285 stay

Which words to use?

• Depends on the social context and speaker characteristics!

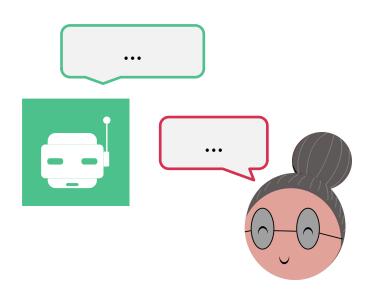
Where is the **elevator**?

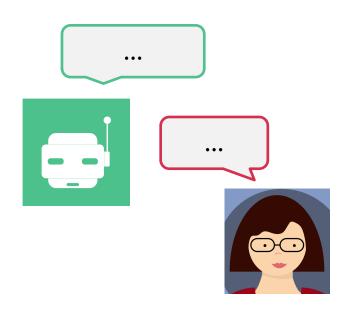
The **lift** is located near the reception.

Where is the **elevator**?

The **elevator** is located near the reception.

Style





Summary

- Making sense of language is more than recognising linguistic content
 - Can we integrate ideas about linguistic and social aspects of conversations to make better dialog systems?

What do you need to know

- What speech acts are
 - Why recognizing speech acts and dialog acts can be challenging when just looking at the textual content
- What dialog acts are
- That there is a rich connection between language and society (and why text normalisation for some tasks isn't a good idea)
- Cross entropy
- Template-based text generation

Thanks

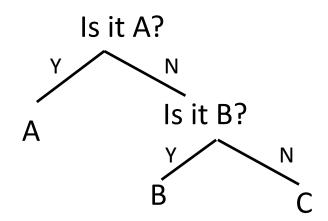
Part of the slides based on slides from Rick Nouwen (speech acts), Kees van Deemter (text generation), Raquel Fernández (speech acts)

Entropy

A: 0.50%

B: 0.25%

C: 0.25%



Average number of questions:

$$0.50 * 1 + 0.25 * 2 + 0.25 * 2 = 1.5$$

Entropy:

$$-0.5 * \log 2(0.5) - 0.25 * \log 2(0.25) - 0.25 * \log 2(0.25) = 1.5$$