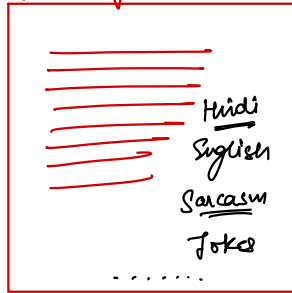


pdf / image ...



Script of
a movie

My Cat ate the Cake.
It is bad.

→ I am sitting on a bank.
→ I am sitting in a bank.

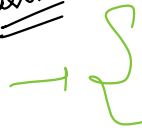
Foundational ML models

→ LLMs
→ translate language
→ analyse sentiment
→ chatbots conversations

Capable of understanding
→ complex data.
→ Identify the relationship b/w words
→ generate new text → coherent
→ grammatically correct

→ Autocomplete a LLM?] NO

Earlier



Once upon a single word

predict the prob of

↳ language model

time → 50%
dream → 15%
reaction → 4%

Human language → bigger

LLM → predict the prob sentence / paragraph / doc

My → name

My name → is

My name is →

is, am, an, the
stop words X

NLP



⊛ Break the text → tokenization



⑥ Understand → Stemming & Lemmatization
⑦ Embedding

~~The~~ Taj Mahal ~~is~~ beautiful

Tokenization

Hello, How are you?

"Hello", "I", "How", "are", "you"

} word tokenization

Don't waste food

English

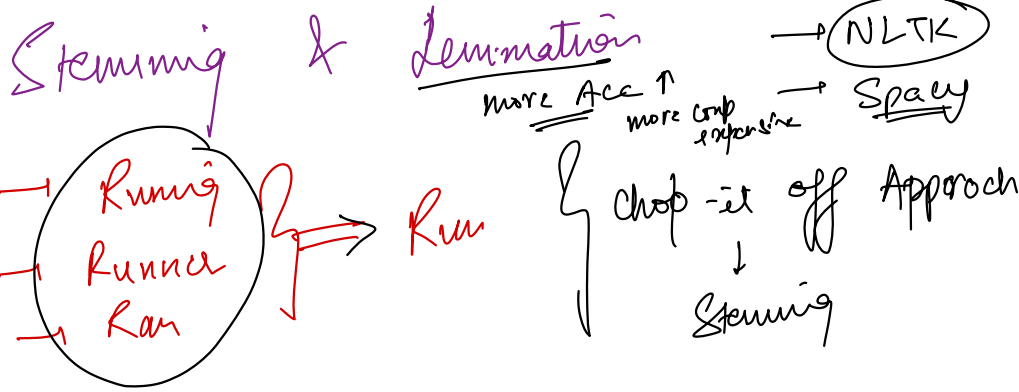
ドント
ウェイスト
フード

Chinese or Japanese

["D", "o", "n't", "waste", "food"]

["D", "o", "n", "'", "t"]

} Sub words
} Character tokenization



Lemmatization
Semantic Connections

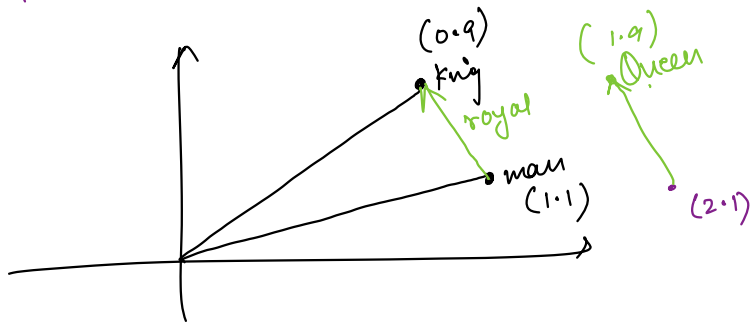
change
changing
changes
changed
change

Change

change

Embeddings

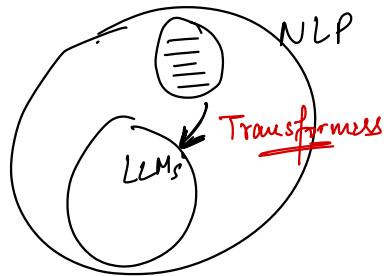
Semantic relation b/w the words



✓ King - man = ✓

King - man + woman = ✓

"word2vec"



Sequential processing

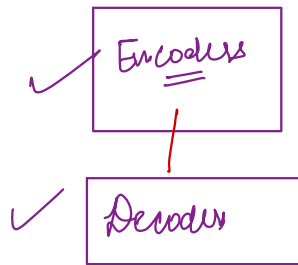
↳ connections b/w words

More context

Sequence ✓✓

Encoder & decoder architecture

Auto-encoders

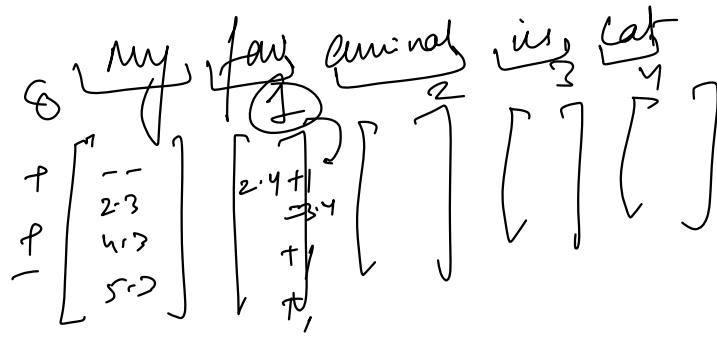


NN layers that process input sequence

Embedding

output

Token embedding + positional embedding



Learning in a transformer

① Encoder learning] (*) Compress the input info into compact representation

- (*) Identifies info features
- (*) Disregards the irrelevant & redundant info.
- (*) Create a condensed summary

② Decoder learning

⊛ reconstruct from encoder's compact rep.

Training

- Under compact rep — loss f...
- learn patterns / rel within data
- generate the output (similar to input)

Scoring process

- ⊛ Encoder → rep state
- ⊛ Decoder — generate output
- ⊛ output compared expected output using the loss fctn.

Autoencoders → Remove noise

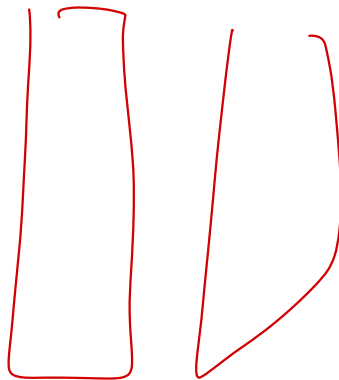
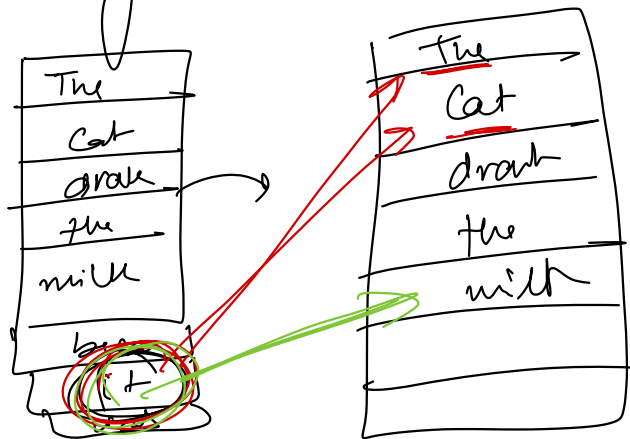


Attention

I played with the bat
The bat flew in the night

+ The cat drank the milk because it was hungry.
it sweet.

Self-Attention



Queries = question you are seeking Library
Key = potential match to your query found in each book.
values = actual piece of info \rightarrow when key match your query.

\rightarrow The Cat sat on a mat.

Step 1 for each word in the sentence, model with create a query, Key, value

"Cat" \rightarrow query \Rightarrow looking for actions association

Step 2 Summary
 \hookrightarrow how related the words are.

Step 3 weighted Summary

