

The (r)evolution of generative models

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About me

VP of AI at Onfido

ex-Criteo (2015-2020)

MIT DARPA Urban Challenge Finalist

ENSTA 2002, MIT PhD 2010 (computer vision)

Published @ CVPR 2007, IJFR 2008, ICCV 2009, ICRA 2010



[oakfr](#)



[@olivkoch](#)

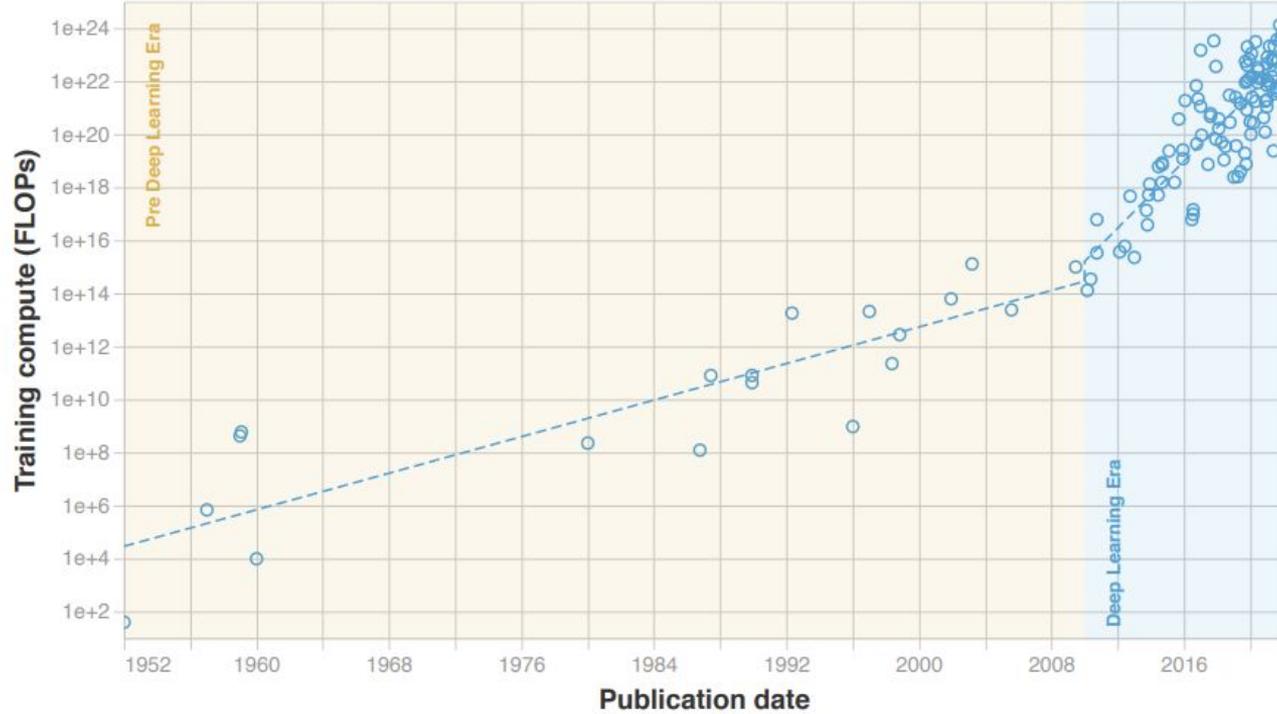


[olivierkoch.net](#)

1. Historical perspective
2. The Transformer shake-up
3. The new landscape

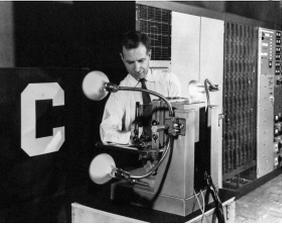
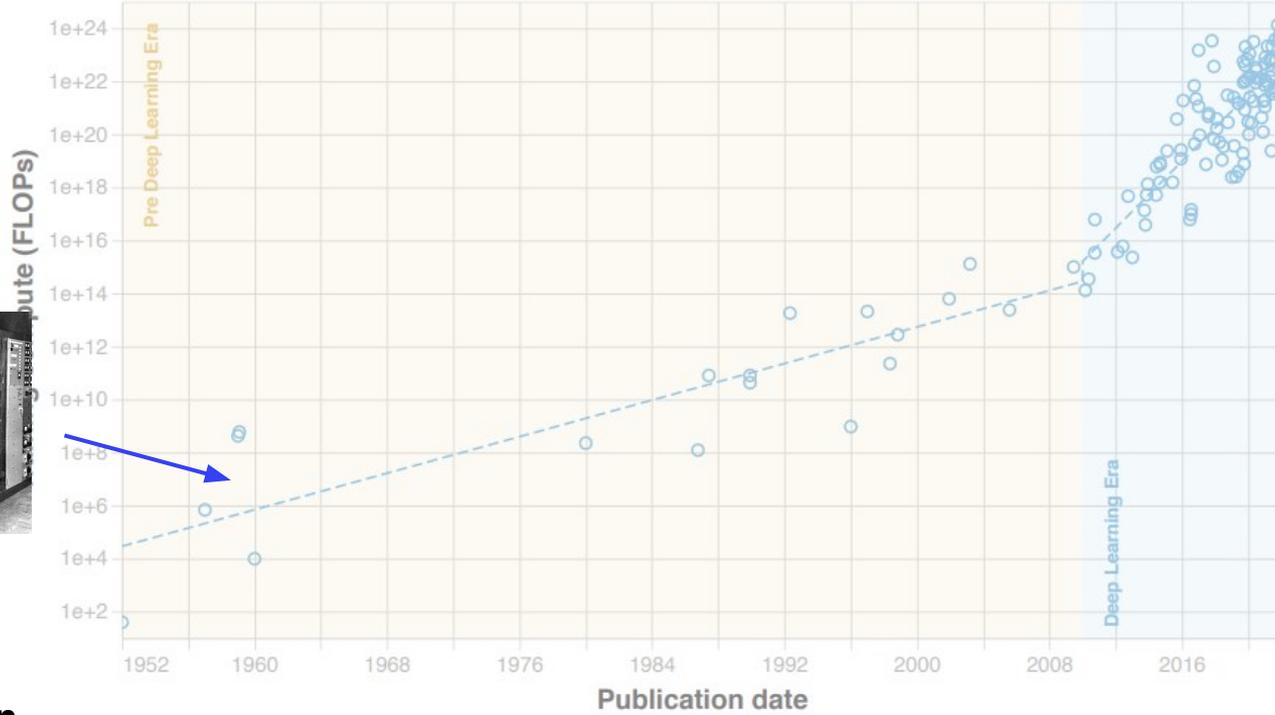
Training compute (FLOPs) of milestone Machine Learning systems over time

n = 121



Training compute (FLOPs) of milestone Machine Learning systems over time

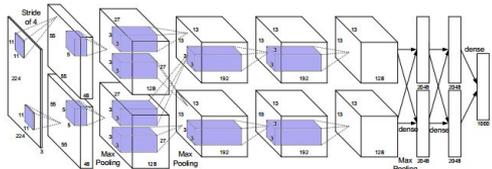
n = 121



1958

The
Perceptron

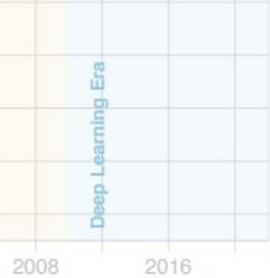
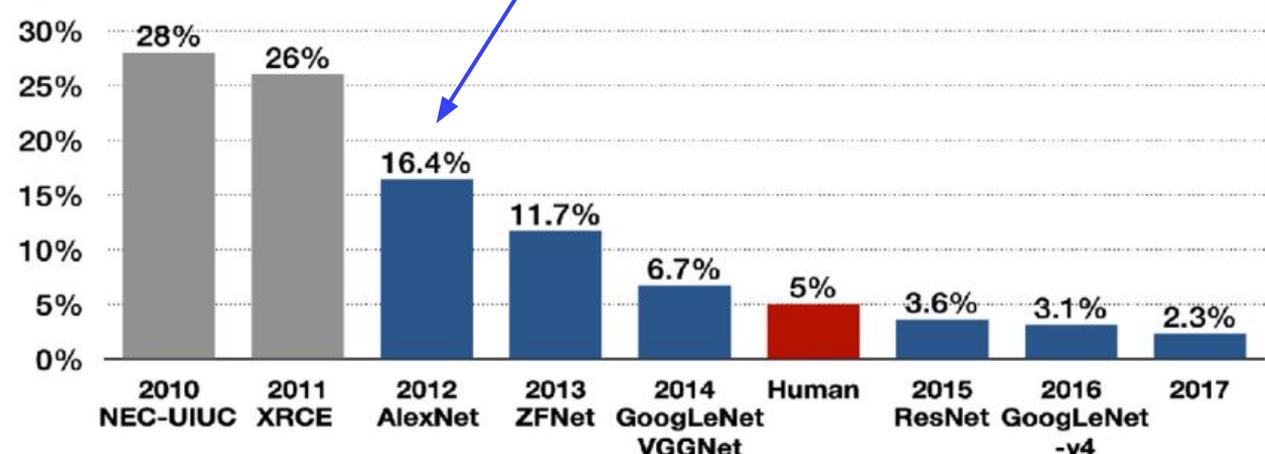
Training com
n = 121



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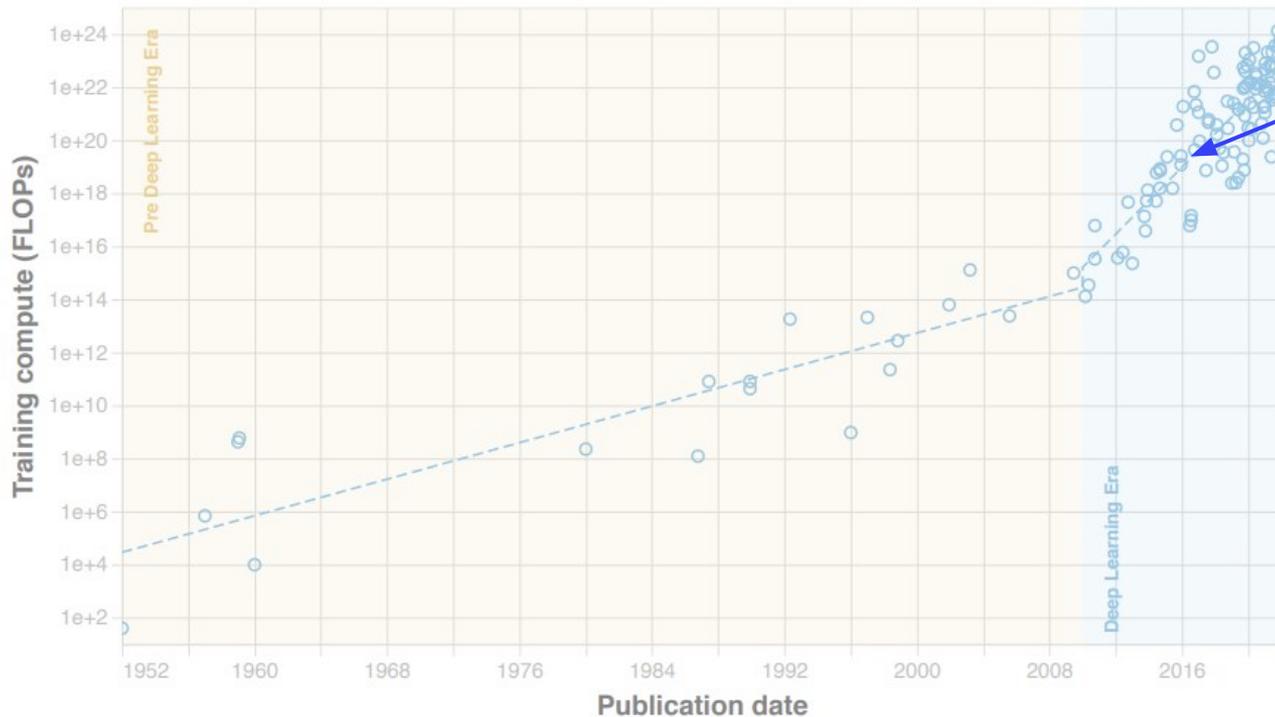


Top-5 error



Training compute (FLOPs) of milestone Machine Learning systems over time

n = 121



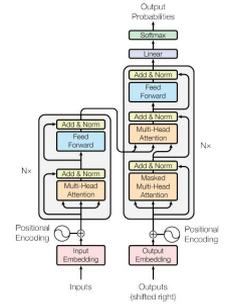
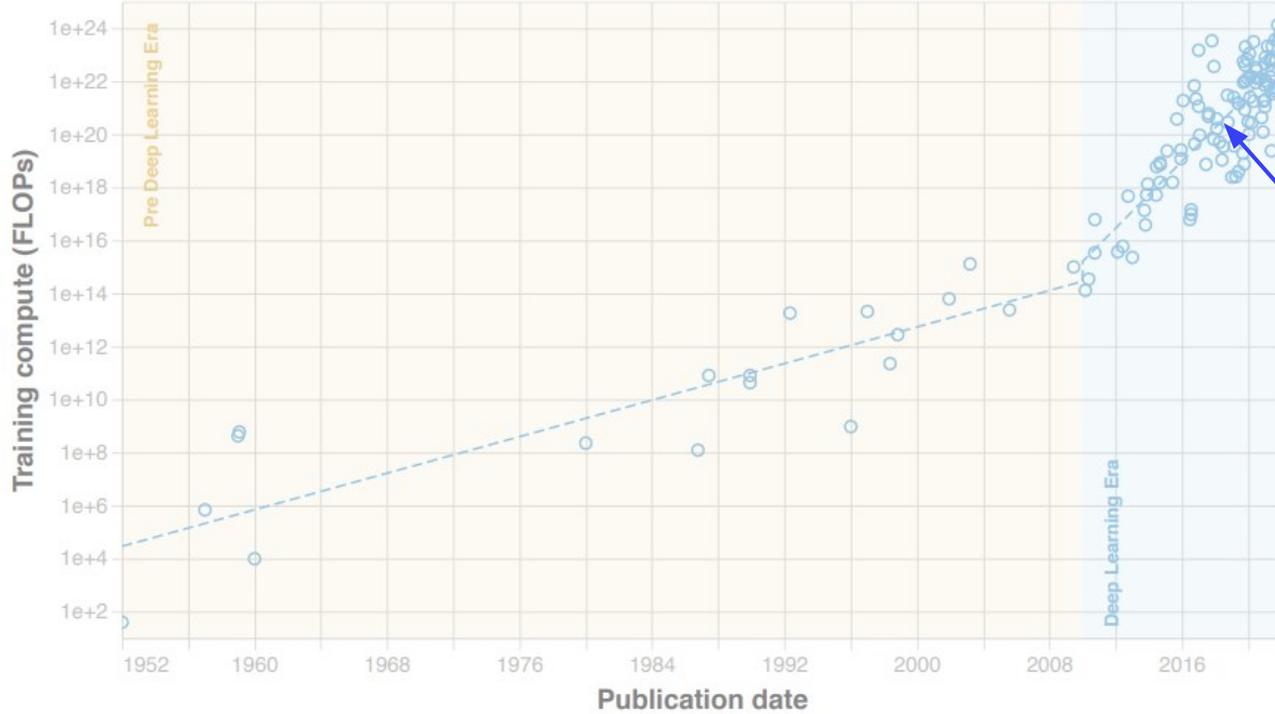
2016

AlphaGo

Reinforcement Learning (RL)

Training compute (FLOPs) of milestone Machine Learning systems over time

n = 121

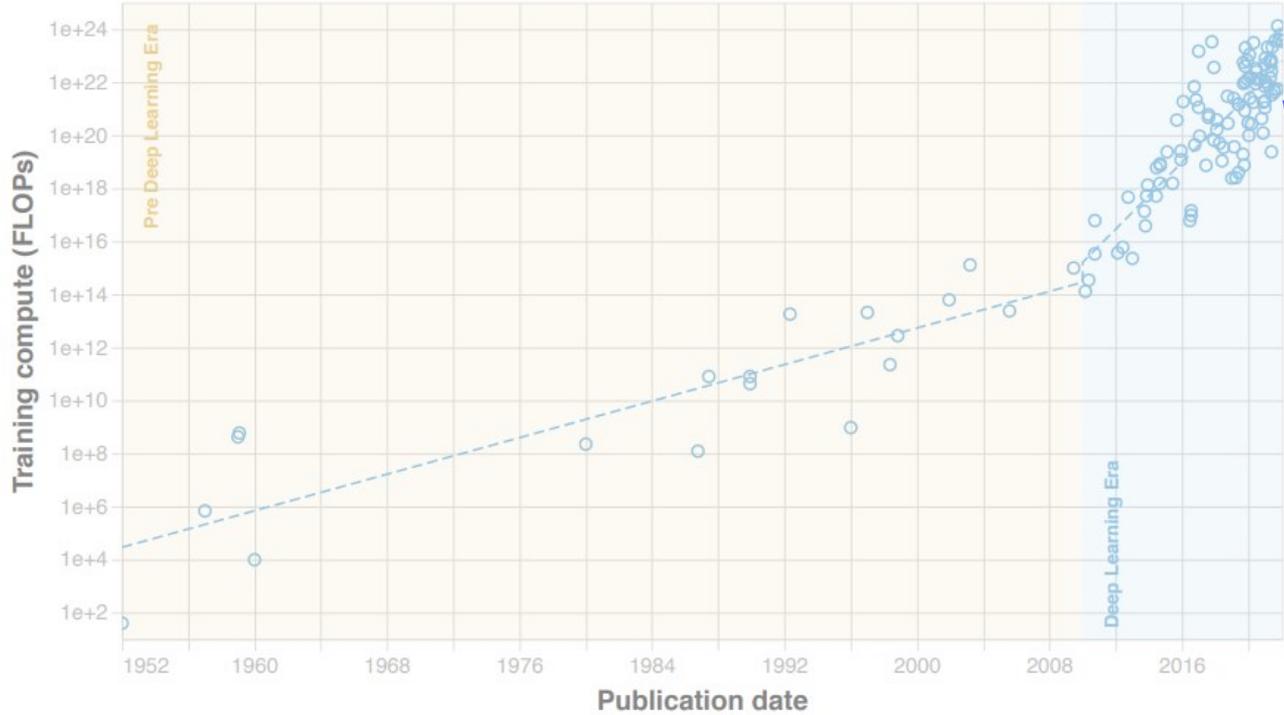


2017

Transformers

Training compute (FLOPs) of milestone Machine Learning systems over time

n = 121



2023

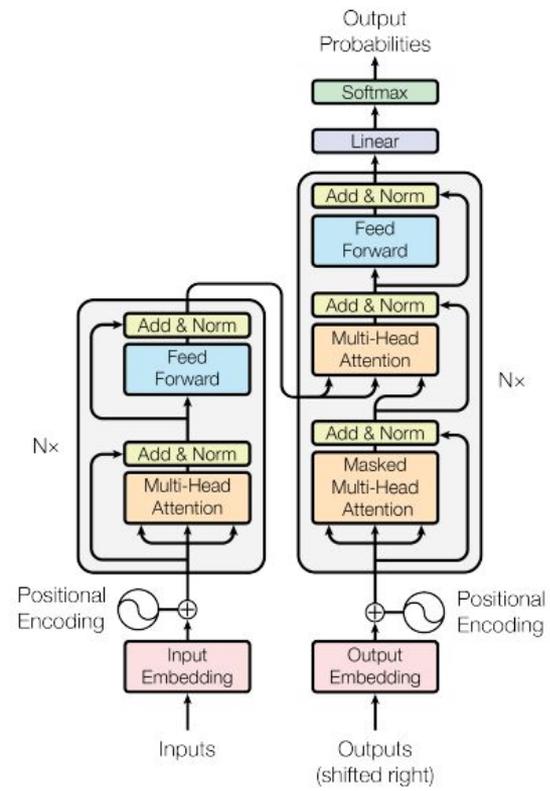
GPT

1. Historical perspective

- 2. The Transformer shake-up**

3. The new landscape

What is a Transformer?



No, seriously... what is a Transformer?

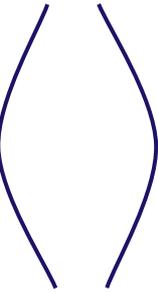
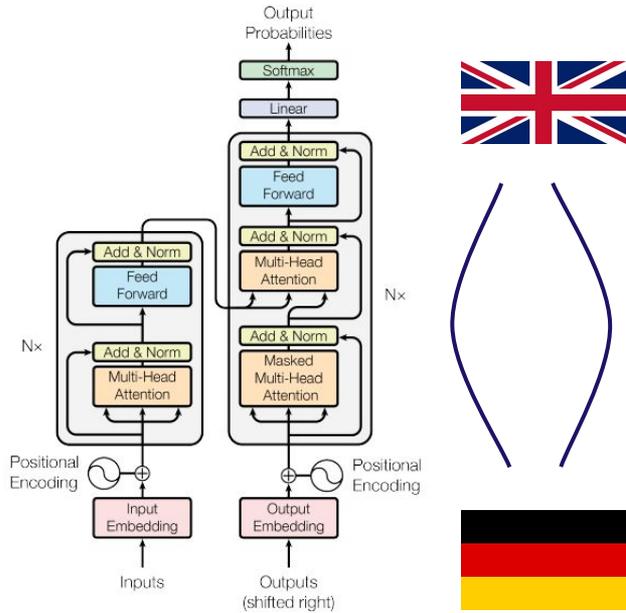
Take the (free) [youtube course](#) by Andrej Karpathy

Read Lilian Weng's [blog post](#)

Attention is a communication mechanism between tokens in a document, with a positional embedding allowing to incorporate ordering into sets of tokens.

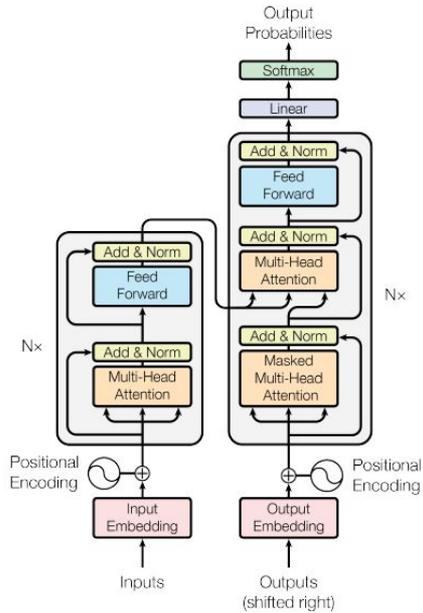
$$\text{Attention}(\mathbf{Q}, \mathbf{K}, \mathbf{V}) = \text{softmax}\left(\frac{\mathbf{Q}\mathbf{K}^\top}{\sqrt{d_k}}\right)\mathbf{V}$$

Transformers shake up translation



Abraham Lincoln (February 12, 1809 – April 15, 1865) was an American lawyer, politician and statesman who served as the 16th president of the United States from 1861 until his assassination in 1865. Lincoln led the Union through the American Civil War to defend the nation as a constitutional union and succeeded in abolishing slavery,...

Abraham Lincoln (12. Februar 1809 – 15. April 1865) war ein amerikanischer Anwalt, Politiker und Staatsmann, der von 1861 bis zu seiner Ermordung im Jahr 1865 als 16. Präsident der Vereinigten Staaten diente. Lincoln führte die Union durch den Amerikanischen Bürgerkrieg, um die Nation als verfassungsmäßige Union zu verteidigen, und war erfolgreich bei der Abschaffung der Sklaverei.



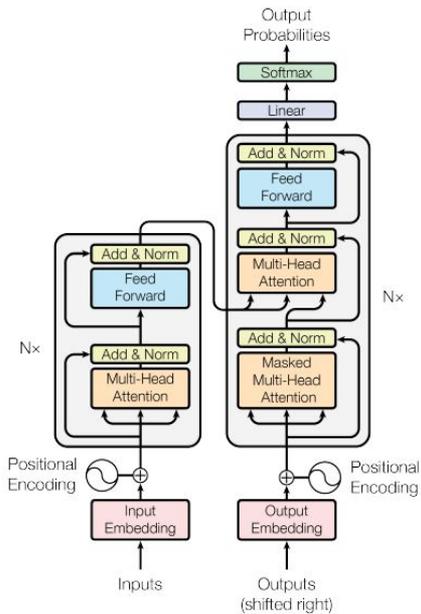
Abraham Lincoln (February 12, 1809 – April 15, 1865) was an American lawyer, politician and statesman who served as the 16th president of the United States from 1861 until his assassination in 1865. Lincoln led the Union through the American Civil War to defend the nation as a constitutional union and succeeded in abolishing _____,...



Likelihood of “slavery”: 99.2%

Likelihood of “making”: 0.1%

Likelihood of “food”: 3%



Context length

Abraham Lincoln (February 12, 1809 – April 15, 1865) was an American lawyer, politician and statesman who served as the 16th president of the United States from 1861 until his assassination in 1865. Lincoln led the Union through the American Civil War to defend the nation as a constitutional union and succeeded in abolishing _____,...

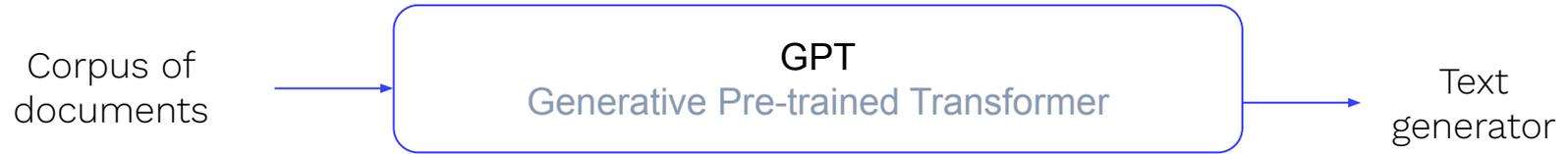


Likelihood of “slavery”: 99.2%

Likelihood of “making”:

Likelihood of “food”:

Transformers for text generation



Try it yourself!



Write With Transformer

Get a modern neural network to
auto-complete your thoughts.

This web app, built by the Hugging Face team, is the official demo of the
`🤗/transformers` repository's text generation capabilities.



<https://transformer.huggingface.co/>

Christopher Columbus discovered America in 1492. When he arrived, he saw a land full of unknown natives that would eventually become the United States . And while it's true that this discovery is one of the most famous events in our history, he didn't come across all the people he would meet , nor all the artifacts he would find .

OK, this is cool, but it hallucinates 100%.

The ChatGPT epiphany

But we know how to train a machine to learn from feedback!

It is called Reinforcement Learning (RL).



Go World Champion Lee Sedol looking confused at move 37 by AlphaGo...

The ChatGPT epiphany

“This whole thing was just an experiment. We had no idea it would work so well”, Ilya Sutskever
[\(full interview\)](#)



The main breakthrough of LLM is the discovery that training a very large model on the simplest task (predict the next token) could yield remarkable intelligence.

<philosophy>

Transformers are putting humankind through an existential crisis.

With Copernicus, we discovered that we were not at the center of the universe.

With GPT, we discover that (maybe) we are not that intelligent after all.

</philosophy>



Public API

Human analysts



RLHF

Reinforcement Learning from Human Feedback

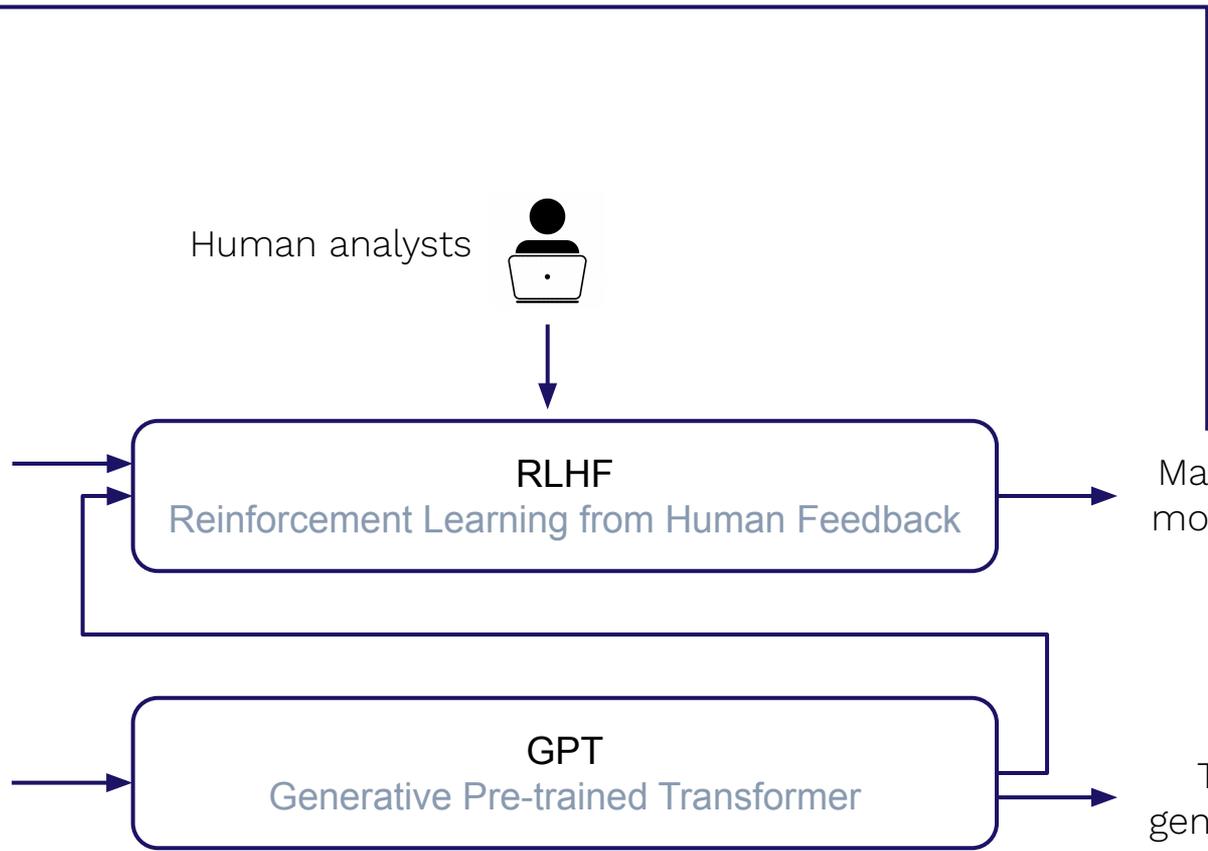
Magic model

Corpus of documents

GPT

Generative Pre-trained Transformer

Text generator





GPT

RLHF

Midjourney

“Man hoarding tigers with a whip, black and white drawing”

The science of human feedback (alignment)

Step 1

Collect demonstration data and train a supervised policy.

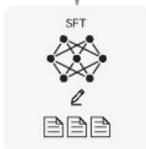
A prompt is sampled from our prompt dataset.



A labeler demonstrates the desired output behavior.



This data is used to fine-tune GPT-3.5 with supervised learning.



Step 2

Collect comparison data and train a reward model.

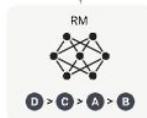
A prompt and several model outputs are sampled.



A labeler ranks the outputs from best to worst.



This data is used to train our reward model.



Step 3

Optimize a policy against the reward model using the PPO reinforcement learning algorithm.

A new prompt is sampled from the dataset.



The PPO model is initialized from the supervised policy.



The policy generates an output.



The reward model calculates a reward for the output.



The reward is used to update the policy using PPO.



1. Historical perspective
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The Cambrian explosion of foundational models



Mistral AI

Experimentation is getting much easier



2010

Write your own backprop
manually



2018

```
loss.backward()
```



2023

```
from transformers  
import SwinModel
```

Data is (relatively) small

Crowd thinking: "GPT-3 was trained on a huge chunk of the internet"

The reality:

GPT-2 was trained on 40GB of text

GPT-3 was trained on 570GB of text (filtered from 50TB of data)

Web2Text weighs 65GB

Stable Diffusion was trained on LAION-400M (240TB). Relatively small.

Raw data is cheap, clean data is expensive

Lots of data cleaning tricks⁽¹⁾

Clean data beats more data

Data requires significant ML effort

(1) [DALL-E pretraining mitigations](#), OpenAI

Complex models are getting cheaper

New GPUs provide 5-10x speed-up every few years (H100 >> A100)

- Stable Diffusion cost \$600K to train in 2021.
- Today, the same model would cost \$60K.

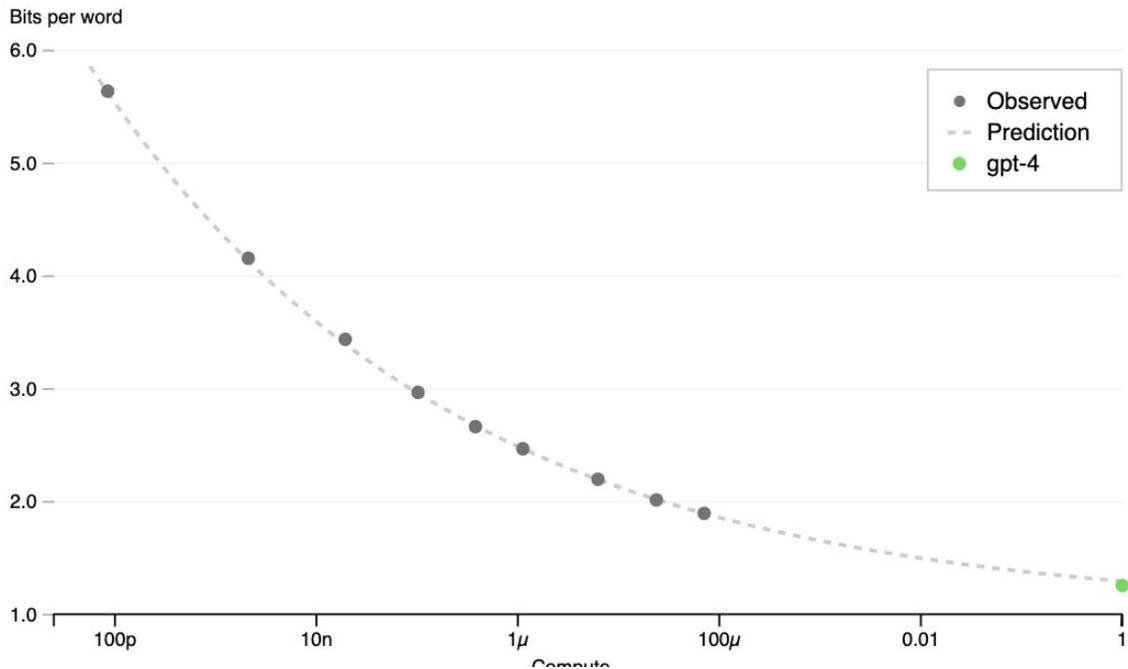
More efficient Transformers are coming.

- Transformers are quadratic in complexity.
- $n \cdot \log n$ Transformers are on their way.
- A \$50M job becomes a \$100K job.

You can finetune a 7B LLM on a single GPU in 1-2 h using techniques like low-rank adaptation ([LoRA](#)).

Model performance can be predicted with 1/1,000th of the compute.

OpenAI codebase next word prediction



Pre-training is still very hard

Goal: get a 175B dense model up and running by any means necessary.

Solution:

[114 pages of OPT175B logbook](#)

Research is far from over



Durk Kingma @dpkingma · Jun 9



Why do LLMs hallucinate? TL;DR:

- LLM pretraining generally results in a well-calibrated distribution $p(\text{completion}|\text{prompt})$
- Obviously, a single sample from this distribution doesn't express the uncertainty contained in it
- This is fixable, e.g. through sufficient RLHF/RLAIF



Yann LeCun @ylecun · Jun 10



Replying to [@dpkingma](#)

I disagree. I don't think it's fixable within the auto-regressive prediction paradigm.



Summary

Data is small. Clean data is expensive.

Experimentation is (much) easier

Models are getting cheaper

Pre-training is (still) hard and expensive

Research is far from over



Take-aways for CTOs

1

System complexity

2

Hallucinations

3

Pre-training is hard

4

What is the ROI?

A photograph of the Golden Gate Bridge in San Francisco, taken from a low angle on the left side of the bridge. The bridge's red-orange towers and cables are silhouetted against a deep blue, twilight sky. The water of the bay is dark blue, and the distant hills are visible in the background. The overall mood is serene and quiet.

Thank you!

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Foundational models vs EU AI Act

Stanford Institute for Human-Centered Artificial Intelligence (HAI)

Grading Foundation Model Providers' Compliance with the Draft EU AI Act

Source: Stanford Center for Research on Foundation Models (CRFM), Institute for Human-Centered Artificial Intelligence (HAI)

	OpenAI	cohere	stability.ai	ANTHROPIC	Google	Bloom	Meta	AI21 labs	ALEPH ALPHA	EleutherAI	Totals
Draft AI Act Requirements	GPT-4	Cohere Command	Stable Diffusion v2	Claude	PaLM 2	BLOOM	LLaMA	Jurassic-2	Luminous	GPT-NeoX	
Data sources	● ○ ○ ○	● ● ● ○	● ● ● ●	○ ○ ○ ○	● ● ● ○	● ● ● ●	● ● ● ●	○ ○ ○ ○	○ ○ ○ ○	● ● ● ●	22
Data governance	● ● ○ ○	● ● ● ●	● ● ● ○	○ ○ ○ ○	● ● ● ●	● ● ● ●	● ● ○ ○	○ ○ ○ ○	○ ○ ○ ○	● ● ● ○	19
Copyrighted data	○ ○ ○ ○	○ ○ ○ ○	○ ○ ○ ○	○ ○ ○ ○	○ ○ ○ ○	● ● ● ○	○ ○ ○ ○	○ ○ ○ ○	○ ○ ○ ○	● ● ● ●	7
Compute	○ ○ ○ ○	○ ○ ○ ○	● ● ● ●	○ ○ ○ ○	○ ○ ○ ○	● ● ● ●	● ● ● ●	○ ○ ○ ○	● ○ ○ ○	● ● ● ●	17
Energy	○ ○ ○ ○	● ● ○ ○	● ● ● ○	○ ○ ○ ○	○ ○ ○ ○	● ● ● ●	● ● ● ●	○ ○ ○ ○	○ ○ ○ ○	● ● ● ●	16
Capabilities & limitations	● ● ● ●	● ● ● ●	● ● ● ●	● ○ ○ ○	● ● ● ●	● ● ● ○	● ● ● ●	● ● ○ ○	● ○ ○ ○	● ● ● ●	27
Risks & mitigations	● ● ● ○	● ● ● ○	● ● ○ ○	● ○ ○ ○	● ● ● ●	● ● ● ○	● ○ ○ ○	● ● ○ ○	○ ○ ○ ○	● ○ ○ ○	16
Evaluations	● ● ● ●	● ● ○ ○	○ ○ ○ ○	○ ○ ○ ○	● ● ○ ○	● ● ● ○	● ● ○ ○	○ ○ ○ ○	● ○ ○ ○	● ○ ○ ○	15
Testing	● ● ● ○	● ● ○ ○	○ ○ ○ ○	○ ○ ○ ○	● ● ○ ○	● ● ● ○	○ ○ ○ ○	● ○ ○ ○	○ ○ ○ ○	○ ○ ○ ○	10
Machine-generated content	● ● ● ●	● ● ● ●	○ ○ ○ ○	● ● ● ●	● ● ● ●	● ● ● ●	○ ○ ○ ○	● ● ● ●	● ● ● ●	● ● ● ●	21
Member states	● ● ○ ○	○ ○ ○ ○	○ ○ ○ ○	● ● ○ ○	● ● ● ●	○ ○ ○ ○	○ ○ ○ ○	○ ○ ○ ○	● ○ ○ ○	● ● ○ ○	9
Downstream documentation	● ● ● ○	● ● ● ●	● ● ● ●	○ ○ ○ ○	● ● ● ●	● ● ● ●	● ● ● ○	○ ○ ○ ○	○ ○ ○ ○	● ● ● ○	24
Totals	25 / 48	23 / 48	22 / 48	7 / 48	27 / 48	36 / 48	21 / 48	8 / 48	5 / 48	29 / 48	

Close vs open source

Two drivers for close source:

1. competitive pressure
2. safety

Open source is usually safer

Open source is going strong

The code is not enough (data matters)

The right question: is it safe?

Carbon footprint of generative AI?

Gen. AI has a significant CO2 footprint

Can the footprint be compensated with the benefits of generative AI for the planet?