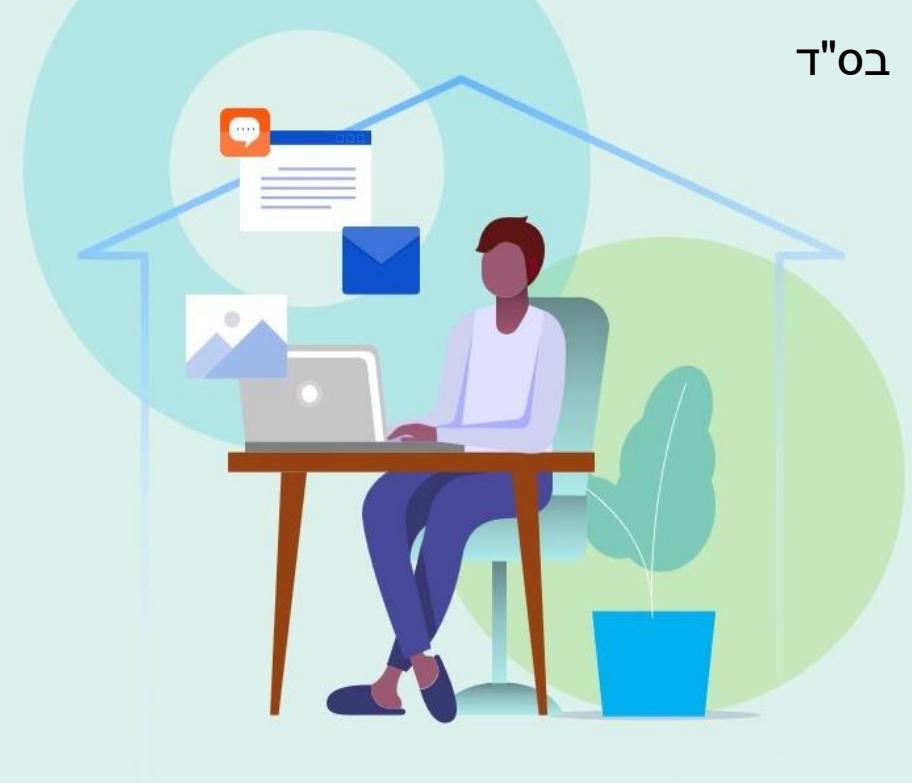


# AI & NLP

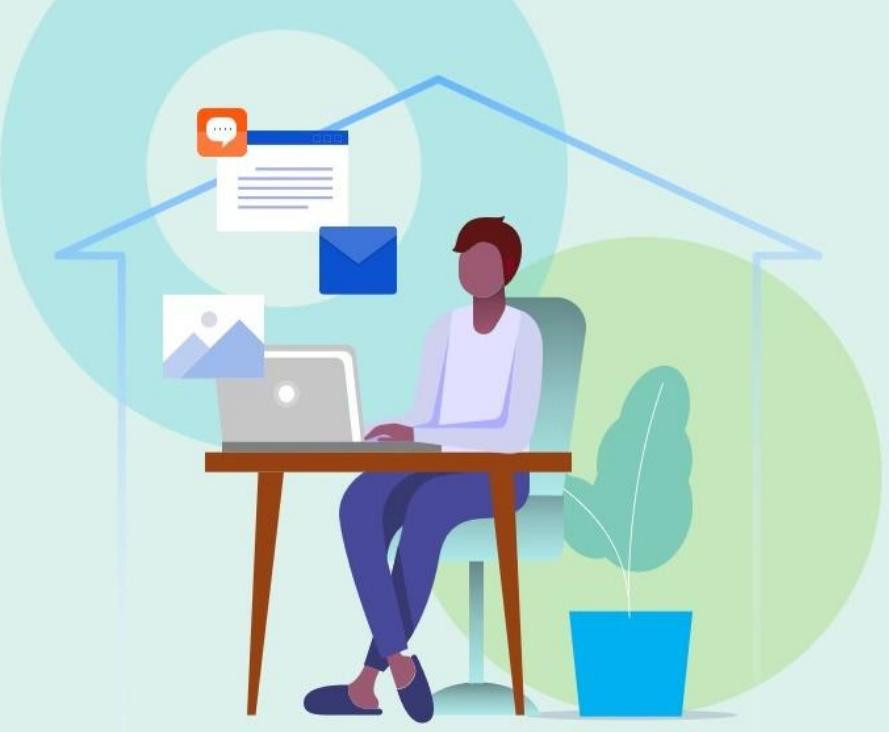
(Natural Language Processing)

# For SW Engineers

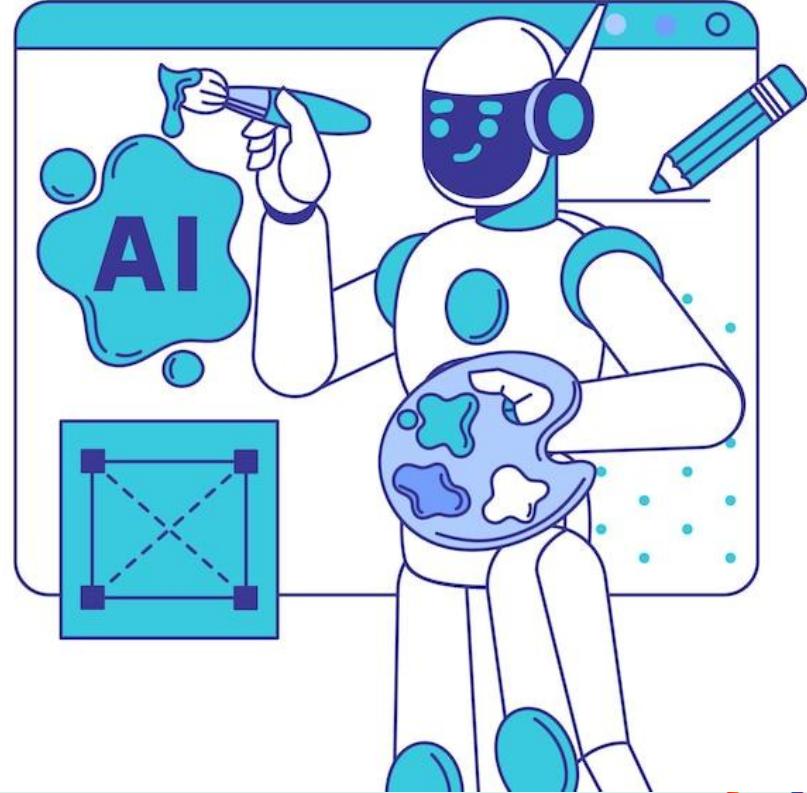


David Berger, 2025  
Software Engineer

1. What is AI?
2. Using AI
3. Review – Activity!
4. What is NLP?
5. An NLP Pipeline
6. Group Activity!



# 1. What is AI? And will it replace me?



# AI vs ML vs Data Science

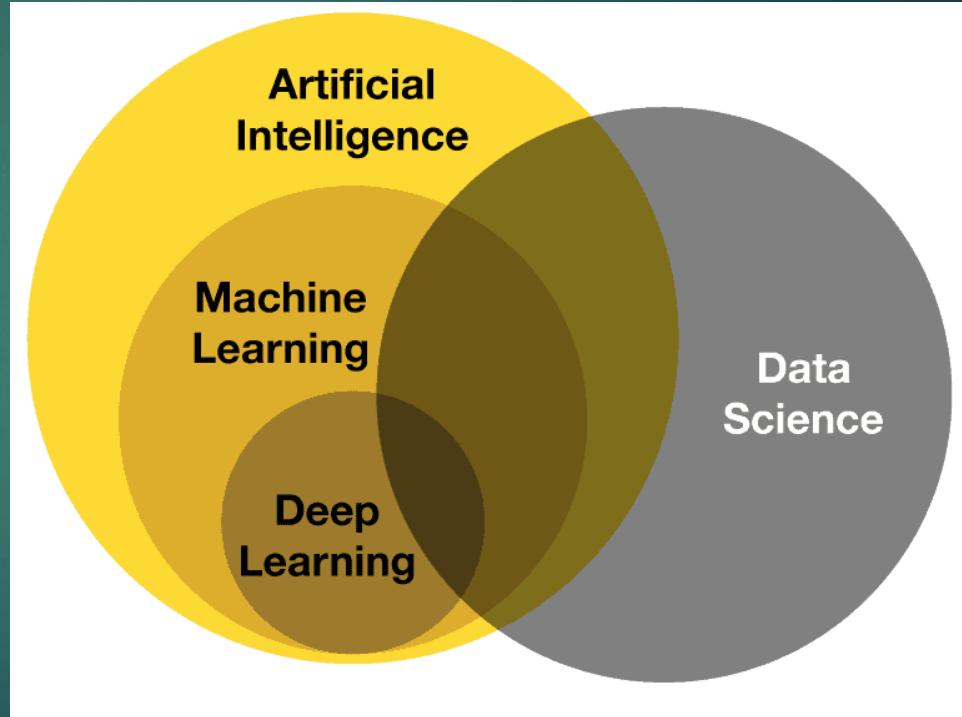
4

**AI**  
Performing human-like operations

**Machine Learning**  
Learning patterns from data

**Deep Learning**  
Complex, multi layered learning

**Data Science**  
Extract insights from Data



# Fields of AI/ML/Data Science

\* Grouping not exact

## Artificial Intelligence

- Natural Language Processing
- Generative AI
- AI Agents
- Knowledge Graphs

## Machine Learning

- Recommendation
- Supervised/  
Unsupervised Learning
- Anomaly Detection

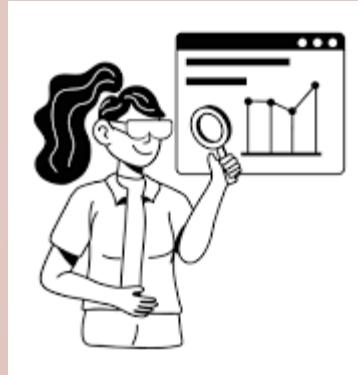
## Deep Learning

- Neural Networks
- Speech Recognition
- Image Recognition
- Transformers

# What is an AI Developer?

## Data Scientist | AI Developer:

- Mathematics
- Research
- Abstract
- Specific
- Accuracy
- Experiments



## SW Developer

- Logic
- Customers, Business
- Tangible / Real World
- Flexible
- Design
- Workflow



# The AI Takeover

"AI is automating tasks in many industries, replacing jobs in customer service, data entry, and even creative fields like writing and design.

As technology advances, companies increasingly rely on AI for efficiency, reducing the need for human workers in repetitive and analytical roles."

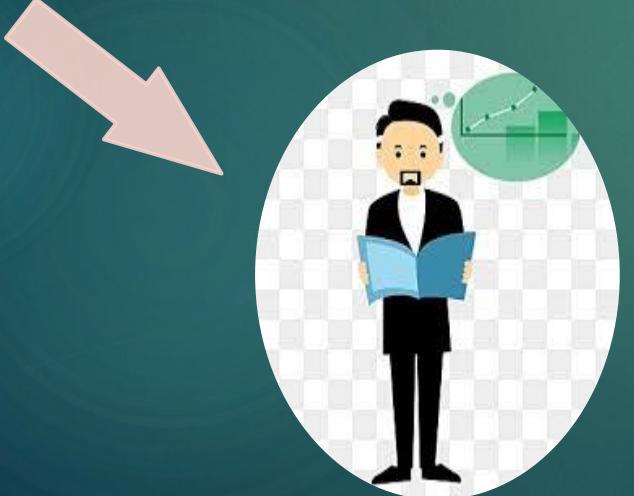


**ChatGPT**

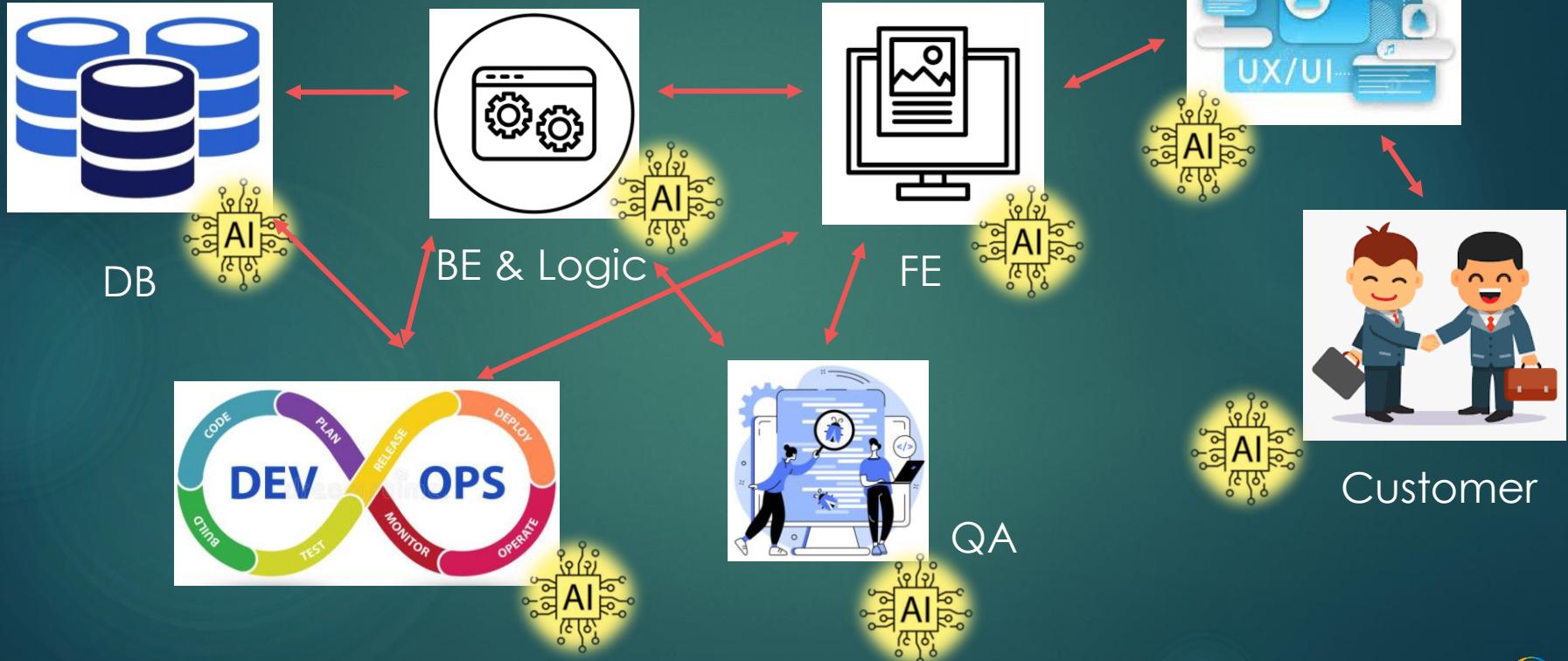
# What about me? (SWEs)

"Our AI model now handles about 50% of software engineering tasks, and we aim for 90% in a year."

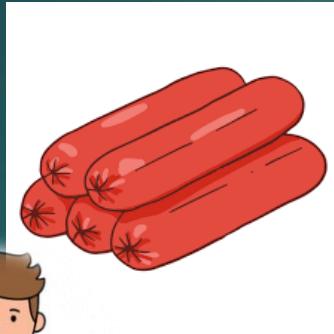
Dario Amodei, CEO of Anthropic (2024)



# Ai Integration



# Complementary Goods ( מוצר משלים )



AI



Software  
Engineers

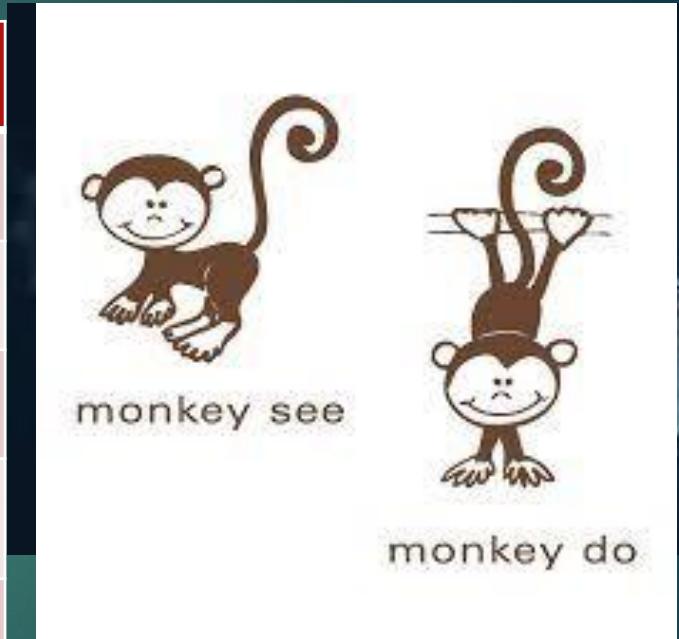


## 2. Using AI (and when *not* to use AI)



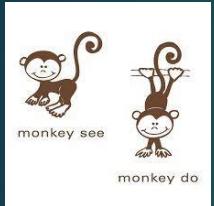
# Gen AI – How it works

Data	Source (context)
Bibi is good	כאן 11
Bibi is bad	ערוץ 12 – קשת
Bibi is evil	ערוץ 13 – רשות
Bibi is a leader	ישראל היום
Bibi is destruction	הארץ



“Bibi is \_\_\_\_ ? ”

# Takeaways

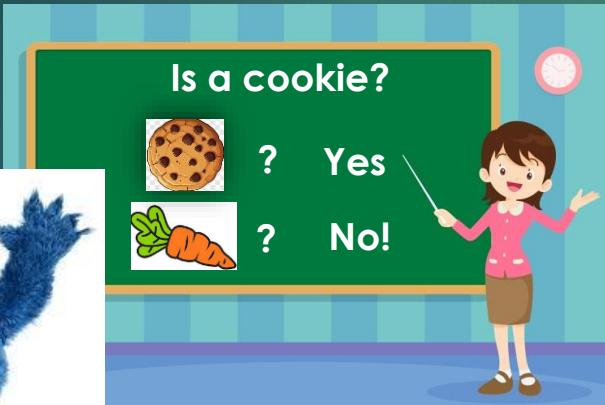


**Gen AI's output:**

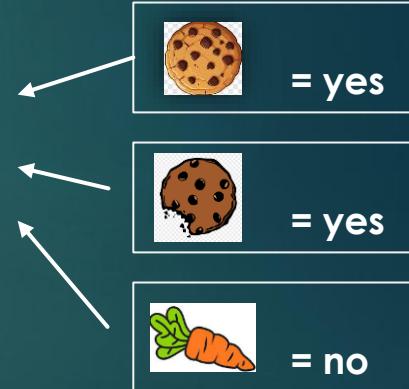
- Context
- Same Input -> Diff Results
- Inaccuracy %

# Machine Learning

## Supervised Learning

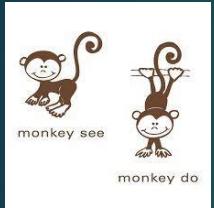


## Unsupervised Learning



IR: Information  
Retrieval

# Takeaways



**Gen AI's output:**

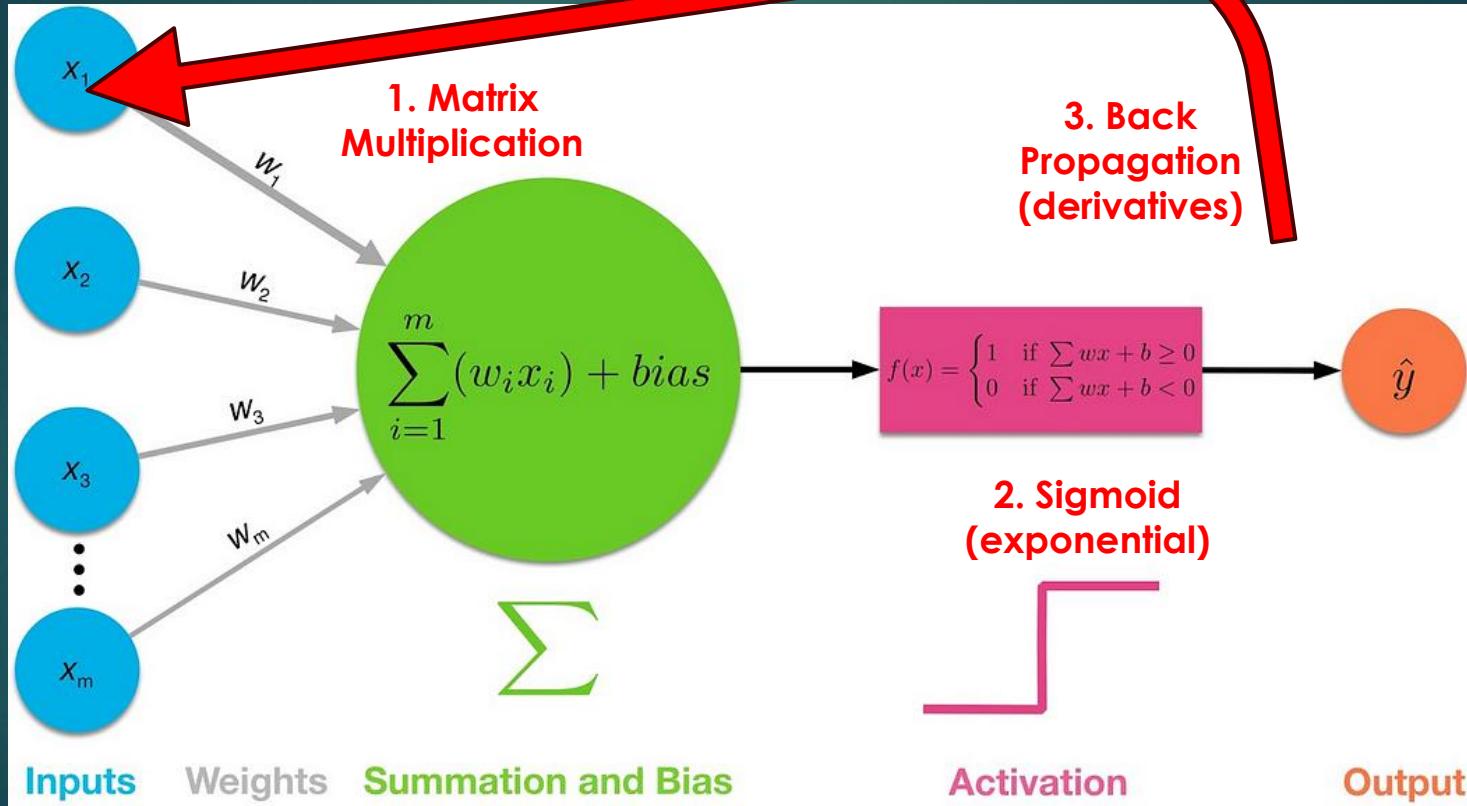
- Context
- Same Input -> Diff Results
- Inaccuracy %



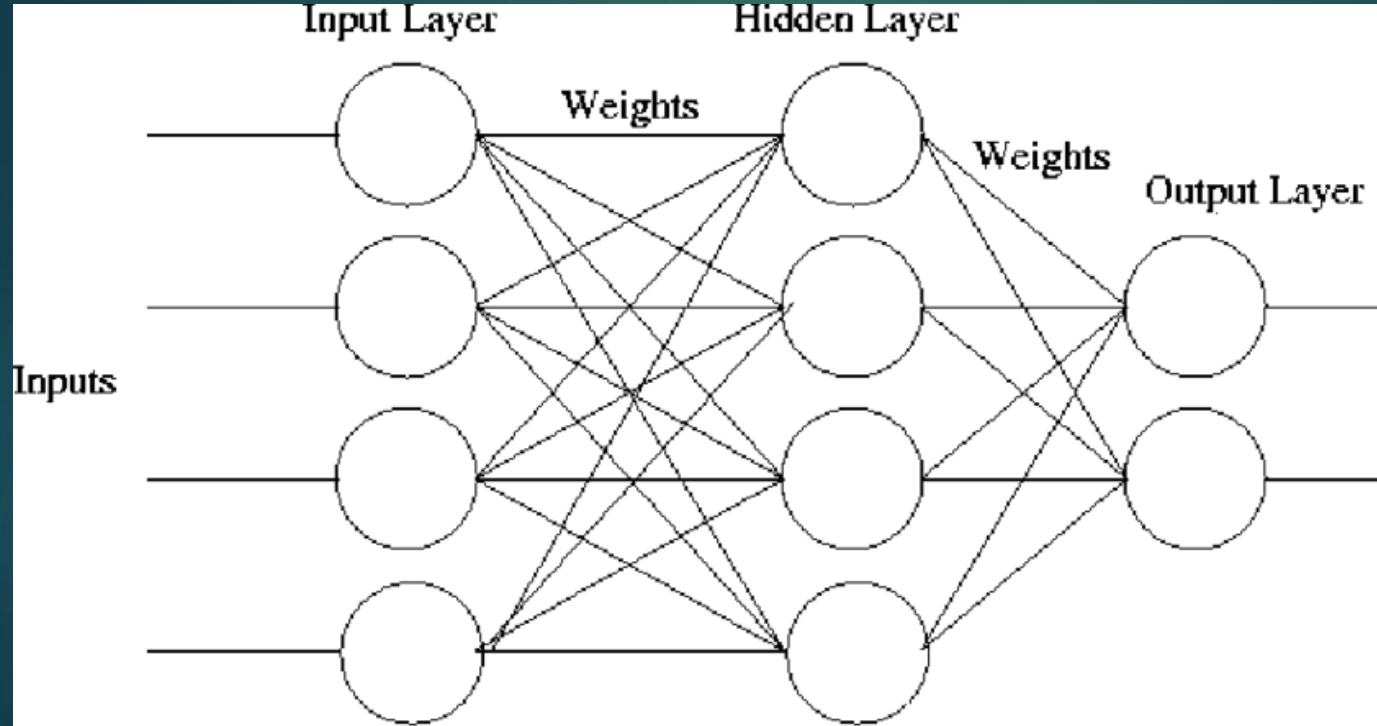
**Data sets** must be:

- Large
- Clean

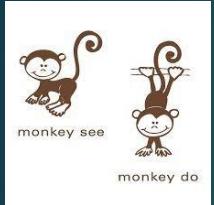
# Neural Networks



# Layered Neural Networks



# Takeaways



**Gen AI's output:**

- Context
- Same Input -> Diff Results
- Inaccuracy %



**Data sets** must be:

- Large
- Clean



**Cool AI is**

- Heavy
- Slow
- Expensive

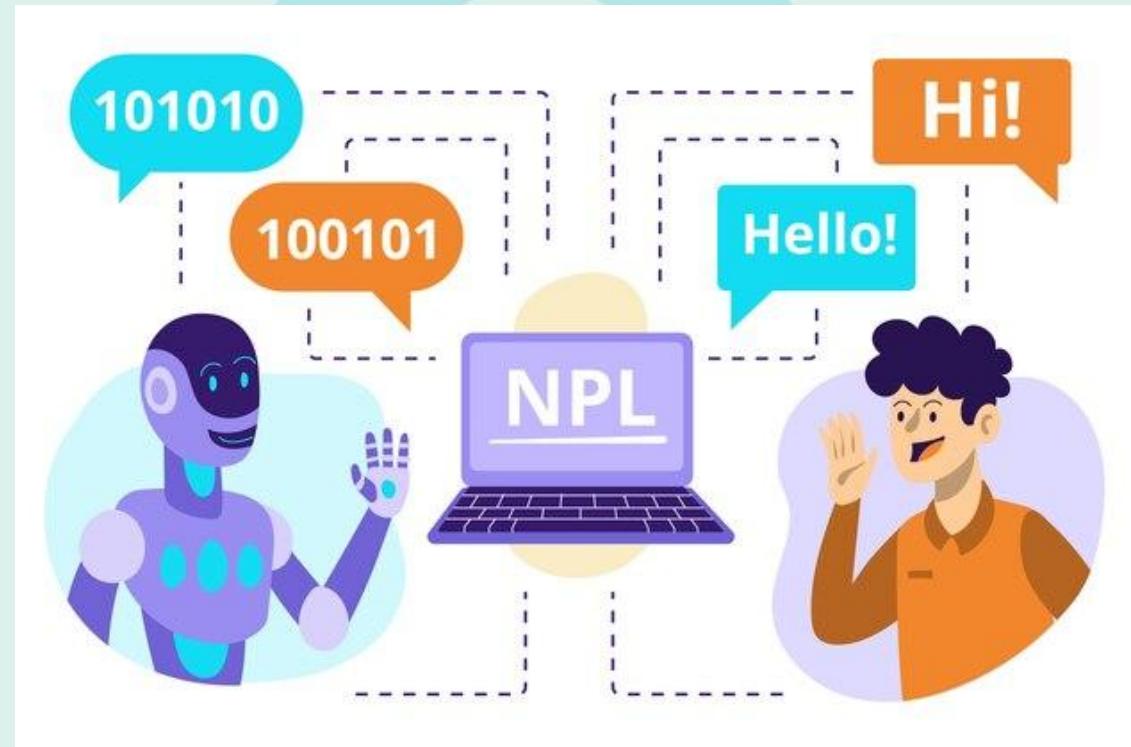


Development Task	Relevant (1-5)	Need Wrapper Code? (Y/N)	Notes
generate back end code			
generate front end code			
Refactor my code			
Review my code			
Design my system			
Finding relevant block of code in my project			



### 3. Review Activity!

## 4. What is NLP? When machines learn to talk



# Math <-> Words



010010010  
110011001  
001000110  
011011000  
100100100  
010010010



"The cat sat on the mat."



"The king and the queen  
met a wise woman in the castle."

### Tokenization



["the", "king", "and", "the", "queen", "met", "a",  
"wise", "woman", "in", "the", "castle"]

### Stop-words



["king", "queen", "met", "wise", "woman", "castle"]

### Lemma



["king", "queen", "**meet**", "wise", "woman", "castle"]

1 Word ≈ .75 token

["king", "queen", "meet",  
"wise", "woman", "castle"]

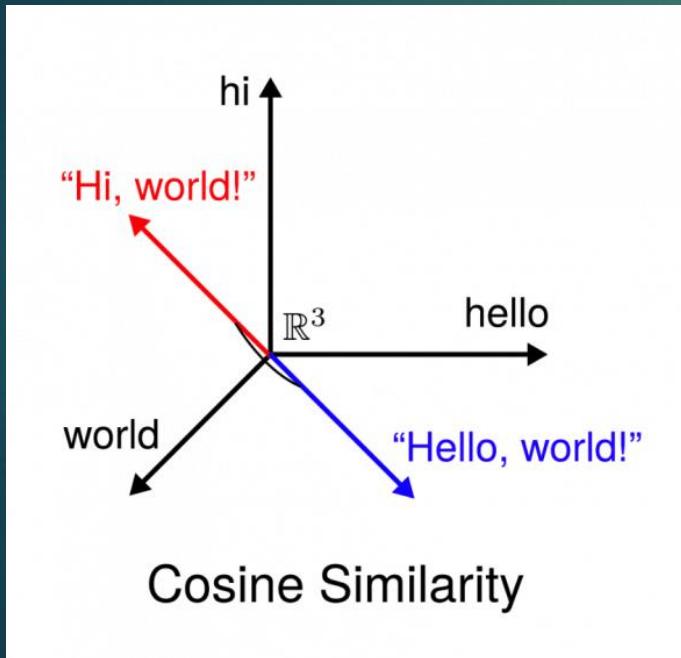


# Word Vectors

Word	Vector (x, y, z)
<b>[Royalty, People, Places]</b>	
king	[0.9, 0.9, 0.2]
queen	[0.9, 0.9, 0.3]
man	[0.4, 0.9, 0.2]
woman	[0.4, 0.9, 0.4]
meet	[0.2, 0.4, 0.3]
wise	[0.2, 0.3, 0.2]
castle	[0.9, 0.2, 0.7]

# Vector Similarity

“Hi, world!” “Hello, world!”



“The king and the queen met a wise woman in the castle.” 1.00

"The wise queen and king met a woman in the castle." ~0.7-0.8

"A castle woman met the wise king and queen." ~0.4-0.5

# Semantic Search



## What about Filtering?

No Metadata Parameters option...

- $k$  = total size. Filter after FAISS's result
- Manage Multiple FAISS indexes

# An NLP Pipeline (so far)

Pre - Processing



Vector

$$\begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

FAISS

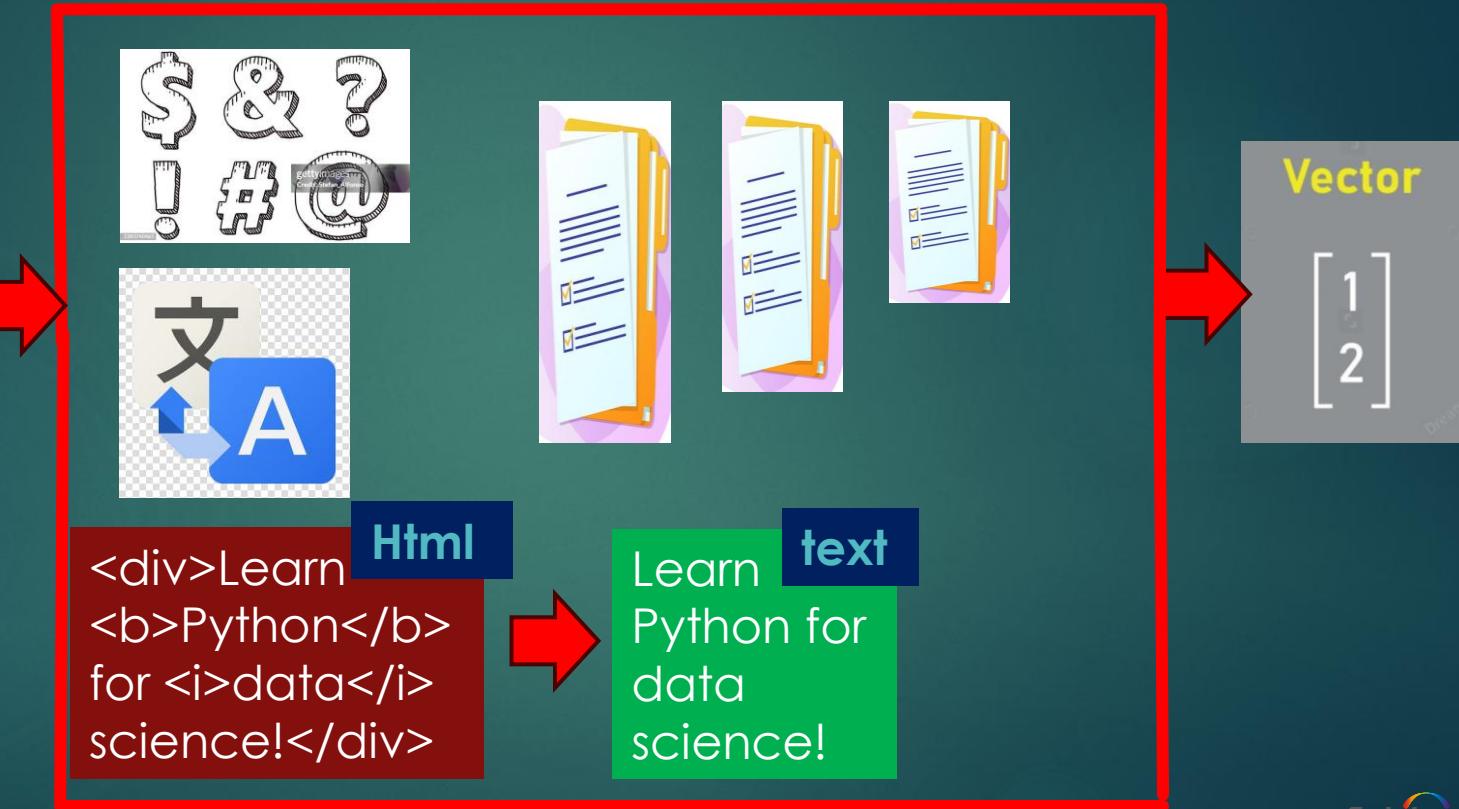


Front End



Post - Processing

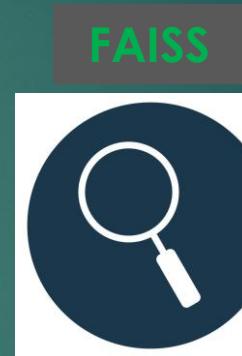
# Pre Processing





Html -> text

## Pre - Processing



## Front End



## Post - Processing

# LLMs

## Large Language Models



ChatGPT

- Broad Knowledge
- Multi – Lingual
- Strong Reasoning



- Open-source
- Run Locally (customizable)



- Long Context
- Good code generation
- Cheaper



- Integrated w/ Google system
- Accept 750k word prompt

# LLMs

## Parameters

Prompt	
(max length depends on Model)	
100-300 Words	1,500+ Words
₪	₪ ₪ ₪

Pricing:  
“\$0.001 per 1k tokens”

1 Word ≈ .75 token



Temperature		
0.3	0.7	1.0
Focused, Factual	Balanced	Creative, Ideas

Max_Tokens	
<100	2000+
Succinct	Large Essays
₪	₪ ₪ ₪

# NER

## Named Entity Recognition

"In September 2023, Microsoft announced a \$10 billion investment in OpenAI to expand their partnership in artificial intelligence research, following similar moves by Google and Amazon in the United States."

Entity	Type
September 2023	DATE
Microsoft	ORG
\$10 billion	MONEY
OpenAI	ORG
Google	ORG
Amazon	ORG
United States	GPE (Geopolitical Entity)

# RE

## Relationship Extraction

"In September 2023, Microsoft announced a \$10 billion investment in OpenAI to expand their partnership in artificial intelligence research, following similar moves by Google and Amazon in the United States."

Entity 1	Relation	Entity 2	Example Label
Microsoft	invested_in	OpenAI	INVESTMENT
Microsoft	investment_amount	\$10 billion	FINANCIAL_VALUE
Microsoft	investment_date	September 2023	TIME_OF_EVENT
Microsoft	partnered_with	OpenAI	COLLABORATION
Google	similar_action_to	Microsoft	COMPARISON
Amazon	similar_action_to	Microsoft	COMPARISON
OpenAI	located_in	United States	LOCATION

# NERs + RE

- LMM gets NERs
- Give LLMs the NERs

python

Precise

```
prompt = """
Extract named entities from this text and label their type.
Text: "Apple acquired Beats for $3 billion in 2014."
Return JSON with 'entity' and 'type'.
"""

Output:
```

json

```
[{"entity": "Apple", "type": "ORG"}, {"entity": "Beats", "type": "ORG"}, {"entity": "$3 billion", "type": "MONEY"}, {"entity": "2014", "type": "DATE"}]
```

python

Free

```
text = get_document()
entities = spacy_model(text).ents
llm_prompt = f"""
Summarize the document.

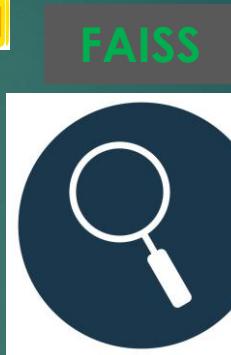
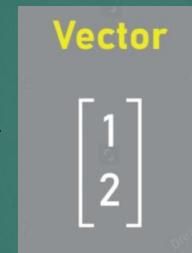
Highlight any organizations and dates: {entities}
"""


```



Html -> text

## Pre - Processing



## Front End



- Collect Metadata
- NERs, RE



LLM

- Summarize
- User Input as Prompt

## Post - Processing



# Group Activity

In groups of 1-3,  
Design your own NLP pipeline.  
See attached sheets

```
{  
  "ref": "Berakhot 2a",  
  "heRef": "ברכות ב' ר' ברוך",  
  "text": [  
    "\u003Cbr\u003E\u003Cbr\u003E\u003Cstrong\u003EMISHNA:\u003C/strong\u003E \u003Cbr\u003EFrom when,\u003C/b\u003E that is, from what time, does \u003Cbr\u003Eone recite \u003Ci\u003EShema\u003C/i\u003E in the evening?  
From the time when the priests enter to partake of their \u003Ci\u003Eteruma.\u003C/i\u003E\u003C/b\u003E Until when does the time for the recitation of the evening \u003Ci\u003EShema\u003C/i\u003E extend?  
\u003Cbr\u003EUntil the end of the first watch.\u003C/b\u003E The term used in the Torah (Deuteronomy 6:7) to indicate the time for the recitation of the evening \u003Ci\u003EShema\u003C/i\u003E is  
\u003Ci\u003Ebeshokhbehka\u003C/i\u003E, when you lie down, which refers  
to the time when individuals go to sleep. Therefore, the time for the recitation of the evening Shema is the time before people go to sleep."}]
```



# Thank you!

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Questions?



David Berger, 2025  
Software Engineer  
Rapido, Ex libris

**ExLibris**  
Part of Clarivate