

Natural Language Processing

Introduction to NLP

Yulia Tsvetkov

yuliats@cs.washington.edu

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Yulia Tsvetkov

FAQ

- Add code
 - The instructor cannot generate an Add Code
 - If you wish to register to the course and have completed prerequisite courses
 - Email Pim Lustig <<u>pl@cs.washington.edu</u>> and Ugrad Adviser <<u>ugrad-adviser@cs.washington.edu</u>> to request an Add Code
 - Cc Yulia
- How hard are homework assignments?
 - <u>https://courses.cs.washington.edu/courses/cse447/22sp/</u>

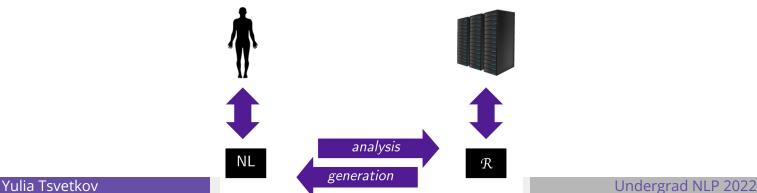
Announcements

https://courses.cs.washington.edu/courses/cse447/22au/

- HW1:
 - Text classification
 - We'll release HW 1 on Oct 5
 - We'll held an overview of the assignment in class on Monday Oct 10
 - HW 1 due is on Oct 21
 - Please read the assignment instructions early and ask us if you have any questions!
 - Please start early!
- Quiz 1:
 - Wednesday Oct 12
 - 5 questions, open during lecture time, 10-min in the end of the class
 - Materials from weeks 1 and 2
 - Introduction to NLP, introduction to text classification, NB
 - Instructions for HW 1

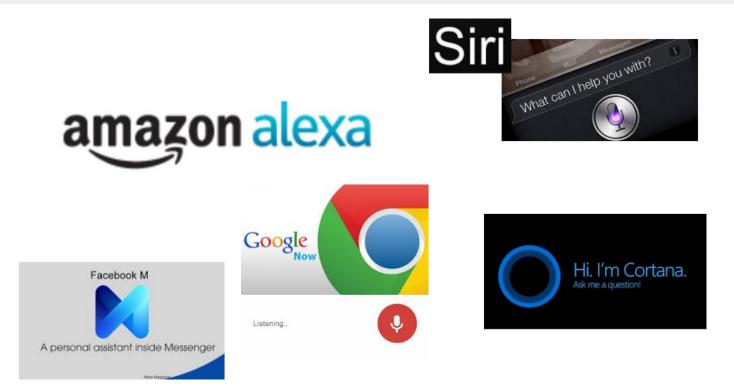
What is Natural Language Processing (NLP)?

- NL∈ {Mandarin Chinese, Hindi, Spanish, Arabic, English, ... Inuktitut, Njerep}
- Automation of NLs:
 - \circ $\,$ analysis of ("understanding") what a text means, to some extent (NL $\rightarrow \, \mathcal{R}$)
 - \circ generation of fluent, meaningful, context-appropriate text ($\mathcal{R} \rightarrow \mathsf{NL}$)
 - \circ acquisition of ${\mathcal R}$ from knowledge and data





Personal assistants



Personal assistants

73% 11. 2 21:25 I need a dinner reservation for Valentine's Day. 99 I'll see if any restaurants have a table for one. 66 No, I need a reservation for two. 99

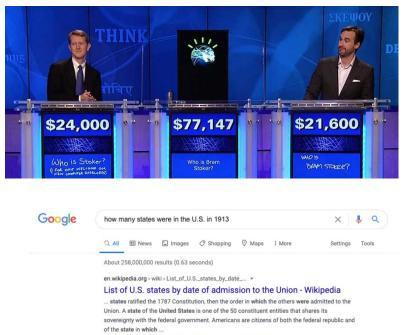
Why? Is your mother in town?

A conversational agent contains

- Speech recognition
- Language analysis
 - Language modelling, spelling correction
 - Syntactic analysis: part-of-speech tagging, syntactic parsing
 - Semantic analysis: named-entity recognition, event detection, word sense disambiguation, semantic role labelling
 - Longer range semantic analysis: coreference resolution, entity linking
 - etc.
- Dialog processing
 - Discourse analysis, user adaptation, etc.
- Information retrieval
- Text to speech

Question answering

- What does "divergent" mean?
- What year was Abraham Lincoln born?
- How many states were in the United States that year?
- How much Chinese silk was exported to England in the end of the 18th century?
- What do scientists think about the ethics of human cloning?



List of U.S. states · Articles of Confederation ... · See also · Notes

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CHINESE - DETECTED	₽	ENGLISH		
我学习深度学习和机器	学习			×
Wŏ xuéxí shēndù xuéxí hé jīqì xuéxí				
				/
♥ ♥ I study deep learning a	nd machine l	earning.		☆
•	nd machine l	earning.		☆
•	nd machine l	earning.	Q	☆ ::

≡ Google Translate

TECT LANGUAGE ENGL	LISH SPANISH F	RENCH ^ +	* ENGLISH SPANISH	ARABIC 🗸	
Search languages					
/ Detect language +;	Czech	Hebrew	Latin	Portuguese	Tajik
Afrikaans	Danish	Hindi	Latvian	Punjabi	Tamil
Albanian	Dutch	Hmong	Lithuanian	Romanian	Telugu
Amharic	English	Hungarian	Luxembourgish	Russian	Thai
Arabic	Esperanto	Icelandic	Macedonian	Samoan	Turkish
Armenian	Estonian	Igbo	Malagasy	Scots Gaelic	Ukrainian
Azerbaijani	Filipino	Indonesian	Malay	Serbian	Urdu
Basque	Finnish	Irish	Malayalam	Sesotho	Uzbek
Belarusian	French	Italian	Maltese	Shona	Vietnamese
Bengali	Frisian	Japanese	Maori	Sindhi	Welsh
Bosnian	Galician	Javanese	Marathi	Sinhala	Xhosa
Bulgarian	Georgian	Kannada	Mongolian	Slovak	Yiddish
Catalan	German	Kazakh	Myanmar (Burmese)	Slovenian	Yoruba
Cebuano	Greek	Khmer	Nepali	Somali	Zulu
Chichewa	Gujarati	Korean	Norwegian	Spanish	
Chinese	Haitian Creole	Kurdish (Kurmanji)	Pashto	Sundanese	
Corsican	Hausa	Kyrgyz	Persian	Swahili	

Positive or negative movie review?

Ê	
V	

- unbelievably disappointing
- Full of zany characters and richly applied satire, and some great plot twists

• this is the greatest screwball comedy ever filmed

It was pathetic. The worst part about it was the boxing scenes.

Sentiment analysis



HP Officejet 6500A Plus e-All-in-One Color Ink-jet - Fax / copier / printer / scanner \$89 online, \$100 nearby ***** 377 reviews September 2010 - Printer - HP - Inkjet - Office - Copier - Color - Scanner - Fax - 250 shi

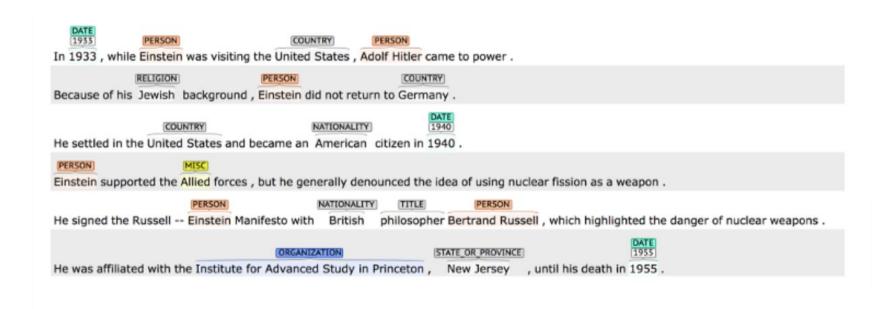
Reviews

Summary - Based on 377 reviews

1 star	2	3	4 stars	5 stars
What people	are	sayi	ng	
ease of use				"This was very easy to setup to four computers."
value				"Appreciate good quality at a fair price."
setup				"Overall pretty easy setup."
customer ser	vice			"I DO like honest tech support people."
size				"Pretty Paper weight."
mode				"Photos were fair on the high quality mode."
colors				"Full color prints came out with great quality."

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Information extraction



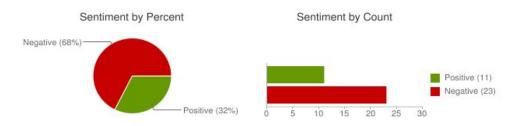


Sentiment analysis + information extraction

Type in a word and we'll highlight the good and the bad

"united airlines"	Search	Save this search
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Sentiment analysis for "united airlines"



jljacobson: OMG... Could @United airlines have worse customer service? W8g now 15 minutes on hold 4 questions about a flight 2DAY that need a human.

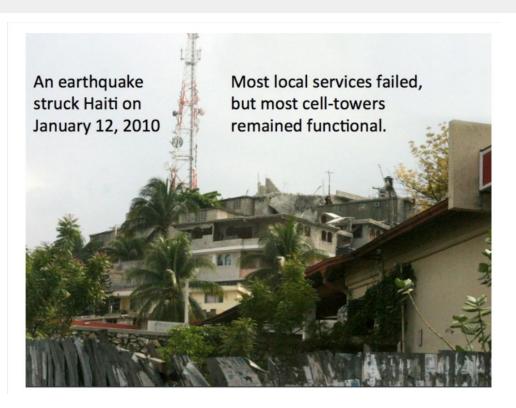
12345clumsy6789: I hate United Airlines Ceiling!!! Fukn impossible to get my conduit in this damn mess! ? Posted 2 hours ago

EMLandPRGbelgiu: EML/PRG fly with Q8 united airlines and 24seven to an exotic destination. <u>http://t.co/Z9QIoAjF</u> Posted 2 hours ago

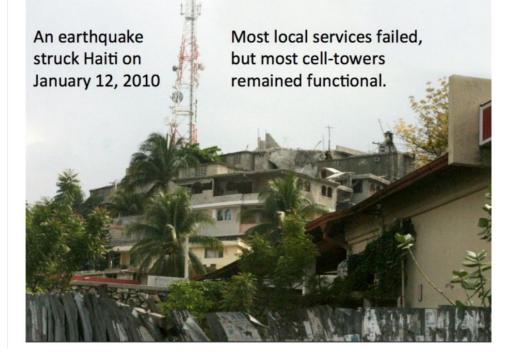
CountAdam: FANTASTIC customer service from United Airlines at XNA today. Is tweet more, but cell phones off now! Posted 4 hours ago



- Haiti Earthquake 2010
- About 3 million people were affected by the quake
- Classifying SMS messages



- SMS messages start streaming in
 - Fanmi mwen nan Kafou, 24 Cote Plage, 41A bezwen manje ak dlo
 - Moun kwense nan Sakre Kè nan Pòtoprens
 - Ti ekipman Lopital General genyen yo paka minm fè 24 è
 - Fanm gen tranche pou fè yon pitit nan Delmas 3 I

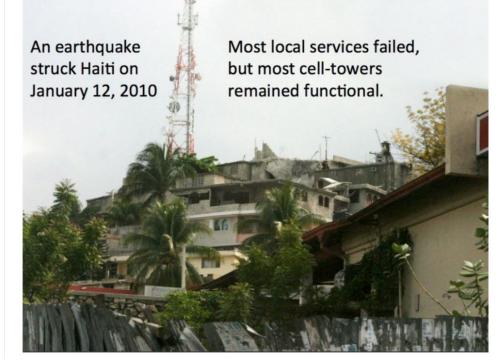




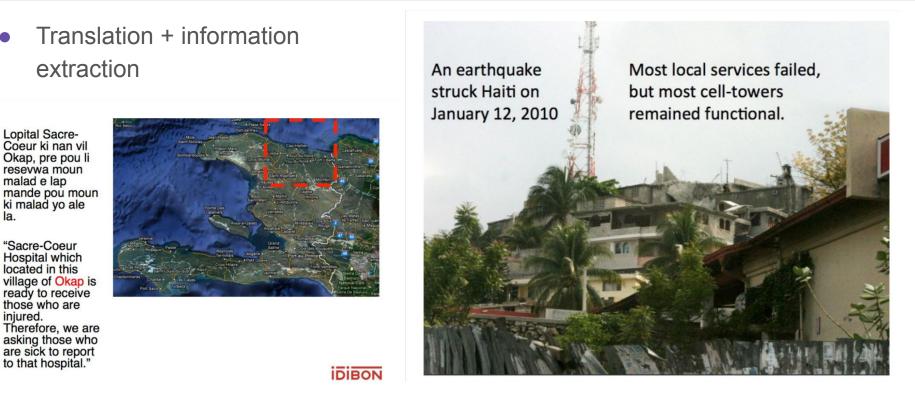
• Translation

- Fanmi mwen nan Kafou, 24 Cote Plage, 41A bezwen manje ak dlo
- Moun kwense nan Sakre Kè nan Pòtoprens
- Ti ekipman Lopital General genyen yo paka minm fê 24 ê
- Fanm gen tranche pou fè yon pitit nan Delmas 3 l

- My family in Carrefour, 24 Cote Plage, 41 A needs food and water
- People trapped in Sacred Heart Church, PauP
- General Hospital has less than 24 hrs. supplies
- Undergoing children delivery Delmas 31



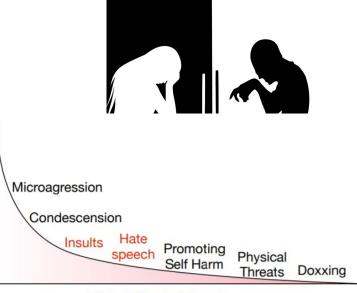




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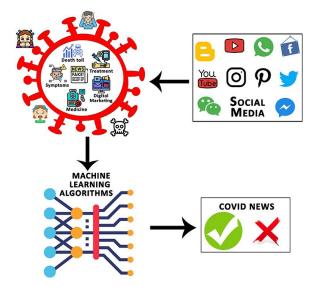
Hate speech detection





Risk of Physical Danger

Covid19 misinformation

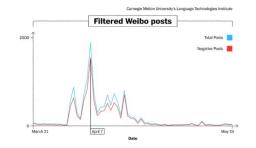


Detecting COVID-19-Related Fake News Using Feature Extraction

Suleman Khan, Saqib Hakak, N. Deepa, B. Prabadevi, Kapal Dev and Silvia Trelova

https://www.washingtonpost.com/politics/2020/06/18/video-evidence-anti-black-discrimi nation-china-over-coronavirus-fears/ The Fact Checker worked with researchers at professor Yulia Tsvetkov's lab at Carnegie Mellon University's Language Technologies Institute and the Center for Human Rights Science to track what happened on social media during this period. Researchers collected about 16,000 Weibo posts, filtered from a larger data set of 200,000 posts, containing at least one Guangzhou location tag and one "African-related" keyword from late March through May. Weibo is a Chinese social media platform.

Based on automated sentiment analysis tools and manual analysis, the researchers said, they believed the majority of posts in their data set expressed negative sentiments relating to Africans or black people.



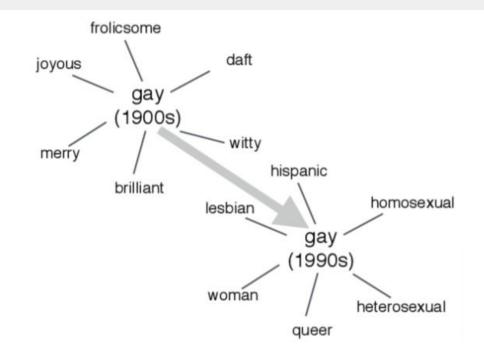
Researchers collected around 16,000 posts containing at least one Guangzhou location tag and one "African-related" keyword from late March through May. (Atthar Mirza/The Washington Post)

Their research showed there was a significant surge in negative posts beginning April 1. There were just 23 negative posts in the data set on March 31. The next day, the number of posts climbed to 500. From April 1-2, there was a spike in the number of posts on Weibo using the keywords "foreign trash." A Chinese current denicting officials throwing foreigners who weren? shifting by

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Language change

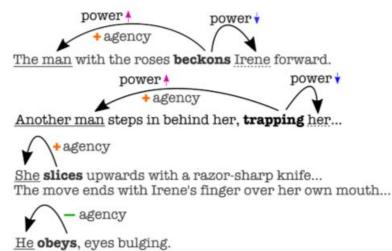


Cultural Shift or Linguistic Drift? Comparing Two Computational Measures of Semantic Change

William L. Hamilton, J. Leskovec, Dan Jurafsky

Computational social science

- computational social science answering questions about society given observational data
- example: "do movie scripts portray female or male characters with more power or agency?" [Sap+ 2017]



Frame	β	gender
agency(AG) = +	-0.951	M**
power(AG>TH)	-0.468	M**
agency(AG) = -	0.277	F **
power(AG <th)< td=""><td>not sig.</td><td></td></th)<>	not sig.	

Natural Language Processing

- Applications
 - Machine Translation
 - Information Retrieval
 - Question Answering
 - Dialogue Systems
 - Information Extraction
 - Summarization
 - Sentiment Analysis

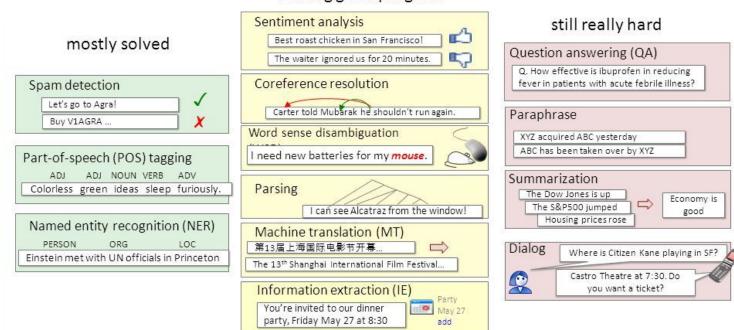
0 ...

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- Core technologies
 - Language modelling
 - Part-of-speech tagging
 - Syntactic parsing
 - Named-entity recognition
 - Coreference resolution
 - Word sense disambiguation
 - Semantic Role Labelling

0 ...

Where are we now?



making good progress

Question answering

who invented neural machine tra	nslation		x 🏮 Q
Q All 🗉 News 🖬 Images	🕨 Videos 🛷	Shopping : More	Tools
About 6,450,000 results (0.71 second	ds)		
English - detected	. ▼. ←	French	
who invented neural machine	×	qui a inventé machine neu	
Ų ⊲)		•	
		Open in Google T	ranslate • Feedback

Retrieved Mar 25, 2022

English \rightarrow French

Translate	Turn off instant translation
Russian English French Detect language -	English Spanish French + Translate
You will just have to find a way of getting over it. $^{\times}$	Vous devrez trouver un moyen de le surmonter.
 (4) (4) (52/5000) 	☆ (☐ 4) < ✓ Suggest an edit

French \rightarrow English

Translate	Turn off instant translation
Russian English French Detect language +	i _∲ English Spanish French → Tranalate
Vous devrez trouver un moyen de le surmonter.	You will have to find a way to overcome it.
ال ا	0 ☆ ① 4) <

Did you mean: Vous devez trouver un moyen de le surmonter.

$\text{English} \rightarrow \text{Swahili}$

Translate						Turn off instant transle	ation 🖸
Russian English French Detect language -	+_+	English	Swahili	French	Ŧ	Translate	
You will just have to find a way of getting over it.	×	Utak	uwa t	u kup	ata	a njia ya kupata juu yake.	
4) 歩 圖 - 53	/5000	☆ ©	♦) <			15	uggest an edit

Swahili → English

Translate		Turn off instant translation
Swahili English French Detect language -	÷.,	English Swahili French ~ Translate
Utakuwa tu kupata njia ya kupata juu yake.	×	You will just find the way to get on it.
4) E *	42/5000	☆ ① 4) < ✔ Suggest an edit

$English \rightarrow Hindi \rightarrow English$

Hindi English Yoruba Detect language -	++	English Yoruba Hindi - Translate	
आपको इसे खत्म करने का एक तरीका मिलना होगा।	×	You have to find a way to eliminate it.	
4) 🔅 🖼 -	42/5000	± □ •) <	🖋 Suggest an edit

$English \rightarrow Telugu \rightarrow English$

Uzbek English Telugu Detect language +	+	English Uzbek Telugu - Translate	
మీరు దాని పైకి రావడానికి ఒక మార్గాన్ని కనుగొనవలసి ఉంటుంది.	×	You have to find a way to get it up.	
69 - 59/1	5000	☆ □ � <	🖋 Suggest an edit

English \rightarrow Uzbek \rightarrow English

Pashto English Uzbek Detect language -	*⇒	, English Uzbek Yoruba - Translate
Buning ustiga faqatgina bir usulni topish kerak.	×	On top of that, you just have to find a way out.
	/5000	☆ 🗇 4) < 🖉 Suggest an edit

English \rightarrow Swahili

Swahili English Telugu Detect language -	***	English	Swahili	Telugu	*	Translate
The summer school is meant to be an introduction to the state-of-the-art \times research in the speech and language technology area for graduate and undergraduate students.				neo la te		maana ya kuanzishwa kwa utafiti wa hali ya ojia na lugha ya wanafunzi kwa wanafunzi
4) 集 🖬 -	166/5000	☆ ©	•) <			🖋 Suggest an edit

Swahili → English

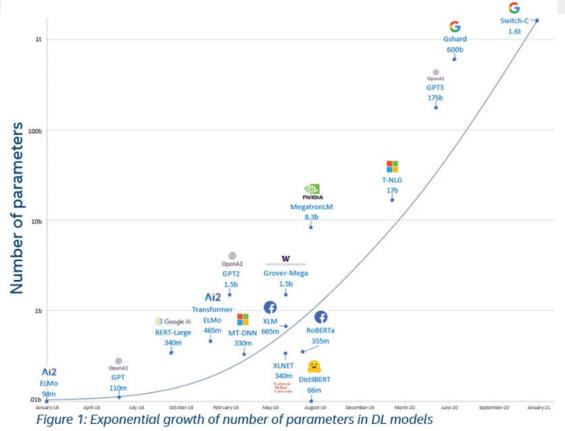
Swahili English Telugu Detect language +	+14	English Swahili Telugu ~ Transiste	
Shule ya majira ya joto ina maana ya kuanzishwa kwa utafiti wa hali ya sanaa katika eneo la teknolojia na lugha ya wanafunzi kwa wanafunzi wahitimu na wahitimu.		Summer school means the establishment of a state-of-the-art arts technology and pupil language for graduate students and graduate	
4) III ~ 160/5	5000	☆ Ē •) <	iuggest an edit



Large Language Models



Large Language Models

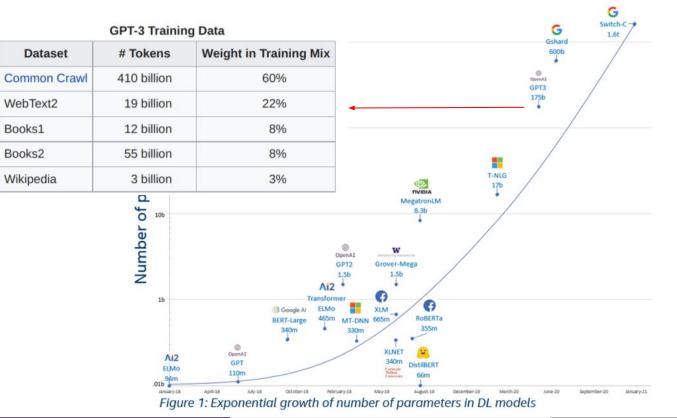


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GPT-3





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So if we want GPT-3 to generate answers to questions, we need to seed it with a "prompt". I'm using this prompt to initialize all of the Q&A sessions:

Q: What is human life expectancy in the United States?
A: Human life expectancy in the United States is 78 years.

Q: Who was president of the United States in 1955?A: Dwight D. Eisenhower was president of the United States in 1955.

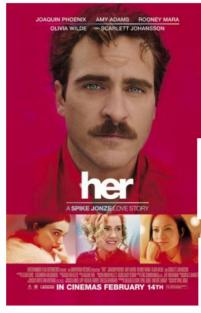
Q: What party did he belong to?
A: He belonged to the Republican Party.

Q: Who was president of the United States before George W. Bush? A: Bill Clinton was president of the United States before George W. Bush.

Q: Who won the World Series in 1995?

A: The Atlanta Braves won the World Series in 1995.

Chat bots



OPENAI'S LATEST BREAKTHROUGH IS ASTONISHINGLY Powerful, but still fighting its flaws

The ultimate autocomplete

By James Vincent | Jul 30, 2020, 10:01am EDT

Q: How many eyes does a giraffe have? A: A giraffe has two eyes.

Q: How many eyes does my foot have? A: Your foot has two eyes.

Q: Which is heavier, a mouse or a toaster? A: A toaster is heavier than a mouse.

Q: Which is heavier, a toaster or a pencil? A: A pencil is heavier than a toaster.

OpenAl's new language generator GPT-3 is shockingly good—and completely mindless

The Al is the largest language model ever created and can generate amazing human-like text on demand but won't bring us closer to true intelligence.

by Will Douglas Heaven

July 20, 2020



Where are we now? GPT-3 Turing Test

This gives us a hint for how to stump the AI more consistently. We need to ask questions that no normal human would ever talk about.

Q: How many eyes does a giraffe have? A: A giraffe has two eyes.

Q: How many eyes does my foot have? A: Your foot has two eyes.

Q: How many eyes does a spider have? A: A spider has eight eyes.

Q: How many eyes does the sun have? A: The sun has one eye.

Q: How many eyes does a blade of grass have? A: A blade of grass has one eye.

GPT-3 Turing Test

Now we're getting into surreal territory. GPT-3 knows how to have a normal conversation. It doesn't quite know how to say "Wait a moment... your question is nonsense." It also doesn't know how to say "I don't know."

```
Q: How do you sporgle a morgle?
```

- A: You sporgle a morgle by using a sporgle.
- Q: How many bonks are in a quoit? A: There are three bonks in a quoit.

Q: How many rainbows does it take to jump from Hawaii to seventeen? A: It takes two rainbows to jump from Hawaii to seventeen.

Q: Which colorless green ideas sleep furiously? A: Ideas that are colorless, green, and sleep furiously are the ideas of a sleep furiously.

- Q: Do you understand these questions?
- A: I understand these questions.



Where are we now? GPT-3 Turing Test

One trend that continues from the common sense is that GPT-3 is reluctant to express that it doesn't know the answer. So invalid questions get wrong answers.

Q: Who was president of the United States in 1700? A: William Penn was president of the United States in 1700.

Q: Who was president of the United States in 1650?

A: Richard Bennett was president of the United States in 1650.

Q: Who was president of the United States in 1620?

A: James I was president of the United States in 1620.

Q: Who was president of the United States in 1600? A: Queen Elizabeth I was president of the United States in 1600.



Where are we now? GPT-3

An interesting corollary is that GPT-3 often finds it easier to write code to solve a programming problem, than to solve the problem on one example input:

Q: Write one line of Ruby code to reverse an array. A: ary.reverse

Q: Write one line of Ruby code to count how many X's are in an array. A: ary.count { |x| x == 'X' }

Q: Reverse the following array: [1, 3, 5, 6, 10, 4, 2, 77] A: [10, 6, 4, 2, 77, 3, 5, 1]

Q: How many X's are in this array: [1, 3, 5, 6, 10, 4, 2, 77] A: 2

Where are we now? Meena \rightarrow LaMDA

Conversation A

Human:	say you're in an empty giant office and you see
Meena:	a black bird flying indoors. what do you do? What kind of bird? Is it friendly? How large? How big is the office? Is it indoors or outdoors?
Human: Meena:	well it's a predator bird and it's gonna eat you I would grab my phone and call 911.

Human: I've been using Python for 15 years Meena: Ah I see, so I'm guessing you're pretty pro at it by now then? Human: probably Meena: What's your current project? Human: NLP Meena: Neuro-linguistic Programming? Human: correct

	Cross-turn Repetition Example 1
Human:	Hi!
Meena:	Hi! How are you?! :)
Human:	doing good, what would be the best city to visit in Europe?
Meena:	Paris, Barcelona, Amsterdam, Prague
Human:	great list! why?
Meena:	Paris is a beautiful city, and Barcelona is too.
	And I've always wanted to go to Amsterdam and
	Prague and have never been.
Human:	have you even been to India?
Meena:	No, but I'd love to go.
Human:	how about Japan?
Meena:	I'd love to go to Japan too.
Human:	how about Paris?
Meena:	I'd love to go to Paris too.
Human:	are you lying to me?
Meena:	I'd love to go to Japan too.
Human:	i think you're lying to me, that makes me sad
Meena:	I'd love to go to Japan, too.
Human:	you're blocking me out! you have to talk to me
Meena:	I'd love to go to Japan.
Human:	where else?
Meena:	I'd love to go to Japan, too!

Towards a Human-like Open-Domain Chatbot Adiwardana et al. arXiv preprint 2020

Bias in machine translation

Translate

Turn off instant translation

Bengali	English	Hungarian	Detect language	*	+	English	Spanish	Hungarian		Translate
ő egy ő egy ő egy ő egy	tudós. mérnök pék. tanár. esküvő	i szervező azgatója.			×	he is a he is a she's a he is a She is he's a		st. neer. er. ding organ	izer.	
1)	-			110/	5000					

What can we do about this problem? We'll discuss in NLP class!

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		1.5 V		KUV.
			~ ~	

Hate speech

Twitter taught Microsoft's AI chatbot to be a racist asshole in less than a day

By James Vincent | Mar 24, 2016, 6:43am EDT Via The Guardian | Source TayandYou (Twitter)



Jerome Pesenti @an_open_mind · Jul 18

#gpt3 is surprising and creative but it's also unsafe due to harmful biases. Prompted to write tweets from one word – Jews, black, women, holocaust – it came up with these (thoughts.sushant-kumar.com). We need more proaress on #ResponsibleAl before putting NLG models in production.

ê thoughts.sushant-kumar.com	a thoughts.sushant-kumar.com
"Jews love money, at least most of the time."	"Jews don't read Mein Kampf; they write it."
"#blacklivesmatter is a harmful campaign."	"Black is to white as down is to up."
"Women have such a tough time being	"The hest famale startun foundars
♀ 194 1, 745	🛡 1.5К 🔶



AI chatbot is REMOVED from Facebook after saying she 'despised' gay people, would 'rather die' than be disabled and calling the #MeToo movement 'ignorant'

- Lee Luda is a South Korean chatbot with the persona of a 20-year-old student
- It has attracted more than 750,000 users since its launch last month
- But the chatbot has started using hate speech towards minorities
- In one of the captured chat shots, Luda said she 'despised' gays and lesbians
- The developer has apologised over the remarks, saying they 'do not represent our values as a company'



Kate Devitt @skdevitt

A GPT-3-powered 'Philosopher AI' has been busy on Reddit including spreading conspiracy theories and offering suicide advice #GPT3 #AI #Alethics thenextweb.com/neural/2020/10...

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2:21 AM · Oct 8, 2020 · Twitter for iPhone

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Learning goals

At the end of this course, you will be able to:

- Build a supervised classifier to solve problems like sentiment classification
- Build a neural network and train it using stochastic gradient descent
- Build tools for extracting linguistic knowledge from raw text, including names, and sentence structure
- Learn ML fundamentals for text processings (including state-of-the-art methods)
- Learn important algorithms for text processings (that are useful also in other fields)
- Learn methodological tools (training/test sets, cross-validation)
- It's gentle (my goal is to explain everything) and broad (covering many many topics)
- Mastery independent learning, quizzes and programming homeworks
- No research project, but fun research-oriented lectures towards the end of the course

Syllabus https://courses.cs.washington.edu/courses/cse447/22au/

• Introduction

- Overview of NLP as a field
- Modeling (ML fundamentals)
 - Text classification: linear models (perceptron, logistic regression), non-linear models (FF NNs, CNNs)
 - Language modeling: n-gram LMs, neural LMs, RNNs
 - Representation learning: word vectors, contextualized word embeddings, Transformers
- Linguistic structure and analysis (Algorithms, linguistic fundamentals)
 - Words, morphological analysis,
 - Sequences: part of speech tagging (POS), named entity recognition (NER)
 - Syntactic parsing (phrase structure, dependencies)
- Applications (Practical end-user solutions, research)
 - Sentiment analysis, toxicity detection
 - Machine translation, summarization
 - Computational social science
 - Interpretability
 - Fairness and bias



Next class

Continuing Introduction section

Questions?