

Natural Language Processing

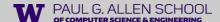
Introduction to NLP

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Announcements

https://courses.cs.washington.edu/courses/cse447/24au/

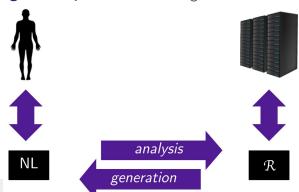
- HW0 is designed to help you setting up the environment. Not for submission
- HW1:
 - Text Classification and N-gram language models
 - Implementing Naive Bayes and Logistic Regression for text classification
 - Training, evaluating, and sampling from n-gram Language Models
 - We'll release HW 1 on Oct 7
 - We'll held an overview of the assignment in class after releasing the HW
 - HW 1 due is on Oct 28
 - Please read the assignment instructions early and ask us if you have any questions!
 - Please start early!
- Quiz 1:
 - Friday Oct 11
 - 5 questions, open during lecture time, 10-min in the beginning of the class
 - Materials from weeks 1 and 2
 - Introduction to NLP, introduction to text classification, NB
 - Instructions for HW 1

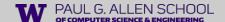
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What is Natural Language Processing (NLP)?

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





Personal assistants

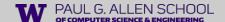




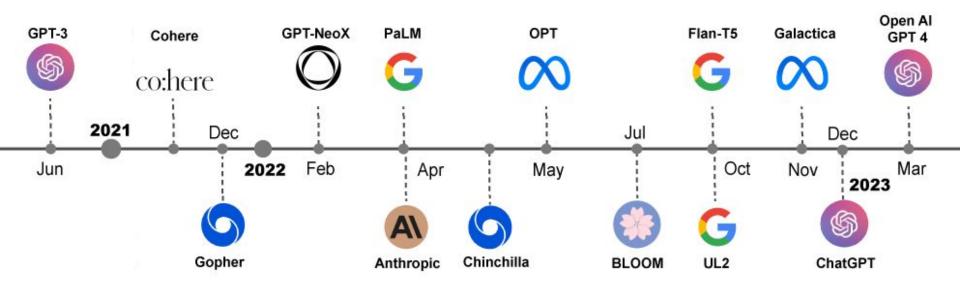








Large Language Models (LLMs)



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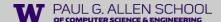


Natural Language Processing



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
 - Deeper semantic analysis: reasoning, knowledge, ect.
- Dialog processing
 - Discourse analysis, user adaptation, etc.
- Information retrieval
- Text to speech



Common tasks LLMs are trained to solve

Answering questions

Creative tasks

Virtual assistants

Content localization

Text generation

Educational applications

Communication

Text translation

Sentiment analysis

Conversational tasks

Writing assistance

Document summarization

Text classification

Intelligent chatbots

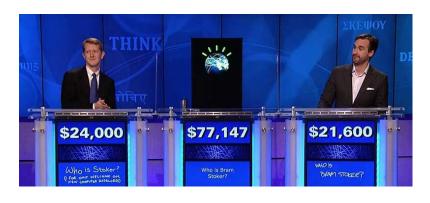
Data analysis Content categorization

Spam filtering

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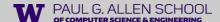
Question answering

- 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?

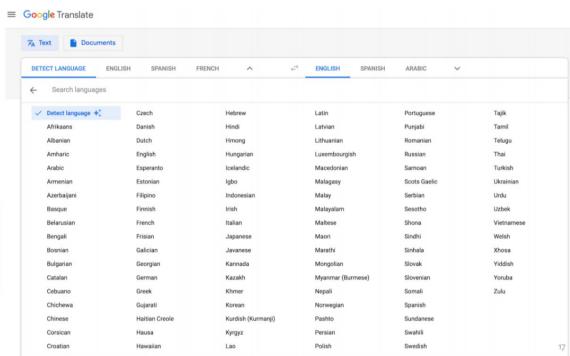




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Positive or negative movie review?



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 sh

Reviews

Summary - Based on 377 reviews

1 star	2	3	4 stars	5 stars
What peop	ole are	sayi	ng	
ease of us	e			"This was very easy to setup to four computers."
value				"Appreciate good quality at a fair price."
setup				"Overall pretty easy setup."
customer	service			"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."	

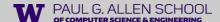
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Information extraction

DATE 1935 PERSON COUNTRY PERSON					
In 1933 , while Einstein was visiting the United States , Adolf Hitler came to power .					
RELIGION COUNTRY					
Because of his Jewish background , Einstein did not return to Germany .					
COUNTRY NATIONALITY 1940					
He settled in the United States and became an American citizen in 1940 .					
Einstein supported the Allied forces , but he generally denounced the idea of using nuclear fission as a weapon .					
PERSON NATIONALITY TITLE PERSON					
He signed the Russell Einstein Manifesto with British philosopher Bertrand Russell , which highlighted the danger of nuclear weapons .					
ORGANIZATION STATE_OR_PROVINCE 1955					
He was affiliated with the Institute for Advanced Study in Princeton , New Jersey , until his death in 1955 .					

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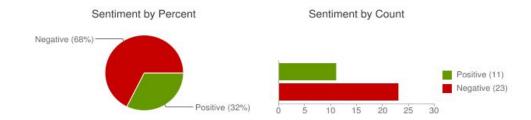


Sentiment analysis + information extraction

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

"united airlines" Search Save this search

Sentiment analysis for "united airlines"



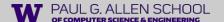
<u>iljacobson</u>: OMG... Could @**United airlines** have worse customer service? W8g now 15 minutes on hold 4 questions about a flight 2DAY that need a human.

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. http://t.co/Z9QloAjF Posted 2 hours ago

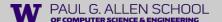
CountAdam: FANTASTIC customer service from **United Airlines** at XNA today. Is tweet more, but cell phones off now!

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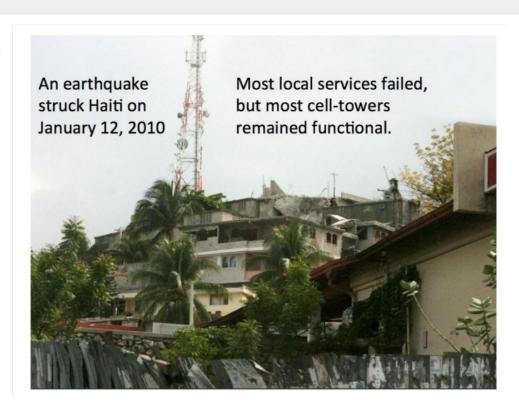


- Haiti Earthquake 2010
- About 3 million people were affected by the quake
- Classifying SMS messages





- 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 fè 24 è
- Fanm gen tranche pou fè yon pitit nan Delmas 3 I

- My family in Carrefour, 24 Cote Plage, 41A needs food and water
- People trapped in Sacred Heart Church, PauP
- General Hospital has less than 24 hrs. supplies
- Undergoing children delivery Delmas 3 I



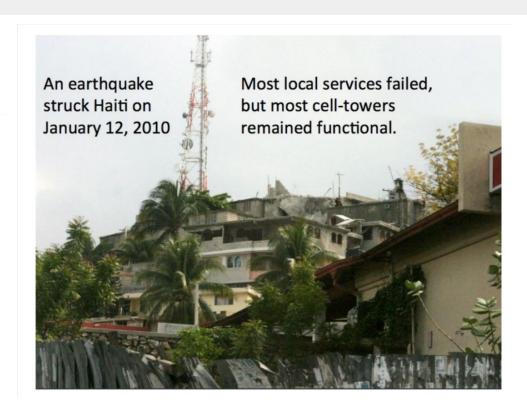
 Translation + information extraction

Lopital Sacre-Coeur ki nan vil Okap, pre pou li resevwa moun malad e lap mande pou moun ki malad yo ale la.

"Sacre-Coeur Hospital which located in this village of Okap is ready to receive those who are injured. Therefore, we are asking those who are sick to report to that hospital."



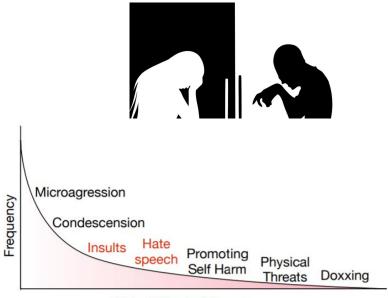






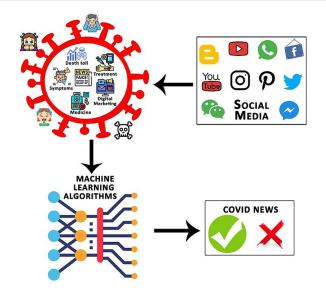
Hate speech detection





Risk of Physical Danger

Covid19 misinformation



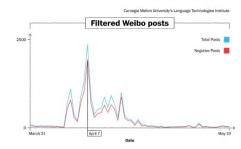
Detecting COVID-19-Related Fake News Using Feature Extraction

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-discrimination-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.

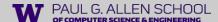
Based on automated sentiment analysis tools and manual analysis, the researchers said, they believed the majority of posts in their data set expressed negative sentiments relating to Africans or black people.



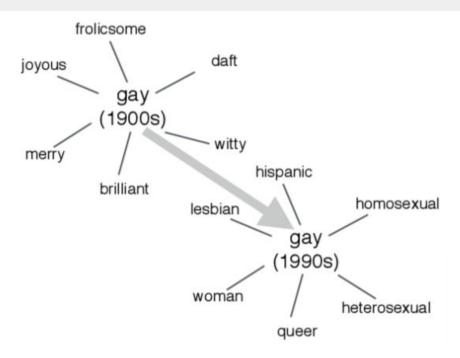
Researchers collected around 16,000 posts containing at least one Guangzhou location tag and one "African-related" keyword from late March through May. (Atthar Mirza/The Washington Post)

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 cartoon depiction officials the wing foreigners who warm? Abiding by

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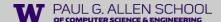
Language change



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

William L. Hamilton, J. Leskovec, Dan Jurafsky

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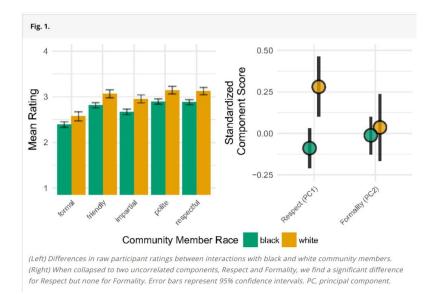


Computational social science

computational social science answering questions about society given observational data

example: "Do police officers speak with Black and White Americans in the same

way?"



Language from police body camera footage shows racial disparities in officer respect

Rob Vojgt 🖾 , Nicholas P., Camp, Vinodkumar Prabhakaran. 🕕 and Jennifer L. Eberhardt 🖾 Authors Info & Affiliations
Contributed by Jennifer L. Eberhardt, March 26, 2017 (sent for review February 14, 2017; reviewed by James Pennebaker and Tom Tyler)

June 5, 2017 114 (25) 6521-6526 https://doi.org/10.1073/pnas.1702413114

Factors changing the NLP landscape

Hirschberg and Manning (2015):

- Increases in computing power
- The rise of the web, then the social web
- Advances in machine learning
- Advances in understanding of language in social context

These days, also:

- Consumer and investor demand
- Emerging ethical questions around deployment

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Where are we now? - before 2022

Sentiment analysis still really hard Best roast chicken in San Francisco! mostly solved Question answering (QA) The waiter ignored us for 20 minutes. Q. How effective is ibuprofen in reducing Coreference resolution fever in patients with acute febrile illness? Spam detection Let's go to Agra! Carter told Mubarak he shouldn't run again. Paraphrase Buy V1AGRA ... Word sense disambiguation XYZ acquired ABC vesterday ABC has been taken over by XYZ I need new batteries for my mouse. Part-of-speech (POS) tagging ADJ NOUN VERB ADV Summarization Colorless green ideas sleep furiously. Parsing The Dow Jones is up Economy is The S&P500 jumped I can see Alcatraz from the window! good Housing prices rose Named entity recognition (NER) Machine translation (MT) PERSON 第13届上海国际电影节开幕... Dialog Where is Citizen Kane playing in SF? Einstein met with UN officials in Princeton The 13th Shanghai International Film Festival... Castro Theatre at 7:30. Do you want a ticket? Information extraction (IE) Party You're invited to our dinner May 27 party, Friday May 27 at 8:30 add

making good progress

Where are we now? - after 2022

ChatGPT 40 ~

what are the weaknesses of NLP today?

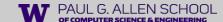


NLP has made great strides but still faces key challenges:

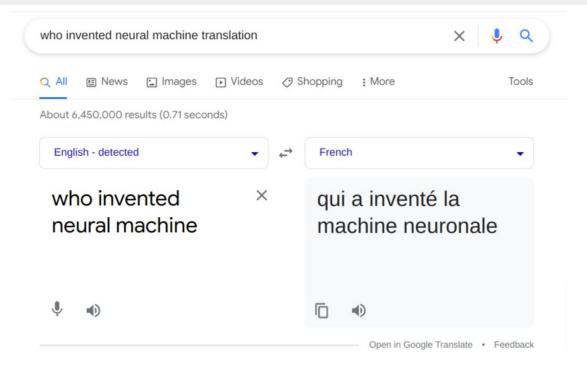
- Lack of Deep Understanding: Models often lack true contextual comprehension, struggling with nuanced reasoning and long-term dependencies in texts.
- Bias and Fairness: NLP systems inherit biases from training data, which can result in unfair outputs. Debiasing remains difficult without affecting performance.
- Data Quality: Models often rely on noisy or low-quality data from the web, and obtaining high-quality, domain-specific labeled data is expensive and time-consuming.
- Domain-Specific Limitations: General NLP models don't perform well in specialized domains like medicine or law without significant adaptation.
- Explainability and Reasoning: Most models lack explainability and struggle with logical reasoning or multi-step inferences, which is critical for high-stakes applications.
- Efficiency and Scalability: Large models are resource-intensive, requiring significant computational power for both training and inference, which limits accessibility.
- Multilingual Limitations: NLP models perform poorly in low-resource languages and codeswitching scenarios, leaving many languages underrepresented.
- Safety and Misinformation: Models can be vulnerable to adversarial attacks and can generate misinformation easily, posing ethical concerns.

These weaknesses highlight the need for improvements in understanding, fairness, domain adaptation, and safety in NLP systems.





Question answering



Retrieved Mar 25, 2022

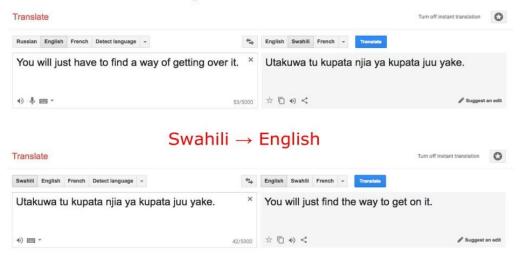
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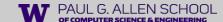


English → Swahili









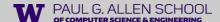
English → Swahili





Large Language Models





Large Language Models

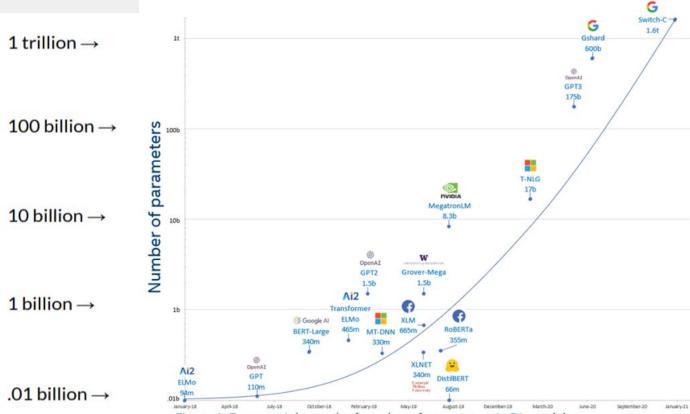


Figure 1: Exponential growth of number of parameters in DL models



GPT-3





Figure 1: Exponential growth of number of parameters in DL models

GPT-3



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.

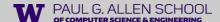
Q: Who was president of the United States before George W. Bush?

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

Q: Who won the World Series in 1995?

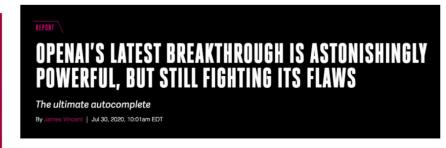
A: The Atlanta Braves won the World Series in 1995.

https://lacker.io/ai/2020/07/06/giving-gpt-3-a-turing-test.html



Chat bots





- 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

https://lacker.io/ai/2020/07/06/giving-gpt-3-a-turing-test.html



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.

https://lacker.io/ai/2020/07/06/qiving-gpt-3-a-turing-test.html

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.

https://lacker.io/ai/2020/07/06/giving-gpt-3-a-turing-test.html

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.

https://lacker.io/ai/2020/07/06/giving-gpt-3-a-turing-test.html

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:

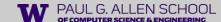
```
Q: Write one line of Ruby code to reverse an array.
A: ary.reverse

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

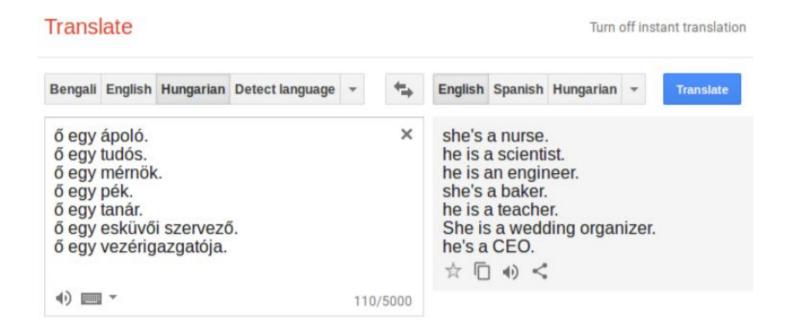
Q: Reverse the following array: [1, 3, 5, 6, 10, 4, 2, 77]
A: [10, 6, 4, 2, 77, 3, 5, 1]

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

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Bias in machine translation



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

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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)







Al 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 leshians
- The developer has apologised over the remarks, saying they 'do not represent our values as a company'



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

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



Next class

Continuing Introduction section

Questions?

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