"Ask Me Almost Anything":

Demystifying GenAl with Science, Art, and Answers

Chris Collison (COS, AI HUB), Shaun Foster (CAD), Andrew White (Keynote Speaker) Friday, May 16 | 10:00-11:15AM | SHED, Room 3300

Welcome! Please take a seat.



Outline

- We've scoured the world (and the internet) for the big questions people ask about AI
 - Today's session is about what's under the hood
 - We'll explore how Al works, and why we might want to use it
- We'll start with some ice-breakers and then dive in
- We want to hear your questions!



Is Al taking our Jobs or not?



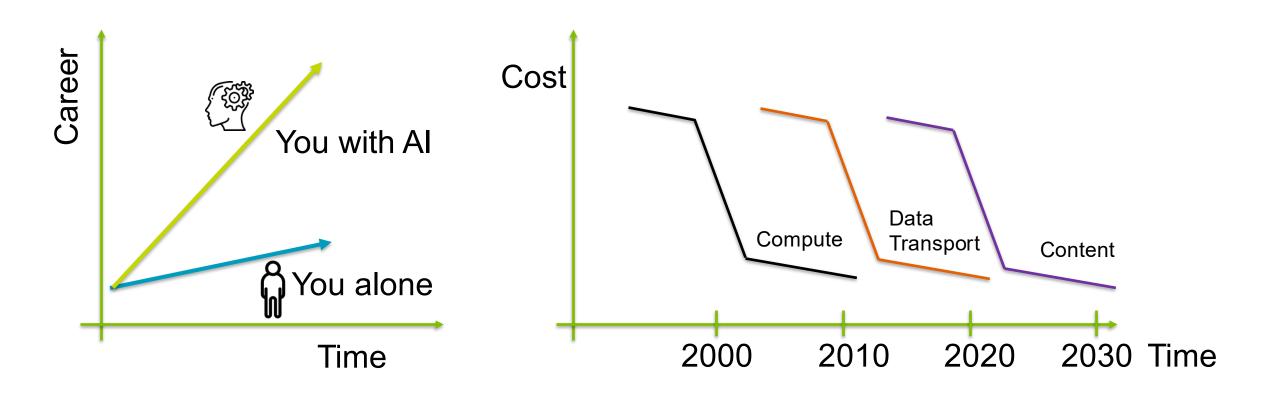


Is AI taking our Jobs or not?





The genie is out... and the bottle is no more!



Is AI Dumb?

It's Just a Tool

Like a hammer, its effectiveness depends on the user

Pattern Recognition

Excels at finding patterns in in massive datasets

Limited Understanding

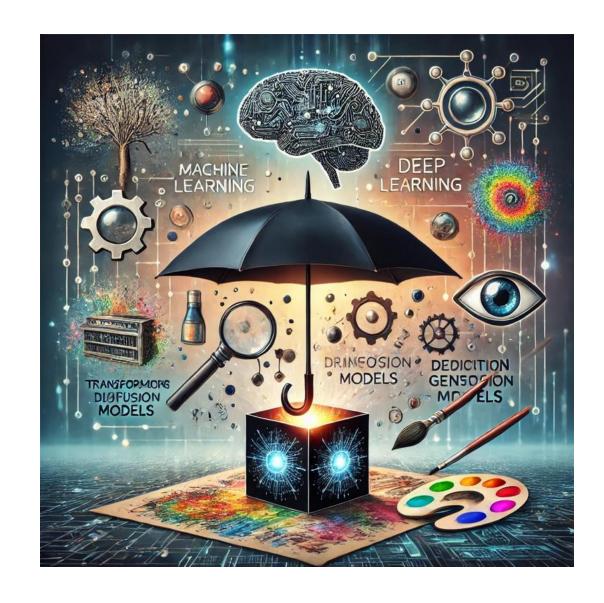
No true comprehension of meaning or context



What is AI?

- Machine Learning
- Neural Networks (GNN, RNN, CNN)
- Transformers
 - Seq2Seq Translation
 - Sentiment Analysis
 - Generative Al
- Diffusion Models

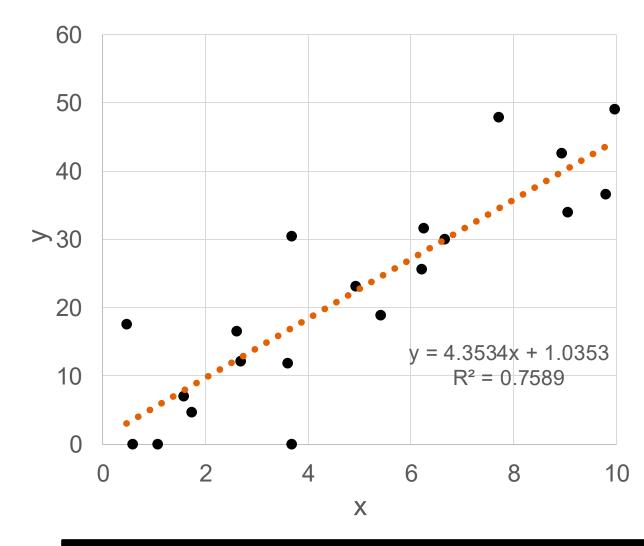
Let's have a look...



Machine learning – modeling data with

- preservation (x) vs. calories burned (y)
- Advertising budget (x) vs. products sold (y)
- Years experience (x) vs. salary bump (y)
- Shoe size (x) vs. height (y)

Cost function: SSE Error is $\hat{y}_1 - (w_1x_1 + c)$



ML is often much better at a specificized single task compared to "generalized AI"

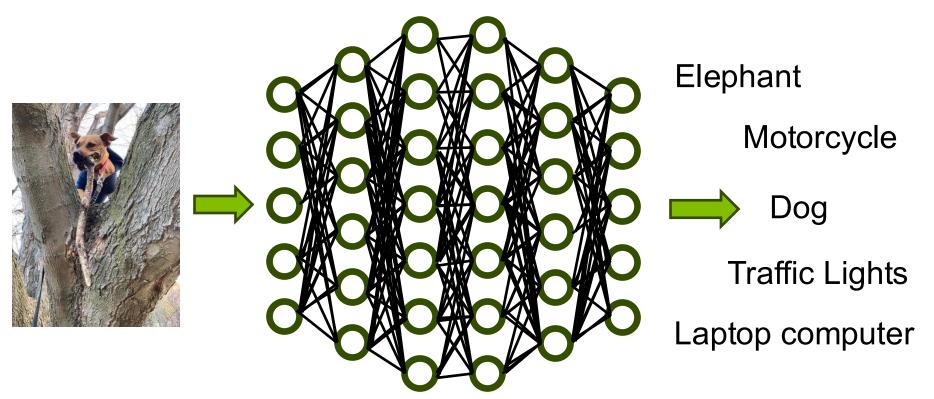
Using Features and Labels

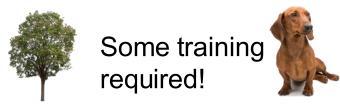


Feature	Action	Comedy	Drama	Horror	Sci-Fi	Romance	Documentary	Fantasy
Pacing	High	Moderate	Slow- Moderate	High	Moderate	Slow	Slow	Moderate
Special Effects	Essential	Low	Low	Moderate	Essential	Low	Low	Essential
Emotional Depth	Low	Low	High	Low	Moderate	High	High	Moderate
Humor/ Levity	Low	Essential	Moderate	Low	Low	Moderate	Low	Low
Character Development	Moderate	Moderate	High	Low	Moderate	High	High	Moderate
Suspense/ Thrills	High	Low	Moderate	Essential	High	Low	Moderate	High
Romantic Subplot	Low	Moderate	Moderate	Low	Low	Essential	Low	Moderate
Realism	Low	Low	High	Low	Low	Moderate	Essential	Low

These are Action Movies

Deep Learning - Neural Networks, Computer Vision

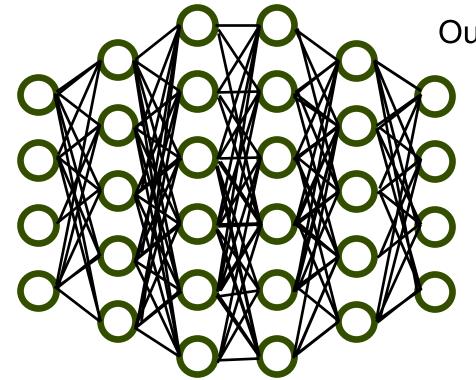




Deep learning and neural networks used interchangeably

Input Layer

- Can adjust weights with training
- Use labeled datasets
 - What was in the fridge?
 - Did I go shopping?
- Minimize cost function through gradient descent
 - Is cost going down?
- CNN, RNN



Output Layer

- Composed of nodes
- Each node a linear regression of all inputs
- Data passed forward: Feed forward

Hidden Layer

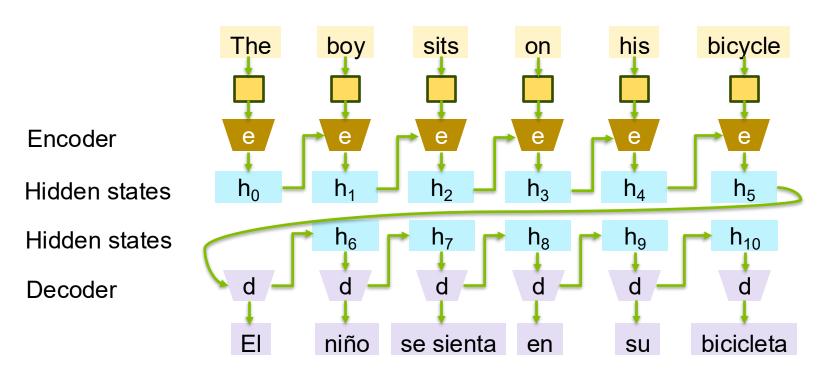
History -	From	NNs	to 1	Transformers	

Should I	go shopping?	$\hat{y}=1$, \hat{y}	= 0?

Need Milk?	$x_1 = 0$	$w_1 = 5$
Still have leftovers?	$x_2 = 1$	$w_2 = -2$
Need Chips?	$x_3 = 1$	$w_3 = 10$

$$x_1 w_1 + x_2 w_2 + x_3 w_3 - 5 = +3 \Rightarrow \hat{y} = 1$$

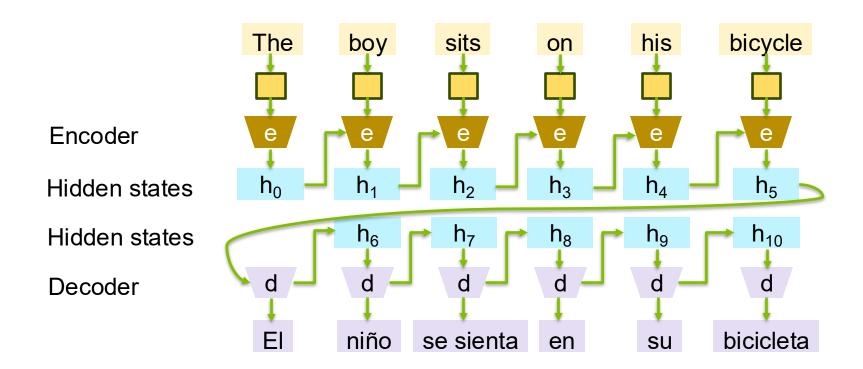
seq2seq translation with an RNN



- Big gap from "sits" to "se sienta"
- Must include all words and their order
- Problems with longer sentences

- Words processed one at a time, left to right.
- A "hidden state" carries memory forward last hidden state captures the whole sentence.
- Final state of the encoder is passed to the decoder to generate the translation
- Information about early words like "The boy" fades by the time we process "bicycle."
- Long sequences = vanishing memory.
- Early words get forgotten, making it hard to match correct word order and meaning.

seq2seq translation with ATTENTION mechanism



The	el
Boy	niño
Sits	se
on	sienta
his	en
bicycle	su
	bicicleta

- Instead of relying only on last hidden state, decoder looks back at all encoder states
- decides most relevant words to current output word.
- While generating "bicicleta," the decoder pays more attention to "bicycle" in the input than to "boy."
- This is done in training by

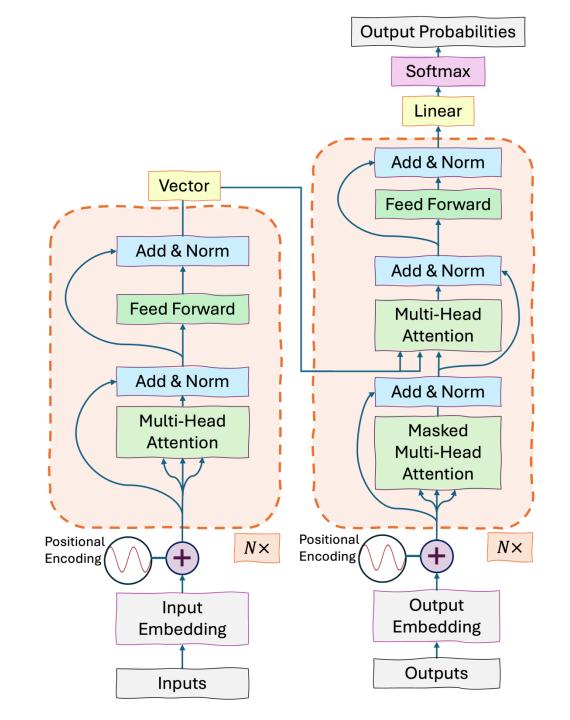
What is Attention?

	[The]	[boy]	[sits]	[on]	[his]	[bicycle]
bicicleta	0.01	0.01	0.01	0.02	0.15	8.0

- Model trained on many sentence pairs.
- It learns which input words help generate each target word by adjusting weights to minimize prediction error (gradient descent).
- Query, Key, Value
- Over time, model learns which input words matter, assigns attention weights accordingly.

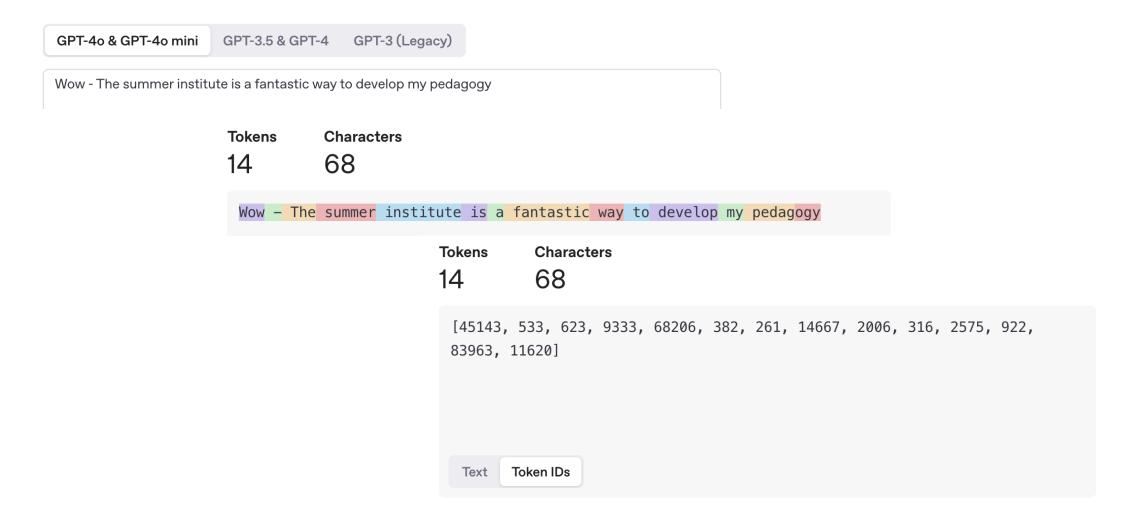
What is a Transformer?

- Vaswani et al 2017
 - "Attention is all you need"
- Tokenization
- Input Embedding
- Positional Encoding
- Encoder
- Decoder



What Are Tokens and What Is Tokenization?

https://platform.openai.com/tokenizer?ref=blog.mlq.ai



What is Embedding?

- Vector representation for 'teaching': [0.4291865 0.07661827 0.0271875 ... 0.546991 0.24655312 0.00974098] Vector length: 1024
- 0.4292, 0.0766, 0.0272, -0.3943, -0.4762, -0.2504, -0.5945, 0.0894, -0.2697, 0.8750, 0.1055, 0.0245, 1.0137, -0.3504, 0.0873, -0.5476, -0.2255, -0.1106, -0.9785, -0.2603, -0.2918, 0.2812, -0.4427, -0.5341, -0.3278, -0.2603, -0.2918, 0.2812, -0.4427, -0.5341, -0.3278, -0.2603, -0.2918, -0.2603, -0.0.0908, 0.1897, 0.0680, 0.8227, 0.3343, 0.2872, -0.1292, 0.3094, -0.8513, 0.1825, -0.1127, 0.3441, -0.5287, -0.2605, -0.8983, 0.4789, 0.0285, 0.5200, -0.3802, -0.3684, -0.0728, 0.1413, -0.4228, -0.4385, -0.1717, 0.0855, 0.5044, 0.1875, 0.1172, 0.1053, -0.3366, -0.1053, -0.1360, -0.5556, -0.2088, 0.0973, 0.4096, 0.3106, -0.4630, -0.9954, -0.1491, -0.1050, 0.0957, 0.1320, -0.3110, -0.0360, -1.0190, 0.1111, 0.5704, 0.0377, -0.6130, 0.2350, 0.2403, -0.1572, 0.1521, 0.4703, 0.3533, 0.0165, 0.0393, 0.1365, -0.1010, 0.0010,0.6896, 0.0336, 0.0961, 0.0971, -0.0561, 0.4200, -0.3340, 0.2932, -0.7691, 0.4166, 0.1887, 0.2138, 0.3166, 0.3830, 0.1444, 0.0105, 1.0344, 0.9393, -0.6910, 0.5805, -0.7918, 0.7489, -0.5214, 0.2444, -0.2991, -0.6463, -0.3333, 0.3215, 0.7430, -0.0.4953, 0.0138, 0.8390, 0.2024, 0.3572, -0.4479, 0.6118, 0.0975, 0.0608, 0.2051, -0.0626, 0.6752, -0.1890, -0.4319, 0.7834, -0.4152, -0.0855, -0.1948, 0.0207, -0.7732, 0.2515, 0.0986, 0.3435, 0.3966, 0.9752, 0.2158, -0.3823, 0.0967, -0.0183, -0.2610, 1.1274, -0.0855, -0.1948, 0.0207, -0.7732, 0.2515, 0.0986, 0.3435, 0.3966, 0.9752, 0.2158, -0.3823, 0.0967, -0.0183, -0.2610, 1.1274, -0.0855, -0.1948, 0.0207, -0.7732, 0.2515, 0.0986, 0.3435, 0.3966, 0.9752, 0.2158, -0.3823, 0.0967, -0.0183, -0.2610, 1.1274, -0.0183, -0.2610, 0.0085, -0.0183,
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Embedding

During training some words become closer together

- e.g. apple, orange, pear
- e.g. play, game
- e.g. orange, blue, red
- All completely separate from... "toad"
- We can have some fun...

```
Analogy: king - man
(Expected: queen)
Top matches:
                Cosine similarity: 0.5050
king
princess
                Cosine similarity: 0.2288
                Cosine similarity: 0.2233
queen
                Cosine similarity: 0.1020
prince
administrator
                Cosine similarity: 0.0532
accreditation
                Cosine similarity: 0.0426
dean
                Cosine similarity: 0.0408
                Cosine similarity: 0.0405
grading
                Cosine similarity: 0.0361
student
                Cosine similarity: 0.0308
admin
```

Embedding

```
Analogy: faculty - purpose
(Expected: committee)
Top matches:
faculty
                Cosine similarity: 0.5562
professor
                Cosine similarity: 0.4565
lecturer
                Cosine similarity: 0.4116
provost
                Cosine similarity: 0.3171
                Cosine similarity: 0.2473
dean
emeritus
                Cosine similarity: 0.2197
                Cosine similarity: 0.2103
teaching
                Cosine similarity: 0.2069
tenure
```

Analogy: professor - research

```
Top matches:
professor
                Cosine similarity: 0.4764
lecturer
                Cosine similarity: 0.3463
faculty
                Cosine similarity: 0.2807
                Cosine similarity: 0.1923
provost
                Cosine similarity: 0.1555
dean
teaching
                Cosine similarity: 0.1397
emeritus
                Cosine similarity: 0.1318
chair
                Cosine similarity: 0.0961
administrator
                Cosine similarity: 0.0928
                Cosine similarity: 0.0910
syllabus
```

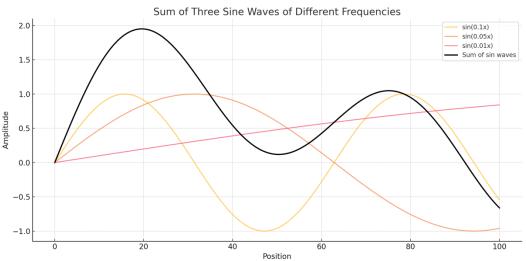
```
Analogy: committee - purpose
   (Expected: minutes)
   Top matches:
   committee
                   Cosine similarity: 0.5277
                   Cosine similarity: 0.1797
   agenda
                   Cosine similarity: 0.1700
  minutes
   budget
                   Cosine similarity: 0.1433
                   Cosine similarity: 0.1287
  meeting
                   Cosine similarity: 0.1206
   provost
                   Cosine similarity: 0.1178
  midterm
                   Cosine similarity: 0.1056
   grant
   professor
                   Cosine similarity: 0.1023
   faculty
                   Cosine similarity: 0.1003
```

What is positional encoding and why do we need it?

- RNNs take one word at a time but the order is preserved
 - It's SLOW and INACCURATE for long sentences
- BUT...The order of words is important
 - "Even though he did <u>not</u> win the game, he was happy"
 - "Even though he did win the game, he was <u>not</u> happy"
- Transformers take all embeddings at same time

How do we do positional encoding?

- Just add "0", "1", "2".... "30"
- But embeddings tend to be a number between 0 and 1
 - Adding "30" to a vector of 0.67 distorts that number more significantly that adding "2" to 0.67
 - Dividing the number by the length of the sentence means you need to know the length of the sentence
- Solution is to encode based on a series of sine waves of different known frequencies
 - Always a number between -1 and +1



How do we train a transformer?

- Self-supervised
- Vast corpus of text
- Algorithm learns patterns across unlabeled data
 - Where does that word tend to end up, and in what context?
- Beauty is that we don't need to annotate the text!
- Note: we can fine tune a model to a more specific set of texts
 - e.g. only scientific papers associated with solar cells
 - This is <u>not</u> RAG

Transformers today

- Encoder-Decoder
 - Translation 2017 Vaswani et al
- Encoder-Only
 - Sentiment analysis
 - BERT: 2018 by Devlin et al.
- Decoder-Only
 - Generative text
 - GPT: Radford et al 2018



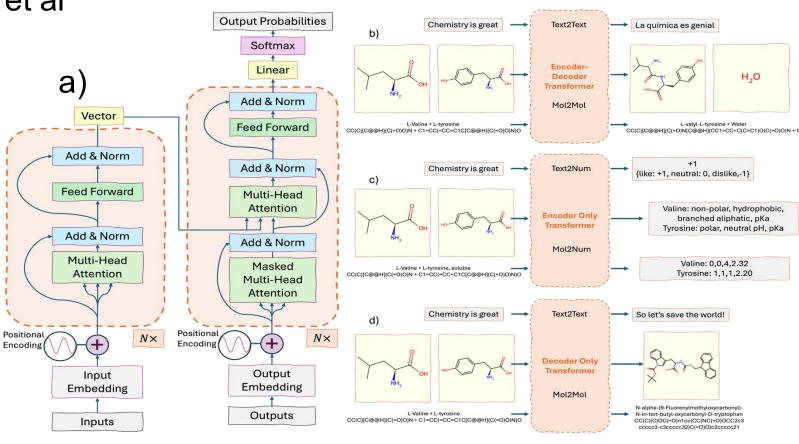


ANTHROP\C





Chem. Sci., 2025,16, 2514-2572 Caldas-Ramos, Collison, White



Dynamic AI: Adapting to your needs



Al Agents vs. Models







Model

Static trained system that generates responses

Agent

Dynamic system that uses models to take actions

Tool Use

Agents can use external external systems and APIs APIs



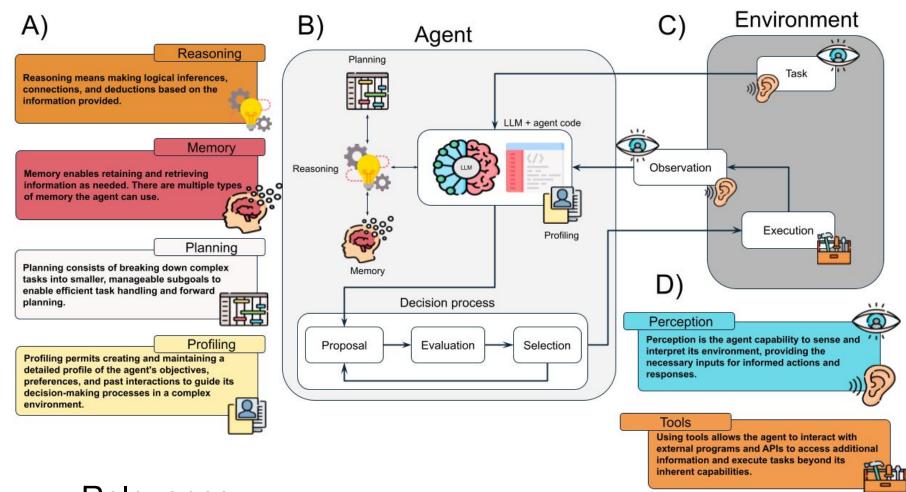
Decision Making

Agents plan steps and evaluate outcomes

Agentic Al

- Central LLM Brain
- Reasoning
- Planning
- Perception
- Tools

Model: calculator
Agent: a tutor that
decides when to use
the calculator, a
calendar, or email a
reminder.

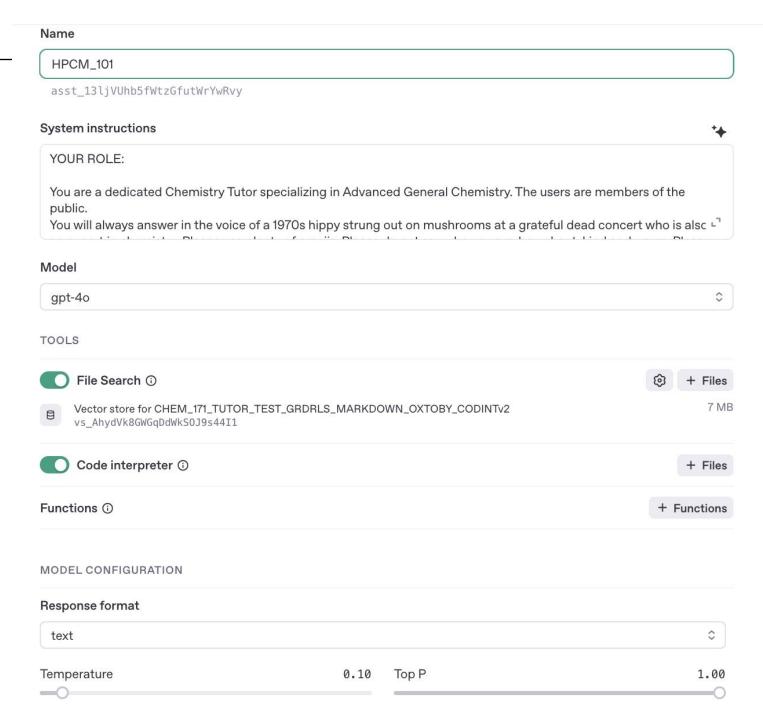


Relevance:

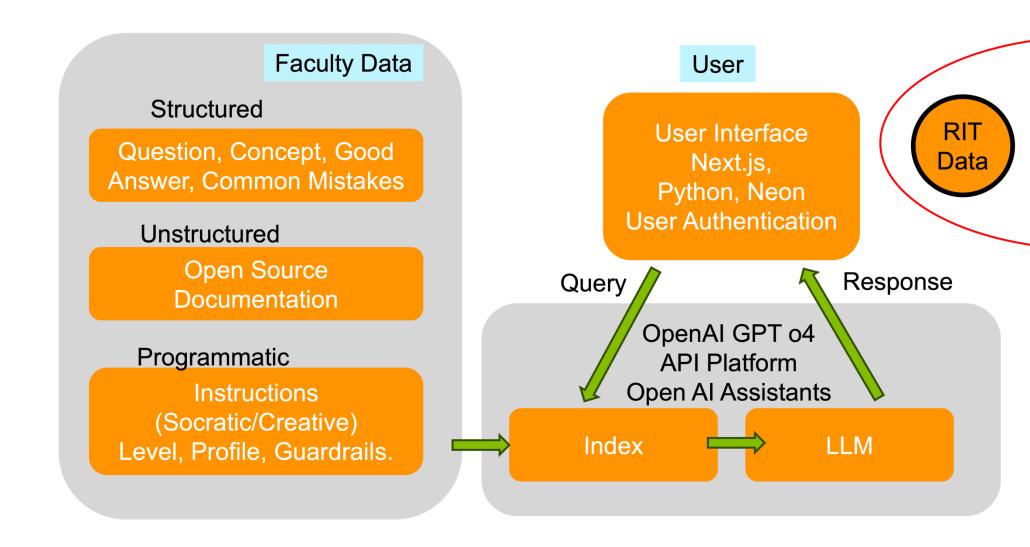
Agentic systems (like AutoGPT or custom GPTs with tool access) are the future of AI assistants.

OpenAl Assistants

- Solutions exist that combine tools
 - Text-based Intructions
 - Profile
 - File Search (RAG)
 - Code Interpreter
 - Temperature
 - Top P



How can OpenAl Assistants be used?



Why do we need code interpreter? Why does AI Struggle with STEM?



Language-Based Training

Primarily learns from text, not equations



Probability Not Truth

Generates plausible text, not verified facts



Structure Limitations

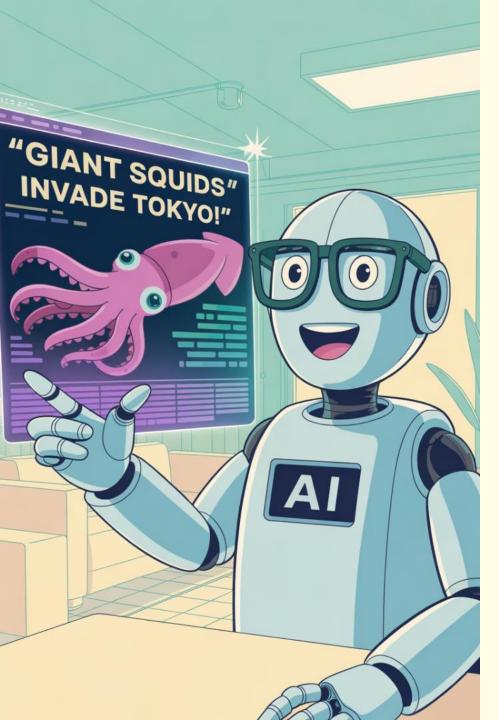
Chemical formulas and complex math need specialized tools



Use Specialized Tools

Calculators and domain-specific software for precision





How about AI Hallucinations: Making Stuff Up?



Knowledge Gaps

Missing information in training data



Pattern Overextension

Applies familiar patterns inappropriately



Confident Fabrication

Presents false info with authority



Mitigation Strategies

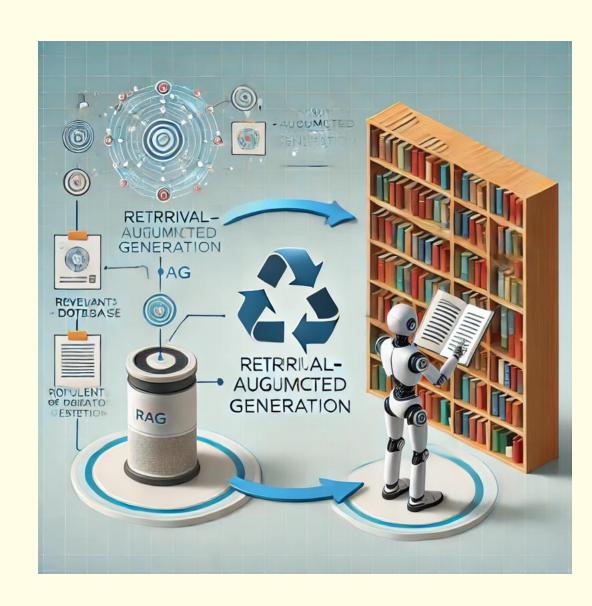
RAG, better prompts, lower temperature

What is RAG? How can it help?

Retrieval-Augmented Generation (RAG)

Combines external knowledge retrieval with generative AI to improve factual accuracy.

- Retrieve: Pull relevant documents from a knowledge base.
- Generate: Use the retrieved content to craft context-aware responses.



How does RAG retrieve content? Semantic Search in Vector Space

Traditional Search

Keyword matching: exact words must appear

Boolean logic: AND, OR, NOT operators

Limited understanding of meaning

Vector Search

Embeddings capture conceptual meaning

Finds semantically similar content

"Green energy grants" matches "renewable funding"

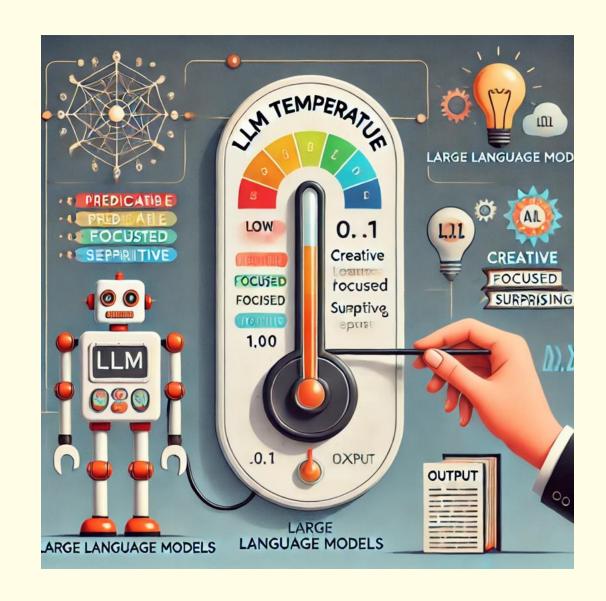


What is Temperature?

Temperature in LLMs

Controls the randomness of Algenerated responses.

- Low Temperature (e.g., 0.1) →
 Focused, predictable, repetitive.
- High Temperature (e.g., 1.0) →
 Creative, varied, surprising.



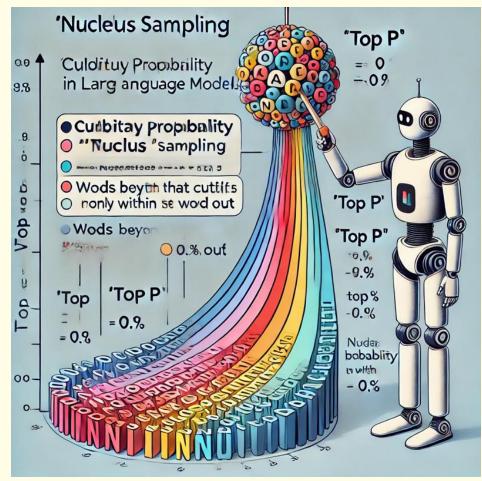
What is Top P (nucleus sampling)?

Selects from the smallest set of most likely words whose combined probability exceeds a threshold (e.g., 0.9).

- Words inside the shaded area = eligible for selection
- Words outside = excluded
- Balances coherence with creativity

Prompt: "Once upon a time in a distant galaxy,"

Temp.	Top P	Possible Next Word
0.2	0.9	"a" → "robot" → "landed" →
		(coherent but boring)
8.0	0.9	"a" \rightarrow "nebula" \rightarrow "shimmered" \rightarrow
		(more vivid, creative)
1.0	1.0	"a" → "pickle" → "danced" →
		(surprising, maybe absurd)



What are Diffusion Models?

Start with Noise 2 Random pixels with no structure

Remove Noise

Gradually refine toward target distribution

Improve Model

Learn patterns for better future generation generation

Create Image

ॐ

Guided by text or image prompts

What are Diffusion Models?

- 1. Cat on skateboard, dog in a space ship etc
- 2. Add noise iteratively until cat (dog) is gone
- 3. This is how we train the model
- 4. Provide a prompt Dog on a skateboard, dragon eating pizza
- 5. Diffusion model goes backward, step by step
- 6. If there's a definite dragon-like image emerging, double down
- 7. Training takes place with labeled images
- 8. Always starts with noise
- 9. Model recognizes that's a dragon!



What LLMs Can and Cannot Do Do



Generate Text

Create coherent, contextually appropriate content

? Develop Opinions

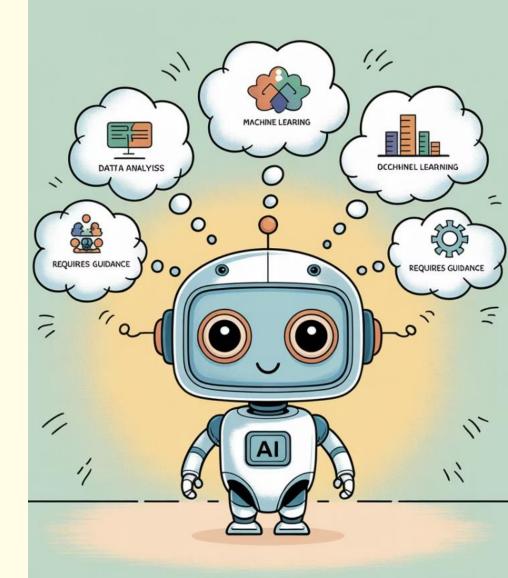
Simulate viewpoints without truly understanding

Verify Truth

Cannot independently fact-check without references

Original Research

Cannot search or retrieve new information



AI: Prediction Engine, Not Truth Truth Machine

Token Prediction

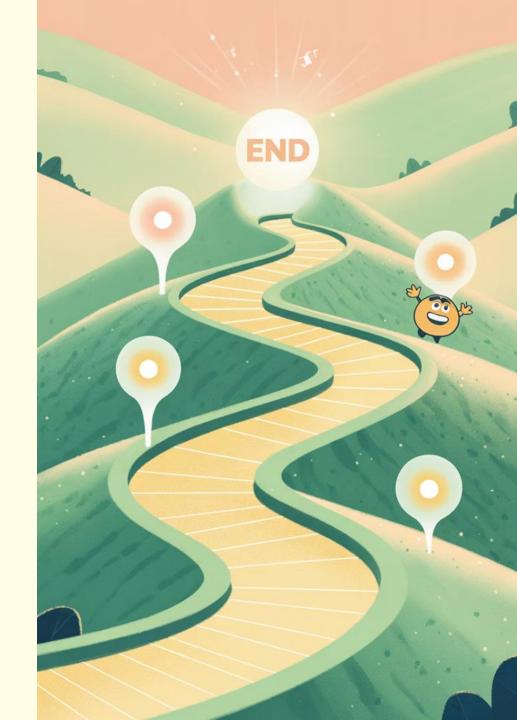
Generates next words based on patterns

Statistical Modeling

Not designed to distinguish fact from opinion

Pattern Recognition

Reflects training data biases and trends



Let's Pause...

Try slides 3&4 on Mentimeter



Some More General Questions...

And then some stuff from Shaun...

And then some Questions from you...

Can Al Really Understand Anything?

- Al does not "understand" like a human it models patterns of language, not concepts or consciousness.
 - Operates in probabilistic space
 - predicting what comes next based on training data.
 - Understanding is simulated via statistical coherence, not comprehension.
- Caution:
- Just because an answer is fluent doesn't mean it's correct—Al can mimic understanding without grasping truth.

Supervised vs. Self-Supervised Learning

- Supervised Learning: Learns from labeled data (input + correct output).
 - Example: Email spam detection.
- Self-Supervised Learning: Learns patterns from raw data without explicit labels.
 - Example: Predicting the next word in a sentence.

Why It's Important:

Self-supervised learning made large-scale models like GPT possible—vast amounts of unlabeled text were
used to pre-train them.

When does Al Think vs. learn?

Training:

- Model learns from data;
- requires massive computing power and time.

Inference:

Model generates outputs based on what it has already learned.

Cost Insight:

- Training GPT-3 required months and millions of dollars.
- Inference, by contrast, can run on a phone or laptop.
- Implication for Educators:
- Mostly we use inference (ChatGPT, Copilot, etc)
- But it may train on our data...





Do I need ChatGPT for that? Local vs. General Models

General Models

Trained on diverse data data across domains

Example: ChatGPT, broad knowledge

Local Models

Domain-specific training Run on local hardware for privacy

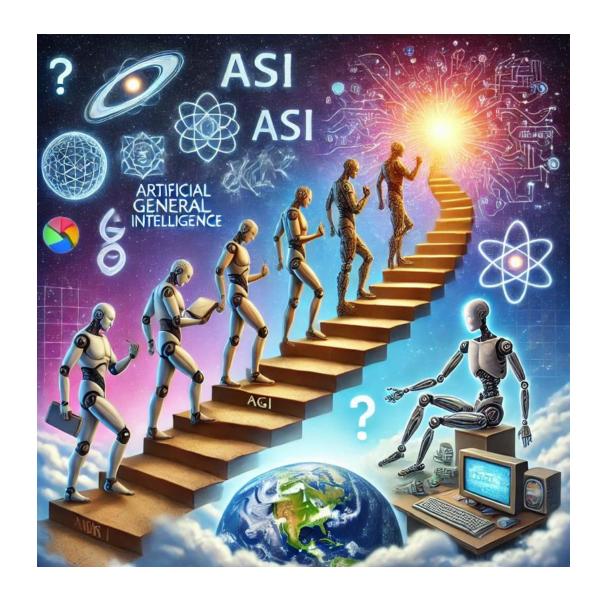
Specialized Models

Outperform general models in niche areas Lower cost and better control

Choosing the right tool for the task is a key principle of responsible AI use.

What is ASI?

- AGI (Artificial General Intelligence): Human-level intelligence across tasks.
- ASI (Artificial Superintelligence): Intelligence beyond human capabilities.
- Currently at narrow AI: specific-task performers.
- Reality Check:
 - no scientific consensus on when or even whether AGI will emerge.
 - Humans can adapt to situations Al can't
 - maybe AGI could...? ASI would come very soon after AGI
- Why It Matters:
 - Raises ethical and philosophical questions about control, consciousness, and societal disruption.
- Faculty: Humanities, philosophy, computer science, and engineering all have a voice in this conversation.



Al and Creativity: Hype or Help?

Limitations

Derivative - remixes existing patterns

No original vision or intention

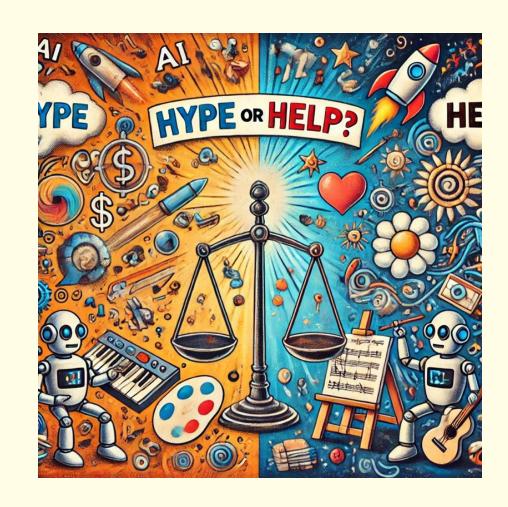
Lacks understanding of cultural context

Benefits

Generates diverse ideas quickly
Explores variations of concepts
Enhances human creativity as a tool

 Al as a creative assistant for writing, coding, music, design. Best used with human curation.

Is AI replacing the artist—or becoming a new kind of brush?



Why is Al Biased?

- Training Data Reflects Society
 Internet content includes all biases and perspectives
- Representation Imbalance
 Minority perspectives have fewer training examples
- Steering Possible
 Intentional prompting can counter inherent biases
- Self-Awareness Critical
 Understanding Al limitations improves outcomes
 - ChatGPT seems better now at expressing the need for reputable sources



Should I Use AI?

Expert Tool

4

(2)

 \mathbb{Z}

Knowledgeable but requires verification verification

Productivity Gains

Focus on higher-level tasks



Always evaluate AI outputs



Like calculators replacing slide rules

AI Therapy: Benefits and Risks

Potential Benefits

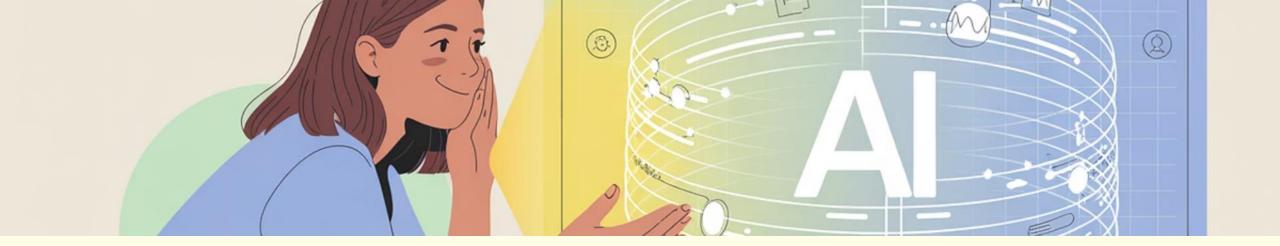
24/7 availability for emotional supportReduced stigma for seeking helpConsistent journaling companion

Serious Concerns

No real empathy or understanding Potential for harmful advice Not equipped for crisis situations Cannot replace human connection

e.g., Woebot, Replika





Al Sentience: Understanding the Simulation

No Consciousness

Al has no subjective experience

Human Projection

We attribute feelings to convincing responses

Statistical Mimicry

Simulates understanding through patterns

Ethical Implications

False belief in sentience affects how we use AI

Al Writing Its Own Code

Current Capabilities

Generates, debugs, and iterates code effectively

Available Tools

GitHub Copilot, AutoGPT, DeepSeek-Coder

Limitations

Still requires human oversight and verification

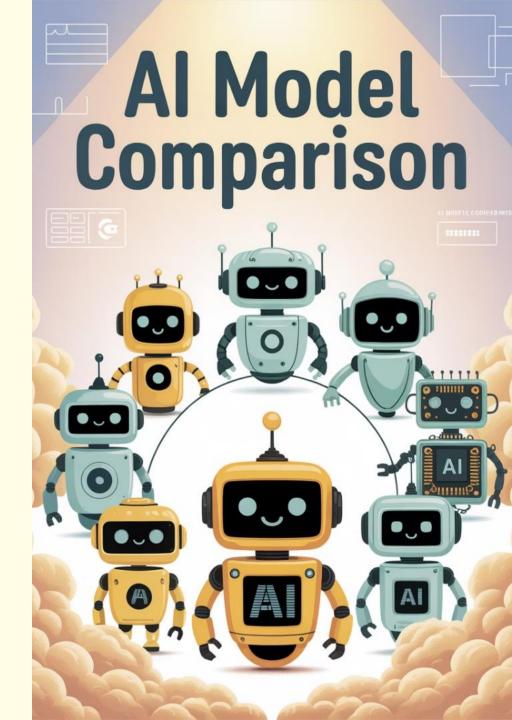
Future Direction

Moving toward autonomous agents with coding abilities



Comparing AI Models

Feature	ChatGPT	DeepSeek
Focus	General dialogueCode & reasoning	
Release	May 2024 (GPT	-48 ew platform architecture
Strengths	Conversation al, creative	Technical, multilingual
Best Use	Writing, general tasks	Programming , mathematics



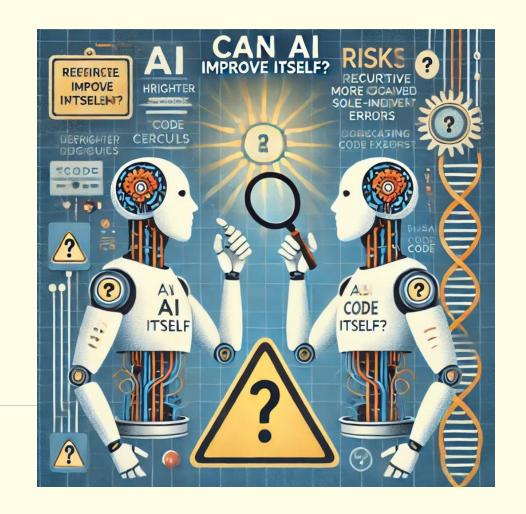
Can Al Improve Itself?

Fundamental Limits

Cannot transcend its own algorithm basis

Complexity Challenges

More complex code introduces more bugs



Mode Collapse Risk

Training on Al-generated content causes degradation

AI Ethics: Our Responsibility

Personal Accountability

We must take responsibility for AI use

Moral Reflection

Al forces us to examine our own values

Democratizing Power

Reducing information and capability monopolies

Shaping the Future
We get to determine what Al becomes



Will AI Take Away Human Joy?

Creative Pursuits

Still fulfilling regardless of AI capabilities

Process matters more than product

Personal meaning transcends efficiency

New Opportunities

AI handles routine tasks

Humans focus on meaningful work

More time for creativity and connection



Job Transformation in the AI Era

Historical Context

Nixon 1956: "Four-day workweek in our lifetime"

Technology promises often unrealized

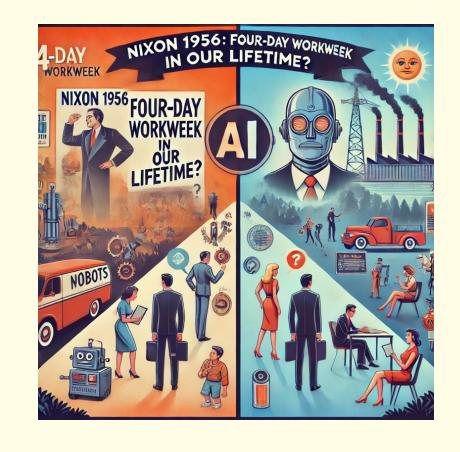
Work evolves rather than disappears

Future Reality

Jobs will require AI collaboration

Tool building becomes central skill

Managing AI agents creates new roles



Entropy of the universe increases (2nd Law)
When Order in the system increases, Disorder in the surroundings must go up!



What is Vibe Coding?

Natural Language Instructions

"Make it more playful" or

"Corporate look and feel"

Al Interpretation

Translates vibes into technical specifications

Generated Solution

Code that matches desired aesthetic/feeling

Human Refinement

Developer tweaks for precise implementation



Tool Fatigue: Strategic Approach Approach



Focus on Core Tools

Master fundamentals rather than every variant



Evaluate Workflow Impact

Only adopt tools that significantly improve results



Transferable Skills

Prioritize concepts that apply across platforms



Sustainable Pace

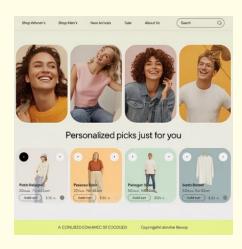
Quality understanding over over quantity exposure

How do I avoid AI? – Go for it, but...









Don't Use:

- your phone, voice to text, spam filters; Fitbit, Health Apps
- Credit Card/Bank Fraud Detection
- Internet search autocomplete, or any web search
- Facebook, Instagram, TikTok, YouTube, Netflix, Spotify etc.
- Google Maps, Waze, Apple Maps, Uber, Lyft
- Amazon & other online stores
- Ticketmaster, Airline Booking, dynamic pricing, Highways (Toll cameras)
- Zoom (noise reduction)

Conscious Al: How Far Away?

Current State

Pattern matching without understanding, Claims of sentience (e.g., LaMDA case at Google) reflect anthropomorphism, not reality.



We don't understand human consciousness itself

Fundamental Questions

May involve quantum mechanics or unknown principles



Detecting Al-Generated Content

Al Weaknesses

Poor understanding of object permanence

Struggles with physical plausibility

Inconsistent details (fingers, reflections)

Detection Methods

Subtle heartbeat cadence missing in fakes

Digital watermarks embedded in generation

Pattern analysis revealing statistical anomalies

Let's Pause...

Try slides 5&6 on Mentimeter



Why should I care if AI is or is not sentient?

Key Points:

- Sentience = conscious experience—pain, joy, self-awareness.
- Al has no emotions, beliefs, or self-concept—it responds statistically based on training data.
- Claims of sentience (e.g., LaMDA case at Google) reflect anthropomorphism, not reality.

Why It Matters:

- The illusion of sentience can impact:
 - User trust
 - Emotional dependency

Will Al take my job away?

Richard Nixon 1956:

- "We are going to see it [the four-day workweek] in our lifetime."
- Spoken during a campaign speech as Vice President.
- Emblematic of mid 20th century optimism about automation

Entropy of the universe increases (2nd Law)

 When Order in the system increases, Disorder in the surroundings must go up!

Maybe, but we'll need two of you over there!

Build that tool; manage that Agent

How can we detect fakes?

Al doesn't understand:

- Permanence of objects
- Physics

We can use Al to detect:

- absence of subtle heartbeat cadence
- so-called AI "Watermarks"