NLP

NLP: Natural Language Processing

-technology used to process, analyze, and create natural language

-subfield of AI concerning interactions between computers and human language

-Google Translate

-chatboxes (think customer service!)

-Amazon Echo

-autocorrect/Grammarly

NLP: Various Languages

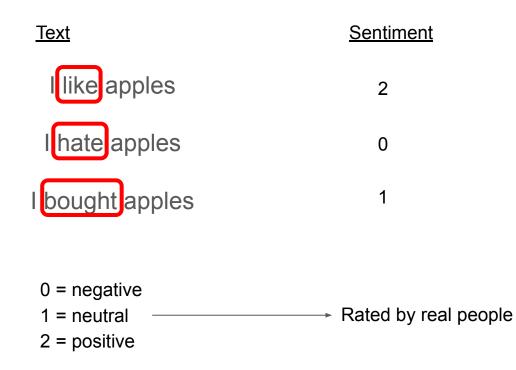
-NLP is not only limited to English

-some of its technology differs across languages that use different structures

-ex: character-based languages like Chinese/Japanese vs alphabet languages of English/French

-ethical concern: NLP techniques developed mainly for English

Tech Behind NLP



Step 1: Convert data into frequencies

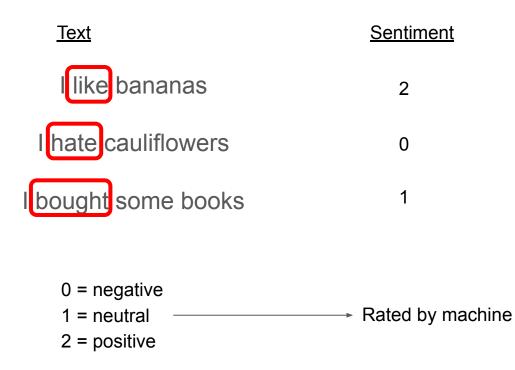
Step 2: Create likelihood table with data

Weather	Play
Sunny	No
Overcast	Yes
Rainy	Yes
Sunny	Yes
Sunny	Yes
Overcast	Yes
Rainy	No
Rainy	No
Sunny	Yes
Rainy	Yes
Sunny	No
Overcast	Yes
Overcast	Yes
Rainy	No

Frequ	ency Tabl	e
Weather	No	Yes
Overcast		4
Rainy	3	2
Sunny	2	3
Grand Total	5	9

Like	elihood tab	le	1	
Weather	No	Yes		
Overcast		4	=4/14	0.29
Rainy	3	2	=5/14	0.36
Sunny	2	3	=5/14	0.36
All	5	9		
	=5/14	=9/14	1	
	0.36	0.64		

Step 3: Use data to figure out sentiment



Naive Bayes is mostly used for:

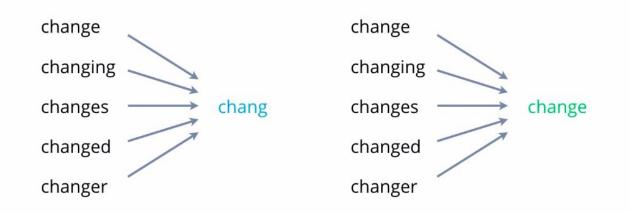
- 1. Text classification/ Spam Filtering/ Sentiment Analysis
- 2. Recommendation System

However, it is also a great example of how machines process words.

Language is complex, how do we simplify it for computers?

Stemming and Lemmatization

Stemming vs Lemmatization



Stemming and Lemmatization

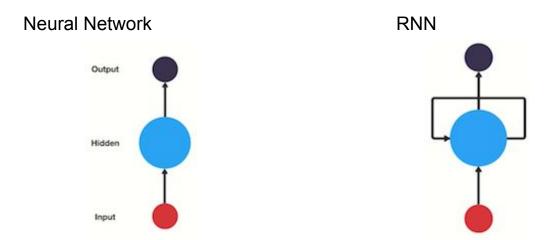
Stemming is the process of reducing a word to its stem or root format.

In lemmatization, the transformation uses a <u>dictionary</u> to map different variants of a word back to its root format.

Stemming is <u>easier and faster</u>, whereas lemmatization is more <u>accurate and</u> <u>preserves context</u> \rightarrow hardly vs hard is a case only lemmatization can process



RNN, aka Recurrent Neural Networks, is a slight twist form the normal neural network.





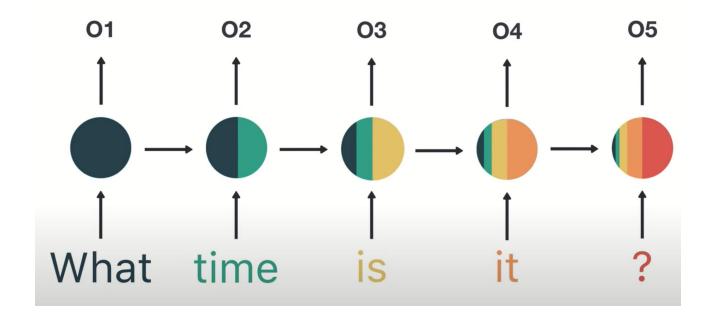
Special thing about RNN is that instead of making all the words numbers like Naive Bayes, it tries to use <u>sequential memory</u>.

Sequential Memory:

ABCDEFGHIJKLMNOPQRSTUVWXYZ ZYXWVUTSRQPONMLKJIHGFEDCBA NOPQRSTUVWXYZABCDEFGHIJKLM

RNN – Chatbot Example

Let's say you ask this question, it will process each word by part like such:



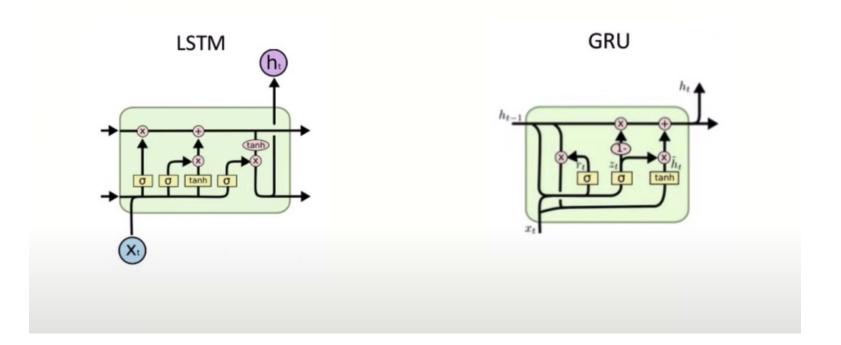
RNN

However, RNN has limited memory, therefore the gradient. As you can see, the

first few words are basically out of the system now.

The lack of context is why you get funny results sometimes.

LSTM and GRU



https://youtu.be/8HyCNIVRbSU?t=180

Watch until 8:30

https://youtu.be/8HyCNIVRbSU?t=581

LSTM and GRU

The video touches on a lot of technical terms, so it is fine if you didn't understand most of it. However, here are some key takeaways:

- 1. LSTM and GRU has better memory and therefore understand context a bit better
- 2. GRU is a simplified version of LSTM but works almost as well
- 3. Both LSTM and GRU are expensive and takes a long time to train

Speaking of context...

As you can tell, A.I., ML, or DL all depend on converting data into numbers, and

despite the complexity of these machines, the context rarely gets carried over. The

lack of understanding often leads to bias and can cause massive damage if it goes undetected.

Demonstration!

https://gpt3demo.com/apps/openai-gpt-3-playground

Discussion Questions: