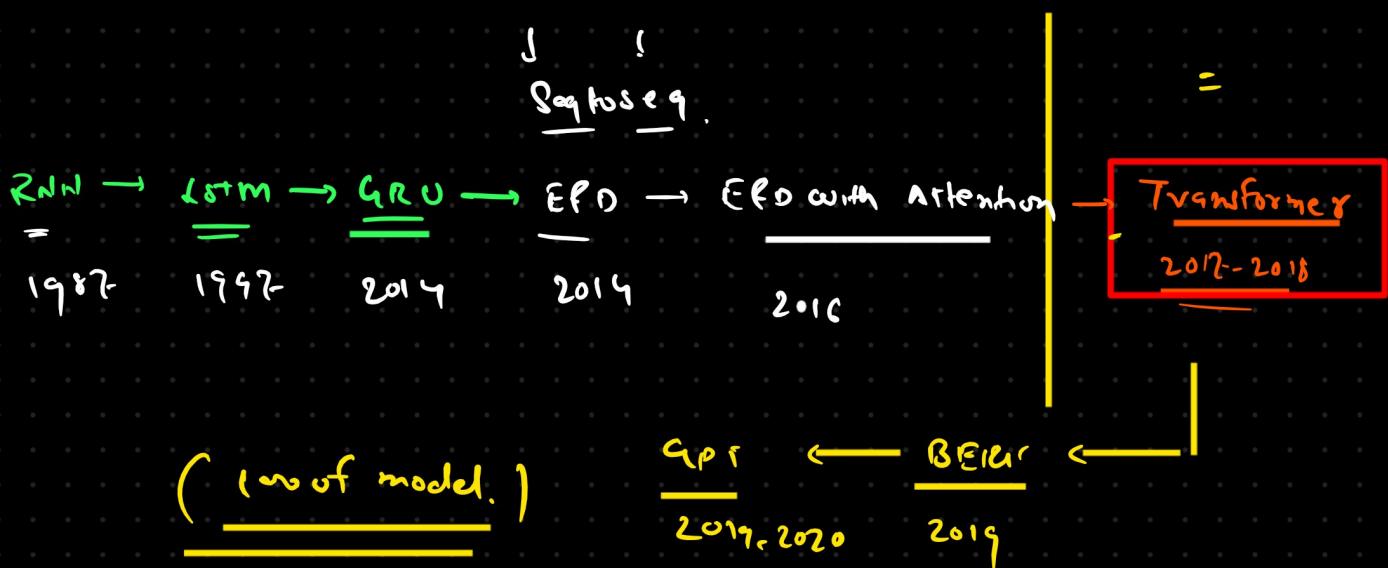
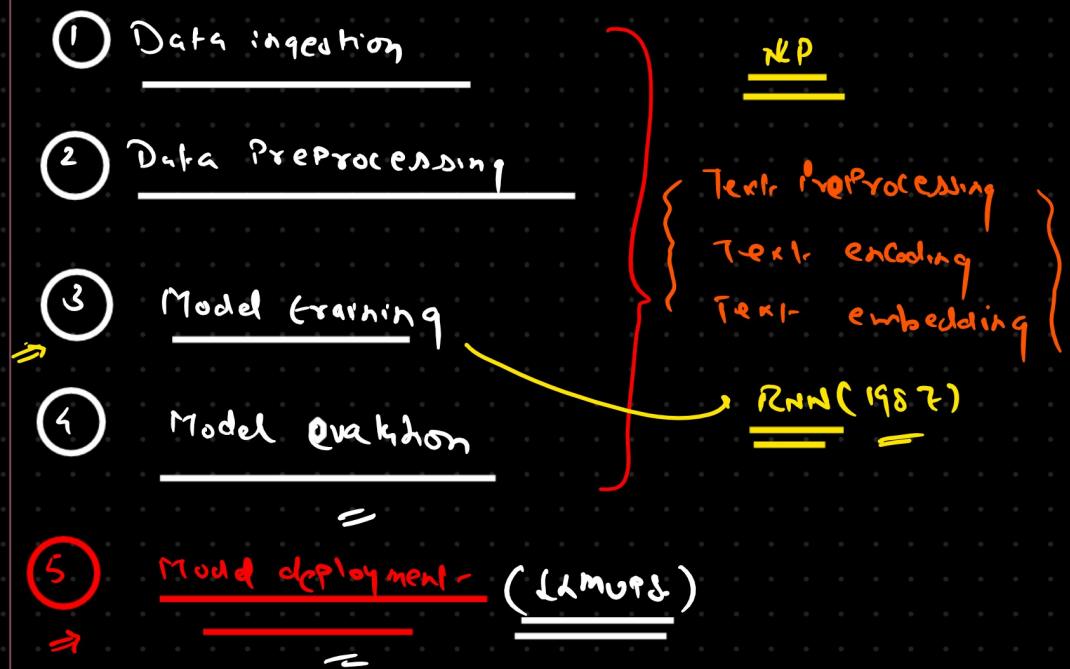
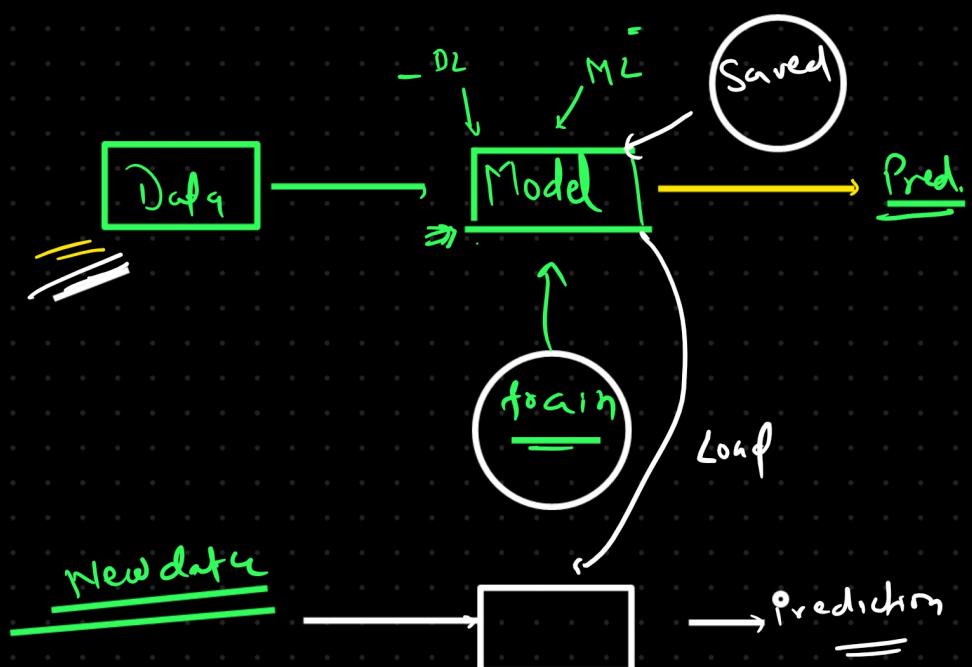
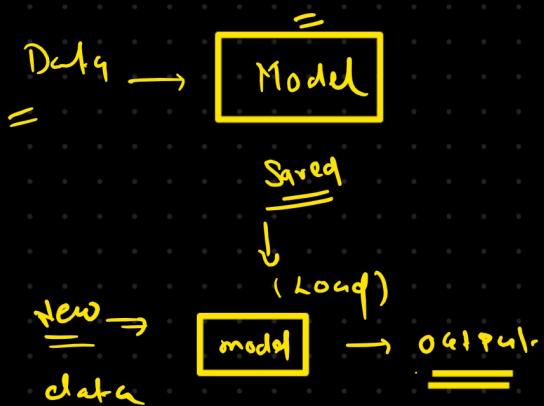


- ① RNN / LSTM / GRU
  - ② Encoder & Decoder
  - ③ Encoder & Decoder with attention
  - = ④ Transformer  $\Rightarrow$  3,4 }  $\Rightarrow$  Bert, GPT, ...  
- ⑤ Pytorch
  - ⑥ Inferencing with LLM
- RAG, Agent, LLMops



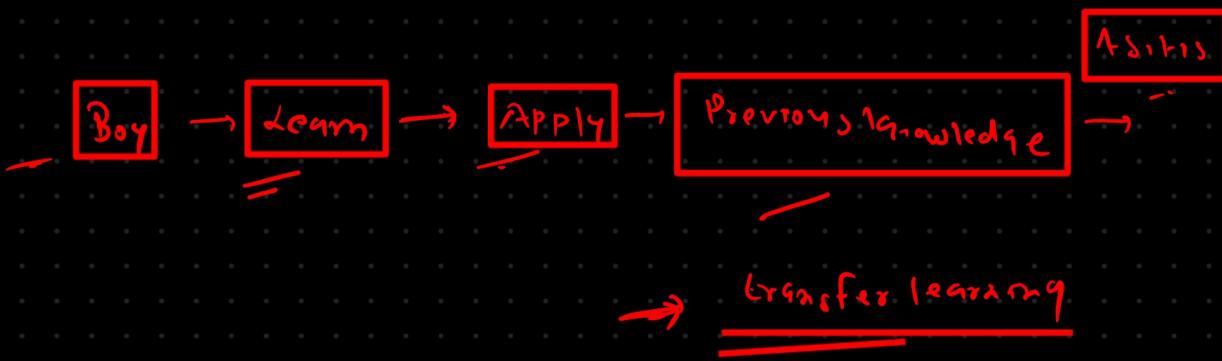
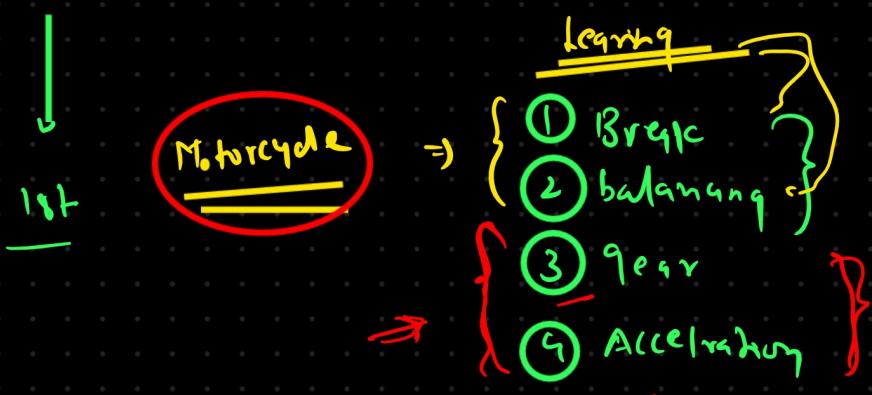
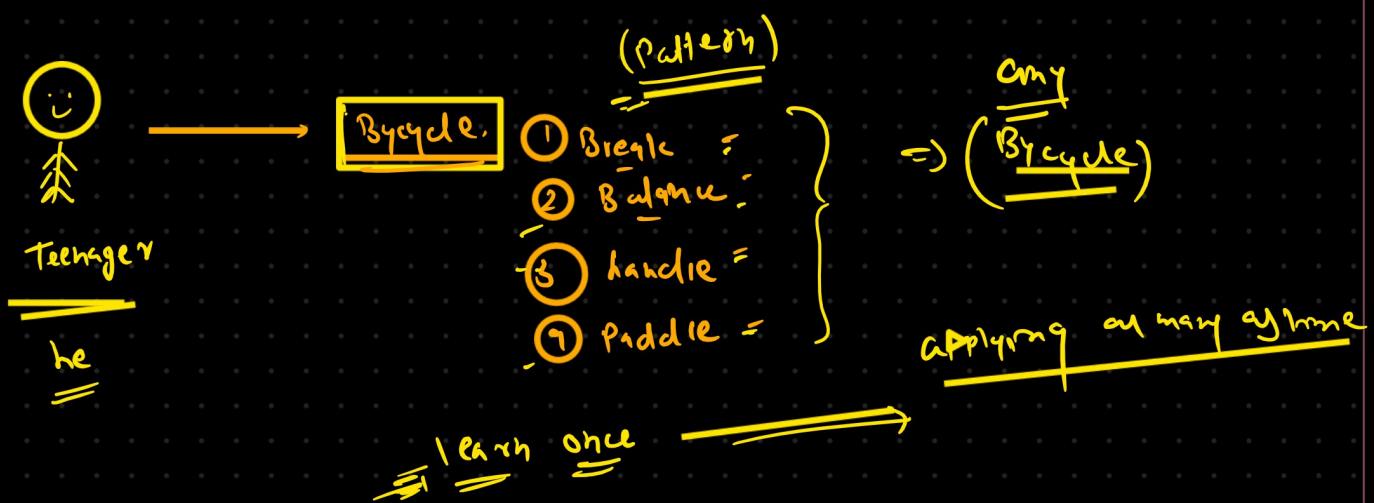
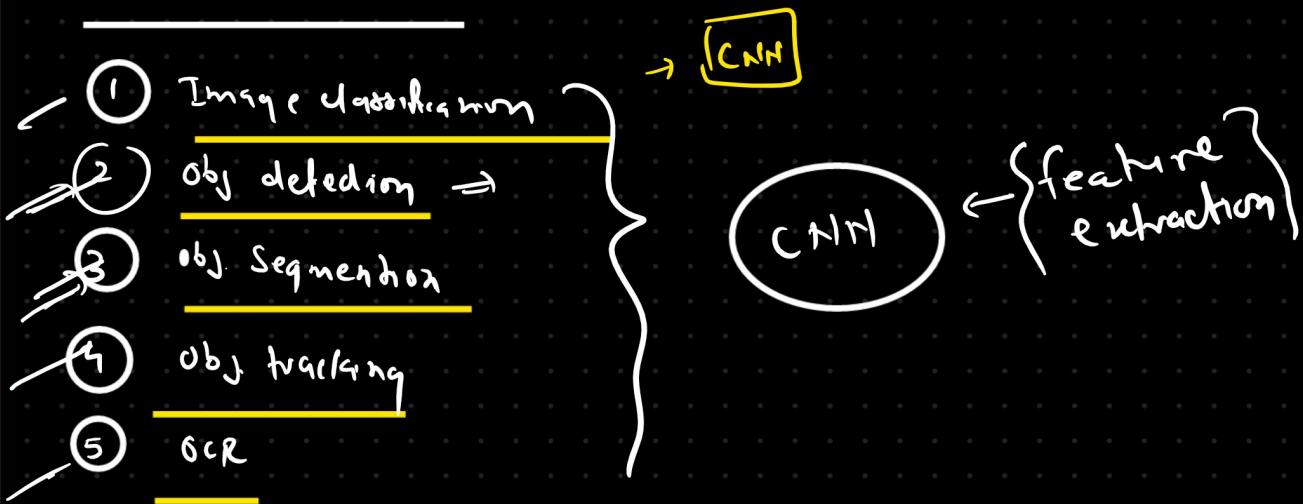
- 1 Why TL was not possible in NLP?
- 2 Gimp research Paper
- 3 Quadratic Transformer
- 4 Self Attention

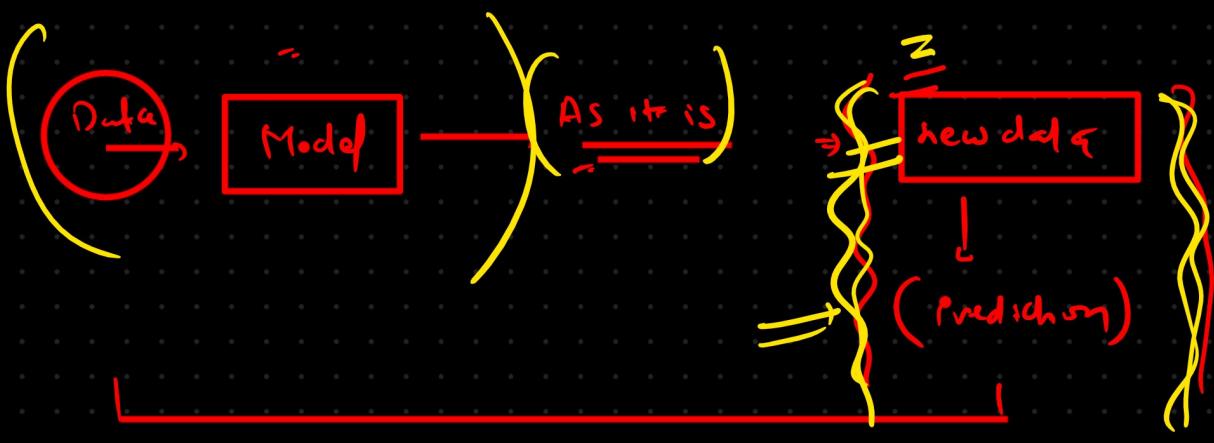
## Transfer Learning





## Computer vision

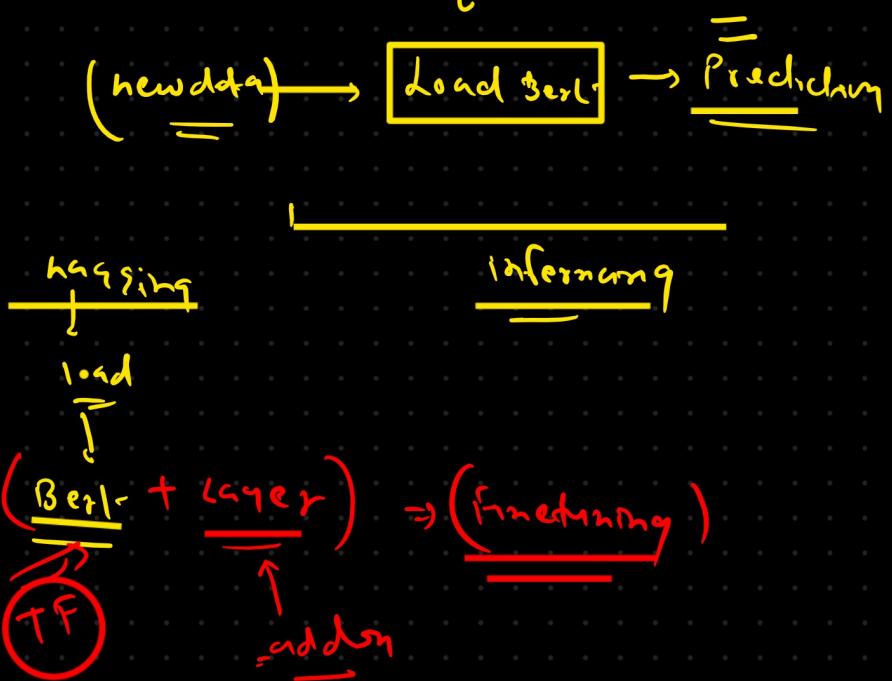




① transfer Learning  
(Inferencing)

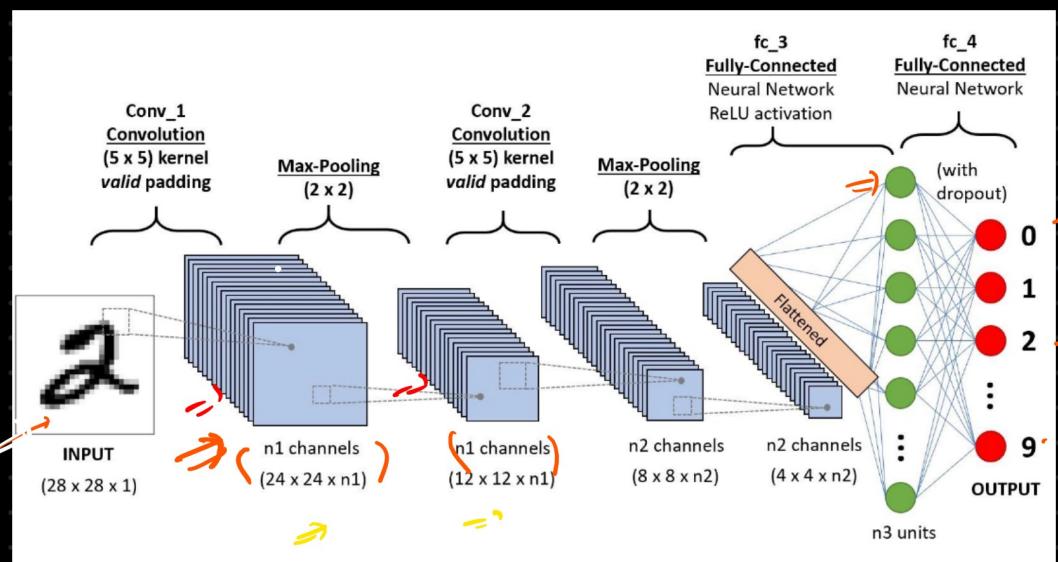
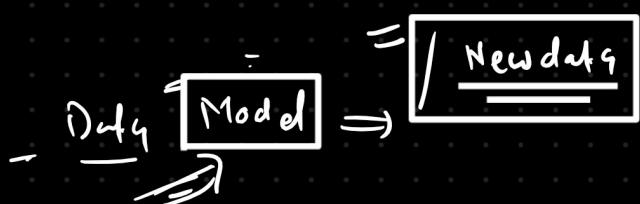
Data  
→ Model ← training

② Finetuning

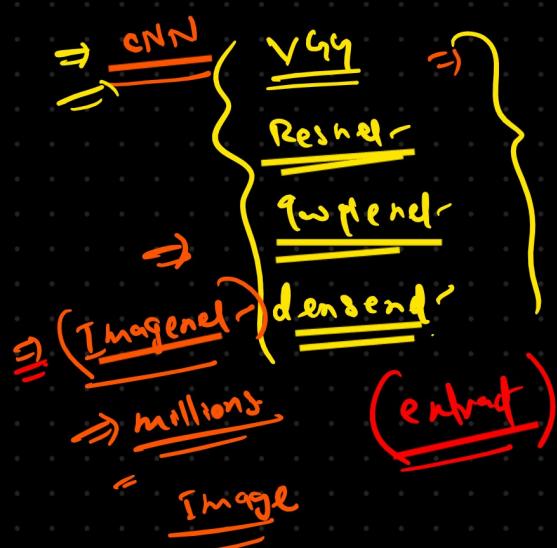


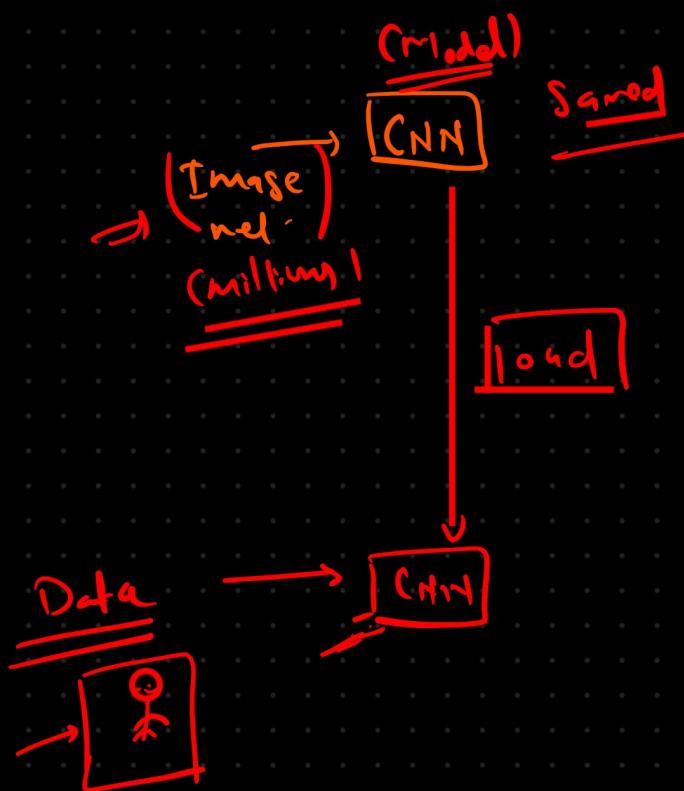
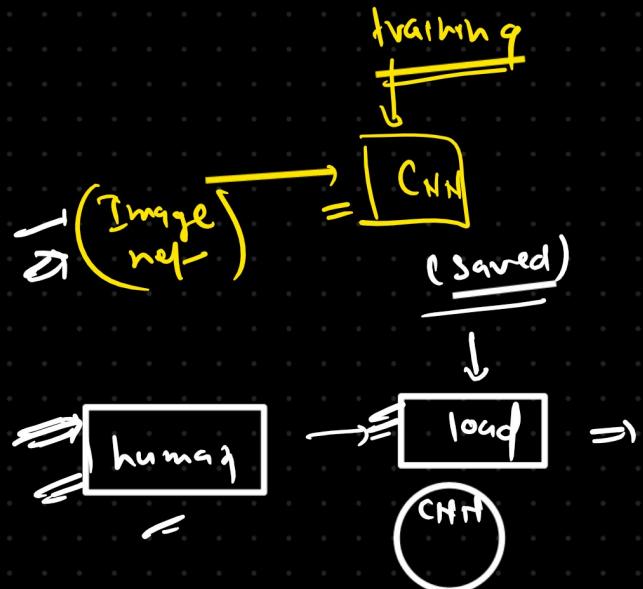
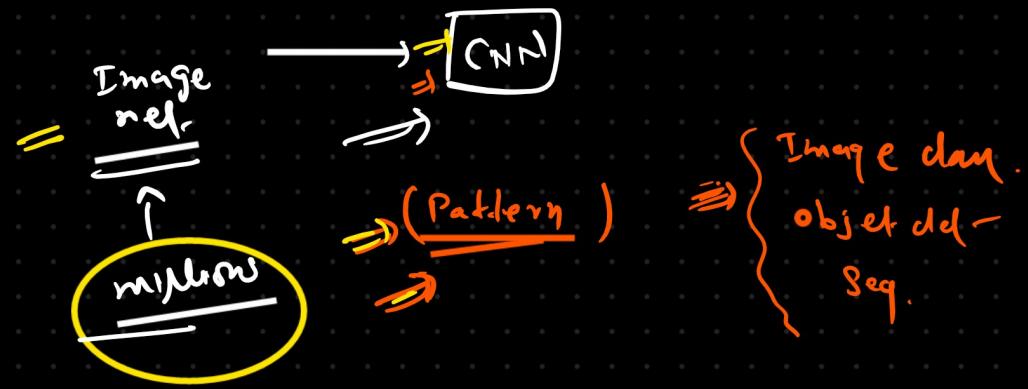
# Google word2vec

↓  
16

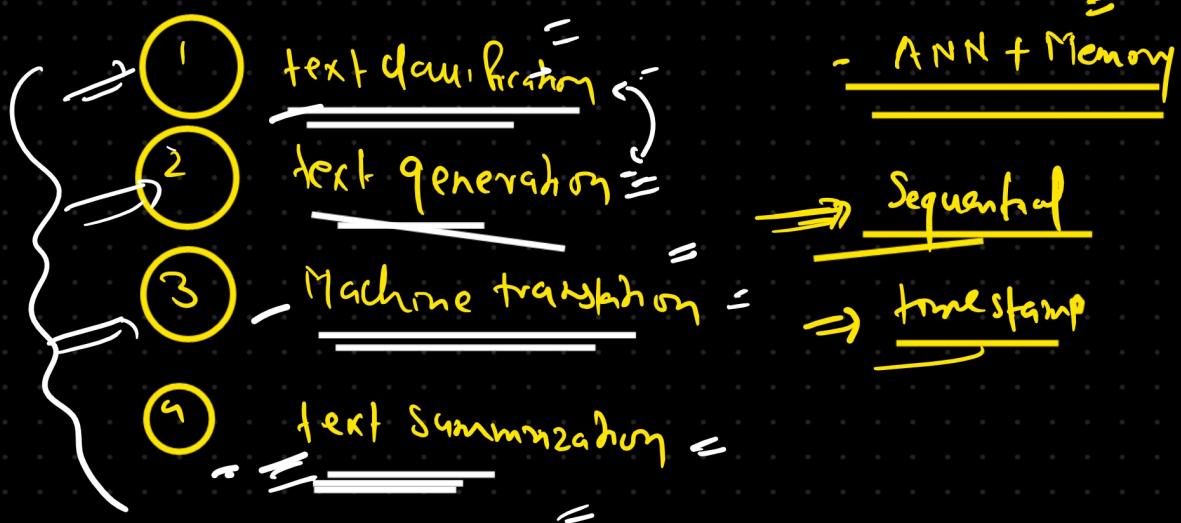


10





NLP = RNN, LSTM, GRU ← ~~lot's of~~

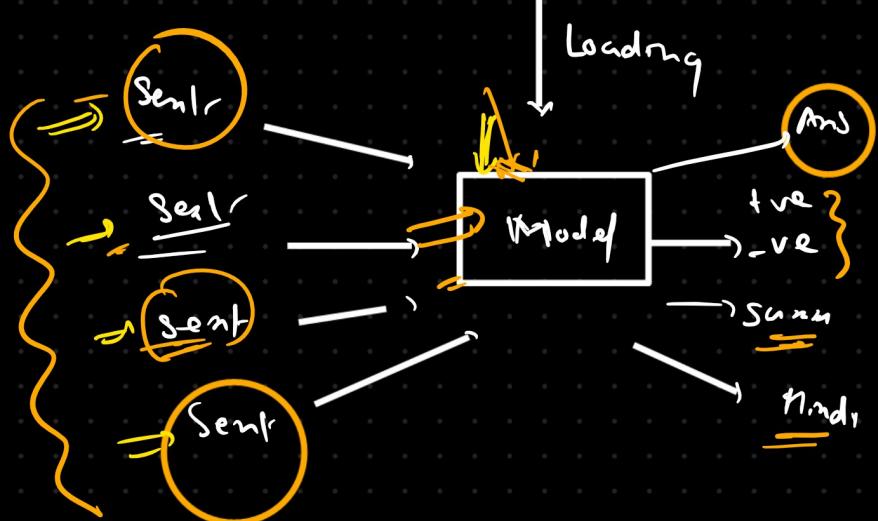
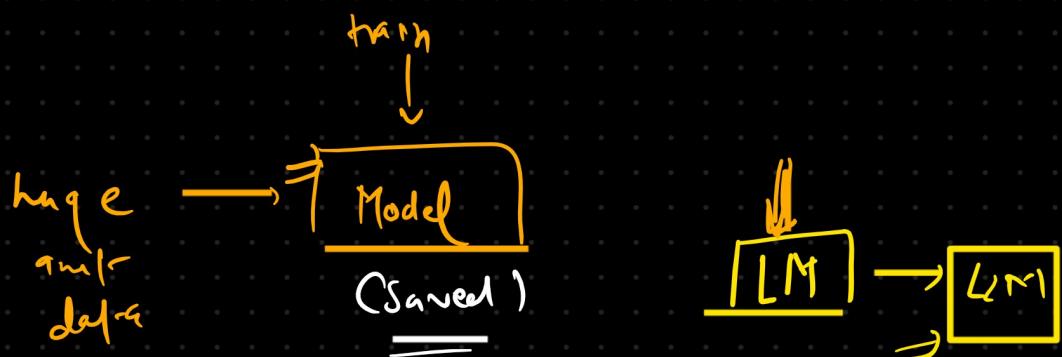
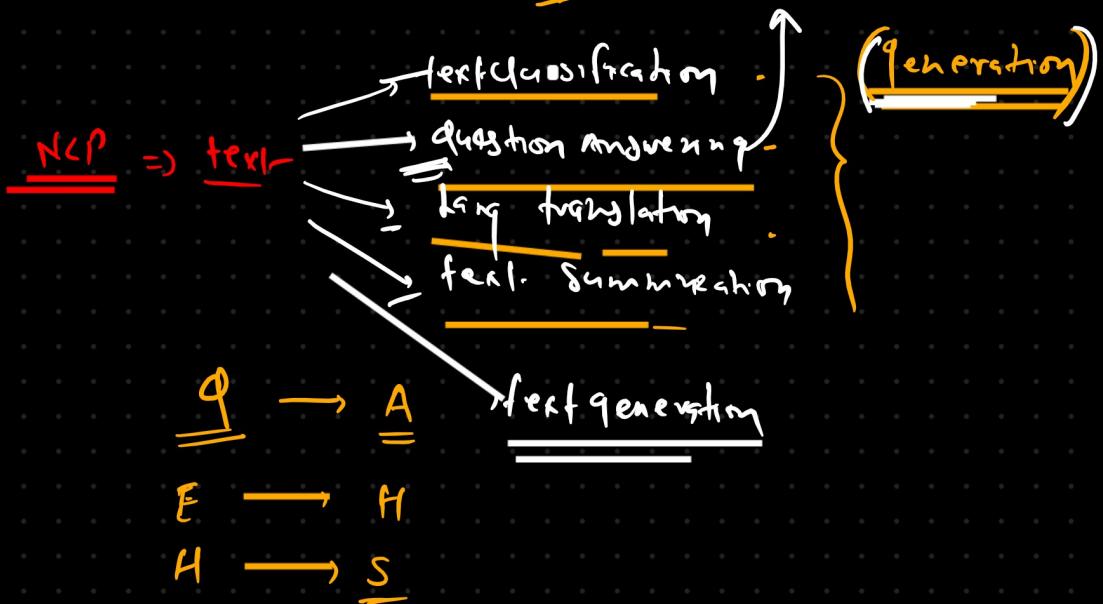


- ⇒ 1 large am. data X  
⇒ 2 task exclusiveness

## Transformer

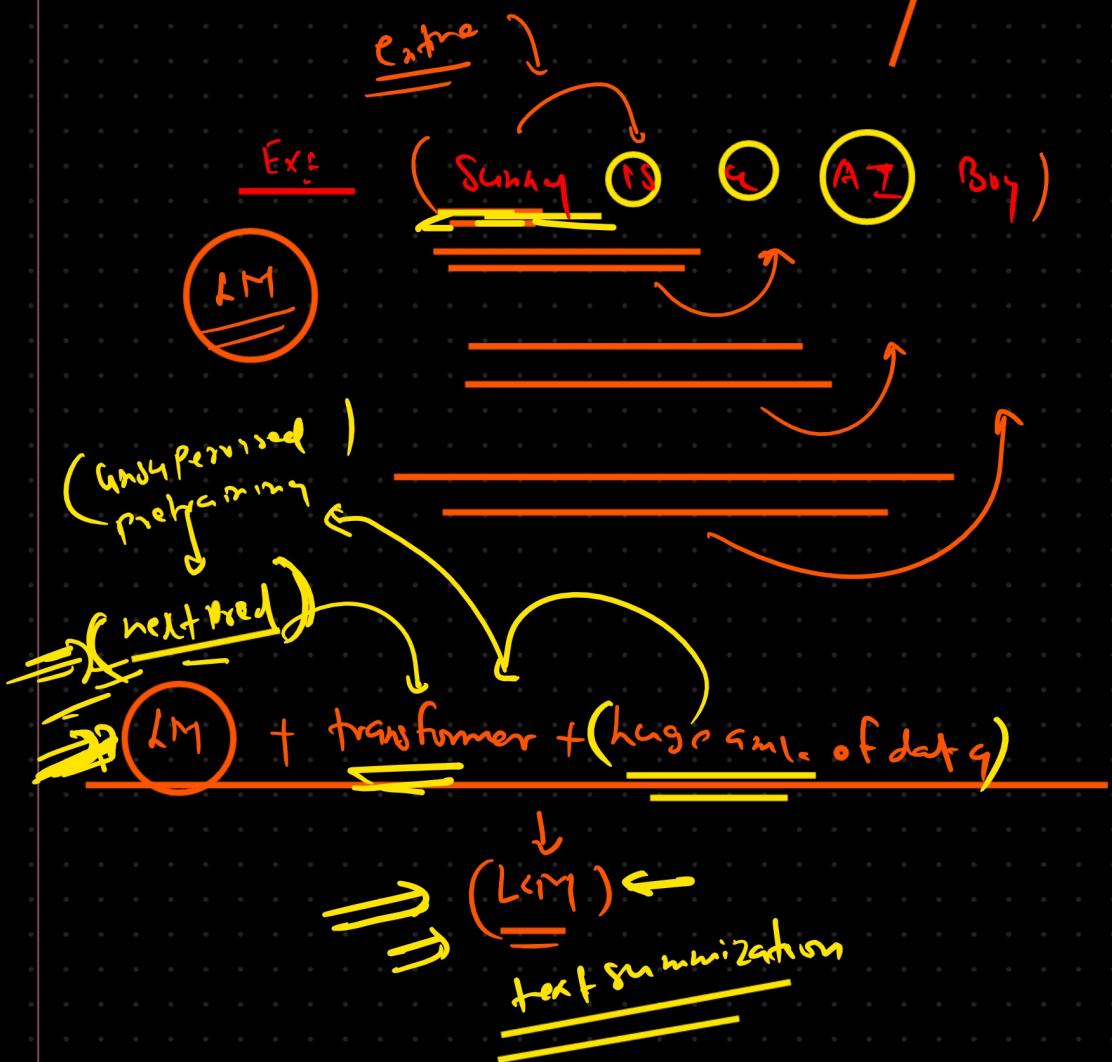
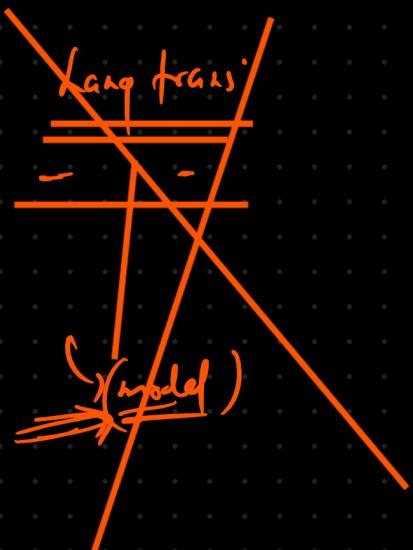
- 1 RNN, LSTM, GRU ✗
- 2 Parallel data ✓ ↘ ↙
- 3 huge contr. ✗
- 4 Training extensive ✓

= Conversational AI



Nextword Pred

Date

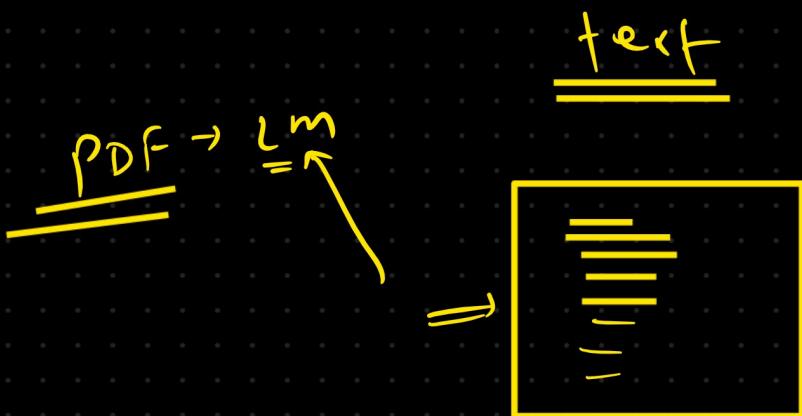
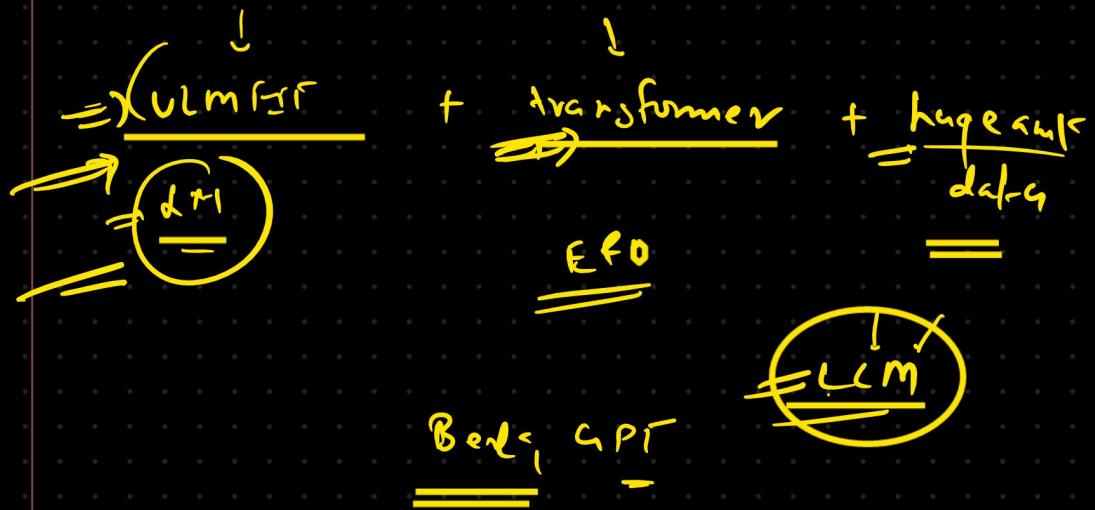


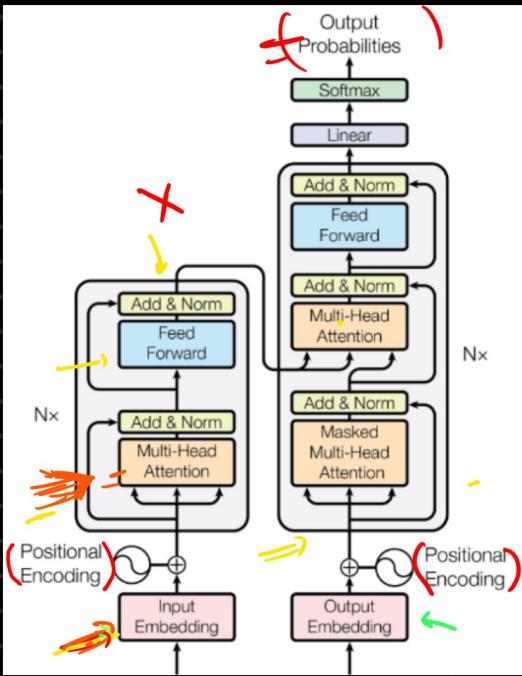
RNN, LSTM, GRU  $\times$

EFLD, EFLD with attention

transfer learning  
finetuning

$\hookrightarrow$  huge amf-





- 1 Positional encoding
- 2 Self Attention / Multihed Self attention
- 3 Encoder working
- 4 Decoder working
- 5 Data flow in transformer

$\Rightarrow \frac{\text{sunny}}{t=1} \frac{\text{is a}}{t=2} \frac{\text{teacher}}{t=3} \frac{\text{}}{t=4}$   $\stackrel{9,4000}{\equiv}$

$t=1$

sent  
 $\Rightarrow \frac{\text{River}}{=}$   $\frac{\text{book}}{=}$   $\frac{\text{flow}}{=}$  (Parallel)  
 $\Rightarrow \frac{\text{Positional}}{=}$   $\frac{\text{encoding}}{=}$   $\frac{\text{layer}}{=}$  8  
 $\Rightarrow \frac{\text{Self}}{=}$   $\frac{\text{attention}}$  layer (Multihead)  
 $\Rightarrow \frac{\text{ANN}}{=}$

$\rightarrow$  BERT  $\rightarrow$  Encoder  
 $\rightarrow$  GPT  $\rightarrow$  Decoder

Seq.  $\rightarrow$  NLP model.  
 $\Rightarrow$  (text) HFE...  
 $\Rightarrow$  (vector)

$\Rightarrow$  vectorization  
 $\Rightarrow$  (embedding) (ANN)  
 $\Rightarrow$  (word2vec) { Skipgram } Cbow  
 $\Rightarrow$  encoding (freq)  
 ① OHE  
 ② BOW  $\rightarrow$  n-grams  
 ③ TFIDF

$\Rightarrow$  (Self attention) -  $Q, K, V$  - value  
Contextual embedding  
dynamic embedding  
query key

## Word embedding

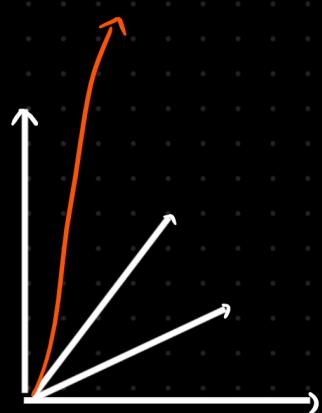
5-dim

Similar { King → [0.8, 0.2, 0.1, 0.05, 0.5] ← vector (word2vec)

King → [0.8, 0.2, 0.1, 0.05, 0.5]

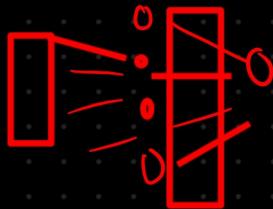
green → [ ]

Sunny → [ ]



- 1 An apple a day keeps the doctor away ↵
- 2 Apple is healthy ↵
- 3 Apple is better than banana ↵
- 4 Apple makes great mobiles ↵

1000 - 10000 ⇒ 3000 F  
and ↑  
Apple → [fruit, tech]  
q.v. 0.85  
1.00 ↵ apple [0.85, 0.15] ↵



(Model word2vec)

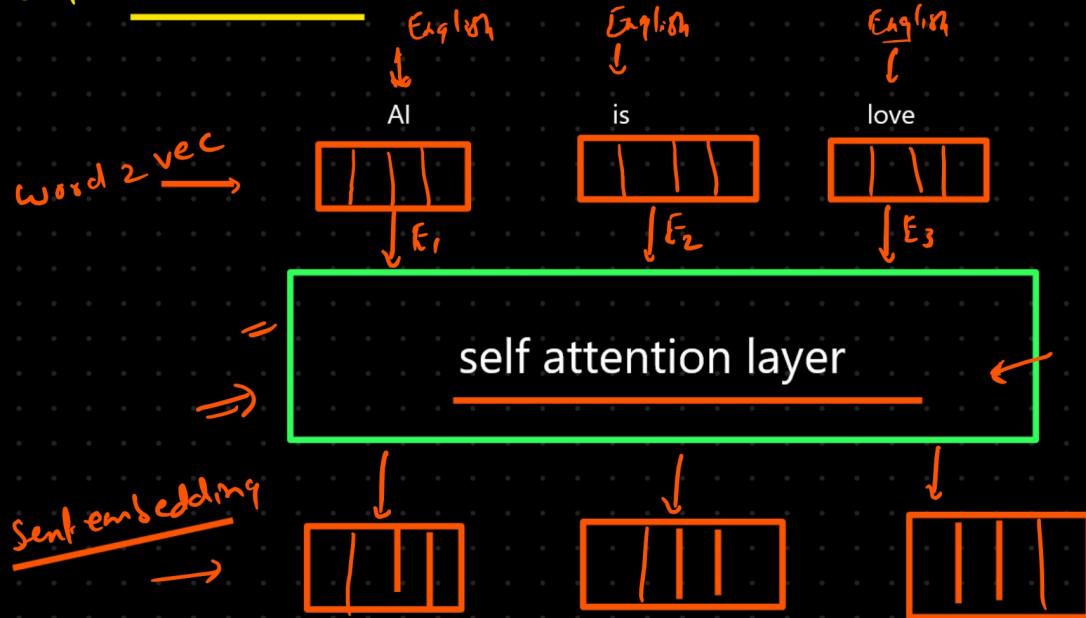
New Sent ↵ Apple [0.85, 0.15] ← but fine  
Launch a new phone while I was  
eating apple ↵  
[0.85, 0.15] ← fine

Contentful embedding & dynamic embedding (self attention)

BANK → ↵ Money Bank  
Bank → [0.5, 0.6] ⇒ [0.5, 0.6]

River Bank  
[0.5, 0.6]

Sent. → AI is love



Translation

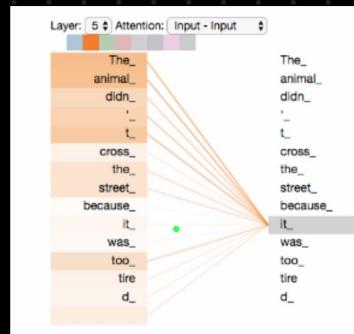
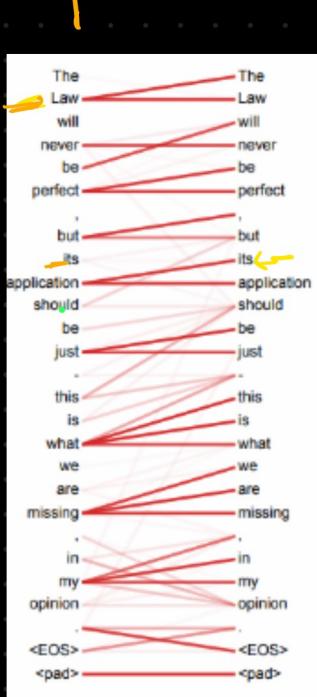
T.V. →  $\overline{\text{ai}}$  chs  $\overline{\text{st}}$

Attention X

(Self attention)

T.V.  
will  
turn  
on  
the  
out  
at  
it  
T.V.

{ Turn      Turn  
on      on  
the      the  
light      light



## Example



$$s_1 = \boxed{\text{bank}} \rightarrow \underline{\omega_1 * \text{money} + \omega_2 * \text{bank} + \omega_3 * \text{grow}}$$

$\downarrow$

$\boxed{\{Q, K, V\}}$



$$s_2 = \boxed{\text{bank}} \rightarrow \underline{\omega_1 * \text{River} + \omega_2 * \text{bank} + \omega_3 * \text{flow}}$$

(Random weights)

self-attention ( $Q, K, V$ )  $\rightarrow (\text{query}), (\text{key}), (\text{value})$   
= (Learning)

$$\left\{ \begin{array}{l} s_1 \rightarrow \underline{\text{bank}} \rightarrow 0.2 * \underline{\text{money}} + 0.7 * \underline{\text{bank}} + 0.1 * \underline{\text{grow}} \\ s_2 \rightarrow \underline{\text{bank}} \rightarrow 0.5 * \underline{\text{River}} + 0.3 * \underline{\text{bank}} + 0.1 * \underline{\text{flow}} \end{array} \right\}$$

$s_1 \quad (Q, K, V) = 1 \quad (\text{Random weight})$

$$\left\{ \begin{array}{l} \text{Money} = 0.7 * \underline{\text{money}} + 0.2 * \underline{\text{bank}} + 0.1 * \underline{\text{grow}} \\ \text{bank} = 0.25 * \underline{\text{money}} + 0.7 * \underline{\text{bank}} + 0.05 * \underline{\text{grow}} \\ \text{grow} = 0.1 * \underline{\text{money}} + 0.2 * \underline{\text{bank}} + 0.7 * \underline{\text{grow}} \end{array} \right\}$$

$s_2$

$$\left\{ \begin{array}{l} \text{River} = 0.8 * \underline{\text{River}} + 0.1 * \underline{\text{bank}} + 0.05 * \underline{\text{flow}} \\ \text{bank} = 0.25 * \underline{\text{River}} + 0.7 * \underline{\text{bank}} + 0.02 * \underline{\text{flow}} \\ \text{flow} = 0.4 * \underline{\text{River}} + 0.01 * \underline{\text{bank}} + 0.01 * \underline{\text{flow}} \end{array} \right\}$$

$\Rightarrow Q, K, V \Rightarrow \text{contextual vector}$

general      specific context vector

Q, K, V  $\rightsquigarrow$  weight  $\Rightarrow$  Learnable

$\Rightarrow$  Positional

$\Rightarrow$  Encoded

(decoder)

$\Rightarrow$  Semantic grid  
Open AI  
Clustering  
Hierarchical  
APL

quantize

$\text{JAN} \rightarrow (\text{finetuning})$

hassing rule

(offline training)

long chain

## Klausunden

