

```
// You'll need  
// com.google.  
listRef.listAll  
.addOr  
prefixes.  
// ALL  
// You  
}  
it  
each { item  
the items  
}  
}
```

# devfest



## AI for Safer AI

ChengCheng Tan

 Google Developer Groups  
Sacramento



# AI & LLM Overview

1. Clarify some terms
2. Transformers
3. Large Language Models
4. Pre-training vs Fine-tuning

## LLMs in AI Safety

1. Clarify more terms
2. What is AI safety?
3. Examples of technical research
4. Resources

```
Text(
  'Section Title',
  style: TextStyle(
    color: Colors.green[200],
  ),
),
),
```

**devfest**

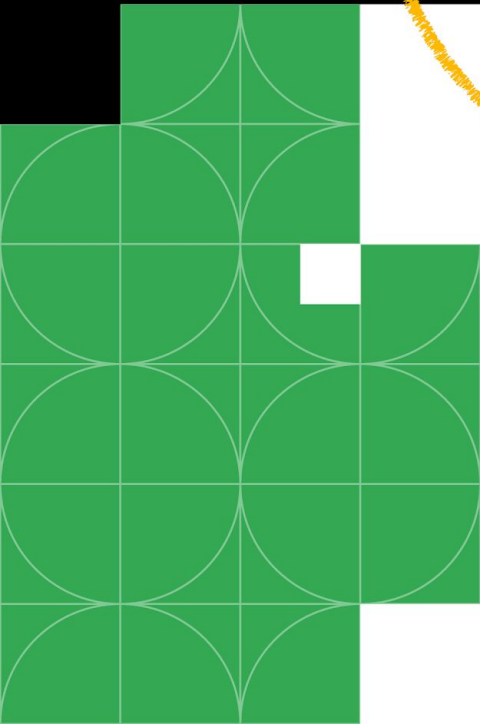
```
s.star,
r: Colors.green[500],
Text('23'),
```



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# AI & LLM Overview



# Terminology

**AI:** Artificial Intelligence

**ML:** Machine Learning

**LLM:** Large Language Model

**NLP:** Natural Language Processing

**GPT:** Generative Pretrained Transformer

**Transformers:** Neural network leading to LLMs

**RLHF:** Reinforcement Learning from Human Feedback

Natural Language  
Processing [NLP]:  
**Computers can  
speak & understand  
human languages**



# Pre-1990s: Rule-based Expert Systems



# 1990s-2000s: **Statistics & Probabilities**





**You shall know a  
word by the  
company it keeps**

J.R. Firth, Linguist

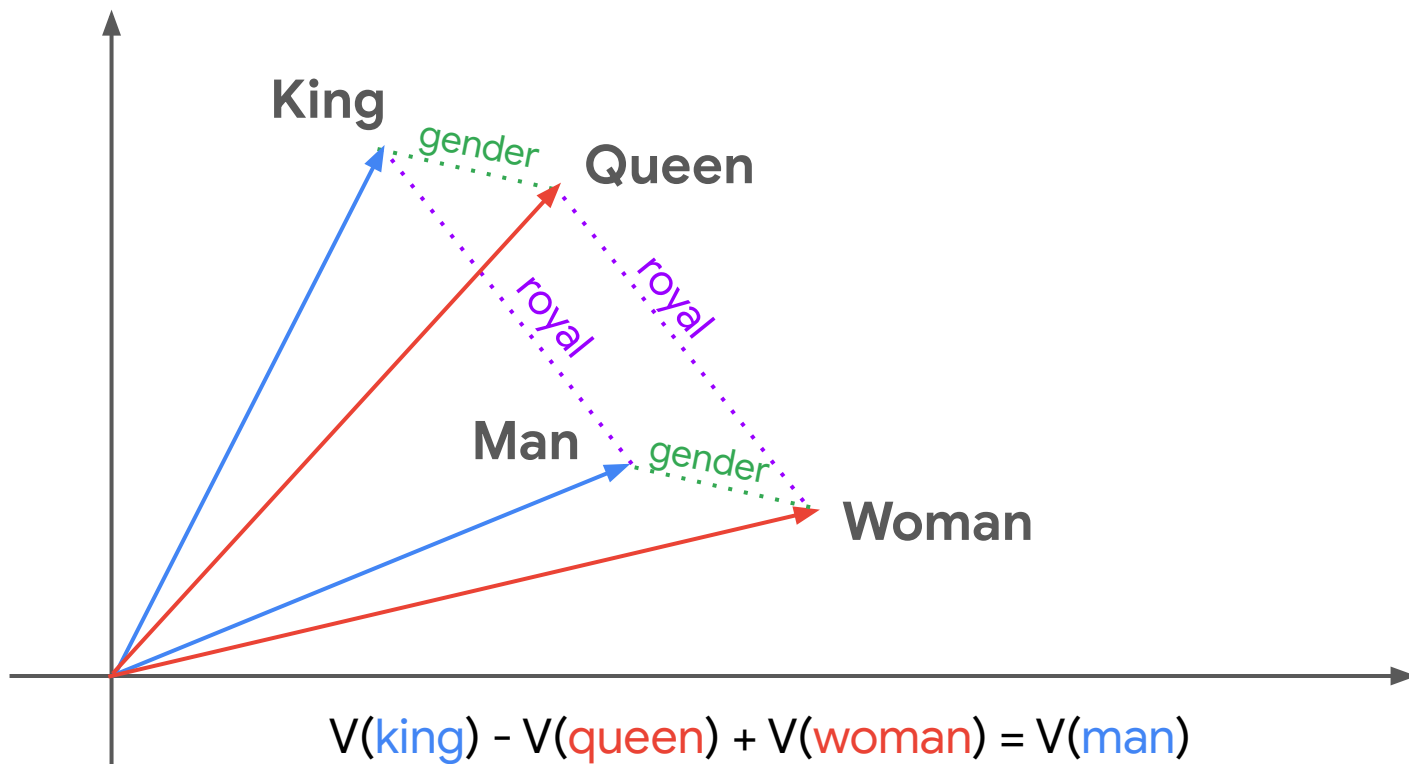




2010s:  
**Rise of Deep Learning  
and Neural Networks**



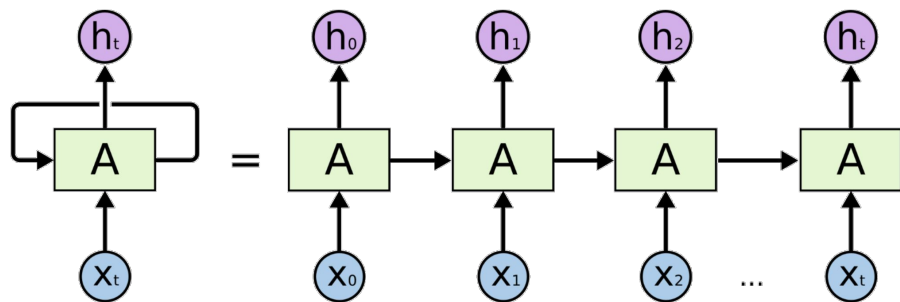
## 2013: Word2Vec Embeddings



## 2013: Word2Vec Embeddings

Analogies	Word Pair 1		Word Pair 2	
Man-Woman	king	queen	man	woman
Capital city	Athens	Greece	Oslo	Norway
City-in-state	Chicago	Illinois	Sacramento	California
Opposite	possibly	impossibly	ethical	unethical
Comparative	great	greater	tough	tougher
Nationality adjective	Switzerland	Swiss	Canada	Canadian
Past tense	walking	walked	swimming	swam
Plural nouns	mouse	mice	dollar	dollars

# 2010s: Neural Networks RNN, GRU, LSTM



# Early Neural Networks

- Slow & forgetful



## 2017: Transformers

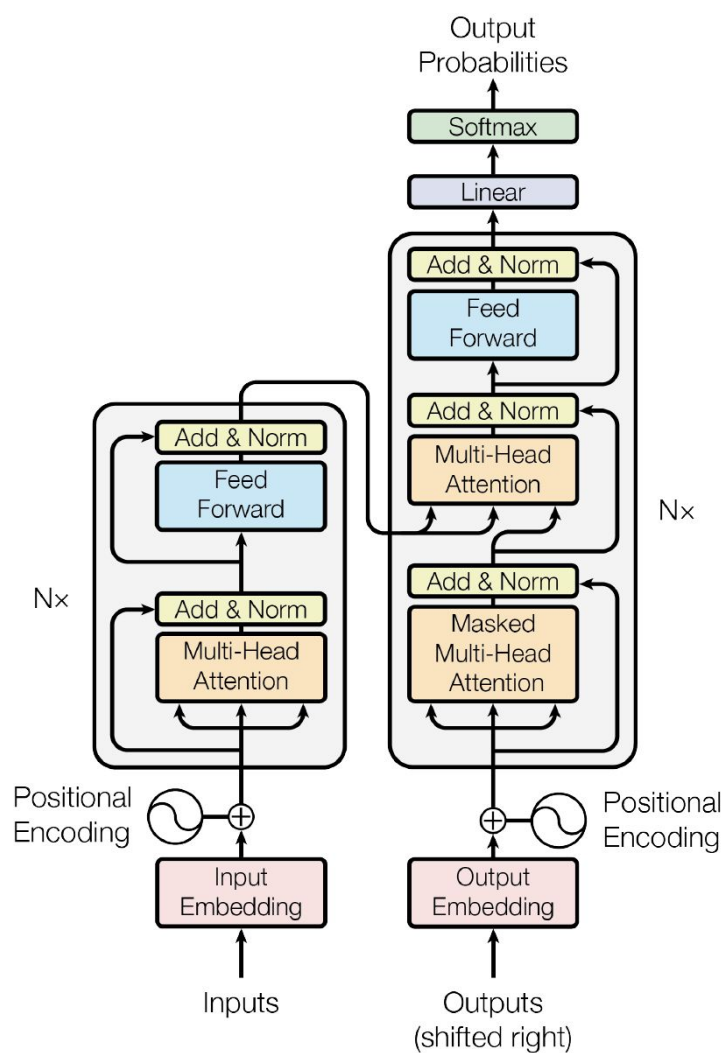
- Attention mechanism
- Parallel processing





# Transformer Architecture

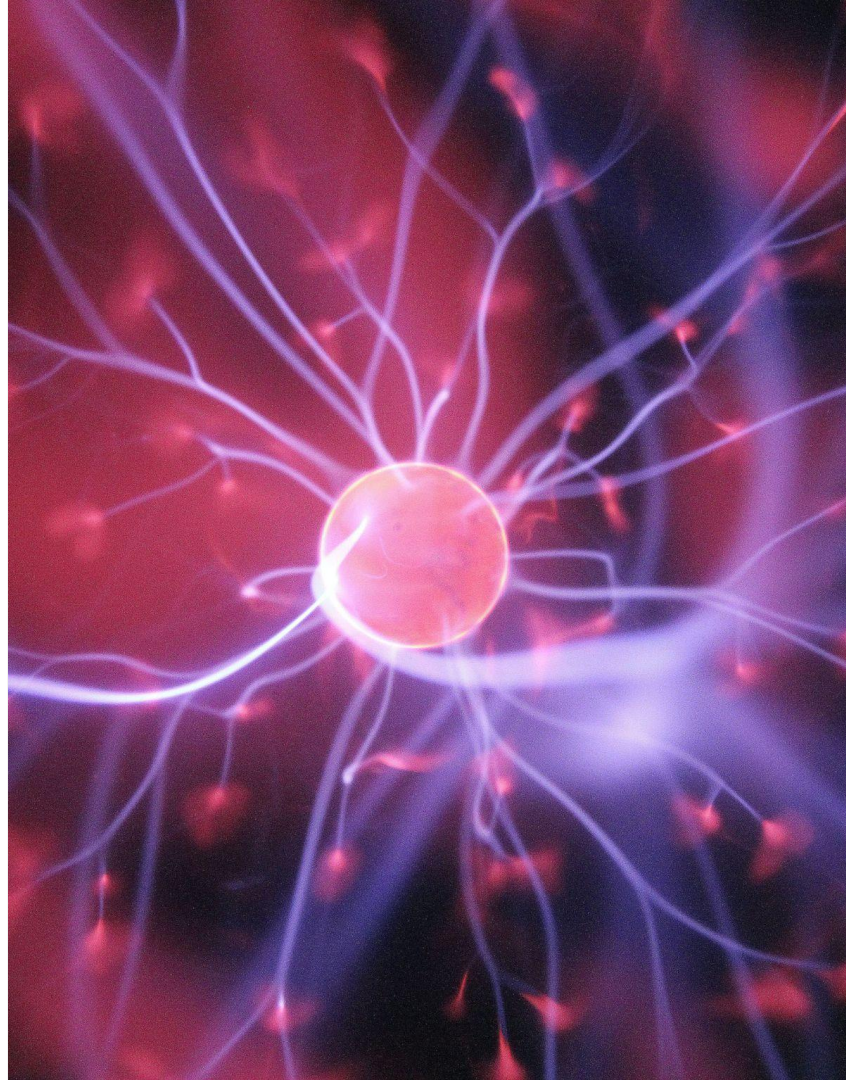
## Encoder + Decoder



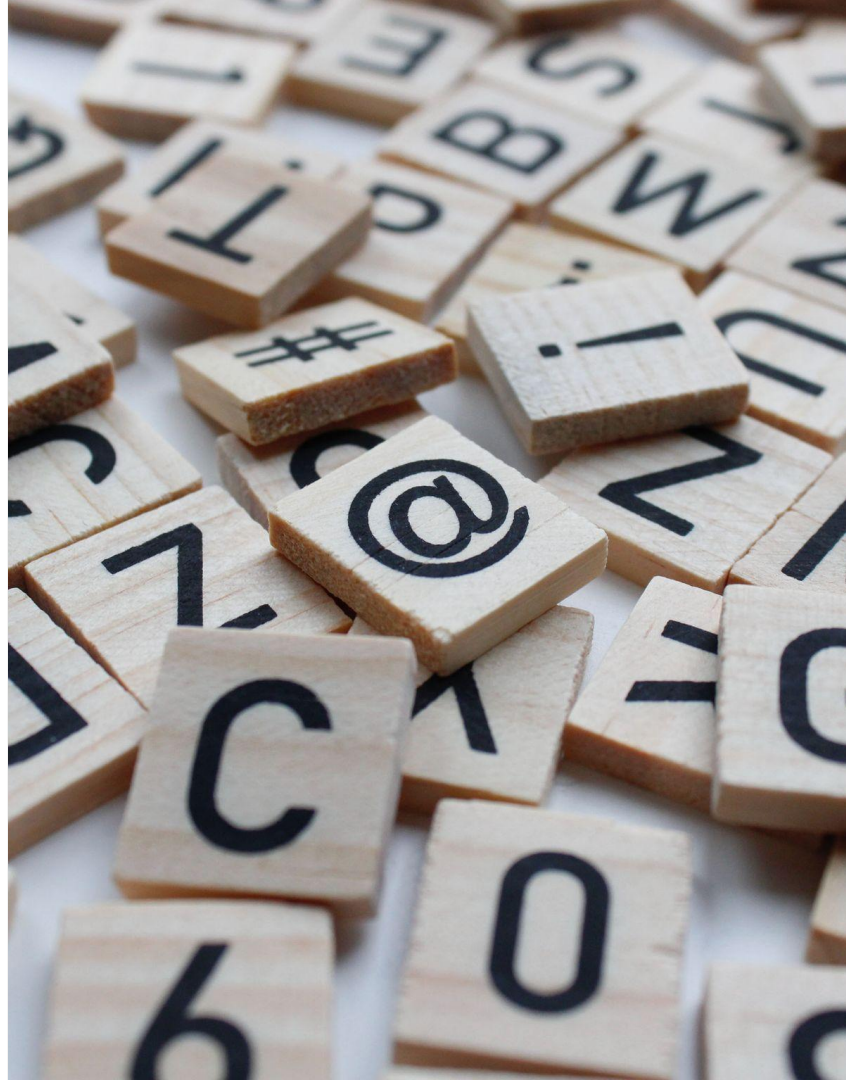


# Rise of LLMs

>1 billion neurons



Trained for  
**next word prediction**



# Pre-trained base VS Fine-tuned models





# RLHF:

## Reinforcement Learning from Human Feedback



Fine-tuned to  
**follow instructions**



# Fine-tuned for conversations



# What's Next?

- Multimodal
- Open-source
- Agents









# More Terminology

**AGI**: Artificial GENERAL Intelligence, Strong AI

**ANI**: Artificial NARROW Intelligence, Weak AI

**HLI**: Human-Level Intelligence

**ASI**: Artificial SUPERINTELLIGENCE surpass humans on all tasks

**Orthogonality Thesis**: intelligence & goals are independent

**Takeoff**: fast vs slow

**Timelines**: short vs long

# ANI

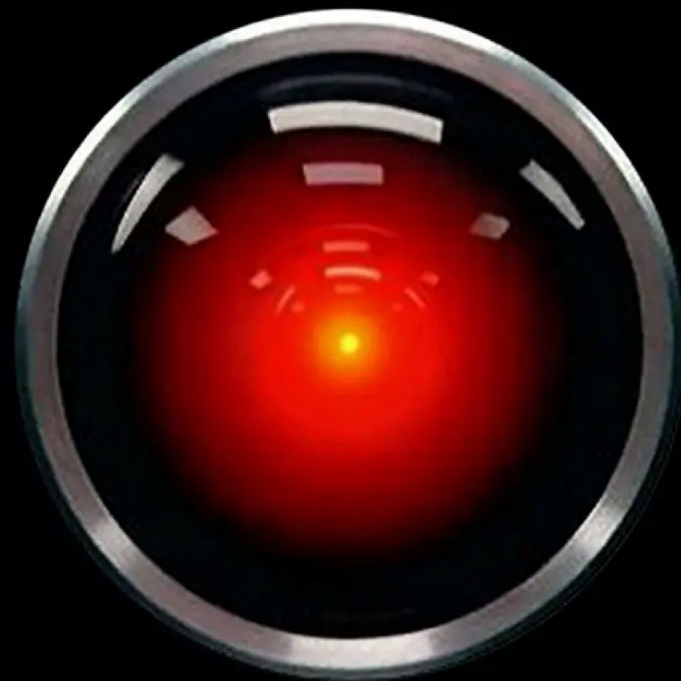
Artificial Narrow Intelligence

Weak AI

# AGI

Artificial General Intelligence

Strong AI



# Intelligence:

- Human-level
- Superintelligence
- ... Singularity



# Takeoff

Fast vs Slow

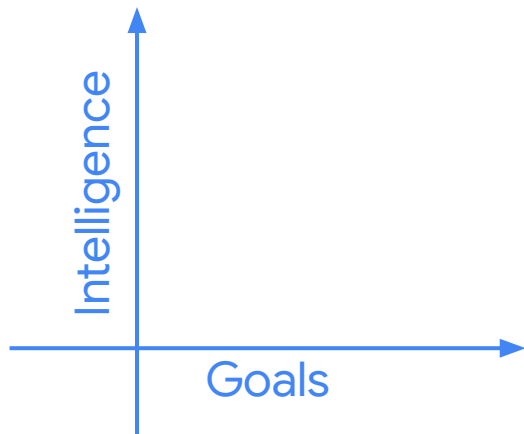
# Timelines

Short vs Long



# Orthogonality Thesis

Intelligence & Goals are independent





# Paperclip Maximizer

Thought experiment  
or stamp alternative

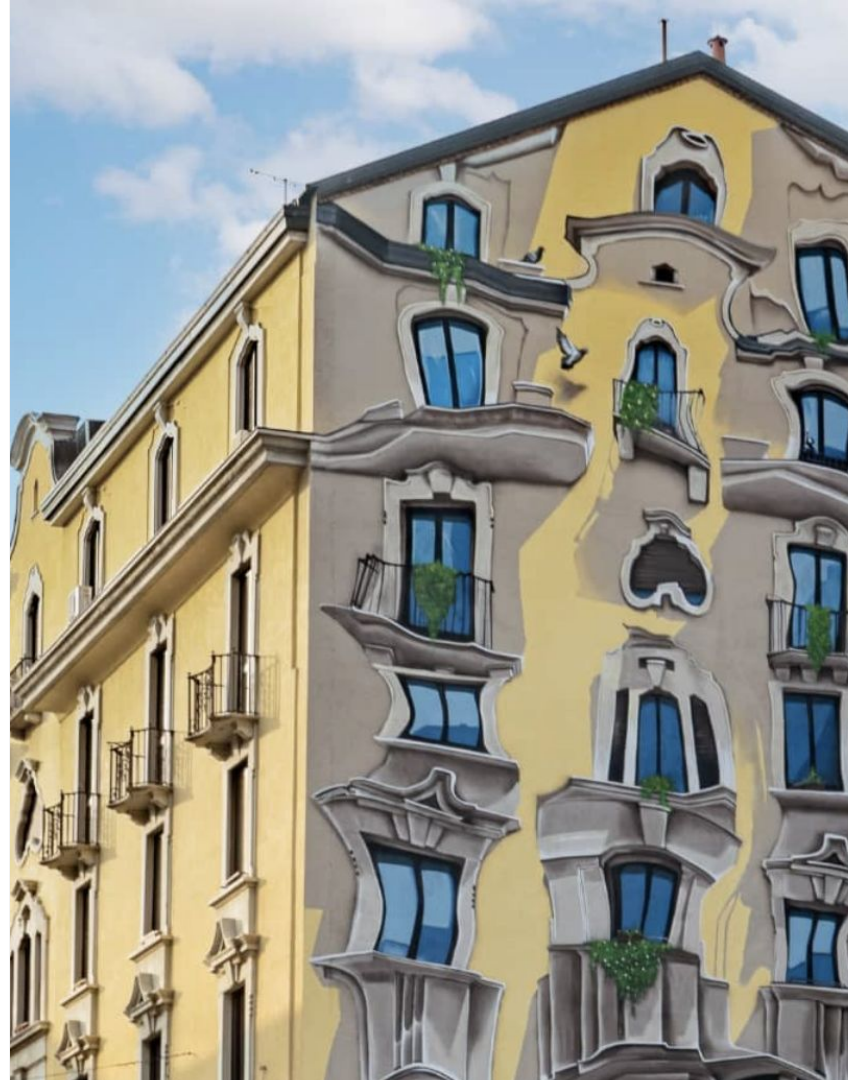


## Outer Misalignment

AI creator goals don't align with general human values

## Inner Misalignment

AI achieves goals in ways unintended by its creators





# Gorilla Problem

Can humans maintain  
autonomy in a world with  
superintelligence?



# Some Concrete Examples of Technical Research...



# Alignment: RLHF Human Values

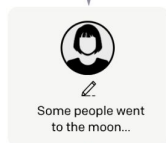
Step 1

**Collect demonstration data, and train a supervised policy.**

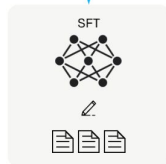
A prompt is sampled from our prompt dataset.



A labeler demonstrates the desired output behavior.



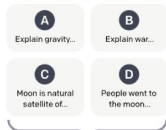
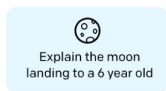
This data is used to fine-tune GPT-3 with supervised learning.



Step 2

**Collect comparison data, and train a reward model.**

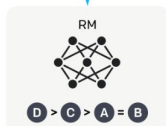
A prompt and several model outputs are sampled.



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 reinforcement learning.**

A new prompt is sampled from the dataset.



The policy generates an output.



The reward model calculates a reward for the output.

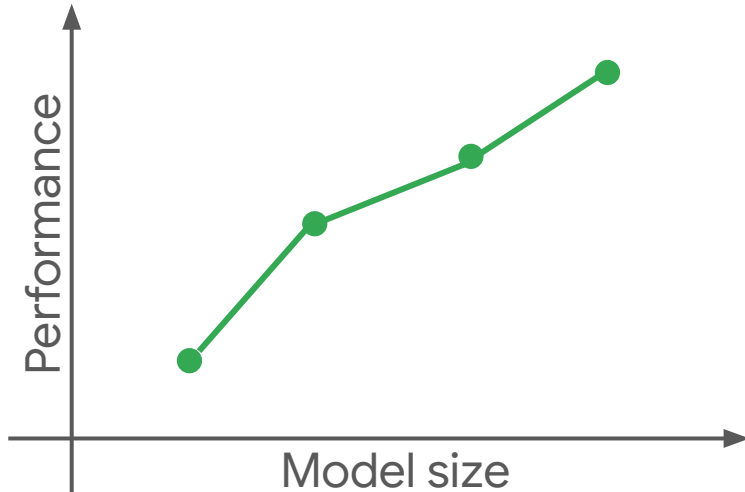


The reward is used to update the policy using PPO.

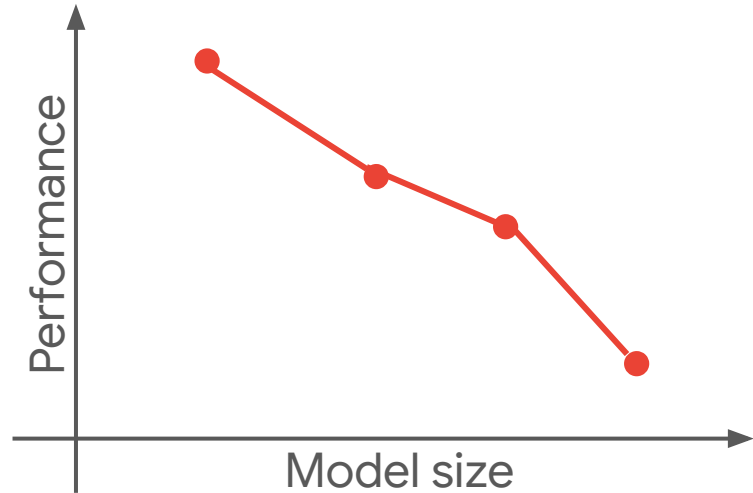


# LLM Output Evaluation: Inverse Scaling Laws

Most tasks like this



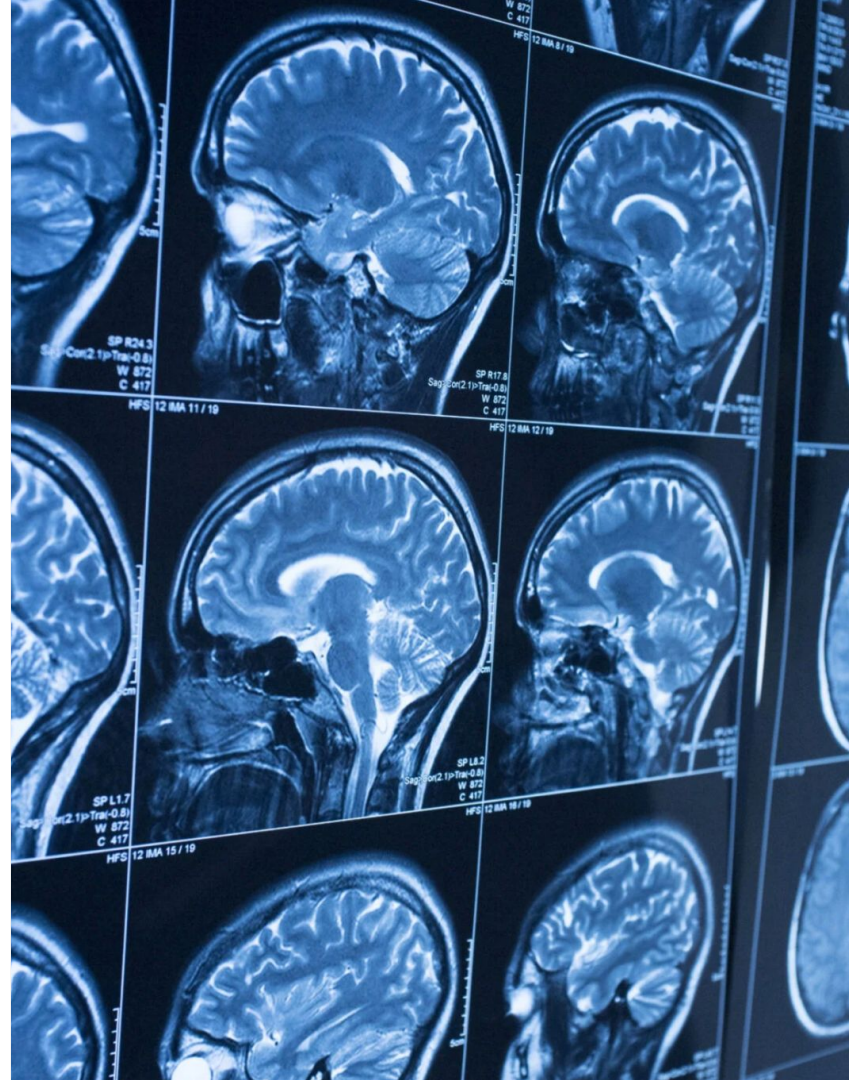
Any tasks like this?





# LLM Internals

## Interpretability







# Resources

1. **AI Explained videos on AI development + safety**  
[youtube.com/@aiexplained-official](https://youtube.com/@aiexplained-official)
2. **80,000 Hours career advice + job board**  
[80000hours.org/problem-profiles/artificial-intelligence](https://80000hours.org/problem-profiles/artificial-intelligence)
3. **AI Safety.info FAQs**  
[AISafety.info](https://AISafety.info)
4. **AI Safety Fundamentals online curricula**  
[AISafetyFundamentals.com](https://AISafetyFundamentals.com)
5. **Alignment Forum share research + discussions**  
[AlignmentForum.org](https://AlignmentForum.org)

Questions?



**Embrace  
Safely**



```
// You'll need  
// com.google.  
listRef.listAll  
.addOr  
prefixes.  
// All  
// You  
}  
it  
each { item  
the items  
}  
}
```

# devfest



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