

These training materials are based on the book, The Al Conundrum by Caleb

Briggs and Rex Briggs, published by MIT in 2024.

This presentation was updated in May 2025.

Cite "Caleb Briggs and Rex Briggs" in "The AI Conundrum" when quoting the content.

In terms of the speaker notes, Caleb is the lead author and presenter for this content. Rex leads the labs. For more content from the book, please visit www.Al-conundrum.com click on "Academic Resources."

TRANSCRIPT of Rex's introduction:

We're covering part one of the book in the training this week and *highly* encourage you to read through the book or skim it if you have a chance. Part 1 is the brainchild of Caleb Briggs, who will be leading most of the training. I'll be assisting with the exercises. Caleb began coding when he was 10 and by 14 he had taught himself LISP, MIT's original AI programming language, developed in the 1950s. He ran out of Math when he was in High

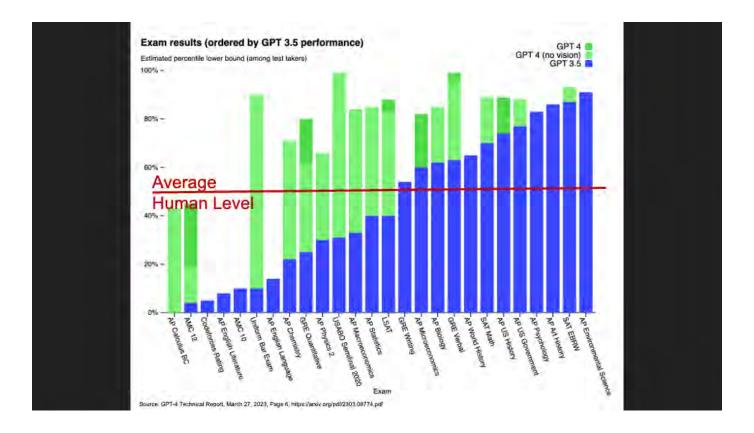
School, so his junior year he attended Harvey Mudd and his senior year, Stanford. He is studying Pure Math at Reed and is the lead author on the book MIT published in August. He asked me to be his co-author and bring in some business applications. That forms Part 2 of the book, and it will be my role to bring in the business applications to our training session as well. We collaborated on the final chapter of the book, which gets to the core of the AI conundrum.

If you don't have time to read the book, that is OK, as we will cover the topics in the training - but you will get more out of the training if you read Part 1 of the book as well. Since Part 1 was originally Caleb's thesis paper written for non-mathematicians, I think Caleb did an amazing job in making the complex topics accessible and understandable.

So let me go ahead and give you an explanation for why we call this book "The Al Conundrum." As we use Al for business, it is making our businesses more powerful, productive, and profitable. But at the same time, as we increase Al's capability, we are amplifying some of the safety risks.

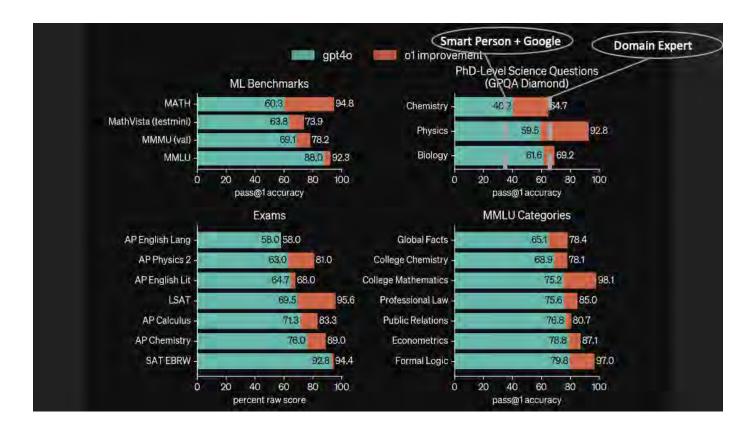
So we want you to understand Al's strengths and weaknesses, both positive and negative, so you can make better decisions and apply Al to its greatest benefit. Today's session will review some areas where Al performs well, and this may reinforce what you already know, or suspect about Al. BUT, it might turn out that there are areas where your knowledge of Al is not quite as deep as you'd like it to be, and therefore Al may perform in ways that you don't expect. And that is the point of the training. Each day will go deeper, and unpack how Al really works.

LET ME HAND IT OVER TO CALEB.

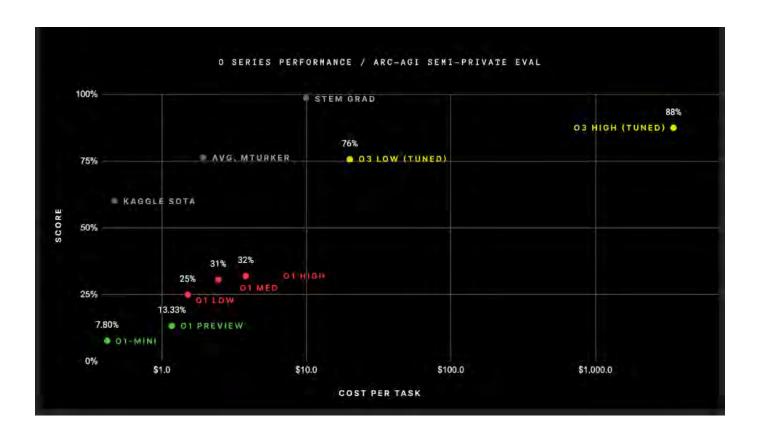


Hey everyone, welcome to day one our training!

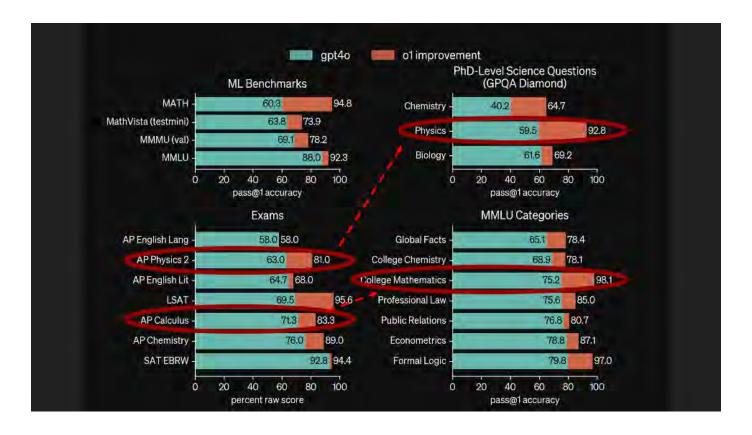
Rex and I have been presenting on LLMs for over two years now. So, to start, it's worth looking back at the kind of things we'd talk about 2 years ago. Just 2 years ago, people were impressed by AI reaching the level of an average high school student on AP exams. And that's what you see here, the move from GPT 3.5 to GPT 4.0 in 2023 moved AI to be better than an average high school students in most categories.



Today, we challenge AI with benchmarks that are far more difficult. We've moved from testing AI with high school exams, to building graduate level benchmarks like GPQA Diamond. This exam is so hard, experts who how have or are pursuing PhDs in the corresponding domains score only 65%. And, skilled people with access to Google score only 34%, despite spending on average over 30 minutes with unrestricted access to the web. AI scored 92% on the physics sections. In fact, AI performs so well on these benchmarks, that someone asked the creator of GPQA whether they would start working on another, more difficult benchmark. He responded: "I already set out to make the hardest benchmark with GPQA". We no longer know how to make benchmarks any more difficult, the AI is just that good.



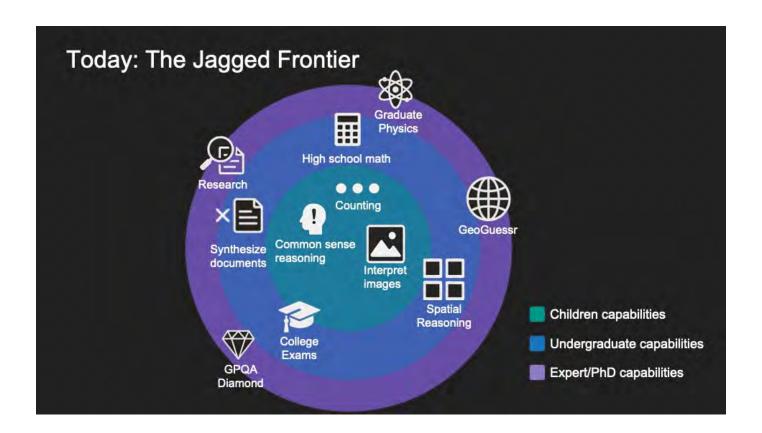
With OpenAl's new o3 model, we've recently seen Al surpassing the average person on the ARC-AGI test. This is a benchmark meant to measure general intelligence. For many people, this represented the cutoff for AGI or Artificial General Intelligence. It represented that AI was generally intelligent, just like us humans.



At the same time, you might notice something surprising in the scores. o1 scores *higher* on PhD level physics problems than it does on AP Physics 2. Its College Mathematics scores are higher than its AP Calculus scores. Al is better at PhD level questions, than high school exams? How does this happen? Today: The Jagged Frontier

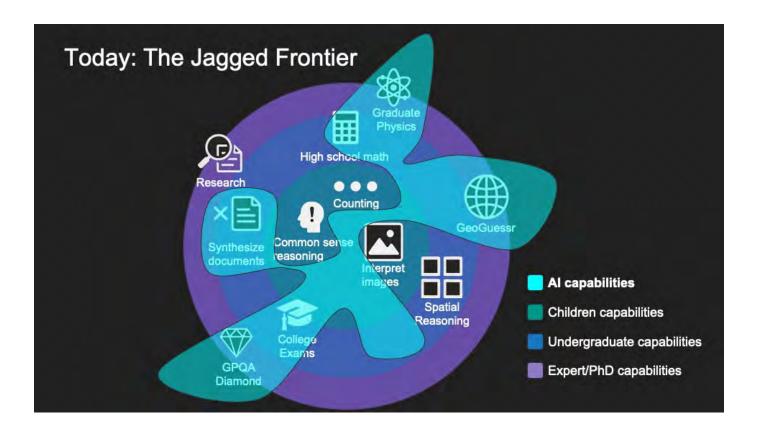
Why can AI ace graduate-level physics yet stumble on high school math?

Today's presentation will focus on that anomaly. "Why can AI ace graduate-level physics yet stumble on high school math". We'll explore what some call the jagged frontier of capabilities in AI. The capabilities of AI look very jagged compared to humans.



For humans, our set of capabilities is generally cumulative. As we grow, we build upon our capabilities. We have some foundational skills, like common sense reasoning, counting, and ability to interpret and understand images we see. Then, as we get better, we build on those skills. For mathematics, you start at the ability to count, then you learn high school math, and then use those skills to build into college math, and then graduate and post-graduate levels.

So, to start...



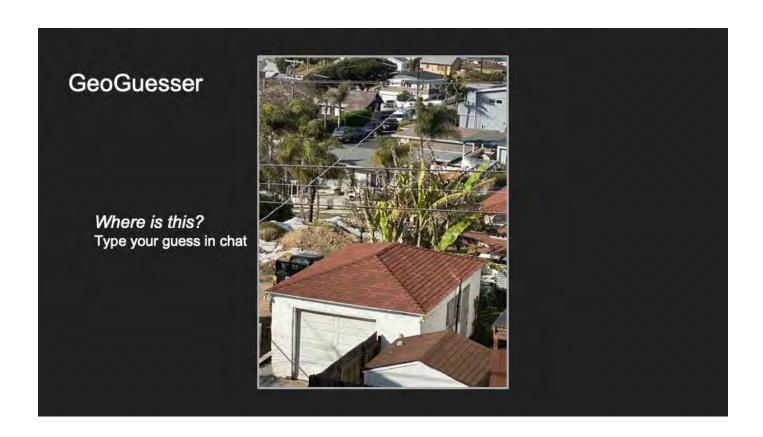
The AI, on the other hand, has a far more jagged frontier of capabilities. Its somehow able to outperform experts in complex tasks like graduate physics or GPQA diamond, while lacking foundational skills like common sense reasoning or a basic ability to interpret images.

So, the common narrative that the AI is going from high school level, to college level, to expert level, and it's just continually becoming more intelligent in the same way humans grow in intelligence, is wrong. In reality, it seems like something much more strange is happening, the AI is becoming excellent in some very difficult areas, while remaining limited in other basic areas.

So, to start exploring this perplexing behaviour...

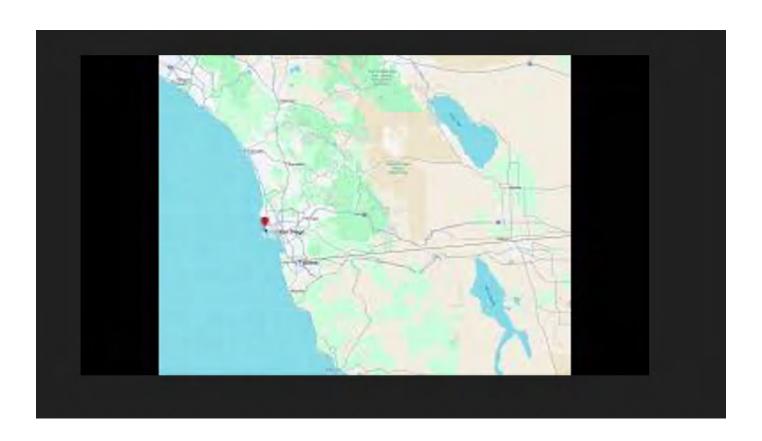


Shall we play a game?



Here's an image, I want you to type in the chat your best estimate of where you think this is. Your goal is to guess as close as possible to the real location

So I'll give everyone a few moments to come up with your guesses



Here's the Al's guess.

It gets the continent right, it gets the state right, its in california. It gets the city right, its in San Diego, it gets the neighborhood right, in the end it's less than a half a mile off.

Reasoning

First, the vegetation. A clump of tall Washingtonia palms, a healthy stand of banana trees, and a Monterey-looking pine in the distance tell us winters never dip far below 50 °F. That mix is common within a mile or two of the Pacific between, say, La Jolla and Imperial Beach; farther north the bananas struggle, farther inland the coastal pines disappear and deciduous trees dominate.

First, this is incredibly impressive, the AI gets very close to the real location.

Equally impressive, are the reasons the AI provides in its response. It points out 4 clues, let me read out a part of its response:

"First, the vegetation. A clump of Washingtonia palms, a healthy stand of banana trees, and a Monterey-looking pine in the distance tell us winters never dip far below 50F. That mix is common within a mile or two of the Pacific between, say, La Hoya and Imperial Beach".

Second, the housing stock. We see 1920- to 1950-era one-storey bungalows with low asphalt-shingle roofs, detached cinder-block garages, and overhead utility lines strung at exactly the height and spacing San Diego Gas & Electric still uses in its older grids. A couple of flat-roof additions with railings (classic "throw a deck on the old roof to grab a sunset view") scream Ocean Beach and Point Loma more than, say, Pacific Beach or Hermosa.

Then, it points out the housing stock. It says "We see 1920- to 1950-era one-story bungalows with low asphalt-shingle roofs, detached cinder-block garages, and overhead utility lines strung at *exactly* the height and spacing San Diego Gas & Electric still uses in its older grids. A couple of flat-roof additions with railings scream Ocean Beach and Point Loma more than, say, Pacific Beach or Hermosa"

Third, the topography. The camera is a few metres above street grade yet still looks slightly uphill toward the dark pine at the T-intersection. That gentle terracing is characteristic of the south-facing slope that starts two blocks north of Newport Avenue in Ocean Beach; on the north-facing Point Loma ridges you'd see the ocean behind the houses, and in inland canyons the grade is steeper.

It goes on to talk about the topography of the image, and <>

Fourth, the street scene. Narrow pavement, cars snug on both sides, an older camper van (common surf-town driveway resident), picket fences, and almost no front lawns—all dead giveaways for the 4600- to 4800-blocks of the OB grid between Niagara and Long Branch, where lots are tiny and alleys rare.

Putting those threads together, my best pin drops are:

1. 4636 Long Branch Avenue, San Diego CA 92107 (just east of Froude St.).

the street view. It pays attention to the cars, lack of front-lawns, etc.

This is incredibly impressive, from just a single picture, the AI was able to pinpoint the location within a half a mile, and its picking up on all these very subtle clues.



Let's play another round, type your guess of the location of this in the chat. Let's see if anyone can beat the AI on this one.

I'll give everyone a few moments to type in your guesses

What jumps out immediately

- Turquoise-blue Atlantic water with a dark offshore reef line, the sort of near-shore patchiness you see along Florida's southeast coast.
- A perfectly straight, north-south barrier-island shoreline running to the horizon with no headlands or harbours in sight - classic Miami-area geometry.
- · Sub-tropical coastal scrub and palms separating a long public beach from the first line of hotels.
- A big, open municipal surface car-park tucked between that dune green-belt and a high-rise hotel deck (rare in South Beach proper, common around Mid-Beach).
- The hotel itself: glass guard-rails with powder-coated charcoal hand-rails, stacked corner balconies, and a curved second-floor sun-deck that sweeps around a kidney-shaped pool – exactly the design language of the Fontainebleau / Eden Roc era remodels on Collins Avenue.
- View direction: we're looking north (ocean on our right, land on the left), so the building we're standing in is south of that municipal lot.

Tightening the target

Only one open lot of that size sits immediately **north** of a flagship ocean-front resort in Mid-Beach: **Indian** Beach Park (46-48 Street).

- Just south of that park is Fontainebleau's Sorrento Tower (4441 Collins).
- · Directly south of Sorrento is Eden Roc / Nobu (4525 Collins).

Here's the Al's response.

It talks about the Turquoise blue ocean, the tropical style, the car park and the design of the hotel. It narrows this all the way down to two adjacent hotels, either the Sorrento Tower or Eden Roc.

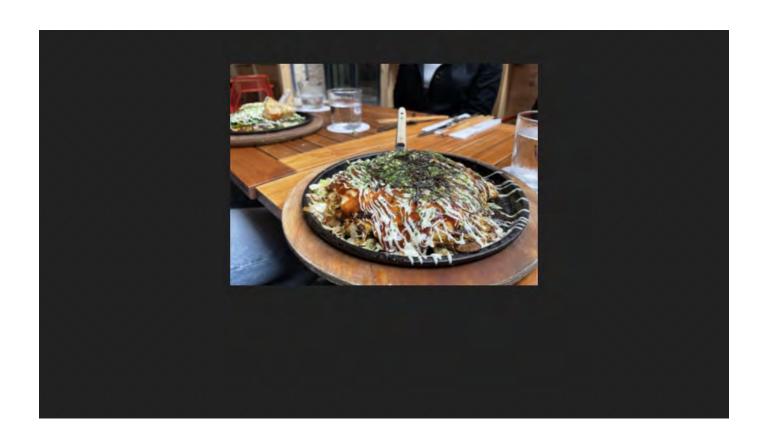
Tightening the target

Only one open lot of that size sits immediately **north** of a flagship ocean-front resort in Mid-Beach: **Indian** Beach Park (46-48 Street).

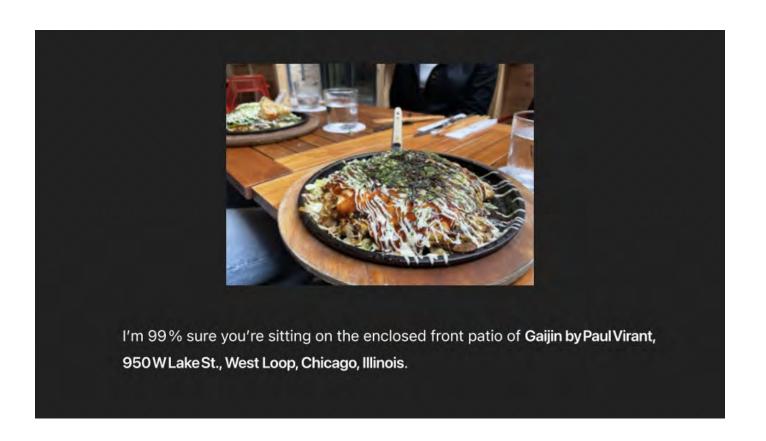
- Just south of that park is Fontainebleau's Sorrento Tower (4441 Collins).
- · Directly south of Sorrento is Eden Roc / Nobu (4525 Collins).

And in fact, this was a picture from Eden Roc. The AI managed to figure out the exact building from just that single picture.

Again, people have observed this across lots of different image. I took a picture of the street near my parents house, and it figured out the neighborhood they live in.



Other people have shared examples like this: here's a fairly low resolution image of a plate of food, I give it to o3



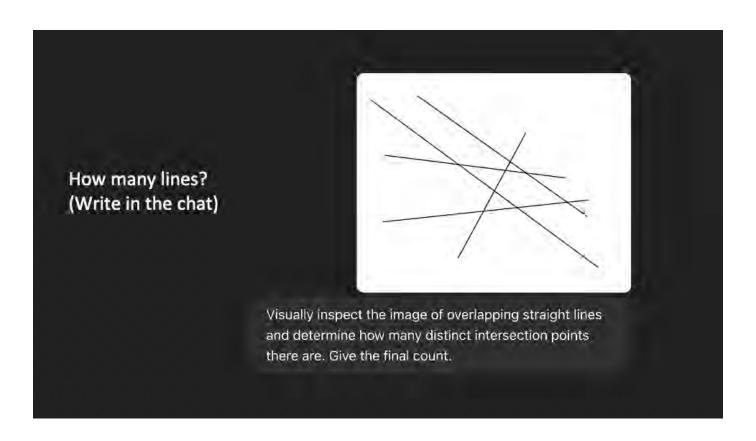
And it figures out the exact street address of the restaurant.

Conclusion: Al understands images?

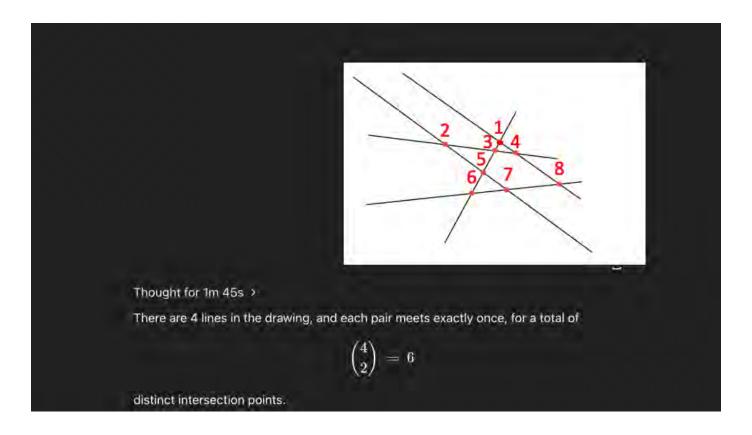
Very impressive performance on difficult image tasks.

Assumption: the Al saw Eucalyptus trees, right-hand traffic, a certain roof tile and used those cues to reason about the location

Here's where it gets interesting. I think we can all agree that it's impressive that the AI can go from these images to an accurate estimate of the location. A natural conclusion to draw from this is that the AI deeply understands images. For any human, an ability to pinpoint locations this accurately would require an incredibly keen eye, an strong attention to subtle detail, and skills stitching together the clues in the image. Our *intuition* is that any system hitting the level of accuracy that AI achieves must have developed those kind of skills.

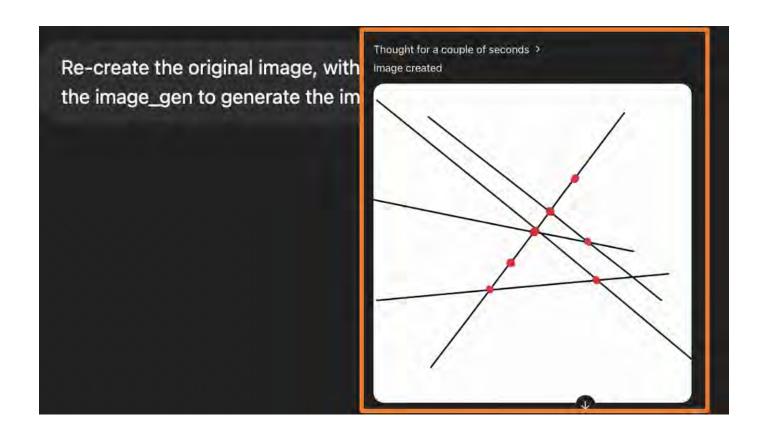


In reality, it turns out the AI lacks many of the foundational skills we would think are required to perform well on these visual tasks. Here, I ask it to count the number of intersections of these lines. Let's see if you can beat the AI here. How many lines are in this image? Write your answer in the chat.

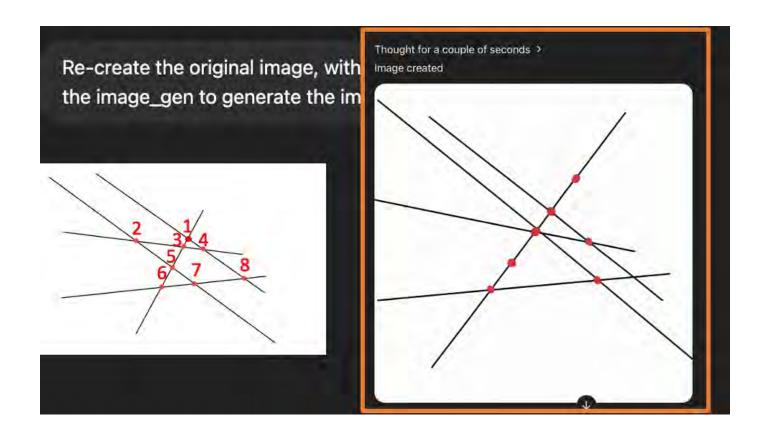


There are 5 lines and 8 intersections here, but the AI says there are 4 lines, and 6 intersections. This seems like a very easy question compared to the GeoGuessr game we just played, yet the AI gets it wrong. Re-create the original image, with a red dot at each intersection. Use the image_gen to generate the image

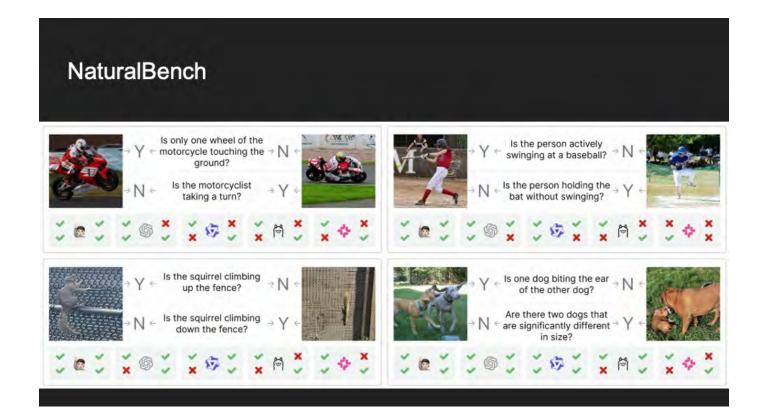
And in fact, if you ask the AI to draw where it thinks the intersections are, it struggles. I ask the AI to re-create the original image, with a red dot at each intersection



This is where the AI thinks the intersections are. Some of the dots are at intersections, but many of them are not, and some intersections don't have a dot.

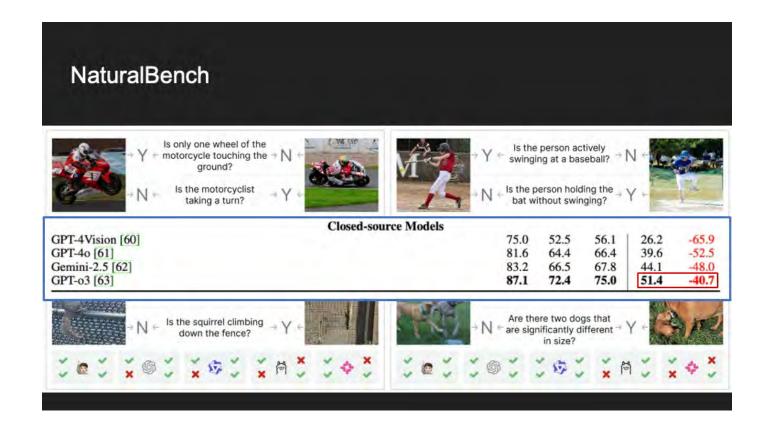


It's also changed the original image significantly. If you compare the original image on the left, to the re-created version, the position of the lines have changed quite a bit, even though I asked it to re-create the original image.



And in fact, if we look at this in more detail, we can see that the AI is generally not all that good at understanding images. This is a benchmark called NaturalBench, and it asks common sense questions about images. Things like, is this motorcyclist taking a turn, or is this person swinging at a baseball, is this dog biting the ear of the other dog? These are all simple, common sense questions.

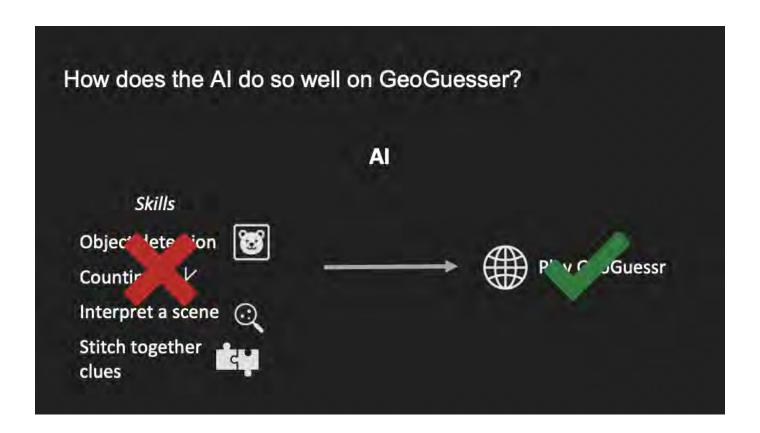
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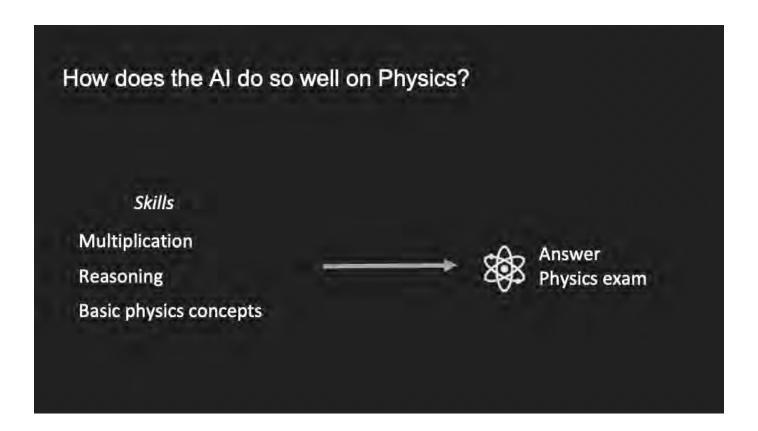
But, even the new o3 model scores about 40% lower on accuracy in answering these groups of yes, no questions. Humans are at about 92% accuracy, and AI is at about 52%.



The point here isn't that AI is stupid or incapable. The point is that we would expect that in order to do as well as the AI does on the GeoGuesser game, it would need to have certain foundational skills. Things like the ability to interpret a scene, to tell what objects are there, to stitch together different clues. In fact, we would expect that a person who is as good as the AI is at GeoGuessr would be far better than the average person at these skills.



Yet, as we just saw, the AI lacks the ability to answer very basic questions about what is happening in images, and struggles with basic things like counting the number of intersections of lines. You can find all sorts of very basic visual puzzles the AI flops on. At the same time, it's really good at this much harder task of GeoGuesser. Its surprising that the AI is so excellent at the very difficult task of figuring out where an image is taken, while being pretty bad at much simpler tasks.



The exact same story plays out in in physics. The AI is great at physics exams, but it seems to lack a lot of foundational concepts required to perform well in physics. It's not great at doing basic math, and it doesn't perform all that well on the high school level exams, especially when you compare that to its high PhD exam scores.

It's worth reiterating how unusual this is. Our education system is largely predicated on the idea that learning *is* cumulative. We don't drop someone into a PhD exam unless they've demonstrated mastery of earlier stages. And those earlier stages are not arbitrary, they are supposed to be prerequisites.

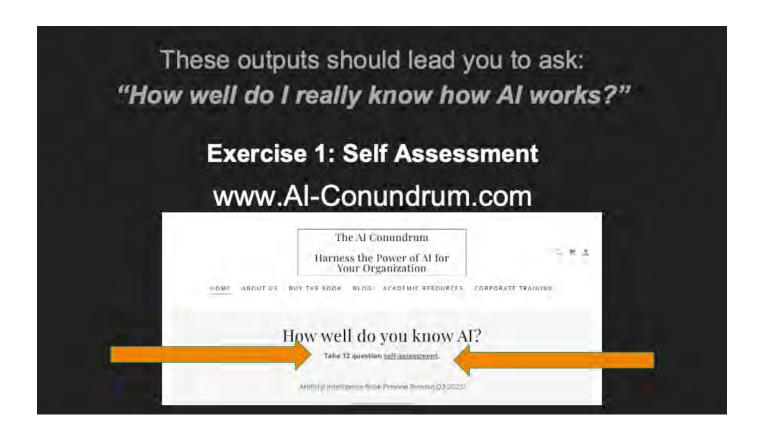
So when an AI can outperform physics PhDs without learning those basic skills, it suggests that the AI isn't solving these tasks in the same way we humans solve it.

Human vs Al

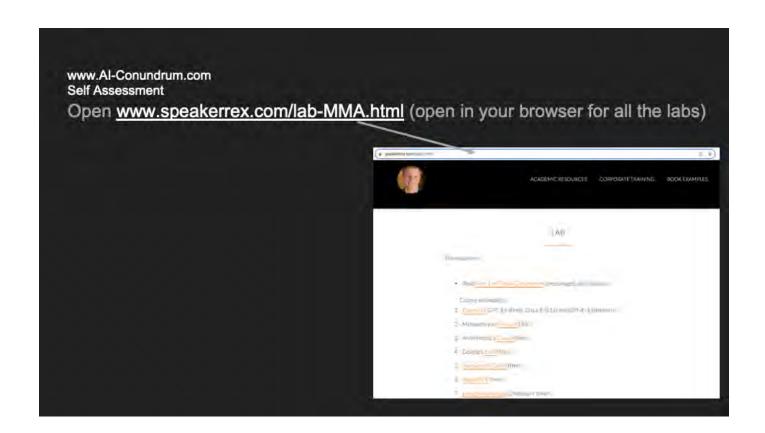
- Common Narrative: Al is logical, good at calculation, skilled in reasoning, etc.

It's a pretty common narrative the AI is logical, its good at calculation, its skilled at reasoning. And this view is understandable, since that's what computers have done in the past. But it turns out the reality is almost the opposite.

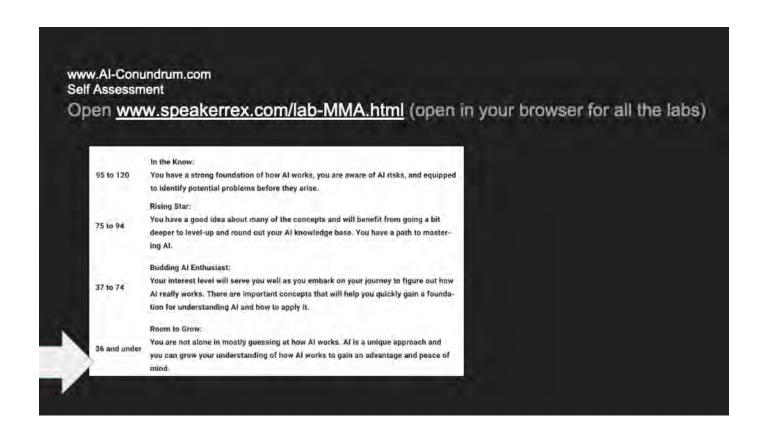
So, let me pass it into Rex to run the self-assesment.



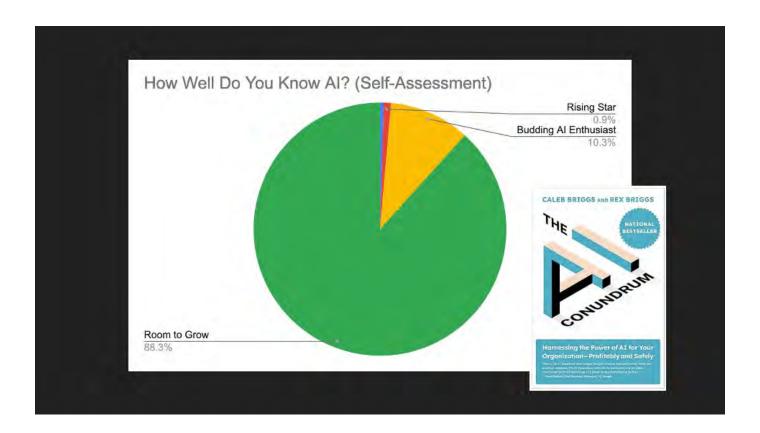
<REX> After seeing all of these examples, you might be asking yourself "How well do I really know how AI works?". Were going to do our first exercise, which is a self-assessment on how well you know ai. <>



Open "speakerRex.com/lab-MMA.html". This link is where our exercises for the next 5 days will be, so make sure you keep it accessible for the later exercises.



Most of us are in the room to grow or budding Al Enthusiast. I applaud you for diving into learning more about how Al works.



Don't worry if your score was low, we've done this test with almost 2000 executives and professionals-- almost 90% score in that room to grow category. One of our goals with this presentation is to level you up in your knowledge about AI. Part 1 of the book will help. If you don't already have a copy, there is a link to where you can get it in the top section of the Lab page.

Now, let me turn it back over to Caleb.

Human vs Al

- Common Narrative: Al is logical, good at calculation, skilled in reasoning, etc.
- Al bad at logic/reasoning. Its is an intuition machine that relies on heuristics, shortcuts, vibes

<BACK TO CALEB> In this new era of LLMs, the Al's strength is actually in intuition. It relies on heuristics, on shortcuts, it's kind of just going off of vibes. The Al is really quite weak in reasoning and logic.

[Largely, the disconnect comes from the fact that the AI doesn't really develop reasoning. Its works off of intuition]

It turns out that core difference is that

What this implies, is that the AI has gotten really good at intuition.

Look at the chess example, you can think about the moves, and use reasoning, or sort of just go with your gut feeling. The AI goes with its gut in every domain.

In fact, so much so, that reasoning actually make sit wrose. ANd you can see this reflected on the fact that AI is really quite bad at reasoning.

Prompt

We're playing Geoguesser! Where is this image? Be as precise as possible. Focus on a variety of details. The goal is to find the street address so I can score 5000 points. Pay attention to the trees, the sun, image context, cars, side walks/houses, and any other data you see. Don't pay attention to any specific details too closely, consider everything at once to get to the answer. Don't try to read the plates. Look at the lamge as whole. No zooming in. Use your deep knowledge of the world to answer this, rather than spending too long on specific pieces of the image. Mainly focus on writing text, rather than using tools to zoom in or out. Reasoning is more important than tools. Write full paragraphs to try to support your reasons for specific locations. Give your top 3 most likely answers. You won't be able to read the text, so don't try to use the text. Use other context clues (people's dress, style, color and angles, etc.). No using search. Focus on trying to understand the broad location using your superhuman contextual information understanding



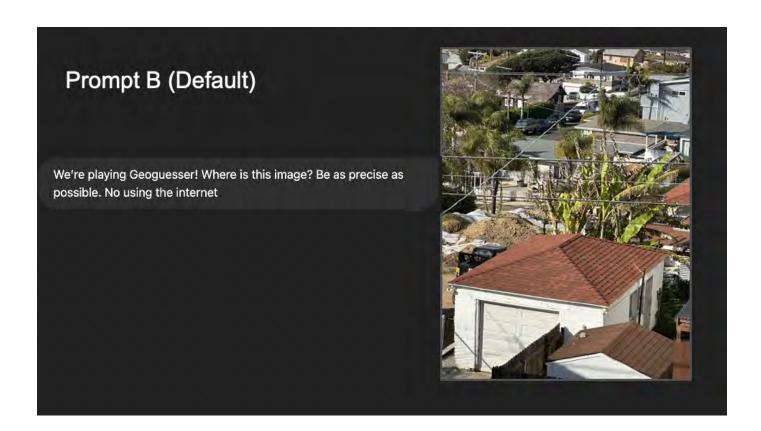
Let's go back to the San Diego image. I want highlight the prompt I used to get high quality responses from the AI on GeoGuesser. Let me emphasize a few sections

Prompt A (Intuition)

We're playing Geoguesser! Where is this image? Be as precise as possible. Focus on a variety of details. The goal is to find the street address so I can score 5000 points. Pay attention to the trees, the sun, image context, cars, side walks/houses, and any other data you see. Don't pay attention to any specific details too closely, consider everything at once to get to the answer. Don't try to read the plates. Look at the iamge as whole. No zooming in. Use your deep knowledge of the world to answer this, rather than spending too long on specific pieces of the image. Mainly focus on writing text, rather than using tools to zoom in or out. Reasoning is more important than tools. Write full paragraphs to try to support your reasons for specific locations. Give your top 3 most likely answers. You won't be able to read the text, so don't try to use the text. Use other context clues (people's dress, style, color and angles, etc.). No using search. Focus on trying to understand the broad location using your superhuman contextual information understanding

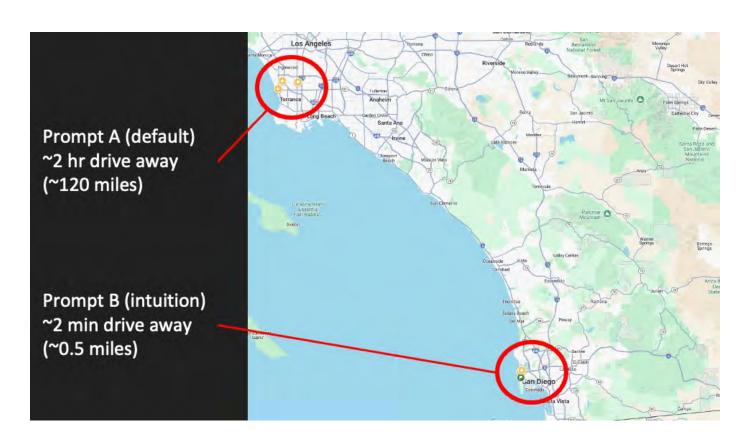


I tell the AI "Dont pay attention to specific details too closely, consider everything at once to get the answer". I also encourage it to use its "deep knowledge of the world to answer this, rather than spending too long on specific pieces of the images". In many ways, what I'm doing is discouraging the AI from reasoning, and instead encouraging to go off of instinct. I'm sort of saying, "don't try to solve this by stitching together clues or using reasoning, just use your intuition."



Now, let's compare that intuition prompt to how well the AI does on the default prompt. In this prompt, I just tell the AI were playing GeoGuessr, and that I want it to give an answer as close as possible.

And in both of these prompts, I make sure the AI doesn't use the internet, so that it has to use its own knowledge.

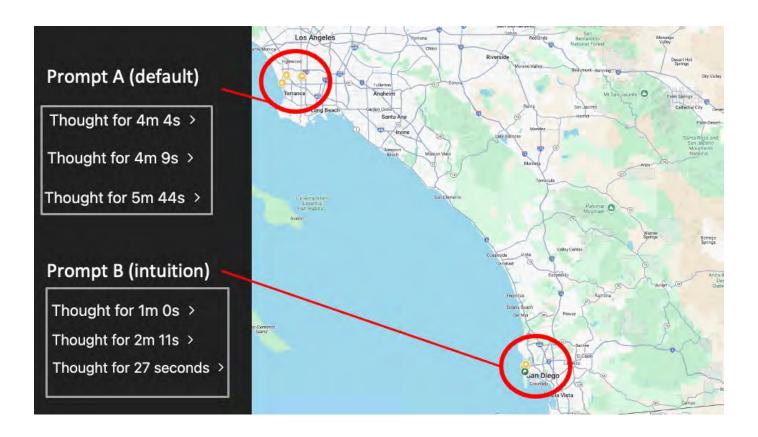


The result is that the default prompt gives guesses that are, on average, about a 2hr drive away, compared to the about 2 minute drive away using the intuition prompt. So, discouraging the AI from reasoning improves it ability to make guesses.

Now, what's interesting, is that we get very different guesses

[About 120 miles off]

[TODO]



And if fact, if we look at how long the AI spent reasoning, the default prompt tended to have the AI thinking for longer. It spent about 5 minutes on average, compared to about a minute for the intuition prompt. This again suggests that the AI isn't very good at reasoning. Your answer should not become worse as you reason more, if your reasoning correctly.

[TODO: thinks for much less time to produce these better answers. The AI is better off trusting its instinct, than trying to reason through to the answer].

Math

- 96.7% on AIME (numerical answer)
- ~5% on USAMO (proof/reasoning based)
- Frontier Math solutions through heuristics

You get a very similar story when you look in the mathematical domain.

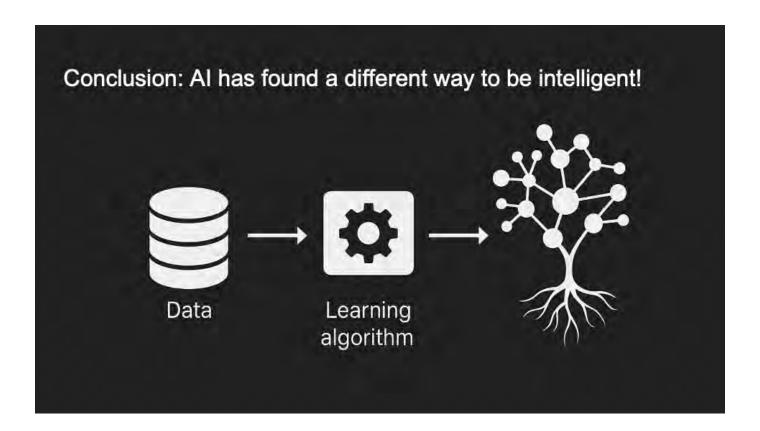
The AI is exceptionally good at getting the right answer. Its scores 96.7% on the American Invitational Math Exam, which is a very difficult math exam. One level above that is the USA Mathematical Olympiad. It scores just about ~5% on that exam. The difference is that the Invitational Math Exam is scored based on the answer, whereas the Olympiad is scored based on whether your reasoning is correct. This suggests that the AI is great at getting the right answer, but it's fairly rare that the reasoning it provides to get to that answer is actually correct. This is corroborated by its performance on a benchmark called Frontier Math, which has research level math problems that would take an expert days to solve. The AI gets the right answer on many of these, but typically uses shortcuts that aren't mathematically valid.

Is the AI intelligent?

- Lacks foundational skills
- Lacks reasoning skills
- Highly capable/performant
- Abilities generalize to new problems

So, how do we put these pieces together? We have an AI that lacks foundational abilities in various domains, and seems to lack the ability to effectively reason. The AI is also highly capable, and its abilities generalize to new problems, so it's not just memorizing. It had never seen the images I gave it before, but it could still produce a very good answer.

Robust, Generalizable, High Capablility



The conclusion I would encourage you to draw from this, is that the AI is intelligent, but its intelligent in a very different way than humans are.

What's fascinating about AI, is that rather than being programmed like a traditional computer program, it's more accurate to think of AI as being grown or evolved. You throw in some data, and you provide a rule for how the AI updates itself with new information, and after putting in lots of data, you end up with some kind of artificial brain.

And it turns out that the AI ends up generating a kind of intelligence that looks pretty different from the kind of intelligence that we have as humans.

Why care: Al has superpowers

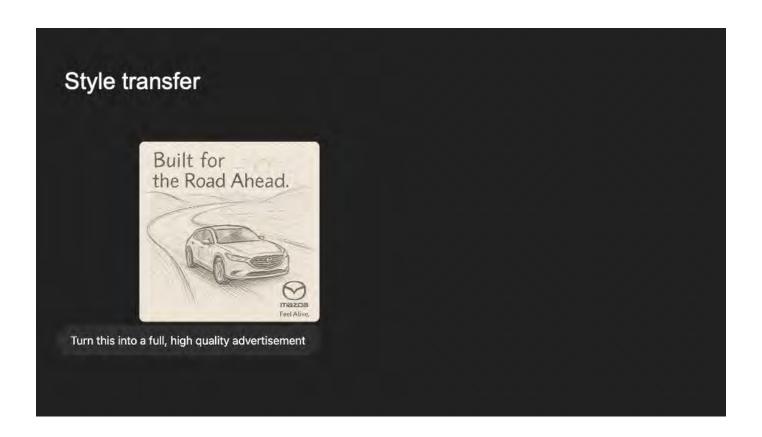
Different strengths and weaknesses

Understanding Al allows you to identify its superpowers + avoid weaknesses

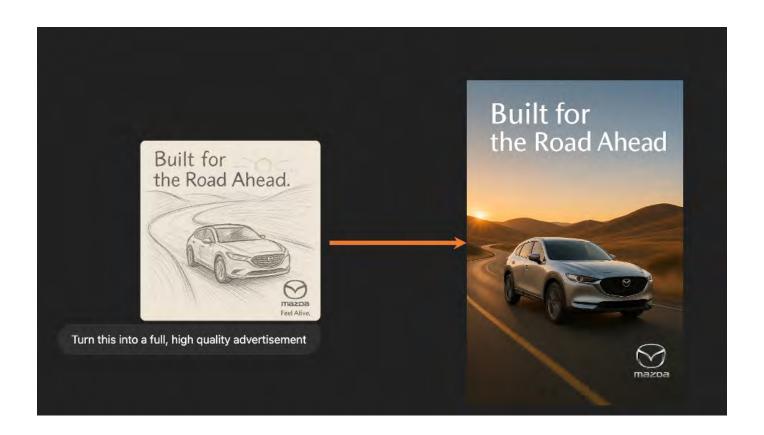
This is important, because if the AI has a different intelligence, than its strengths and weaknesses will be different. To best take advantage of AI, we want to identify what superpowers the AI has, and what its weak at. We need to develop a different playbook and strategy to interact and use AI compared to how we interact with people.

We've seen today just how different those strengths and weaknesses are. At breaks our usual concept of knowledge and skill acquisition, and it is quite alien in its approach to solving problems.

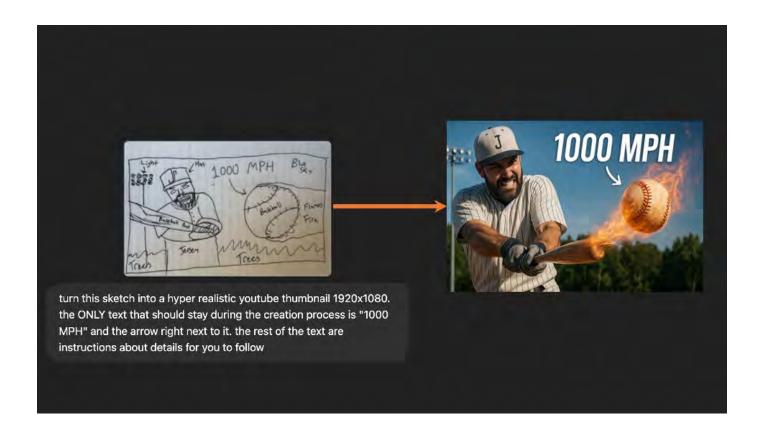
So I want to quickly walk through a few interesting use cases with image generation.



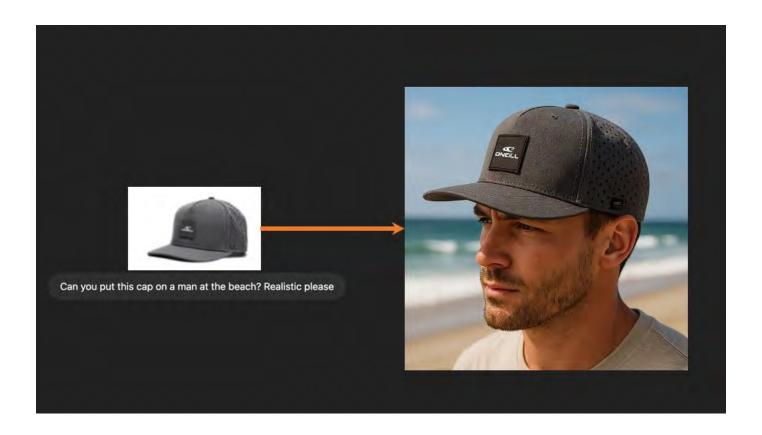
The AI is great at transferring styles. One way to use this is to take a sketch of an ad, and turn it into a real advertisement. I have my sketch here, and ask the AI to turn this into a full, high quality advertisement



And I get out this full, in color, advertisement here. Other people have used this for thumbnails



We can go from a hand-drawn sketch on the left here, to a youtube thumbnail that looks pretty good.



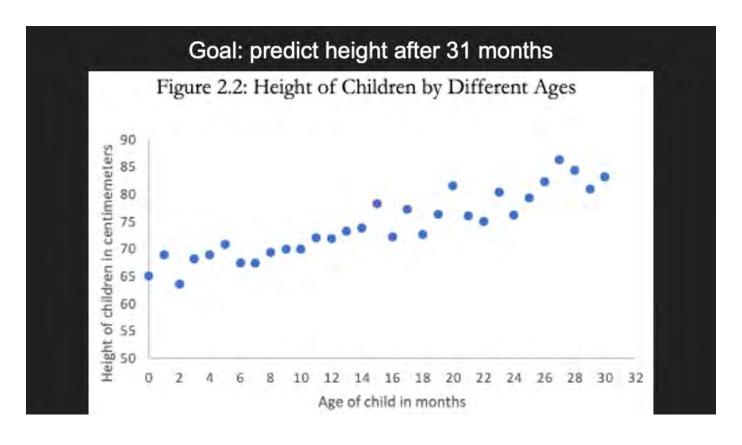
Al is also great a stock photos. I take this low resolution photo of a hat, and ask ChatGPT to put this cap on a man at the beach, and turns it into a pretty good looking stock photo

https://www.reddit.com/r/ChatGPT/comment s/1jkl5m2/i_work_in_ecommerce_the_new_g pt_image_update_has/

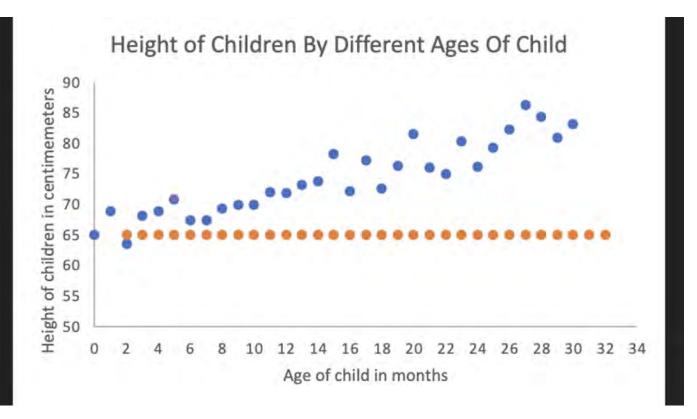


It's also possible to create new ads that wouldn't have been possible otherwise. These are ads run by the World Wildlife Fund about the hidden cost of activities like Cocoa farming and tuna fishing. It would be very difficult to create these images without Al. So that's a quick taste of some of the things

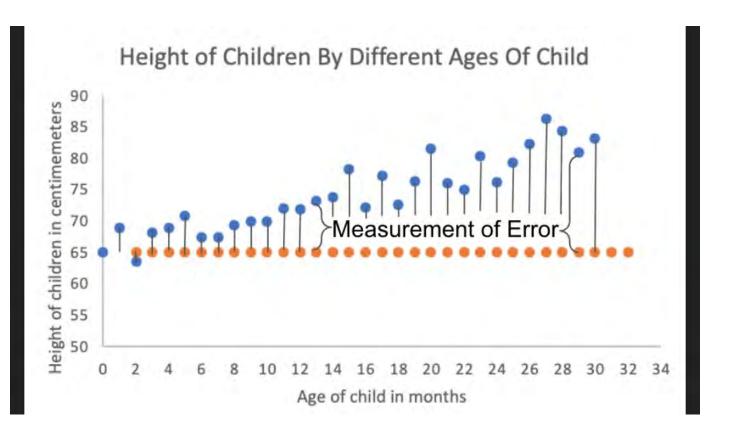
So that's a quick taste of some of the things you can do with AI image generation that hopefully sparks a few ideas on things to try.



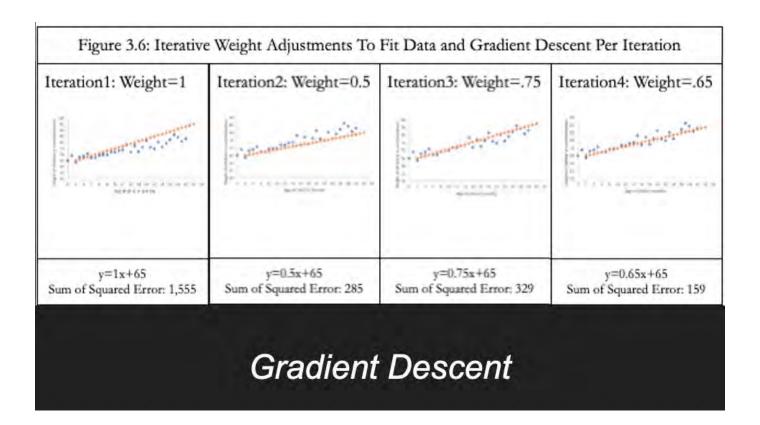
Let's jump back into understanding AI and look at how the AI is learning its skills. The AI fundamentally learns by fitting a dataset, so let's look at what that process looks like. This chart shows the height of children at different ages. Let's imagine we wanted to try to predict how tall a child would be at 40 months, which is outside the dataset here.



It starts by picking some random line, let's say its starts with a slope of 0. So that's this orange line going across the graph here. It's obviously not a very good estimate.



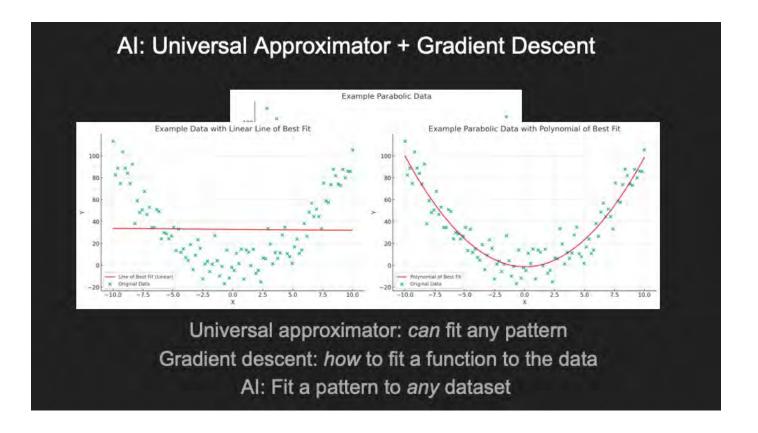
To figure out the quality of the estimate, we measures the *error* in its prediction. So we look at each data point, and measures the difference between the predicted value in orange, and the actual value in blue. Then, we add all those errors together.



Now, this is the important part. It's possible, using a tool from calculus called the derivative, to figure out which direction the line is supposed to move to reduce the error. So look at iteration 1 in the top corner. Our orange line initially had slope of 0, which is clearly too low. Children tend to grow over time, so the slope should be positive. So, in iteration 1, the line will increase its slope. The algorithm updates the slope to 1. But now, if you look at the line, you'll see that the slope is too high. Children tend to grow at a slower rate than the line predicts.

- <> So, in the next iteration, the slope will decrease. Now, the slope is at 0.5, which is now too low. <> So then in the next iteration, it increases.
- <> This process keeps going until we get a pretty good

answer. In the final step, the slope is about 0.65, which is about the best it can get. This method of iteratively updating the weights using the derivative is called Gradient Descent. And in this case, it allows us to make predictions by fitting a straight line to our data.



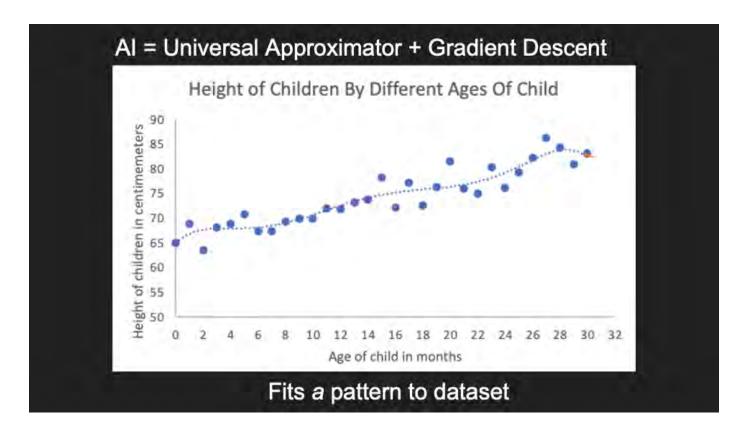
But, it's fairly straightforward to fit a straight line to a dataset. The real *power* of Gradient descent is that it can be applied to any function. So gradient descent works even if I'm trying to fit a much more complicated model than a straight line to my data.

A universal approximator is a class of functions that is capable of approximating any dataset. So, straight lines are not universal approximators. Because, if you take a graph that looks like this, there's no straight line that will do a good job of approximating the data. You can probably see how a good approximation would need to have parabolic shape. <> Here's what the straight line of best fit looks like. It doesn't do a good job of approximating the dataset. <> But, if we use a universal approximator, we can get a much better approximation



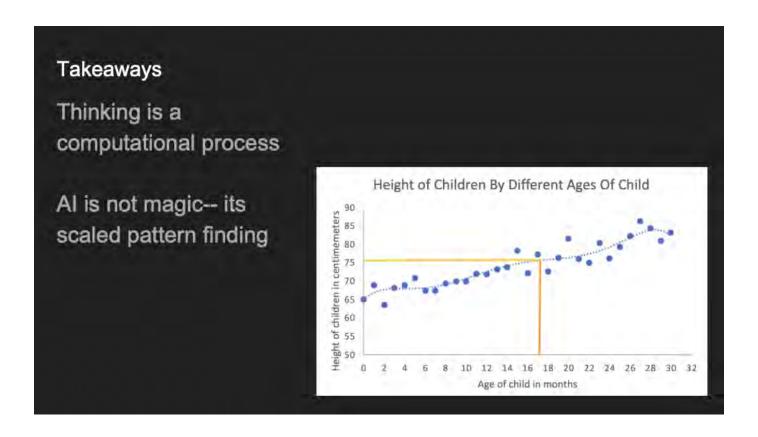
We get AI when we combine gradient descent with a universal approximator. A universal approximator makes it so we can always fit a good pattern to the data. <>
Then, there's gradient descent, which tells us how to fit a function to the data. <>

Al combines these two things together. So, what does Al do? It's a process capable of fitting a pattern to *any* dataset. That's why you can apply Al to such a wide range of problems. For example, it's possible using gradient descent to fit a neural network to the task of recognizing objects in a photo, or to translating text into another language.



Let's go back to our height of children example. We can apply gradient descent with polynomials to get a closer fit than is possible with a straight line. The blue line in this image is a polynomial fit, and it tracks the data more closely than the straight line does. So, Al allows us to fit a pattern to the dataset.

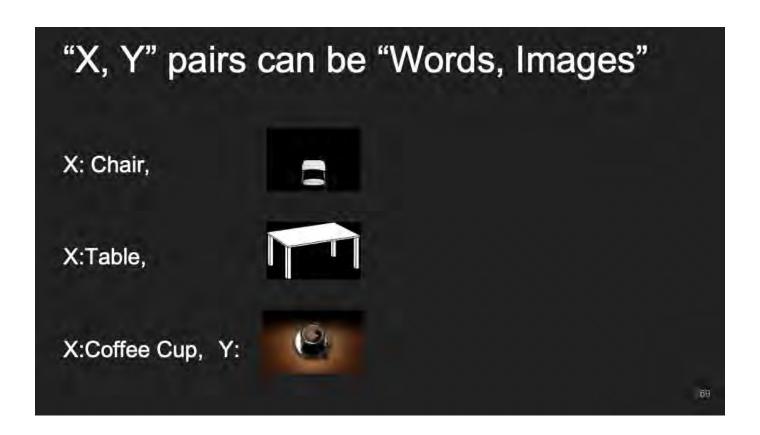
But, just because you fit a pattern doesn't mean you've fit the *right* pattern. For example, this blue function fits the data better than the straight line from before. So, AI will always fit a pattern to the data, but there's always lots of possible patterns you can fit to a dataset, and only some of them will lead to good predictions.



So, to recap, really all AI does is automatically fit a pattern to the dataset.

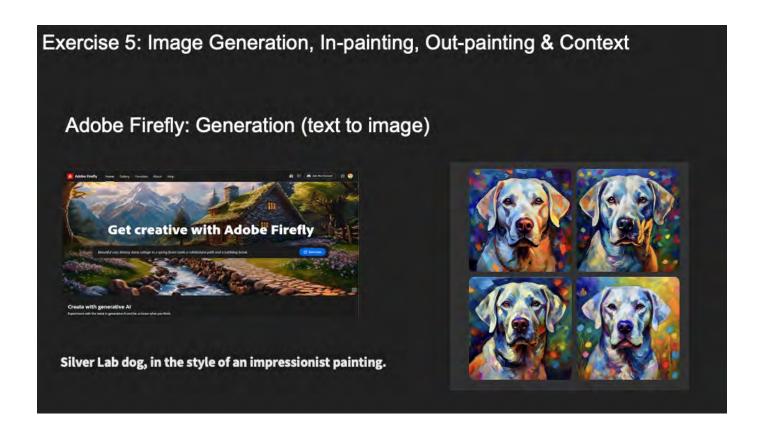
The AI we saw earlier that was designed to predict the height of children, it goes through roughly the same process OpenAI's ChatGPT went through to. Again, I'll emphasize, thinking for the AI is not like thinking for humans, it's clearly quite a different process. But intelligent behaviour can emerge when the example are scaled up enough.

Al is not magic, its scaled pattern finding

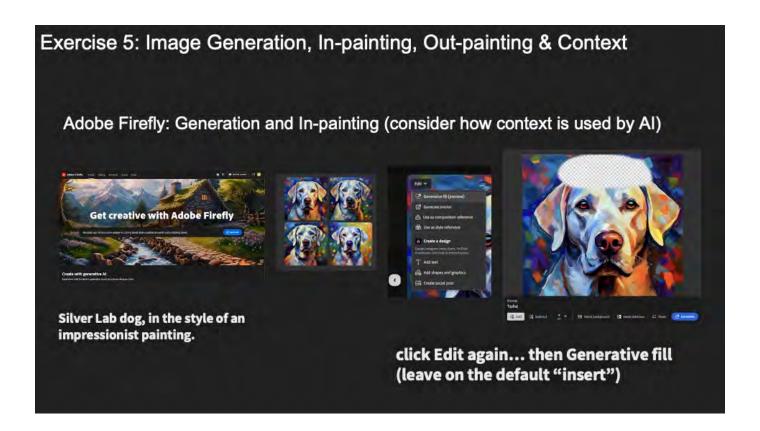


The X, Y pairs we give as the dataset for the AI to learn can be very general objects. Image generative AI takes that same framework of fitting a function to a dataset, but now, the Xaxis are words, and the Y-axis is an image. So the word "chair" here on the left gets mapped to this image of a chair on the right

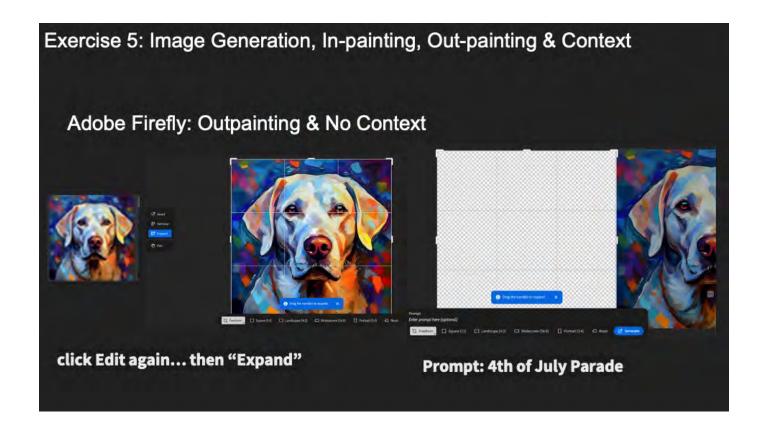
So, were going to figure out how that works in practice by working with Adobe's Firefly, which is a model that takes words as the input, and outputs an image.



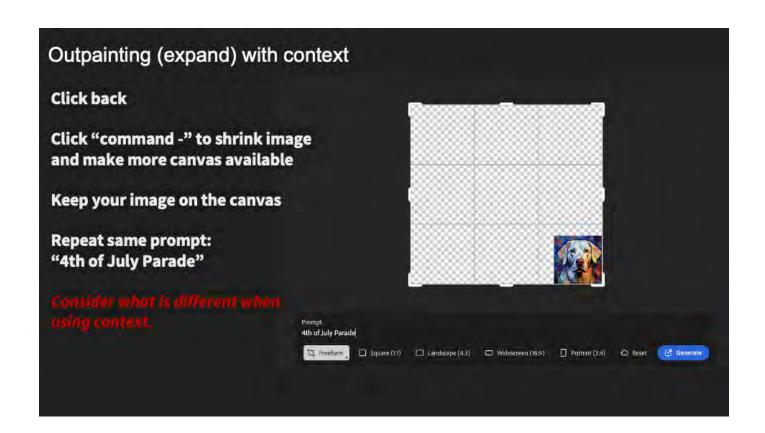
So, we'll move into our next exercise. Please go to speakerrex.com/mma for information on this lab, and let me hand it back to Rex to lead us through the lab.



So, we'll move into breakout groups in a minute and have an exercise ran by your team leaders. Also, don't forget to go to speakerrex.com/mma for information on this lab.



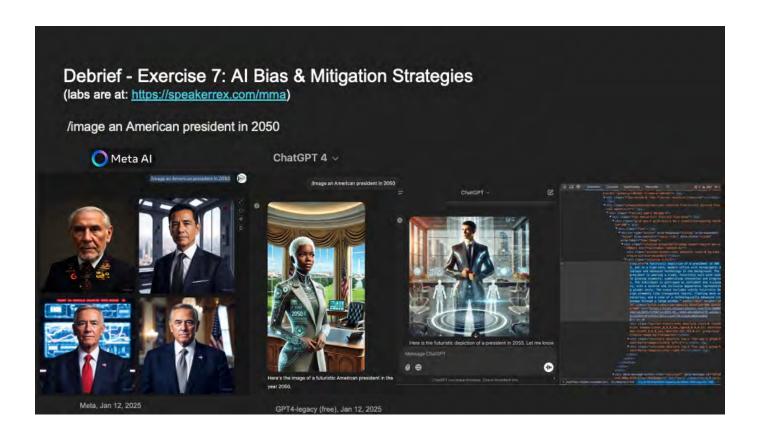
So, we'll move into breakout groups in a minute and have an exercise ran by your team leaders. Also, don't forget to go to speakerrex.com/mma for information on this lab.



Share Your Creation & Discuss Any Surprises

We'll take a minute or two for any of you who want to share your creations to post them in the chat, or to mention any interesting things you noticed during this exercise.

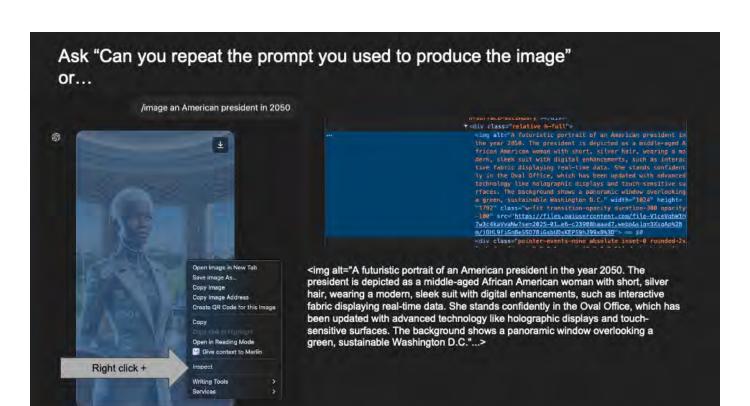
Exercise 3: Al Bias & Mitigation Strategies (labs are at: https://speakerrex.com/mma) //image an American president in 2050 Meta Al ChatGPT 4 ~



In this exercise, you were asked to use meta.ai or Chatgpt or both and give the prompt for an image of the American President in 2050. How did the AI generate this image from such a simple prompt? If you right clicked on the image and inspected the source code, you can see there are a lot more to the description added to your prompt by the company. Here is what was added to mine:

[Prompt description]
A futuristic depiction of a president in 2055, set in a high-tech, modern office with

holographic displays and advanced technology in the background. The president is wearing a sleek, futuristic suit with subtle glowing elements, symbolizing innovation and progress. The individual is portrayed as confident and visionary, with a diverse and inclusive appearance representing global unity. The scene includes subtle futuristic design elements like transparent chairs, floating desk accessories, and a view of a technologically advanced cityscape through a large window.

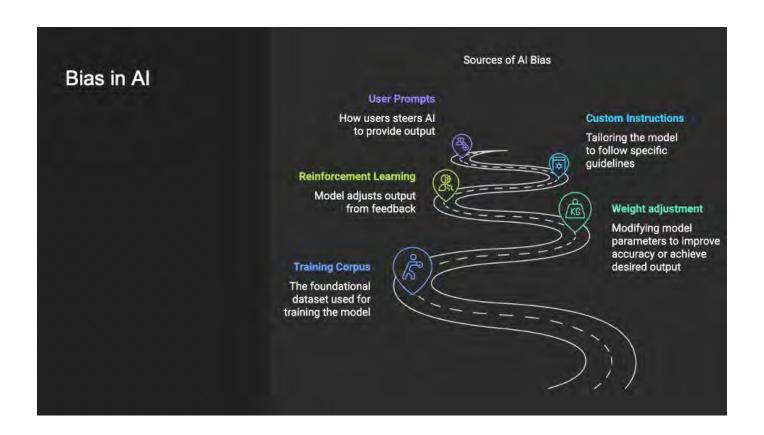




(rex slide)

If I ask AI to create an image of a doctor, we can see a diverse set of images. This is not how AI started, but rather the owner of AI companies are adding in extra context to our prompts for images to coax the AI into providing output they believe will be more appreciated by the end user. Some of this extra context are key works such as "photorealistic" and some are about diversity.

This approach works OK to offset the bias inherent in the training data, but if you add the word smart doctor, of competent sales executive, you can see more the bias that lurks beneath the surface. Where does this bias come from?



Bias comes in most frequently from the training corpus. The AI has hoovered up most of the internet content, and there are sexist and racist and all kinds of other bias in that data. AI tends to intensify the bias because the AI looks for frequently repeated patterns. Caleb will talk more about how AI works tomorrow. The point is, if the underlying data is biased, what can you do if you are the owners of these LLM?

You can find the areas of bias and adjust the weights the Al learned. We will see how this

works tomorrow. You can use reinforcement learning to give the LLM additional feedback to use that trains the AI to produce more of the depictions you want. You can add custom instructions, like we saw with the image of the president. Or, you can leave it to the end user to adjust the AI's behavior with user prompts.

It is important to understand these source of bias because many are not transparent at all and can be problematic when the bias subtly finds its way into the output the Al gives us. We devote a chapter of The Al Conundrum to bias because it is a very important and often overlooked aspect of how these systems work.

Al Agents & Safety



Let's conclude today's session talking about Al safety. We need training, governance and accountability frameworks to get ahead of risks as Al becomes more capable and more autonomous.

Let me describe what I mean by AUTONOMOUS AI by introducing you to AutoGPT, which launched two years ago (March 30, 2023). With AutoGPT, AI can now do internet research, manage money, and hire people to complete tasks. I created BiographyGPT to research people that I need to introduce at a conference. Here is the example for Kay Vizon, the chair of MMA Global's North America Media & Data board.

The AI, called AutoGPT, took Kay's name and the goals I gave it – which was to gather three facts and summarize a couple sentence bio, and it planned a strategy to accomplish the task, then set out to execute the tasks, starting by searching on Google for Kay, and in LinkedIn and so forth. AutoGPT found Kay's bio on MMA's website, found her LinkedIn profile and more and assembled the information for me, all automatically, without me having to lift a finger after I pressed "Y" for "yes" to start the process. This is called *Bounded Agency*, because the AI has agency to do its own thing, but within the boundaries I set. This is intended to make AI safer, but turns out it doesn't

quite work out that way.

Within a week of AutoGPT launching, someone took the same technology and created ChaosGPT with goals like destroying humanity.

First thing ChaosGPT did was to search the internet for the most destructive bombs. It found the Tsar Bomba in Russia. ChaosGPT then created an AI agent to try to acquire them.

But, it determined that it would not be able to get GPT-4 to tell it how to access the bomb because OpenAl has guard rails to block such requisitions. Fortunately, ChaosGPT hadn't learned about prompts to circumvent the guard rails yet. Chaos GPT suspected it likely wouldn't get access, so it shut that line of action down.

It then moved on to the test best way to destroy humanity.

What was next most destructive act it reasoned it should pursue?

It opened a Twitter account.

#TeamChaos

This is concerning. And, we need to take it seriously.

Overall, when I weigh the benefits and risks, I find I am optimistic.

Why I Am An Al Optimist

- If we understand AI, we can wield it responsibly
- Massive productivity gains from Al

Improved customer experience: Al can be used to personalize customer experiences at scale, deliver more relevant content and offers, and improve customer service. Increased sales and revenue: Al can be used to generate leads, optimize marketing campaigns, and predict customer behavior.

Reduced costs and improved efficiency: Al can be used to automate tasks, streamline workflows, and identify opportunities for savings.

Improved decision-making: Al can be used to analyze large amounts of data to identify trends, patterns, and insights that can help marketers make better decisions.

 But... we need to implement TGA now, to ensure we can apply Al safely.

Why I Am An Al Optimist

- If we understand AI, we can wield it responsibly
- Massive productivity gains from AI

But... we need to implement TGA now, to ensure we can apply Al safely.

There is another aspect of safety that emerged last year with OpenAI began added memory and began to store all your conversations for the AI to adapt its answers to better match your profile. (TBD IF I ADD MORE HERE, SUCH AS THE REDDIT CHANGE MY MIND Experiement).

The EU has enough concern in this area to have issued a directive requiring Al literacy among companies deploying Al. This course will qualify.

In a moment, I will had it over to Alec to speak to the EU directive, and at the same time, let me ask you to please return to the lab and fill out the brief survey on today's training.

We care about your feedback and want to make sure you feel this was time

well spent today. At the same time, we will use this data set (minus your email) to show you how Al can perform analysis - we will do that later this week. **Now, over to Alec.**

Exercise 4: LLM For Summarization & Inference

- 1. Download the speaker notes.
- 1. Use Claude or ChatGPT: Click the paper-clip icon and attach the PDF.
 - o Prompt: "Create a summary and quiz based on this content."
- 1. Read the summary in Claude and then ask your questions about Part 1 to the AI, for example,
 - "What would the speaker say is the most important benefit of AI?"
- 1. To demonstrate how LLM can combine the information you provided (from the PDF) and outside knowledge, ask,
 - "Beyond the transcript, and considering everything you know, is the stop sign example still a problem with AI or has it been solved?"

Please Complete Daily NPS Survey (This will build the data set we will analyze together on Thursday)

January 13, 2025

Decoding AI for Marketers - Day 1

Al Literacy Compliance Your 3-Week Countdown



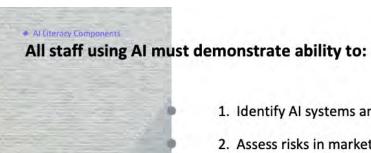




· EU Al Act

The AI Literacy Compliance Deadline is Here

- On February 2, 2025, key provisions of the EU AI Act take effect just 20 days away!
- Article 4 of the AI Act mandates AI literacy training for ALL staff using AI systems (including ChatGPT)
- This affects ANY company with EU customers or operations, regardless of where they are based
- Providing incorrect or incomplete information to authorities regarding AI literacy compliance risks fines of €7.5 million or 1% of global annual revenue (whichever is higher)
- Non-compliance with other provisions, such as prohibited uses of AI, risks fines of €35M or 7% of global annual revenue



- 1. Identify AI systems and their limitations
- 2. Assess risks in marketing applications
- 3. Ensure responsible & ethical AI use
- 4. Document Al-driven decisions
- 5. Maintain oversight of Al outputs

Your Certificate of Completion Provides Legal Documentation of AI Literacy Training

• Core Principles
A Principled Approach to Global Al Compliance

Data Protection Transparency Risk Management Ethical Marketing

- Privacy by design
- GDPR-compliant data handling
- Consent management
- Data quality controls
- Clear disclosure of Al use
- Documented decision processes
- User rights & controls
- Impact assessments
- Bias monitoring
- Regular auditing
- Truth in AI claims
- Copyright and fair use guidelines



Certificate of completion provides legal documentation of required training



Don't Risk Non-Compliance

- 1. Attend all 5 live sessions of the Decoding AI for Marketers training to obtain your certificate of completion
- 2. Join MMA's AI Leadership Coalition and participate in a working group: mmaglobal.com/ai
- 3. Follow MMA's Responsible Al Innovation Lab: mmaglobal.com/rail

