

# Generative AI: Transformers: Architecture

## PART II: Encoder

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## 2: Generative AI: LLM: Transformers: Architecture



*AI Deep learning (Source: mindovermachines.com)*

# **3: Generative AI: LLM: Transformers: Architecture**

1. Generative AI: Transformers: Architecture: Overview
2. Generative AI: Transformers: Architecture: Encoder-Decoder Structure
3. Generative AI: Transformers: Architecture: Revolutionary Core Features
4. Generative AI: Transformers: Architecture: Input Embedding & Positional Encoding
5. Generative AI: Transformers: Architecture: Attention & Multi-Header Attention
6. Generative AI: Transformers: Architecture: Encoder & Encoding Stack of Layers
7. Generative AI: Transformers: Architecture: Decoder & Decoding Stack of Layers

# 4: Generative AI: LLM: Transformers: Architecture

## Artificial Intelligence: Generative AI

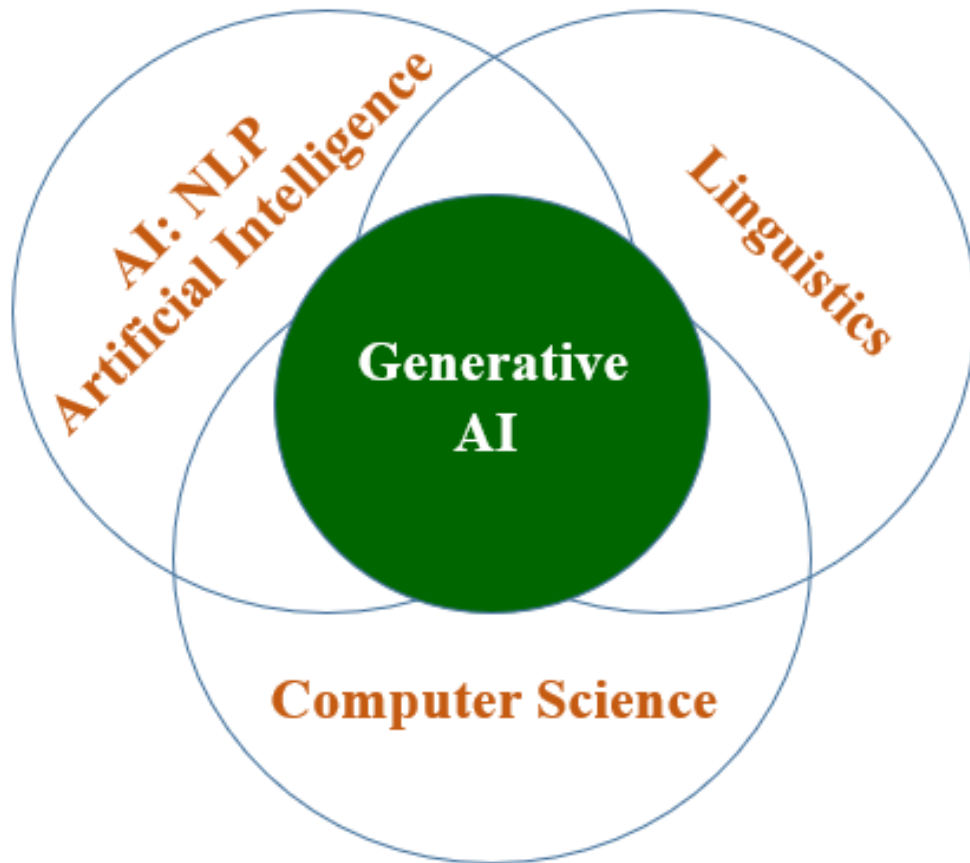
### What is It?

Generative AI: A category of artificial intelligence focused on using AI deep learning models to generate new contents, including text, images, audio, video, and more. The contents are novel but look realistic and may be indistinguishable from human-created ones.

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Artificial Intelligence: Generative AI: LLM

Foundational Sciences & Technologies



**Generative AI** is based on the NLP technologies such as Natural Language Understanding (NLU) and Conversational AI (AI Dialogues) - Those among the most challenging tasks AI needs to solve.

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## Artificial Intelligence: Generative AI: LLMs

### Large Language Models

**Large Language Models (LLMs)** are **revolutionary AI Deep Learning neural networks** that excel in **natural language understanding (NLU)** and **content generation**.

- “LARGE” in LLMs refers to the vast scale of data and parameters used to train them, allowing LLMs to develop a comprehensive understanding of language.
- Being particularly transformer-based models trained on massive text datasets using deep learning techniques,
- Able to learn complex language patterns, capture nuances like grammar and tone, and generate coherent and contextually relevant text



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Artificial Intelligence:  
Generative AI: LLMs

Large Language  
Models: Transformers

*Photo Source: Prompt by Thuan L Nguyen –  
Generated by **Generative AI Text-To-  
Images**: Google-DeepMind ImageFX*

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**“Attention Is All You Need!”**

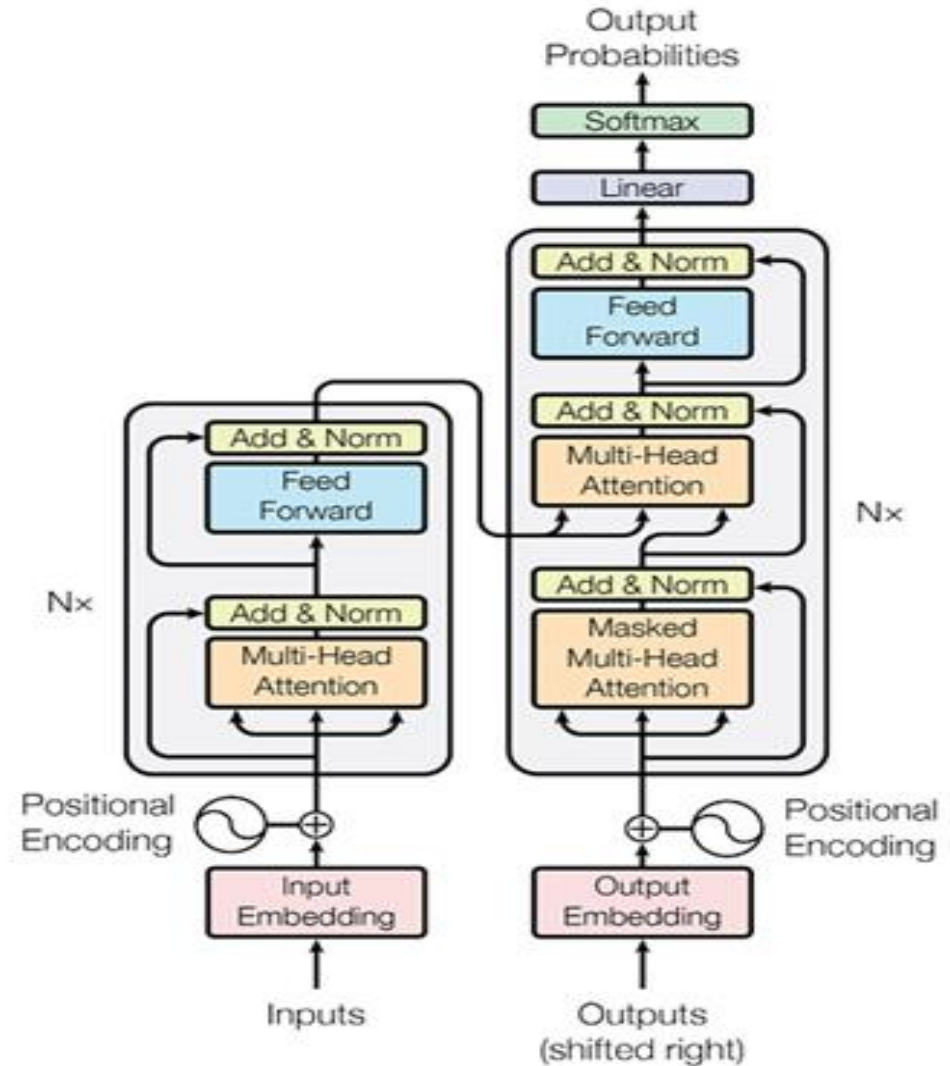
*Photo Source: Prompt by Thuan L Nguyen –  
Generated by Generative AI Text-To-  
Images: Google-DeepMind ImageFX*





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## LLM: Transformers: Architecture Encoder-Decoder

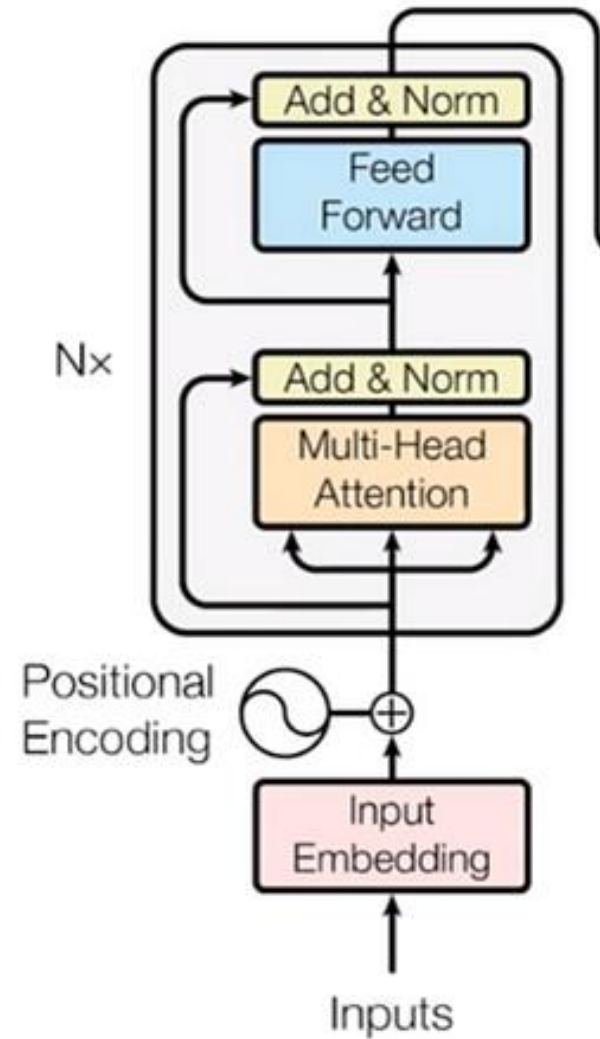


The Transformer - model architecture.

*source: arXiv:1706.03762*

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## LLM: Transformers: Architecture Encoder



*source: arXiv:1706.03762*

# 11: Generative AI: LLM: Transformers: Architecture

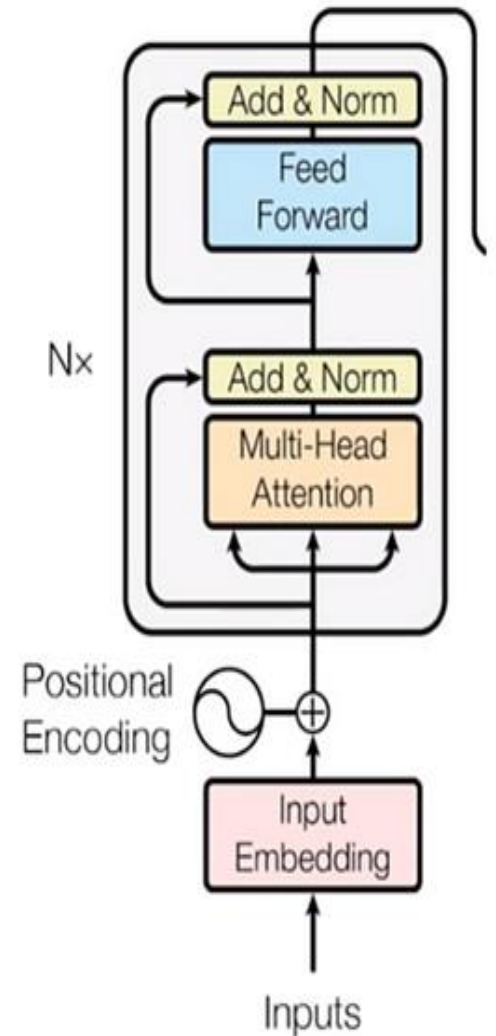
## LLM: Transformers: Architecture

### Encoder

#### Main Layers of the Transformer: Encoder

- **Input Embeddings:** The input words are converted into word embeddings (dense vector representations).
- **Multi-Head Attention (Self-Attention):** Words "attend" to each other within the input sequence.
- **Layer Normalization:** Help stabilization during training.
- **Feed-Forward Network:** Further refine the representations of words
- **Residual Connections:** Shortcut connections add the input to the output of each sub-layer, making training more manageable.

A Transformer has **stacked encoder layers** (usually 6 or more), with **each layer refining the representations of words or contents from the previous**.



*source: arXiv:1706.03762*

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## LLM: Transformers: How Does it Works?

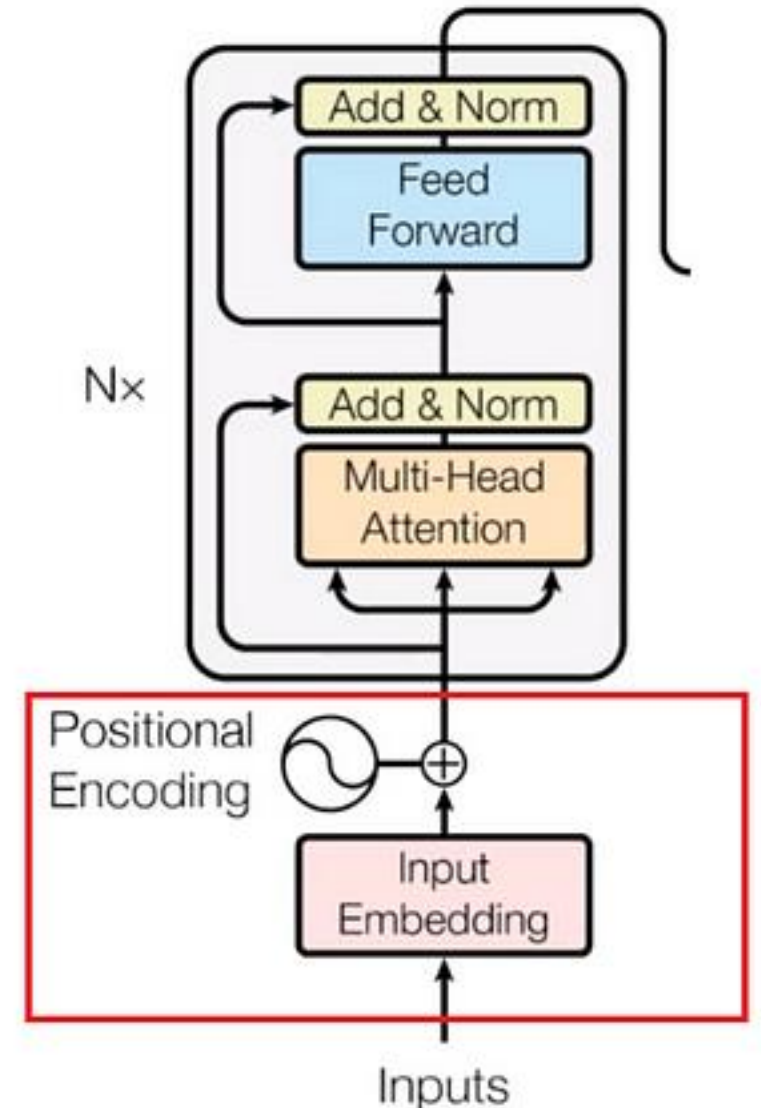
### Encoder

Computers does not understand words. Computers can only works on numbers, vectors or matrices.

Words → Vectors: **Embedding Space**

#### **Embedding Space:**

- A dictionary where words of similar meanings are grouped together and are present close to each other.
- Every word, according to its meaning, is mapped and assigned with a particular value, i.e., a vector.



source: [arXiv:1706.03762](https://arxiv.org/abs/1706.03762)

# 13: Generative AI: LLM: Transformers: Architecture

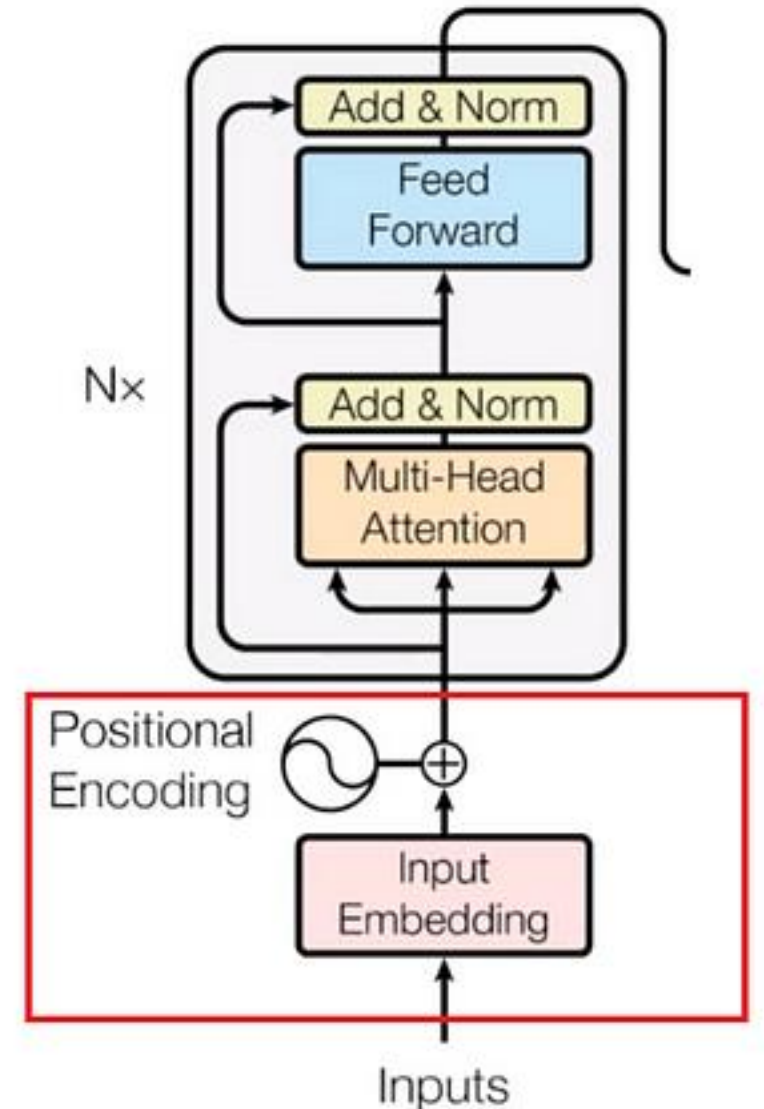
## LLM: Transformers: How Does it Works?

### Encoder

Every word in different sentences has different meanings.

**Positional Encoding** → **Positional Encoders**

**Positional Encoder**: A **vector** that gives the **context** of each word according to the **position** of the word **in a sentence**.



source: [arXiv:1706.03762](https://arxiv.org/abs/1706.03762)



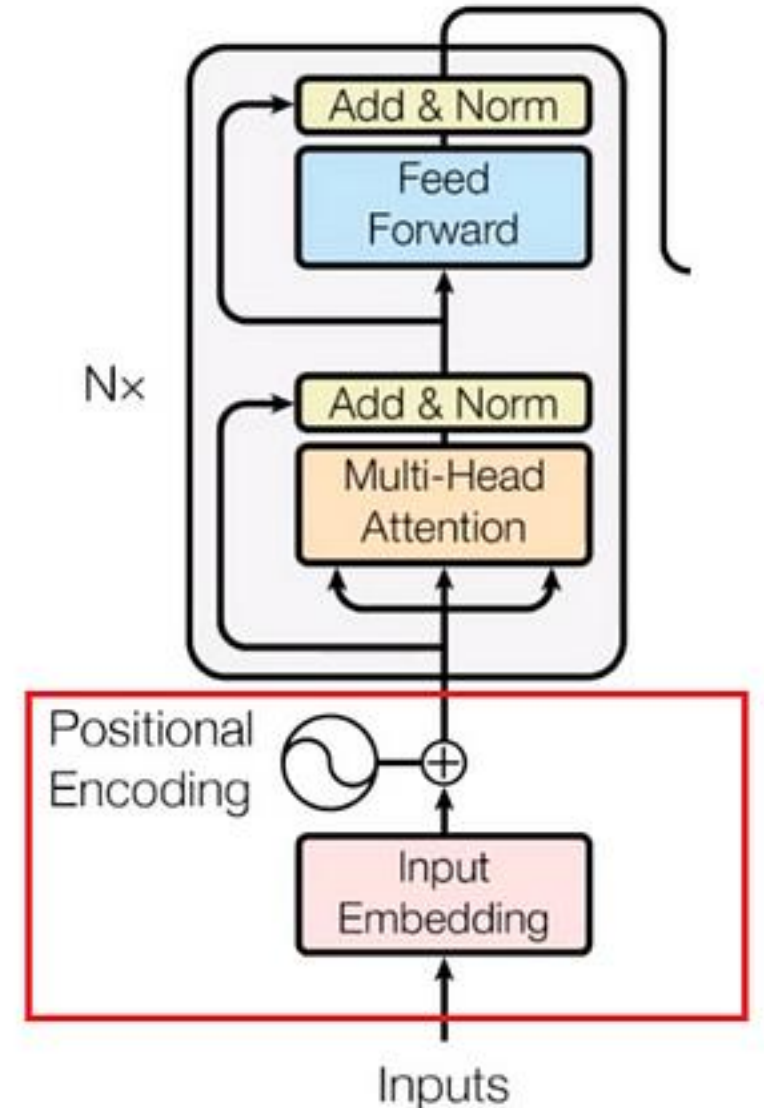
# 14: Generative AI: LLM: Transformers: Architecture

LLM: Transformers: How Does it Works?

## Encoder

**Each Word:**

- **Word Embedding Vector** (Input Embedding)
- **Positional Embedding Vector** (Positional Embedding)
- **Final Vector: Context**
  - Ready to be fed into the encoder block



*source: arXiv:1706.03762*

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## LLM: Transformers: How Does it Works?

### Encoder

#### Multi-Head Attention (Self-Attention):

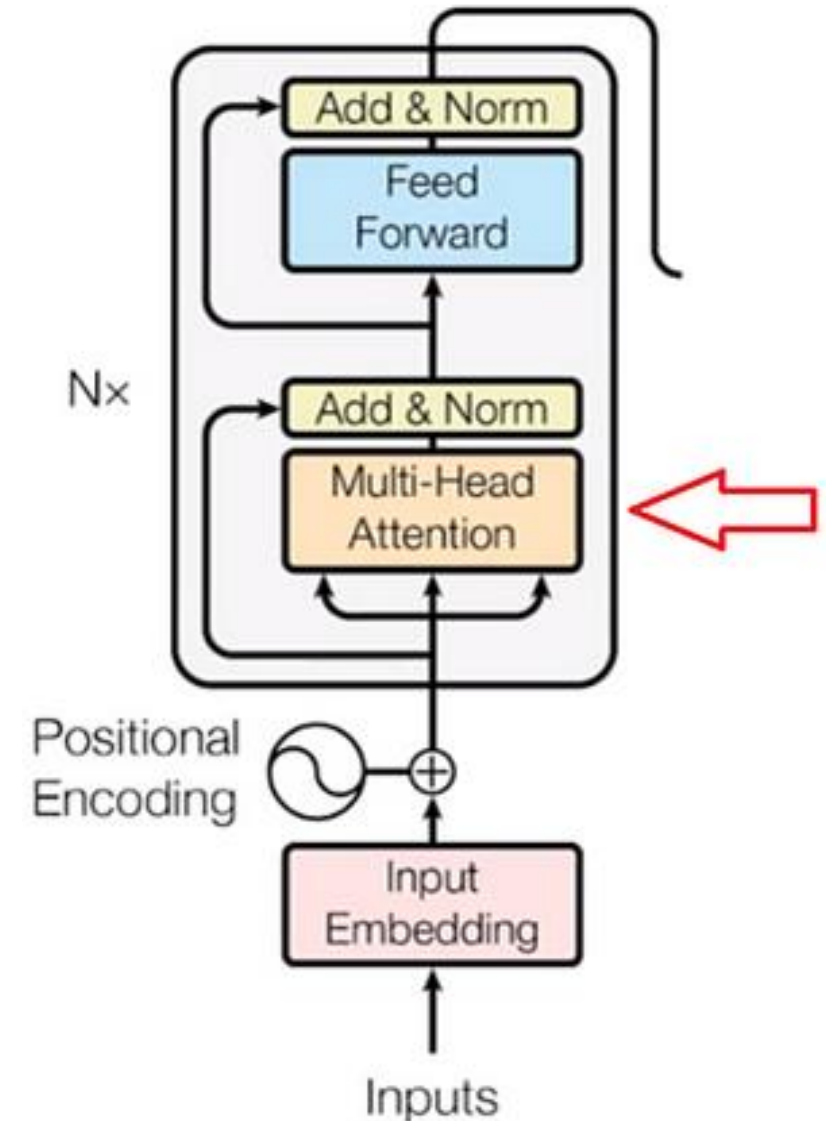
**TO DO:** Find out how relevant a particular word is w.r.t to other words in that sentence.

#### OUTPUT: Attention vectors

- Each vector represents the **relevance** of a **word** w.r.t **other words**.

Every word → An **attention vector**

- Capture the **contextual relationship** between the **word** and **other words** in that **sentence**.



source: *arXiv:1706.03762*

# 16: Generative AI: LLM: Transformers: Architecture

## LLM: Transformers: How Does it Works?

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Attention : What part of the input should we focus?

	Focus		Attention Vectors
The	→	The big red dog	$[0.71 \ 0.04 \ 0.07 \ 0.18]^T$
big	→	The big red dog	$[0.01 \ 0.84 \ 0.02 \ 0.13]^T$
red	→	The big red dog	$[0.09 \ 0.05 \ 0.62 \ 0.24]^T$
dog	→	The big red dog	$[0.03 \ 0.03 \ 0.03 \ 0.91]^T$

Source: <https://www.youtube.com/watch?v=rPFkX5fJdRY>

# 17: Generative AI: LLM: Transformers: Architecture

## LLM: Transformers: How Does it Works?

### Encoder

#### Multi-Head Attention (Self-Attention):

For **every word**:

- The **attention on itself** is much **higher** than that of other words in the sentence.
- **Multiple** attention vectors are calculated **per word** → **Multi-Head Attention**.
- **Final attention vector**: A **weighted average** of all attention vectors is calculated.

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# 18: Generative AI: LLM: Transformers: Architecture

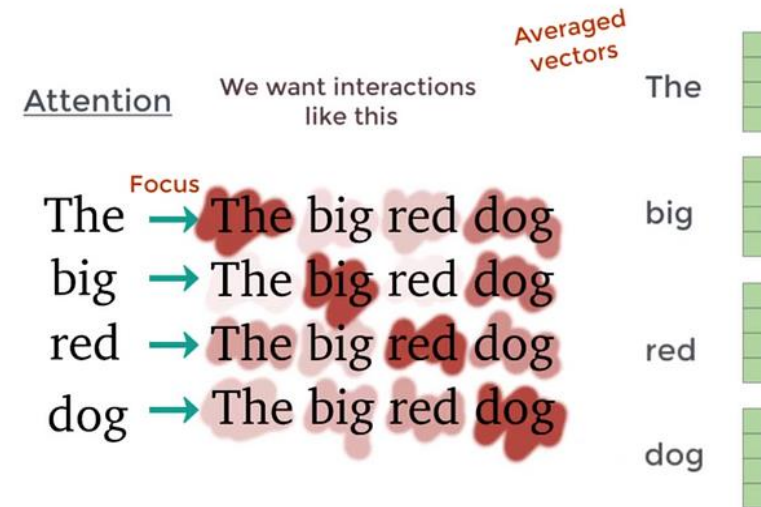
## LLM: Transformers: How Does it Works?

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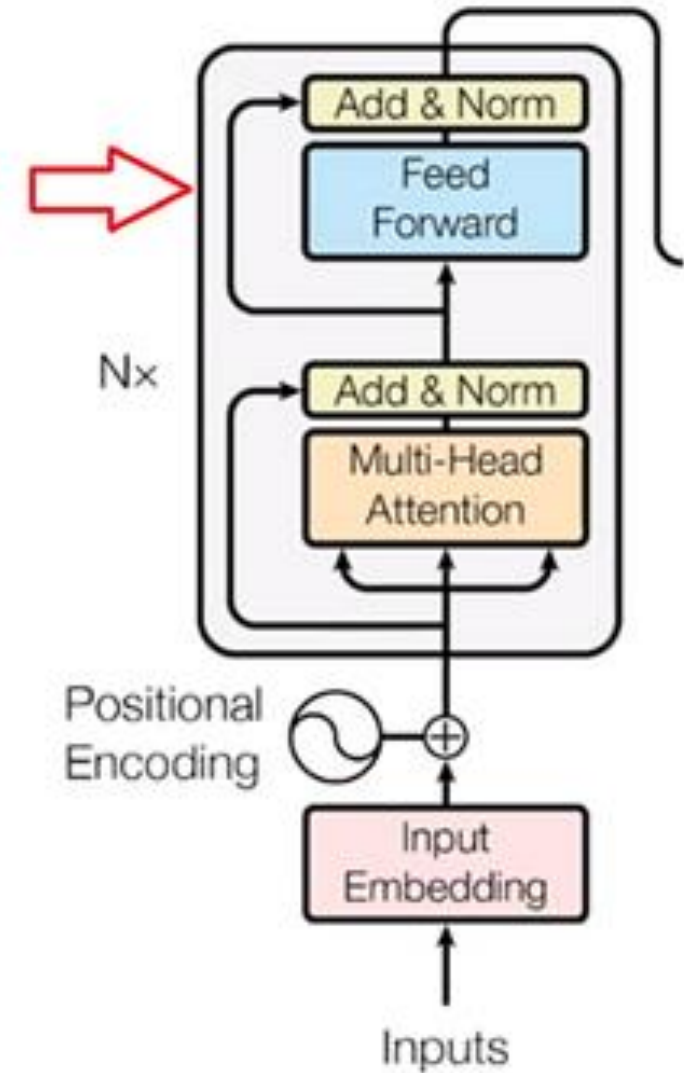
LLM: Transformers: How Does it Works?

## Encoder

### Feed-Forward Neural Network

**TO DO:** Transform each attention vector into a form acceptable to the next encoder-decoder level.

**OUTPUT:** A data structure (Flattened Vector) acceptable to the next encoder-decoder level.



*source: arXiv:1706.03762*

# 20: Generative AI: LLM: Transformers: Architecture

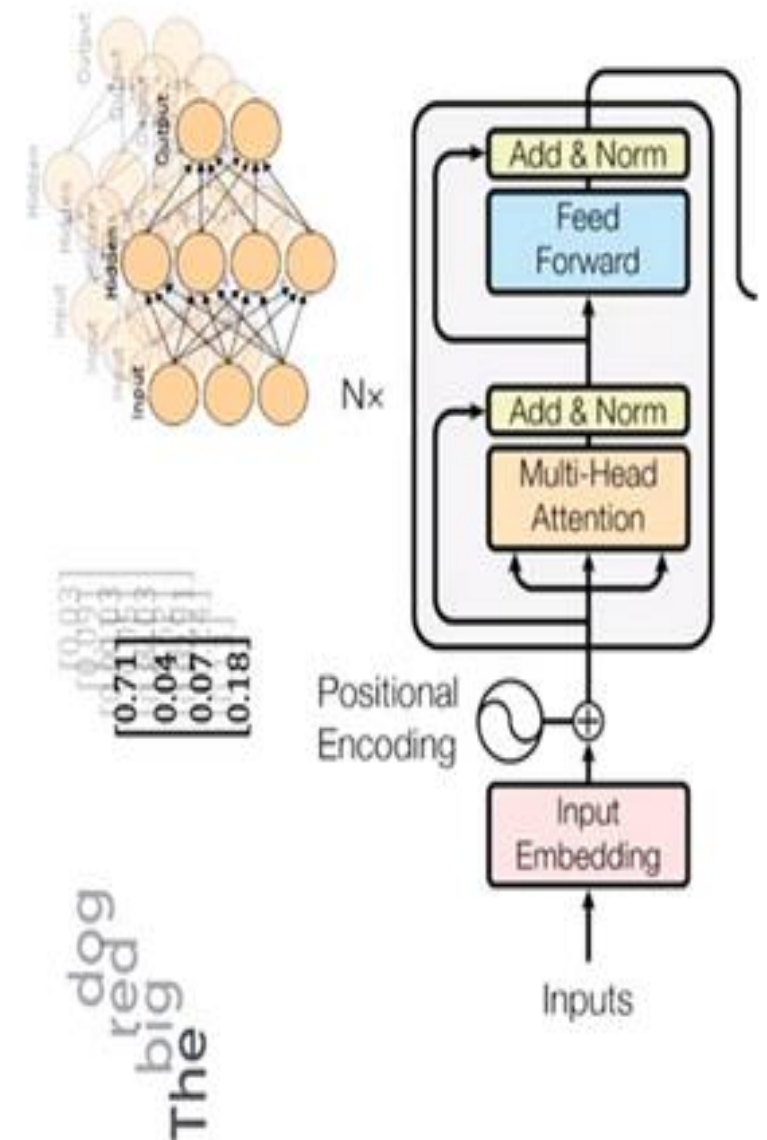
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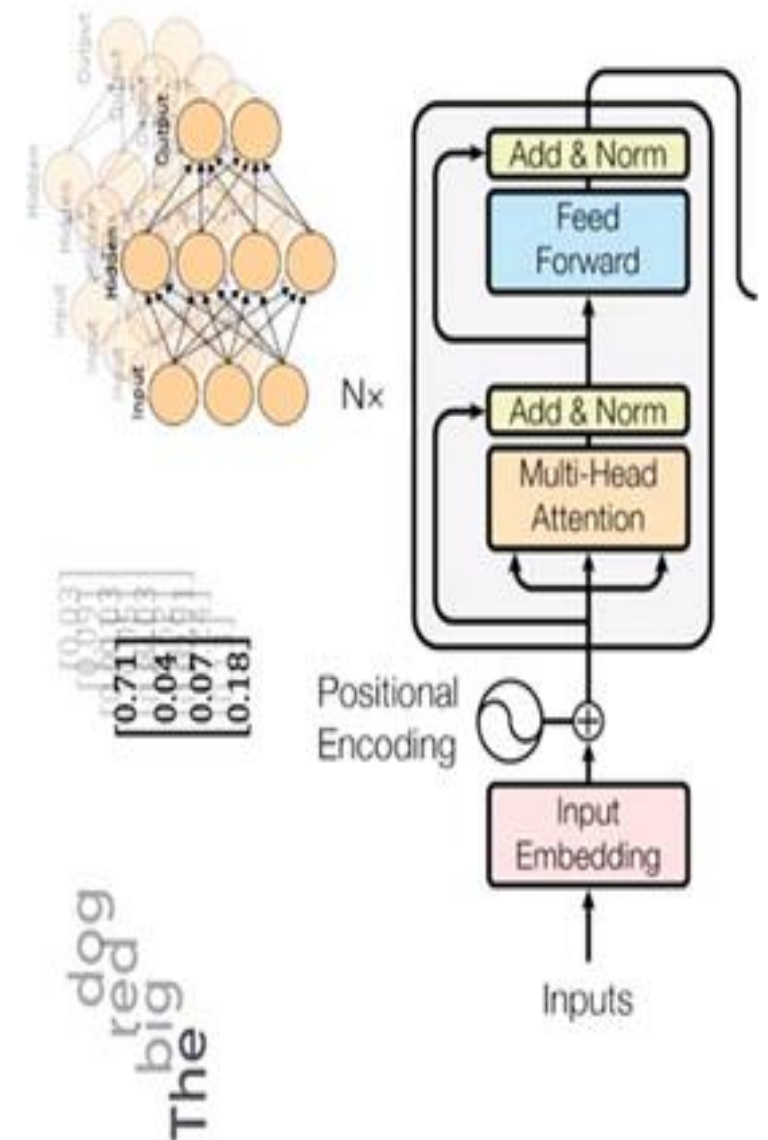
## LLM: Transformers: How Does it Works?

### Encoder

#### Feed-Forward Neural Network

Feed Forward Neural Network **accepts only one** attention vector at a time.

- **Each** attention vector is **independent** of each other.
- **Parallel computing** can be applied here, *and that makes all the difference.*



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## LLM: Transformers: How Does it Works?

### Encoder

#### Encoder: Feed-Forward Neural Network

Feed Forward Neural Network **accepts only one** attention vector at a time. However, **each** attention vector is **independent** of each other.

- **Parallel computing** can be applied.
- Possible to pass **all** the words at the same time into the **encoder** block, and get the set of Encoded Vectors for **every word simultaneously**.

