```
aultPrevented()){var h=a(d);this.activate(b.closest("li"),c),this.a
                                                                                                 {fun
rigger({type:"shown.bs.tab",relatedTarget:e[0]})})}}},c.prototype
                                                                                                   1)
> .active").removeClass("active").end().find('[data-toggle="tab
ia-expanded",!0),h?(b[0].offsetWidth,b.addClass("in")):b.removeC
).find('[data-toggle="tab"]').attr("aria-expanded",!0),e&&e()}va
e")||!!d.find("> .fade").length);g.length&&h?g.one("bsTransition
var d=a.fn.tab;a.fn.tab=b,a.fn.tab.Constructor=c,a.fn.tab.noConf
show")};a(document).on("click.bs.tab.data-api",'[data-toggle="ta
se strict";function b(b){return this.each(function(){var d=a(thi
                                                                                                    at
typeof b&&e[b]()})}var c=function(b,d){this.options=a.extend({}}
",a.proxy(this.checkPosition,this)).on("click.bs.affix.data-api"
ull,this.pinnedOffset=null,this.checkPosition()};c.VERSION="3.3.7";
                                                                                            larget=a
State=function(a,b,c,d){var e=this.$target.scrollTop(),f=this.$elem
                                                                                           osition
bottom"==this.affixed)return null!=c?!(e+this upe
                                                                                           ffix-top
```

## Autonomous Systems: Deep Learning **Transformers**



t a"),f=a.Event("hide.bs.tab ,{relaceururge.com

Prof. Dr.-Ing. Nicolaj Stache

Heilbronn University of Applied Sciences

### Motivation



## So far: RNN based Seq2Seq

- Processes data sequentially
- Captures timely dependencies in sequences

## Problem:

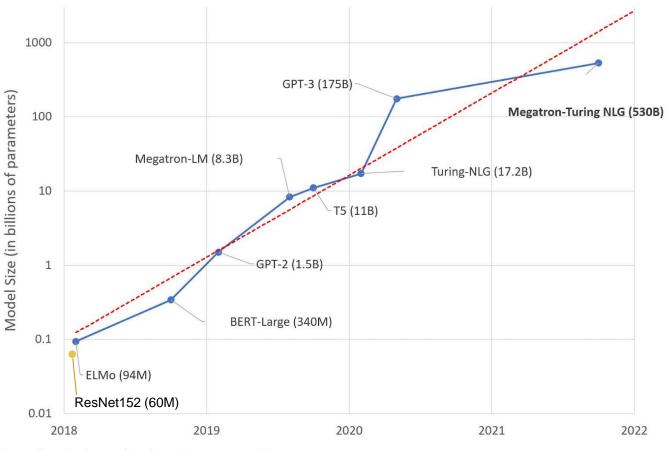
- The sequential nature prevents parallelization
- Struggles with long-range dependencies

### Solution:

Transformers

## **Transformers**



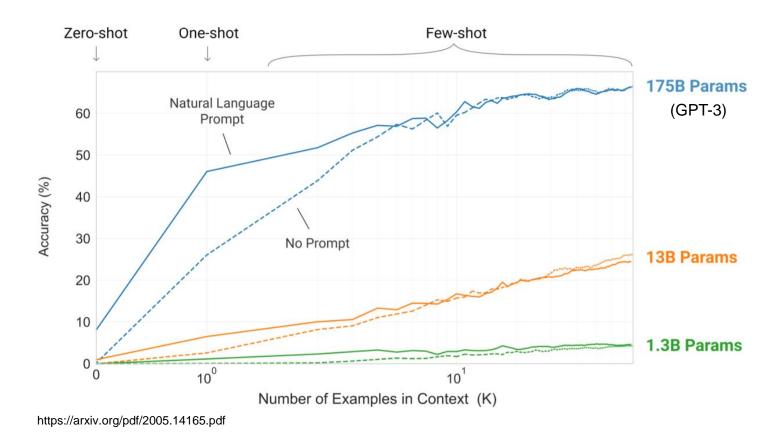


https://huggingface.co/blog/large-language-models 01.02.2022

Size of transformers grows at an exponential rate

## **Transformers**





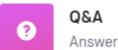
More parameters result in better accuracy

## GPT-3



- 175 billion trainable parameters
- Trained on about 45 TB of text data
- Training would cost \$4.5 million and would take 355 years on a V100 GPU server (28 TFLOPS capacity)

## Example Tasks:



Answer questions based on existing knowle...



Summarize for a 2nd grader

Translates difficult text into simpler concep...



Python bug fixer

Find and fix bugs in source code.



Recipe creator (eat at your own risk)

Create a recipe from a list of ingredients.



Explain code

Explain a complicated piece of code.



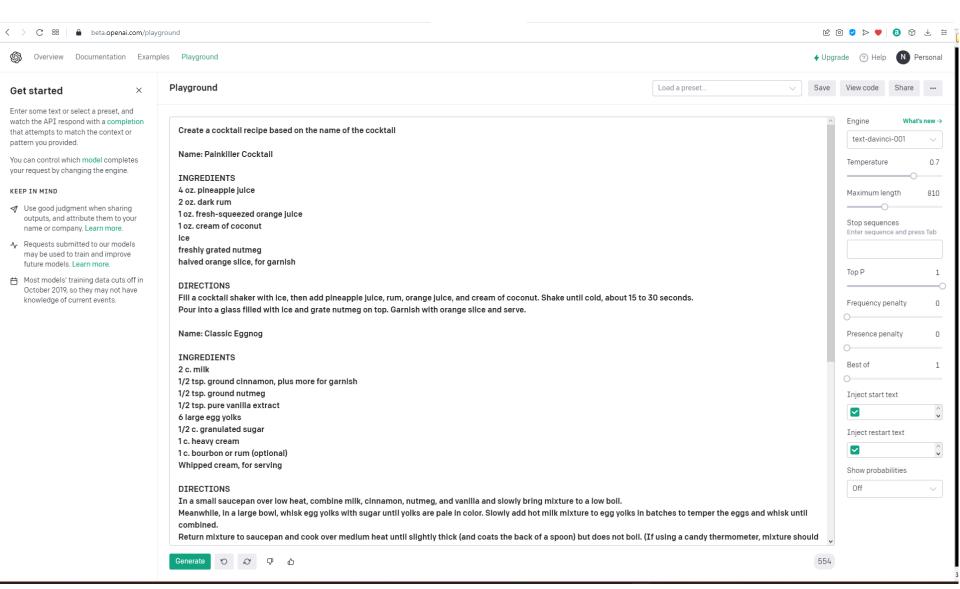
Ad from product description

Turn a product description into ad copy.

→ Try it out on: https://beta.openai.com/

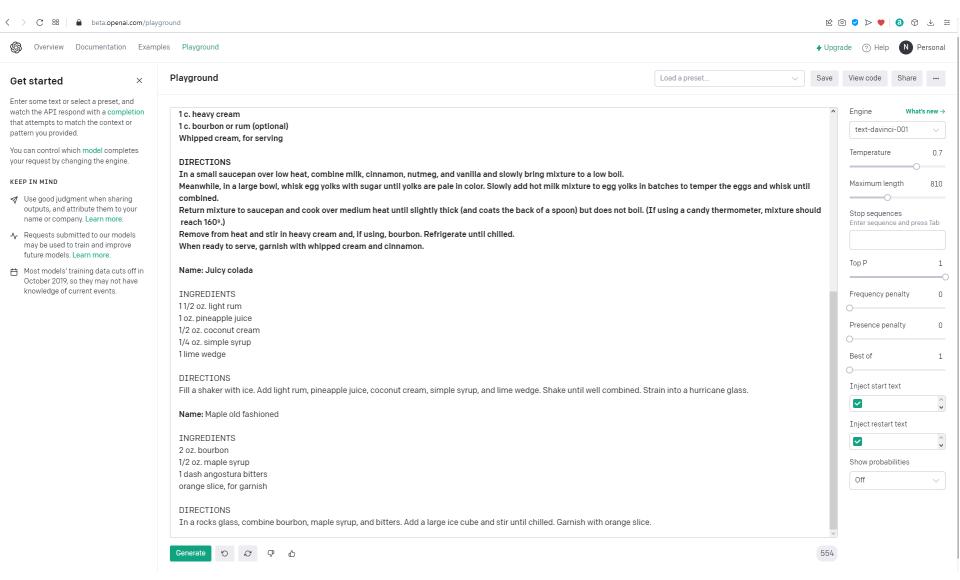
## Example





## Example





## **Applications**



#### **Question Answering**

#### **Passage Sentence**

In meteorology, precipitation is any product of the condensation of atmospheric water vapor that falls under gravity.

#### **Ouestion**

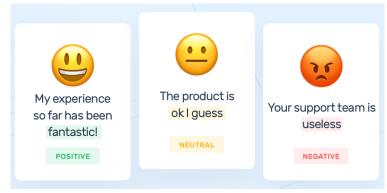
What causes precipitation to fall?

#### **Answer Candidate**

gravity

https://rajpurkar.github.io/mlx/ga-and-squad/ 01.02.2022

#### **Sentiment Analysis**



https://monkeylearn.com/sentiment-analysis/ 01.02.2022

#### Named Entity Recognition

Ousted WeWork founder Adam Neumann lists his Manhattan penthouse for \$37.5 million

[organization]

[person]

[location]

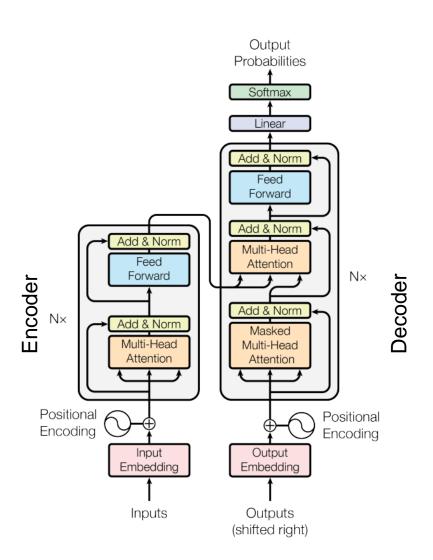
[monetary value]

https://monkeylearn.com/blog/named-entity-recognition/ 01.02.2022

## And many more...

## "Attention is all you need" 2017

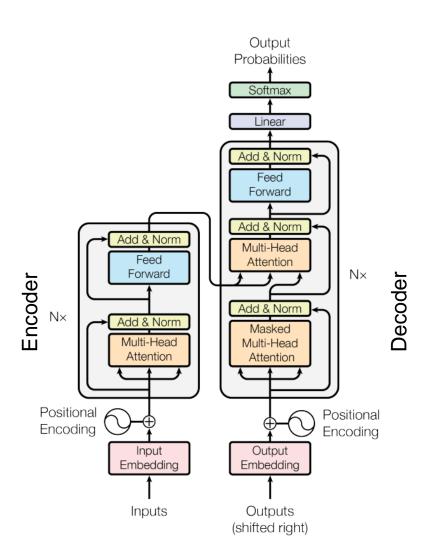




- Encoder-Decoder Structure
- No Recurrence
- Outperformed all SOTA Algorithms in NLP Tasks

## **Transformers**



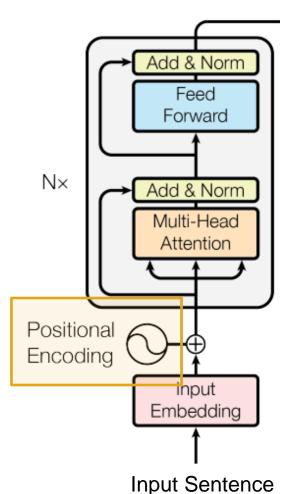


#### Whats new?

- Positional Encodings
- Self-Attention
- Multi-Head Attention
- Masked Multi-Head Attention

## Encoder





**Encoded Representation** 

- Encodes the input into a continuous representation
- Adds attention information
- Helps the decoder to focus on appropriate words

## Positional Encodings



- It should output a unique encoding for each time-step
- Distance between any two time-steps should be consistent across sentences with different lengths.
- It must be deterministic.

$$\overrightarrow{p_t}^{(i)} = f(t)^{(i)} := egin{cases} \sin(\omega_k.\,t), & ext{if } i = 2k \ \cos(\omega_k.\,t), & ext{if } i = 2k+1 \end{cases}$$

$$\omega_k = rac{1}{10000^{2k/d}}$$

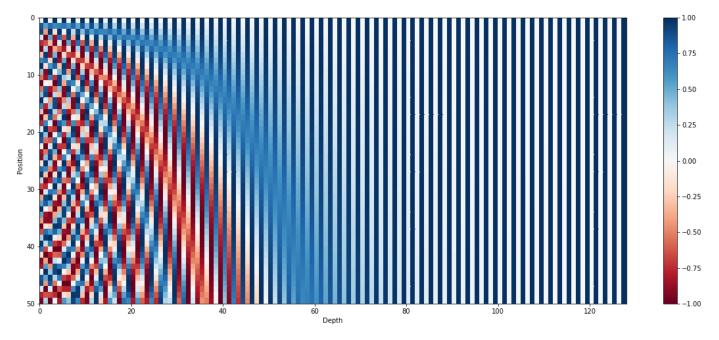
$$\overrightarrow{p_t} = egin{bmatrix} \sin(\omega_1.t) \ \cos(\omega_1.t) \ \sin(\omega_2.t) \ \cos(\omega_2.t) \ & dots \ & dots \ \sin(\omega_{d/2}.t) \ \cos(\omega_{d/2}.t) \end{bmatrix}_{d imes 1}$$

https://kazemnejad.com/blog/transformer\_architecture\_positional\_encoding/ 01.02.2022





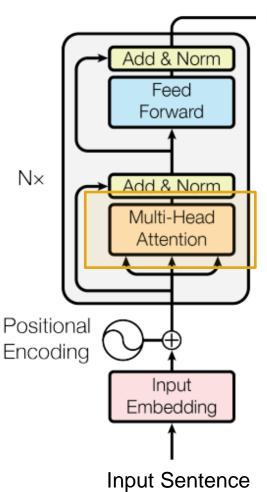
```
0:
    0 0 0 0
                   1 0 0 0
1:
             9:
    0 0 0 1
                   1 0 0 1
            10:
    0 0 1 0
                   1 0 1 0
            11: 1 0 1 1
3:
    0 0 1 1
            12: 1 1 0 0
    0 1 0 0
              13:
    0 1 0 1
                  1 1 0 1
              14:
    0 1 1 0
                   1 1 1 0
7:
    0 1 1 1
              15:
                   1 1 1 1
```



https://kazemnejad.com/blog/transformer\_architecture\_positional\_encoding/ 01.02.2022

## Encoder





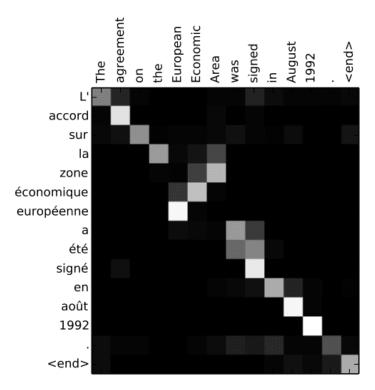
**Encoded Representation** 

- Attention
- Self Attention
- Multi-Head Attention

## Attention

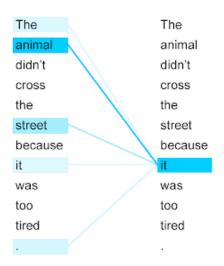


- Mimics the human attention
- Focusing on a few relevant parts
- Relations between Input und Output Sequence
- Enables better processing of very long sequences

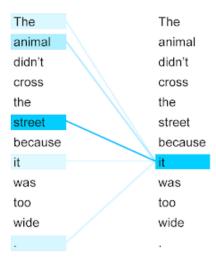


https://arxiv.org/pdf/1409.0473.pdf 01.02.2022



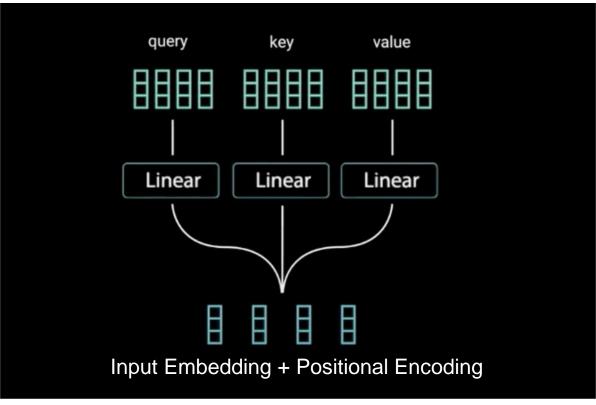


- Relations inside the same sequence
- Better understanding of context



https://ai.googleblog.com/2017/08/transformer-novel-neural-network.html 01.02.2022

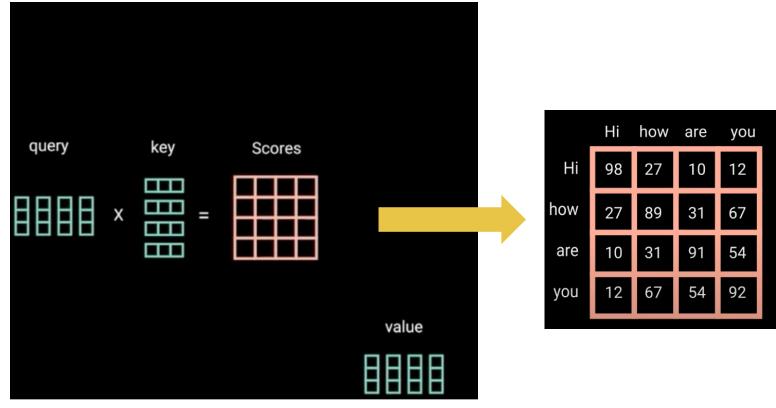




https://towardsdatascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0

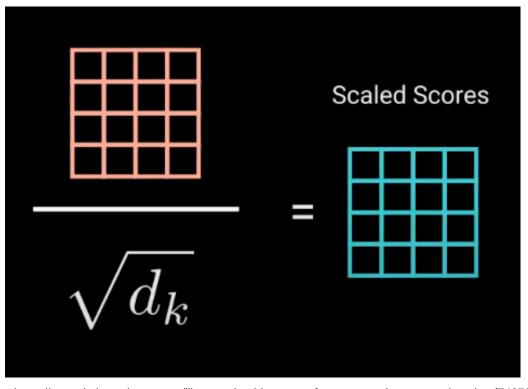






https://towardsdatascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0

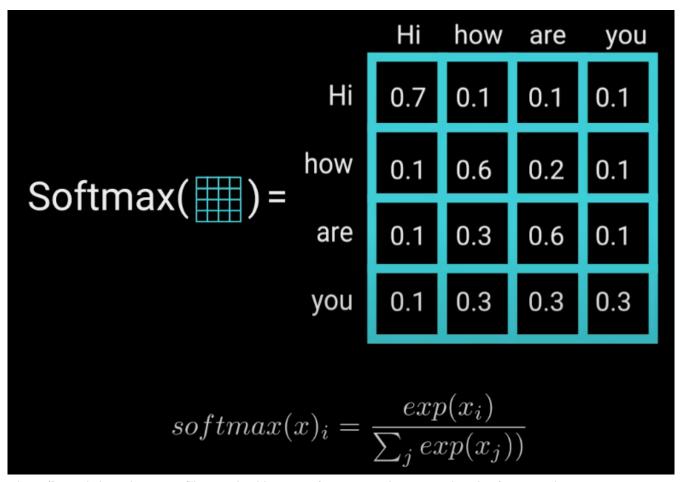




- More stable gradients
- Prevents exploding effects

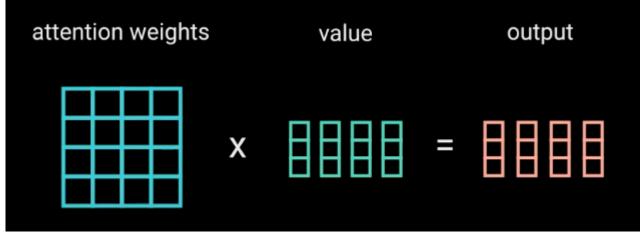
https://towardsdatascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0





https://towards datascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0

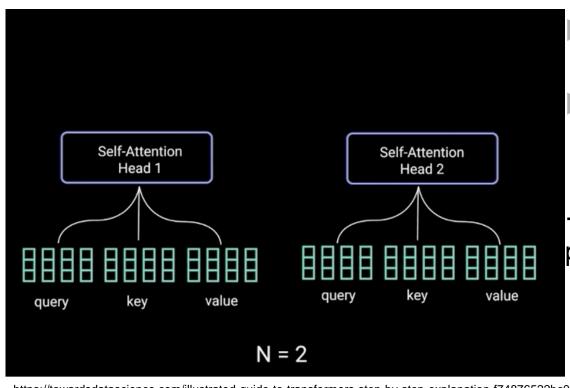




https://towardsdatascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0

## Multi-Head Attention





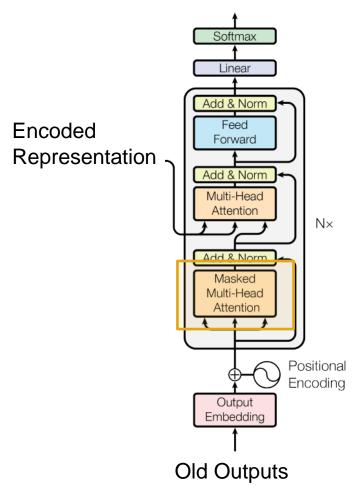
- Calculate Self-Attention multiple times
- Each head learns something different
- → More representation power

https://towardsdatascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0

## Decoder



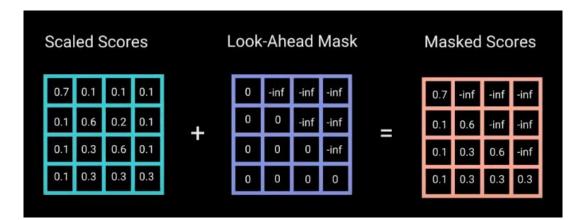
#### **Output Probabilities**



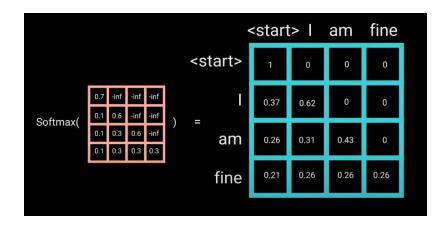
- Decodes continuous representation
- Generates text sequences
- Works autoregeressiv

## Masked-Multi-Head Attention





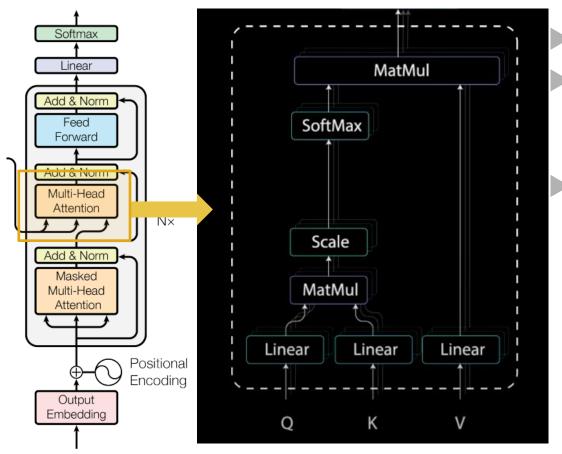
- Only used for training
- Each word gets only the attention score for itself, and the words generated before.



https://towardsdatascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0





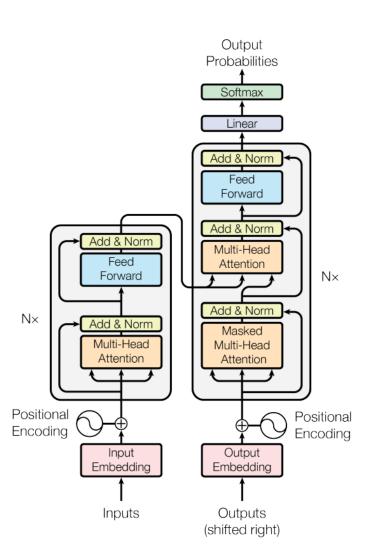


- Q and K from Encoder
- V from Masked-Multihead Attention of Decoder
- Helps to focus on the right parts of the encoders output

https://towardsdatascience.com/illustrated-guide-to-transformers-step-by-step-explanation-f74876522bc0

## **Transformers**





### What you know now!

- Positional Encodings
- Self-Attention
- Multi-Head Attention
- Masked Multi-Head Attention



## Hands-on part: Translation with transformers

Please work through the tutorial 12\_Transformers\_Translation\_Example.ipynb