

transformers and beyond

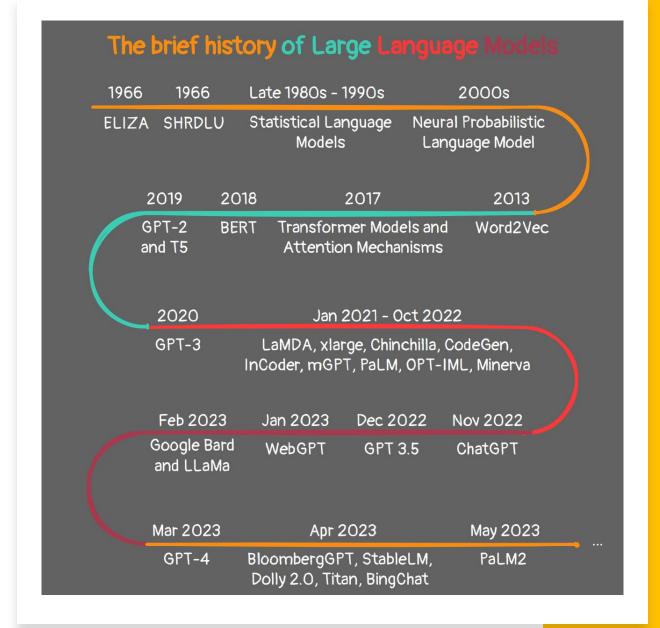
GESIS Fall Seminar 2023

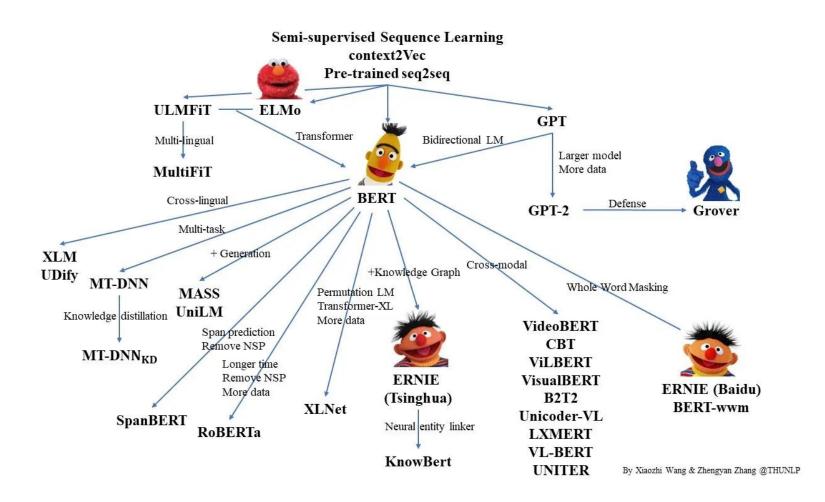
"From Embeddings to Transformers: Advanced Text Analysis in Python"

[day 3, afternoon: attention]

some history

[The brief history of Large Language Models: A Journey from ELIZA to GPT-4 and Google Bard | by Armin Norouzi, Ph.D | Level Up Coding (gitconnected.com)]



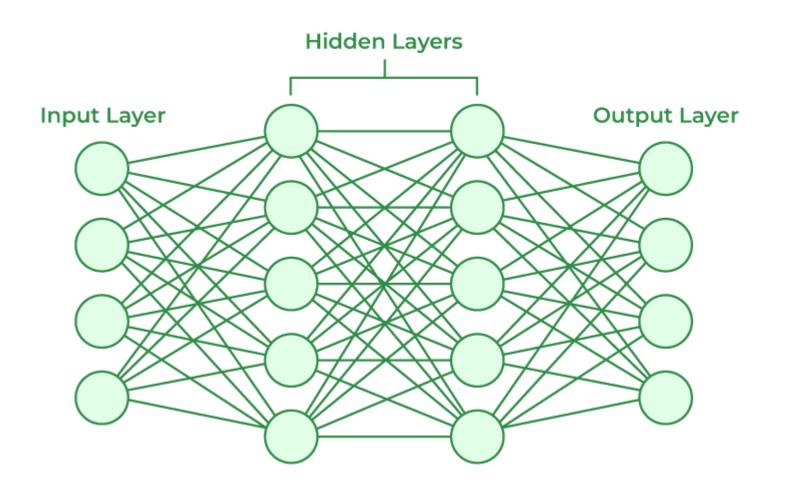


ELMo, BERT, & crew aka GPT

10 Things You Need to Know About BERT and the Transformer Architecture That Are Reshaping the Al Landscape (neptune.ai)

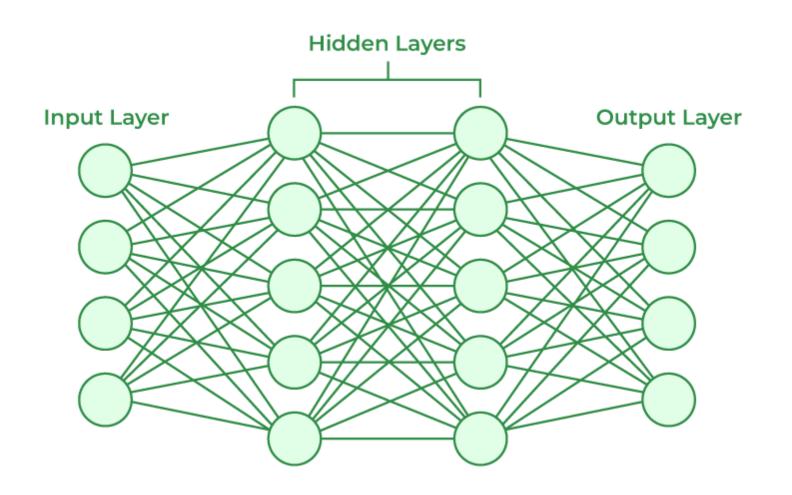
neural networks

Neural Network In 5 Minutes | What Is A Neural Network? | How Neural Networks Work | Simplifearn - YouTube

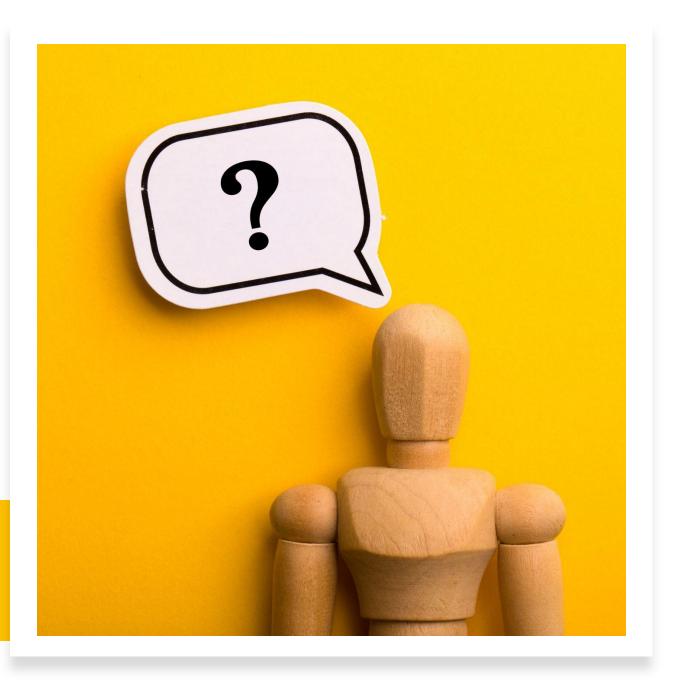


neural networks

[Artificial Neural Networks and its Applications - GeeksforGeeks]



does this look familiar?



don't think too far!

regression?

Lesson 3: Neural Network is nothing but a Linear Regression | by Md. Asifur Rahman | Medium

linear regression!

single neural network path!

$$y = \beta_0 + \beta_1 X + \varepsilon$$

$$y = mx + c$$

$$F(x) = w^T x + b$$

y=mx+c; x is feature vector, c is bias, y is output or dependent
variable, m is weight vector.

Lesson 3: Neural Network is nothing but a Linear Regression | by Md. Asifur Rahman | Medium

multiple linear regression!

simple perceptron!

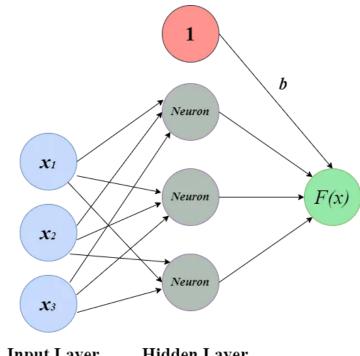
 $Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + ... + \beta_i X_i$

Y: Dependent variable

 β_0 : Intercept

β_i: Slope for X_i

X = Independent variable

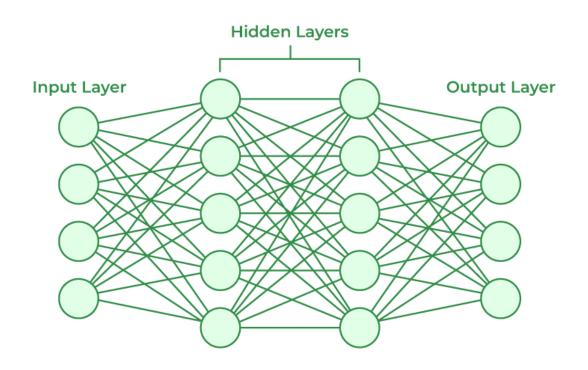


Input Layer

Hidden Layer

Lesson 3: Neural Network is nothing but a Linear Regression | by Md. Asifur Rahman | Medium

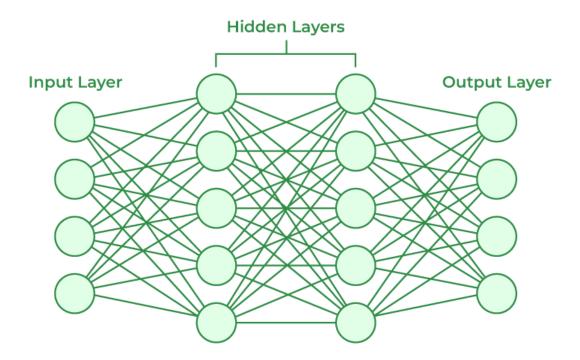
adding additional layers \implies neural network!



but what about linearity and non-linearity?

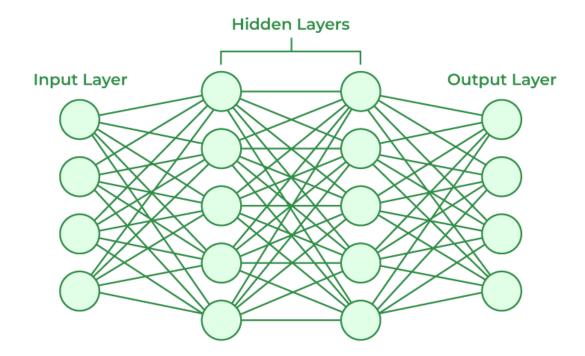
linear regression = linear.

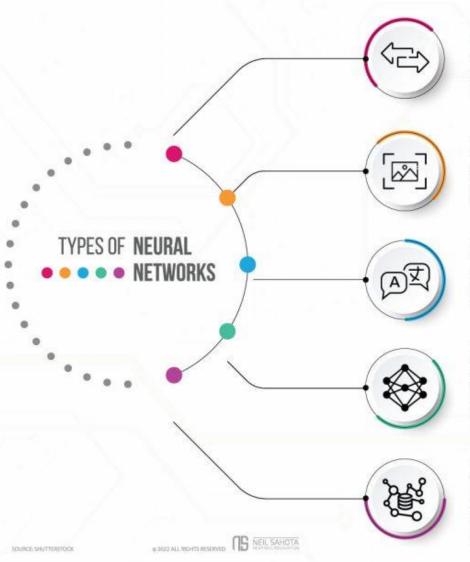
neural networks = solve complex non-linear problems.



remember the activation function

If we exchange a non-linear activation function
with a linear activation function,
the neural network in its function becomes a linear regression.





FEEDFORWARD NEURAL NETWORKS

Feedforward neural networks are good at solving problems with a clear relationship between the input and the output, but may not be as effective at figuring out more complex relationships.

CONVOLUTIONAL NEURAL NETWORKS

Convolutional neural networks are used for tasks that involve data with a grid-like structure, such as image recognition, but may require a large amount of data and be slow.

RECURRENT NEURAL NETWORKS

Recurrent neural networks are used for tasks involving data in a sequence, such as a language translation and speech recognition, but they may need help learning long-term relationships, which can be challenging to train.

GENERATIVE ADVERSARIAL NETWORKS

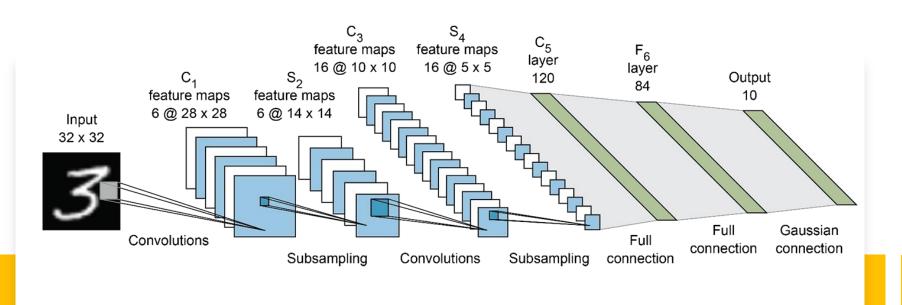
Generative adversarial networks are composed of two neural networks that work together to generate synthetic data that appears real but may be challenging to train and require a large amount of data to perform well. They have been used for tasks such as creating realistic images and

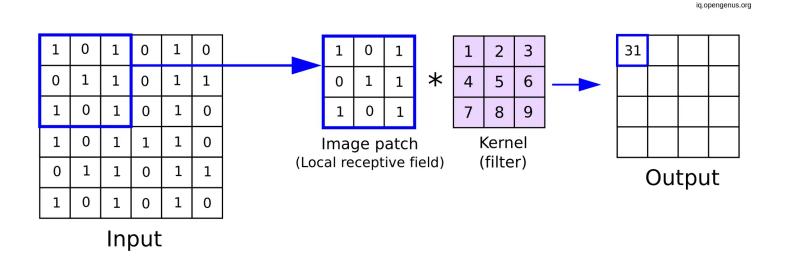
AUTOENCODER NEURAL NETWORKS

Autoencoders are used to reduce the complexity of data and learn important features, but they may be sensitive to the settings used and may not always learn meaningful patterns in the data. They have been applied in tasks such as image and speech recognition.

task-specific types of neural networks

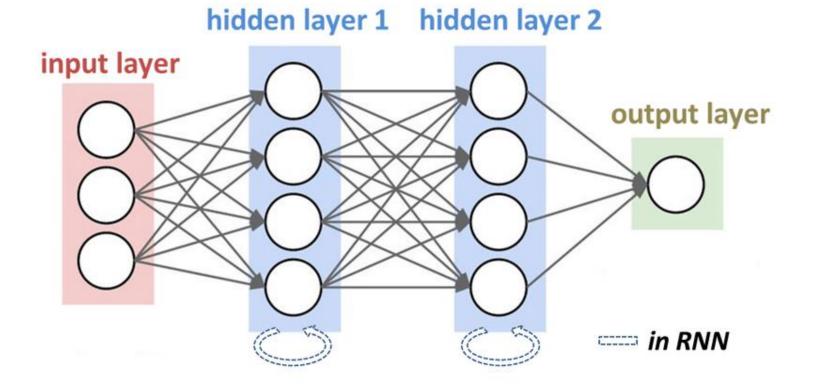
[Neural Networks: Solving Complex Science Problems (neilsahota.com)]





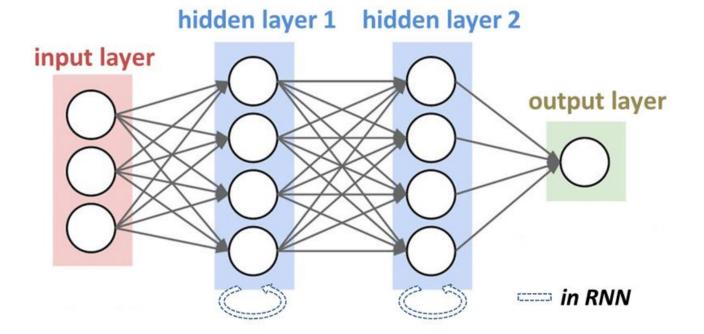
convolutional neural network

[Convolutional Neural Networks: 1998-2023 Overview | SuperAnnotate]



text = sequential processing

RNNs = recurrent neural networks

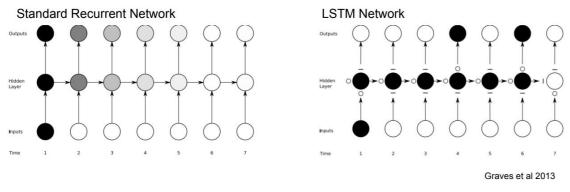


problem: vanishing gradient

RNNs

long shortterm memory LSTMs

LSTMs reduce vanishing gradient problem

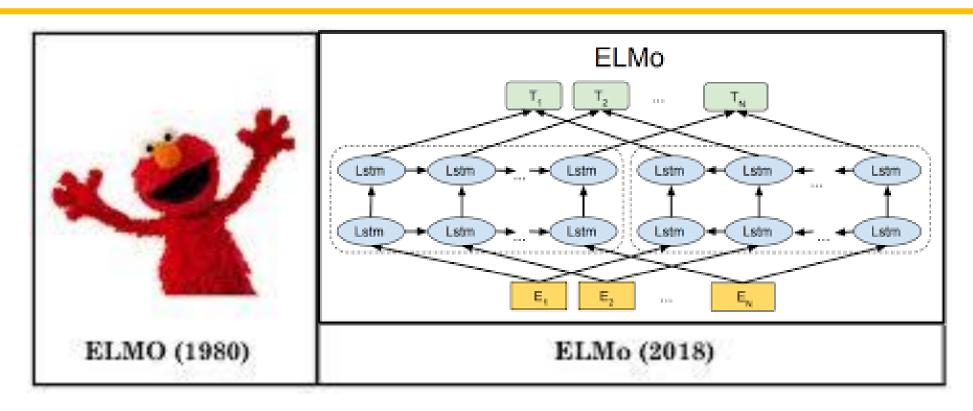


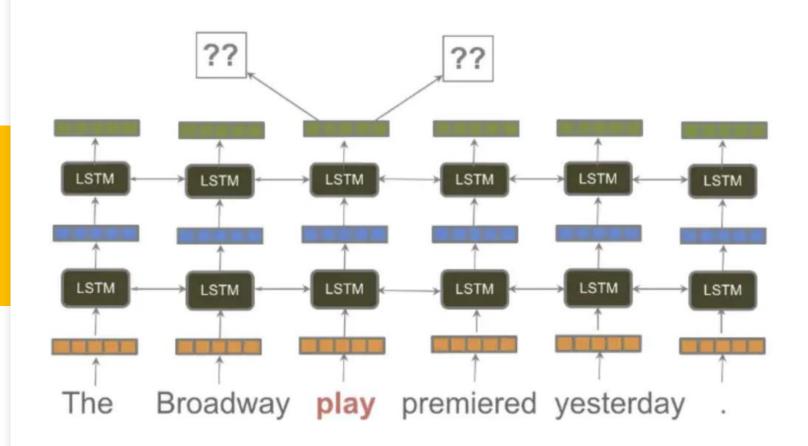
- The darker the shade, the greater the sensitivity
- The sensitivity decays exponentially over time as new inputs overwrite the activation of hidden unit and the network 'forgets' the first input



ELMo

Embeddings from Language Models





ELMo

bidirectional stacked LSTM

ELMo shortcomings

- lack of ability to capture long range dependencies and global context (dynamic context window)
- training complexity
 - computationally expensive and time-consuming
- fixed architecture
 - not adaptable to different tasks and data sizes
- lack of masked language modeling



transformers

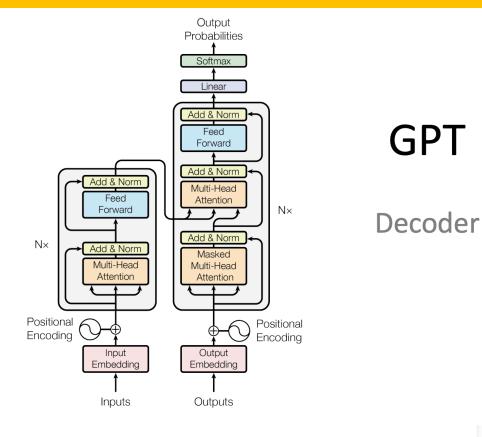
transformer architecture

 Parallel processing: Increases performance and scalability

• Bidirectionality: Allows understanding of ambiguous words and coreferences

BERT

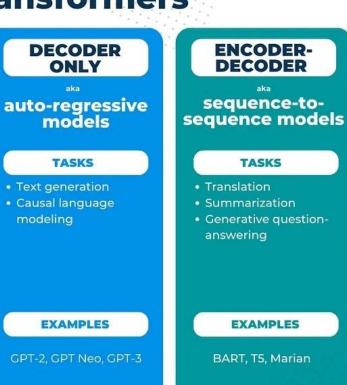
Encoder



transformer architecture

Transformers





A mechanism that computes a weighted sum of values (elements in a sequence) based on their similarity to a given query.

It allows the model to dynamically emphasize or de-emphasize different elements in the sequence, enabling the capture of contextual information and relationships between elements.

attention

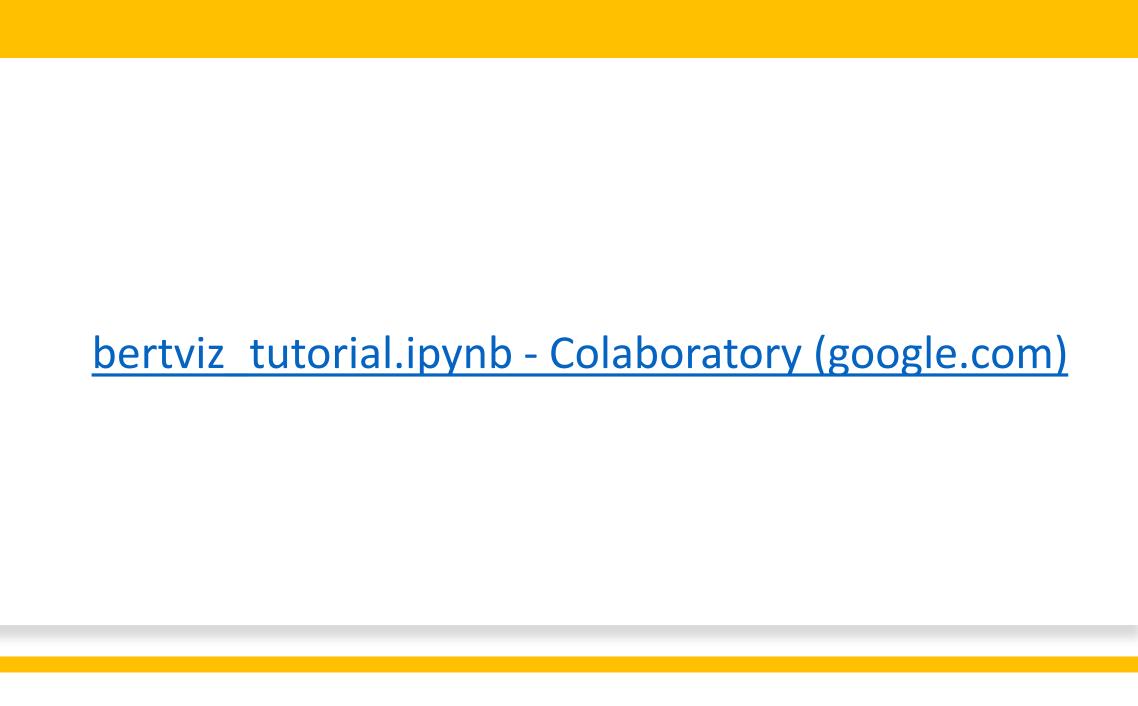
[1706.03762] Attention Is All You Need (arxiv.org)

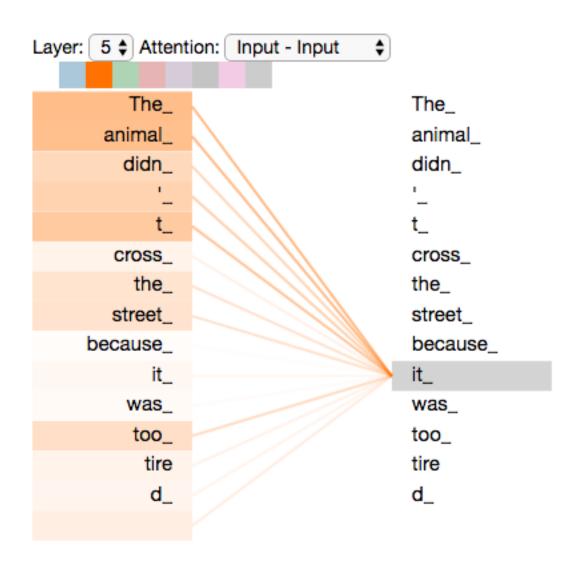
a little help from ChatGPT

Attention Mechanism in Transformers (in Simple Terms):

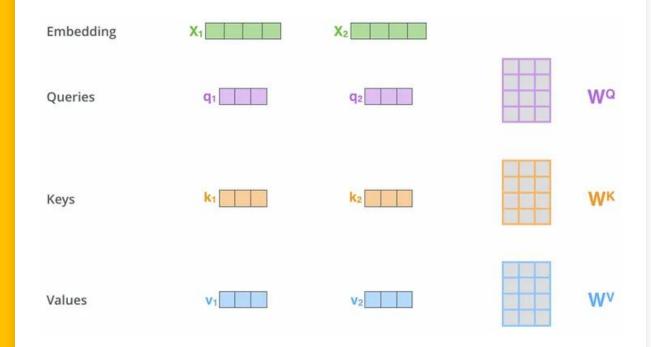
Imagine you're reading a sentence, and you want to understand the meaning of a specific word. Instead of just looking at that one word, you glance at all the words in the sentence. However, you pay more attention to the words that are more relevant to the word you're trying to understand. This way, you get a better grasp of the context and meaning.

The attention mechanism in Transformers works similarly. It helps the model focus on different parts of a sentence when trying to understand a specific word. It's like giving more importance to the words that matter most for understanding that word's context. This allows the model to create richer and more accurate representations of words and their relationships.





attention score needs query and key



attention score needs query and key



why attention

- capturing context
- handling long-range dependencies

- parallel processing
- interpretable representations

Attention for Neural Networks, Clearly Explained!!! YouTube