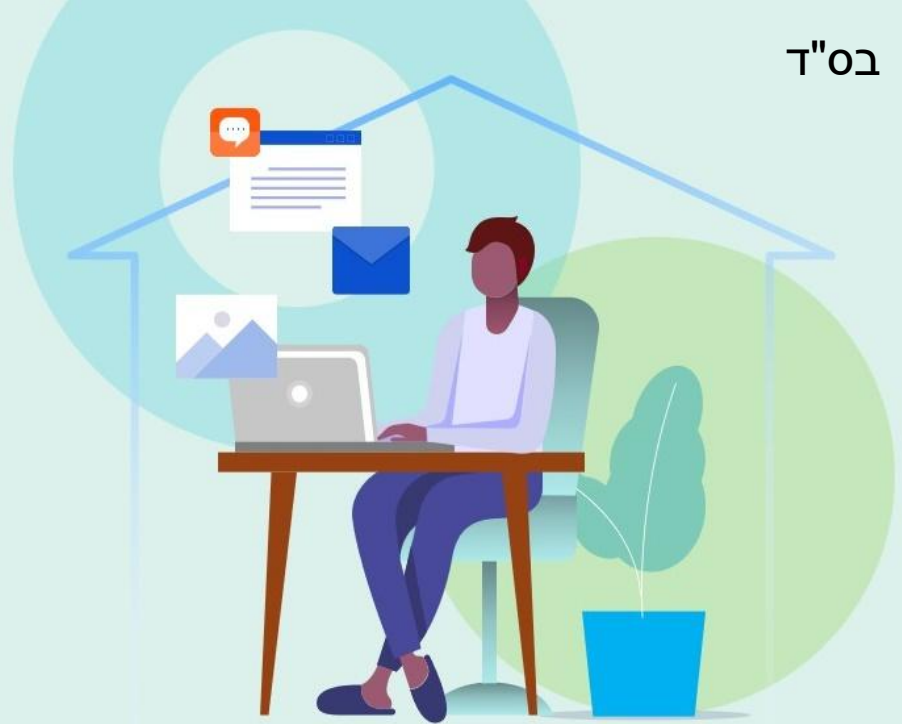


AI & NLP

(Natural Language Processing)

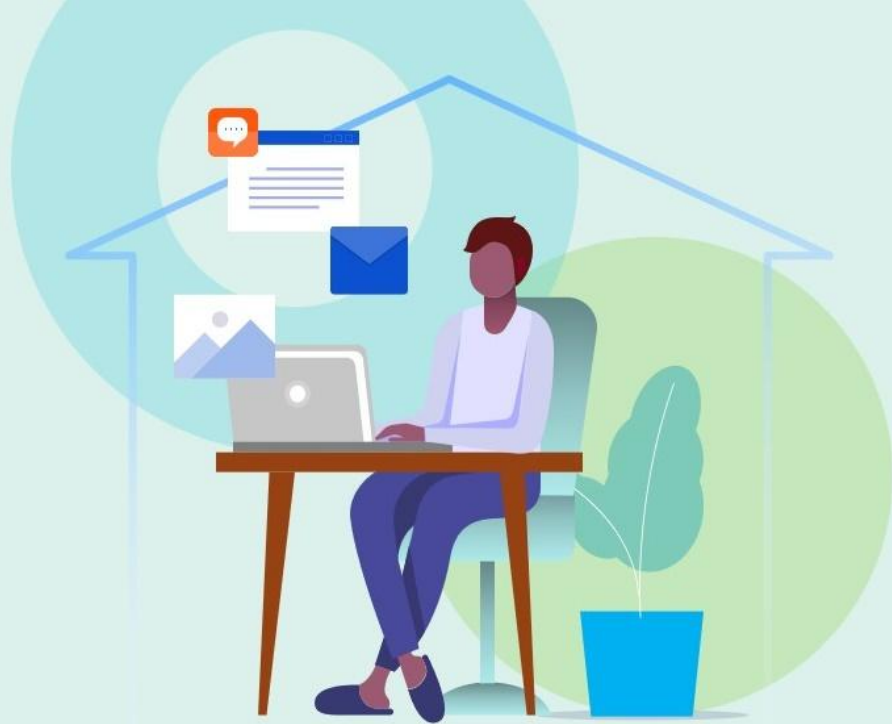
For SW Engineers



David Berger, 2025
Software Engineer

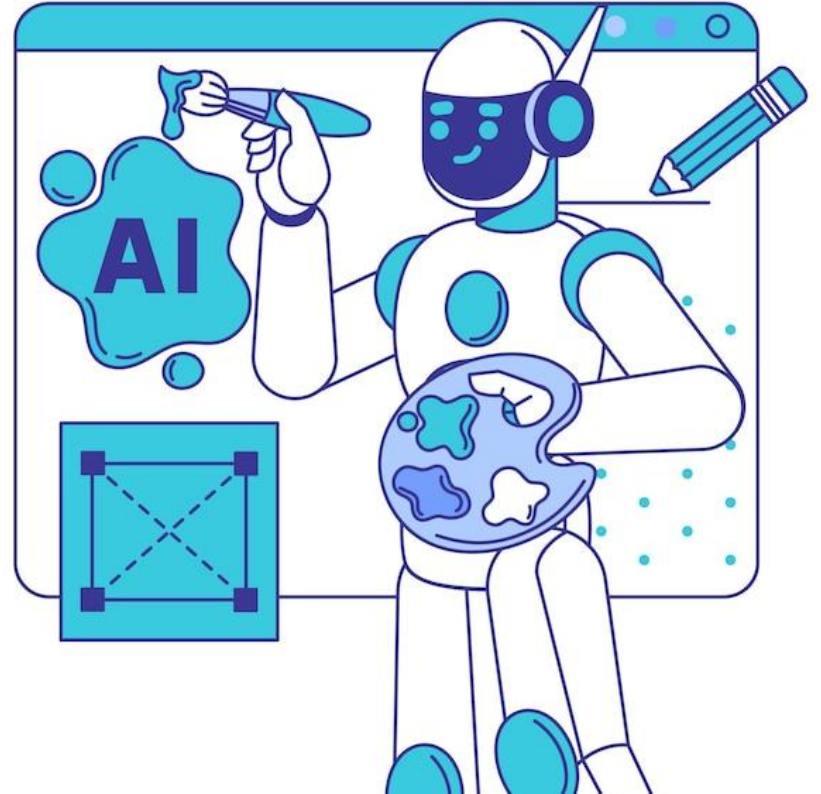
ExLibris
Part of **Clarivate**

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

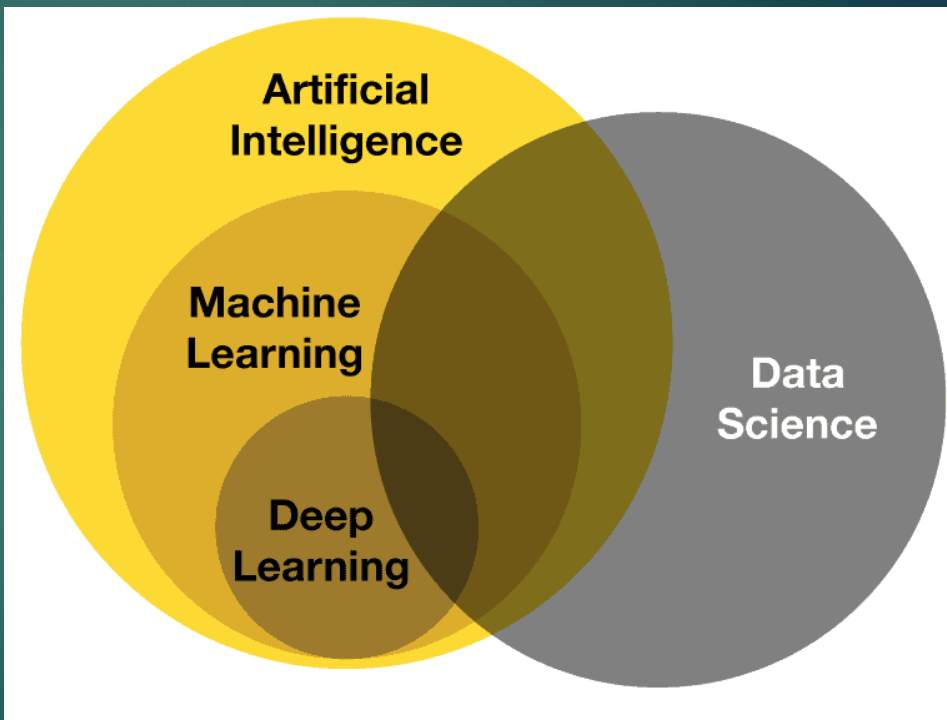
Learning patterns from data

Deep Learning

Complex, multi layered learning

Data Science

Extract insights from Data



Fields of AI/ML/Data Science

5

* Grouping not exact

Artificial Intelligence

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

Machine Learning

- Recommendation
- Supervised/
Unsupervised Learning
- Anomaly Detection

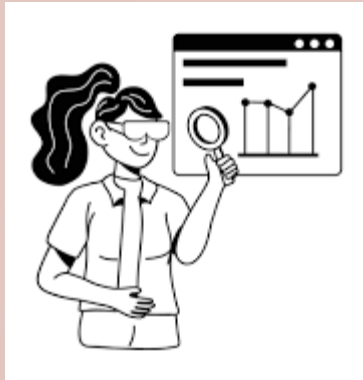
Deep Learning

- 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

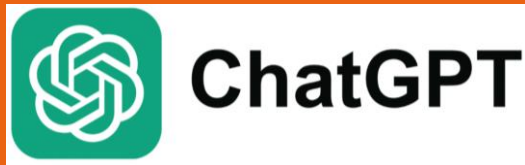


The AI Takeover

7

"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."



What about me? (SWEs)

8

"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)



OpenAI

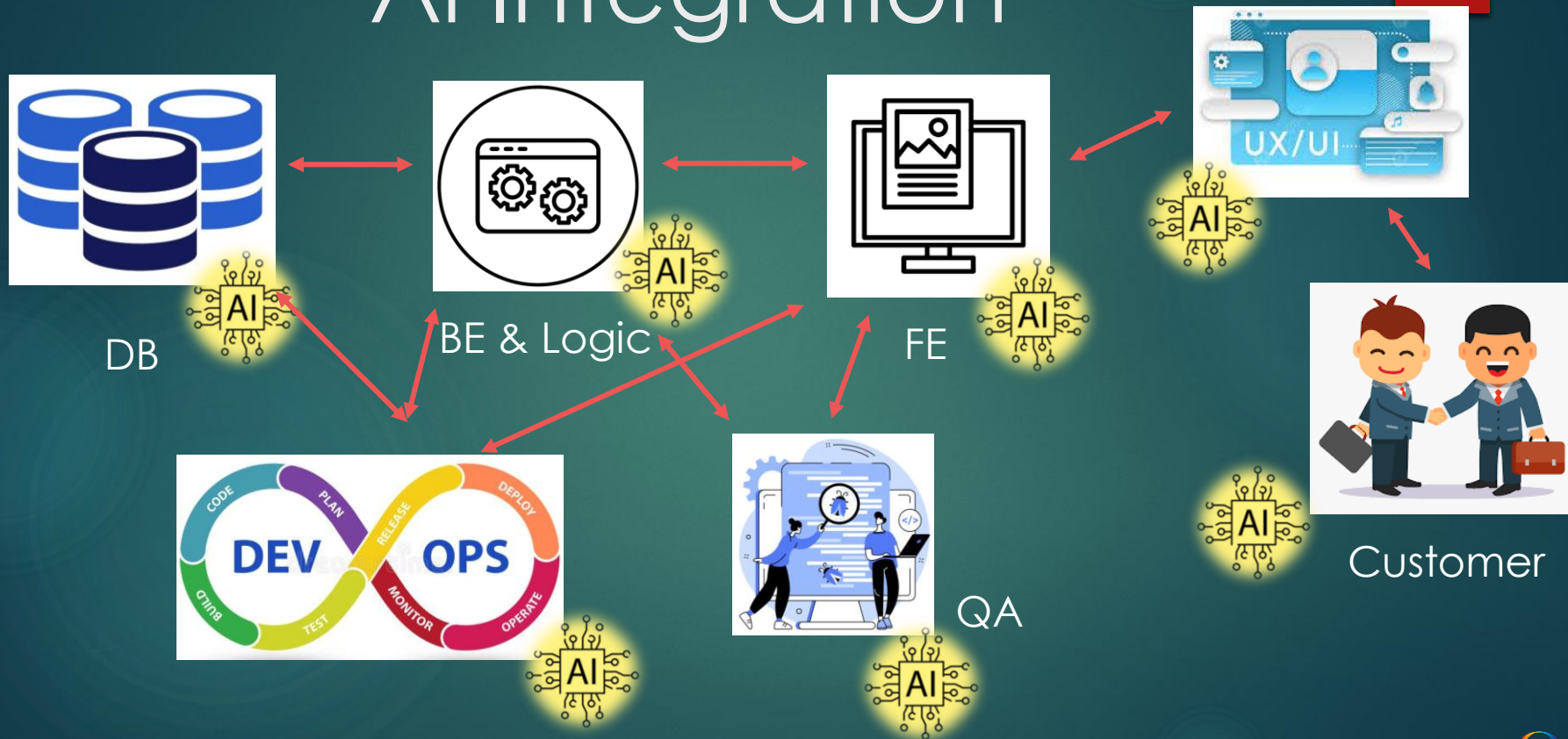
ANTHROPIC



MISTRAL
AI_

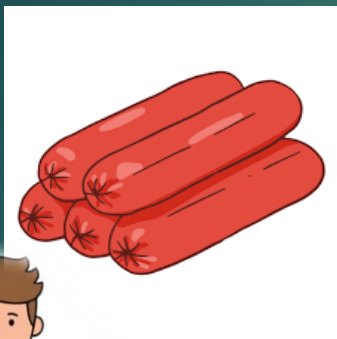


Ai Integration



Complementary Goods

(מוצר משלים)



AI

**Software
Engineers**



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



Gen AI – How it works

12

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



monkey see



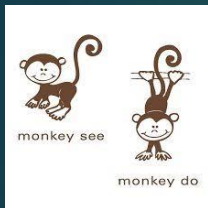
monkey do

“Bibi is _____ ?”

Takeaways

Gen AI's output:

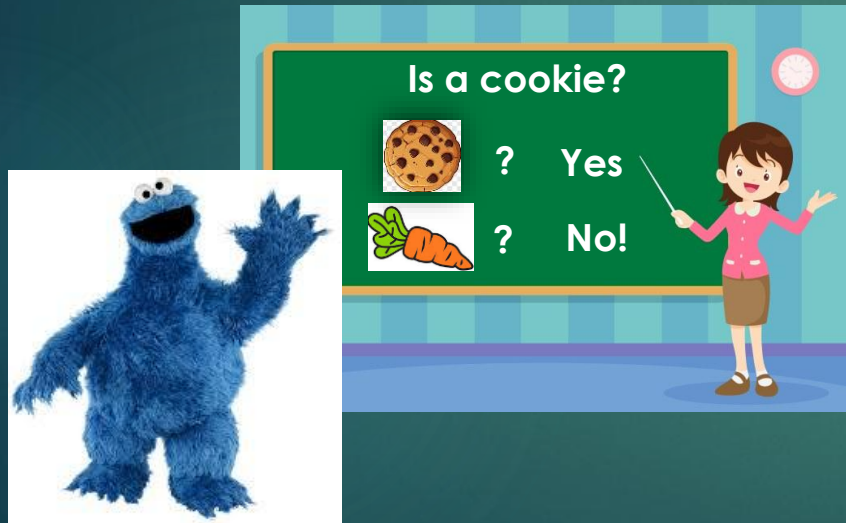
- Context
- Same Input -> Diff Results
- Inaccuracy %



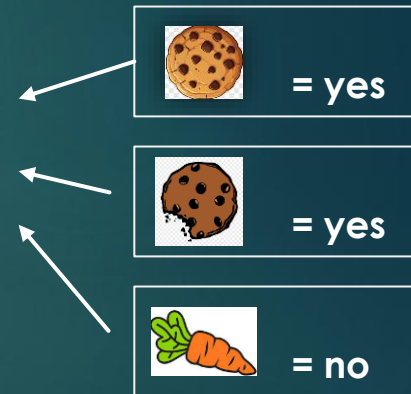
Machine Learning

14

Supervised Learning

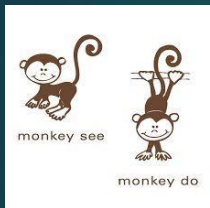


Unsupervised Learning



IR: Information
Retrieval

Takeaways



Gen AI's output:

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

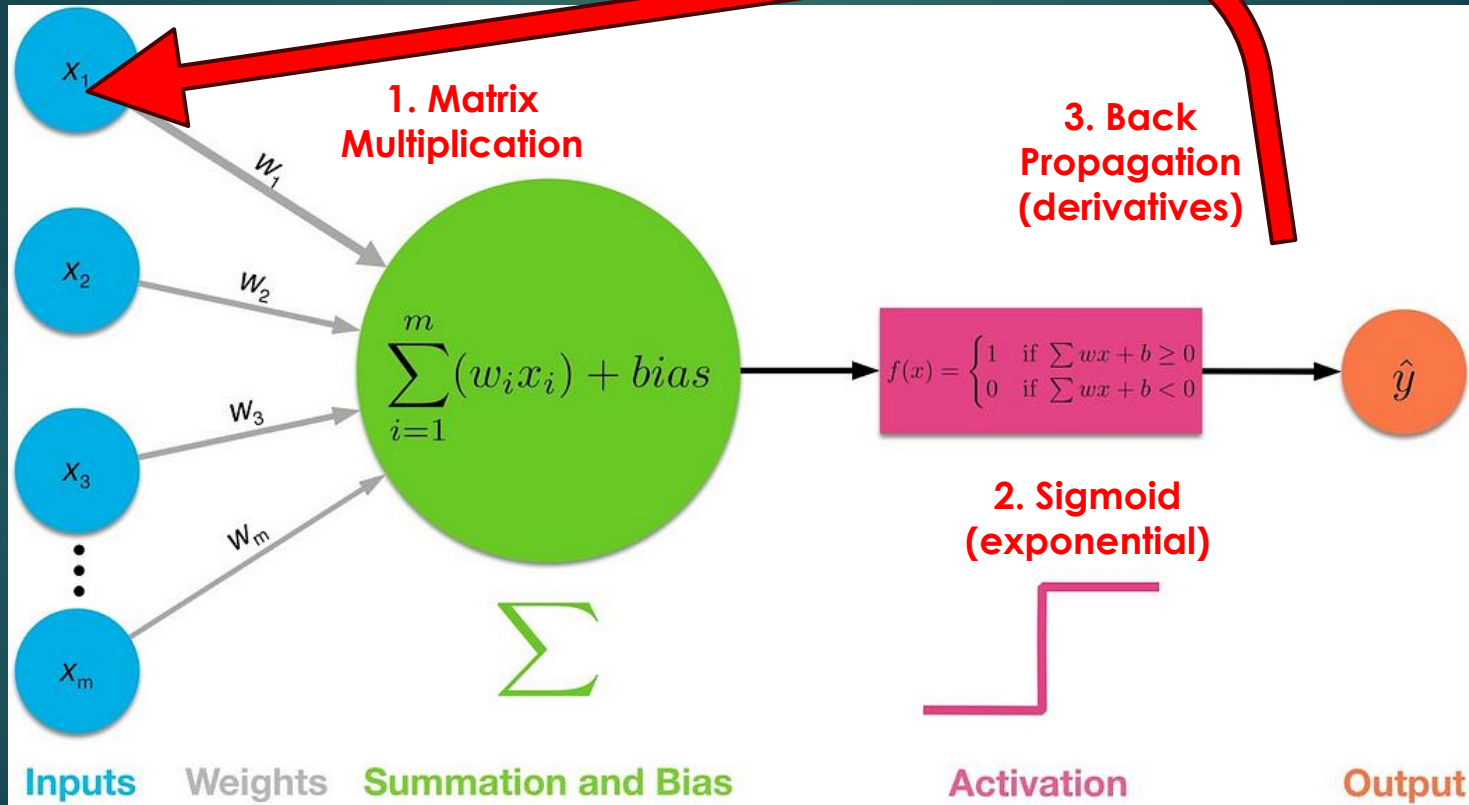


Data sets must be:

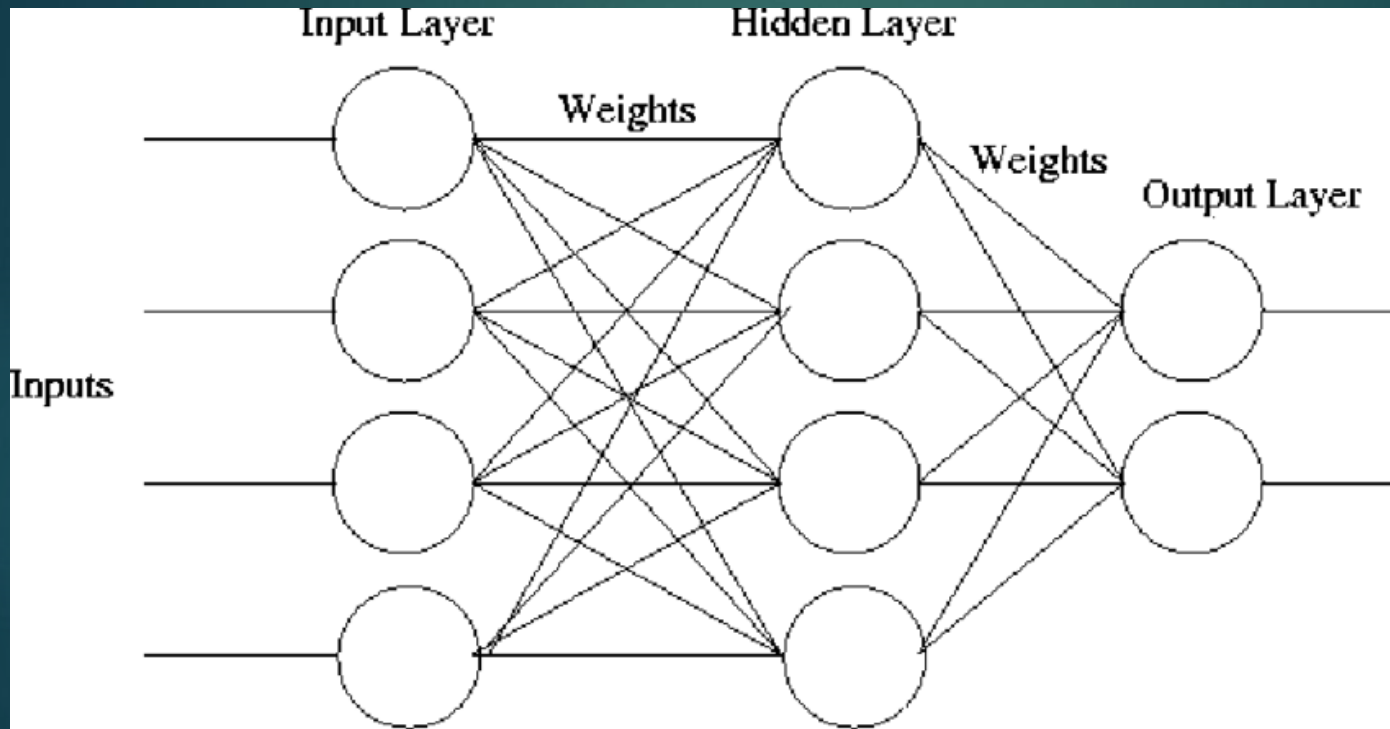
- Large
- Clean

Neural Networks

16



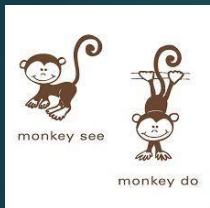
Layered Neural Networks



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

21

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 \approx .75 token

Word Vectors

384d

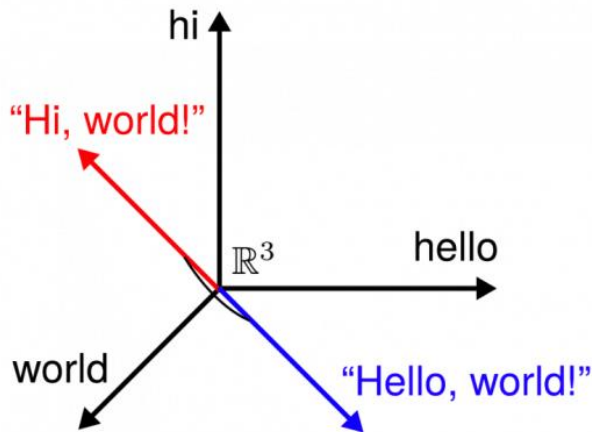
["king", "queen", "meet",
"wise", "woman", "castle"]



Word	Vector (x, y, z)
[Royalty, People, Places]	
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



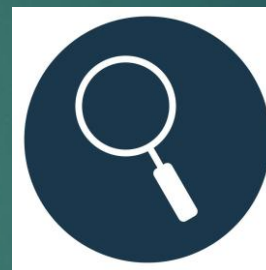
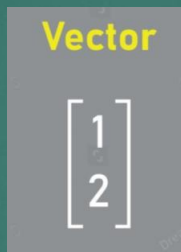
What about Filtering?

No Metadata Parameters option...

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

An NLP Pipeline (so far)

Pre - Processing



Post - Processing

Pre Processing



Vector

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

`<div>Learn
Python
for <i>data</i>
science!</div>`

Html



Learn
Python for
data
science!

text



An NLP Pipeline

28

Html -> text

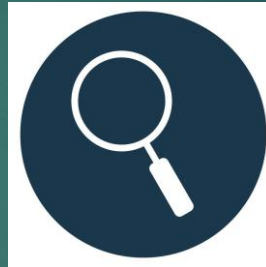
Pre - Processing



Vector

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


FAISS



Front End



Post - Processing

LLMs

Large Language Models



- 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



LLM

Large Language Model

Temperature

0.3

Focused,
Factual

0.7

Balanced

1.0

Creative,
Ideas

Max_Tokens

<100

Succinct



2000+

Large Essays



Pricing:

“\$0.001 per 1k tokens”

1 Word \approx .75 token



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_VALU E
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}
"""
```



An NLP Pipeline

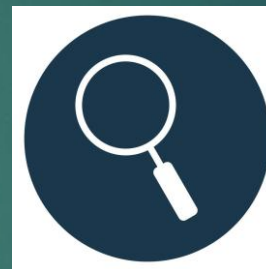
34

Html -> text

Pre - Processing



Vector

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


Post - Processing



- Collect Metadata
- NERs, RE



LLM

Large Language Model

- Summarize
- User Input as Prompt

Group Activity

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



```
{  
  "ref": "Berakhot 2a",  
  "heRef": "ברכות ב' א'",  
  "text": [  
    "\u003Cbr\u003E\u003Cbr\u003E\u003Cstrong\u003EMISHNA:\u003C/strong\u003E  
    \u003Cbr\u003EFrom when,\u003Cbr\u003Ethat is, from what time, does  
    \u003Cbr\u003Eone recite \u003Cbr\u003E\u003Cstrong\u003Eshema\u003C/strong\u003E in the evening?  
    From the time when the priests enter to partake of their  
    \u003Cbr\u003E\u003Cstrong\u003EUntil when does the time  
    for the recitation of the evening \u003Cbr\u003E\u003Cstrong\u003Eextend?  
    \u003Cbr\u003EUntil the end of the first watch.\u003Cbr\u003EThe term used in  
    the Torah (Deuteronomy 6:7) to indicate the time for the recitation of the evening  
    \u003Cbr\u003E\u003Cstrong\u003Eis  
    \u003Cbr\u003E\u003Cstrong\u003E, when you lie down, which refers  
    to the time in which individuals go to sleep. Therefore, the time for the recitation
```

Thank you!

Questions?



David Berger, 2025
Software Engineer
Rapido, Ex libris

ExLibris
Part of **Clarivate**