# AI/ML In Science & Society

G. Watts (UW/Seattle)

GE IGD

# Who am I?

Gordon Watts Professor of Physics <u>gwatts@uw.edu</u>

Joined UW in 1999 (!!)

**Experimental Particle Physicist** 

Interests:

- Intersection of physics, computing, and hardware
- Top Quark Physics
- Searches for Long Lived Data





Machine Learning and AI

# Why Do We Care About AI/ML In Science?



# Why Do We Care About AI/ML In Science?



**Big Data** 

### Non-Trivial Pattern Recognition

# Worth Money and Time!



This change in this plot is due to:

- Increase in data (about 50%)
- Use of more sophisticated Machine Learning technique (about 50%)

This translates into close to a year of running the Large Hadron Collider!

Electricity alone is ~90M USD Everything else is another ~250M USD

## My Personal Opinion

# You should not get a Ph.D. without knowing what ML is, where it is useful, and how to use it.

(and likely how to code one up)

(how long till this is true for an undergraduate? Not long...)

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# Introduction To Deep Learning and Machine Learning

### Artificial Intelligence

Human-like intelligent machines or programs

### Machine Learning

Al algorithms that learn from data instead of being explicitly human programmed

# It is all about the Math!

# But that is all it is about!

# What is this?

Guess where the x = 1.9 value would be!

You can guess this trend in your head!



# Straight Line With Jitter



# How can we determine *m* and *b*?

Straight-line Fit

# Analytically...

What are the values of m and b such that we minimize the distance between the line they describe and all the points?

There is an analytical function for this "simple" problem...

### Segmented Least Squares

#### Least squares.

- Foundational problem in statistic and numerical analysis.
- Given n points in the plane:  $(x_1, y_1), (x_2, y_2), \ldots, (x_n, y_n)$ .
- Find a line y = ax + b that minimizes the sum of the squared error:





Solution. Calculus  $\Rightarrow$  min error is achieved when

$$a = \frac{n \sum_{i} x_{i} y_{i} - (\sum_{i} x_{i}) (\sum_{i} y_{i})}{n \sum_{i} x_{i}^{2} - (\sum_{i} x_{i})^{2}}, \quad b = \frac{\sum_{i} y_{i} - a \sum_{i} x_{i}}{n}$$

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# The Sum Of The Distances...

Plot all the sums of distances for all possible m and b.

We want the minimum value of this plot!



# The Sum Of The Distances...



# Gradient Decent

Each *epoch* is a small step down that surface!

But only small steps to as not to "miss" the lowest point.



Recall the target value is 2.0!

# Gradient Decent



# But that is a straight line...

What is this thing, a Neural Network?

# What Is A Neural Network?

### A mathematical Function!

Modeled after a biological neuron...

This function has 2 inputs and 1 output... And has 7 "internal" functions Numbers flow left-to-right!



# Nuron



(biology)

# Nuron?



# Ok... skipping many many papers...



If I combine enough of the neurons I can model *any* function I want



I can use my gradient decent from before to tune these functions to model whatever I want

# History...

### The History of Deep Learning

Below is a timeline of some of the important events in the history of Deep Learning. They are by no means exhaustive, but they are chosen as they help illustrate how Deep Learning arrived to where it is today.

#### **Deep Learning Timeline:**

- 1958: Perceptron (Simple version of deep learning) Discovered
- 1979: Convolutional Neural Network Invented (Today dominates image recognition)
- 1982: Recurrent Neural Network Invented (Sequence processing used for <u>NLP</u>)
- 📲 👬 🏶 AI WINTER 🏶
- 2009: ImageNet Introduced

My Dad used simple NN's!

ImageNet (2011)



This is the start of the Deep Learning Revolution

# Sophistication...

Me: How can I quickly explain deep learning to a lay audience?

Certainly! Let's break down **deep learning** in simple terms for a lay audience:

### 1.What is Deep Learning?

•Deep learning is a type of artificial intelligence (AI) that mimics the way our brains work.

•It involves training **neural networks**, which are like interconnected virtual brain cells, to learn patterns from data.

### 2.Neural Networks: The Building Blocks

•Imagine a network of interconnected nodes (like neurons in our brain).

•Each node processes information and passes it to others.

•These layers of nodes form a **neural network**.

### **3.Deep Neural Networks**

• "Deep" refers to having many layers in the network.

•Each layer learns progressively complex features.

•The final layer makes predictions (e.g., recognizing cats in images).

### 4. Training the Network

•We feed the neural network lots of labeled examples (e.g., cat images with labels "cat").

•It adjusts its internal connections (weights) to get better at making predictions.

### **5.**Applications of Deep Learning

•Image Recognition: Identifying objects in photos.

•Natural Language Processing: Understanding and generating human language.

•Recommendation Systems: Suggesting movies, products, etc.

#### 6.Why "Deep"?

•Deeper networks can learn intricate patterns.

•They're like solving puzzles with more layers.

Remember, deep learning is like teaching a computer to learn from extamples, stills as we learn from experience!

Bing CoPilot (ChatGPT 4 Turbo)

# Deep Learning In Science

# Is that Signal?

This was late 1980's

- The tau lepton was of intense interest
  - Still is...
- But we had no large samples of it (then)...

A group, including my Dad, decided to try simple Neural Networks to pick them out of the CDF detector!



# Finding Tau's



Meeehhhh... kind-a worked?

Simpler techniques worked better...

# My last published paper...



# It works well!



Most of my signal is here!

# What is deep learning/Neural Networks/ML?

It is function fitting!

# **Function Fitting**



### This is my (crazy) function



This is one event of my data (Argument to the function)

### This is the result of the function (on many pieces of data)



# "Fitting" is called "Training"



These 100K parameters take about 2 hours to train on a nVidia 2–generation-old-high-end card.

Modern Large Language Models take months to train on many many of the latest high-end cards! (~billion bucks?) (~Trillion parameters?)

# How Interesting Are Those Function?



### **How Interesting Are Those Function?**



# This is where Modern Research Is

Training Is Expensive Data is limited

Can you purpose design a function to accomplish the task at hand?

This would short-cut your training!

YES

### This is the future of LLM's...

Attention Is All You Need			
Ashish Vaswani* Google Brain avaswani@google.com	<b>Noam Shazeer</b> * Google Brain noam@google.com	Niki Parmar* Google Research nikip@google.com	<b>Jakob Uszkoreit</b> * Google Research usz@google.com
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The dominant seque convolutional neur performing models mechanism. We p based solely on atte entirely. Experime be superior in quali- less time to train to-German translat ensembles, by over our model establish training for 3.5 day best models from tl other tasks by appl large and limited tr	tence transduction model al networks that included also connect the encoor ropose a new simple n ntion mechanisms, dispe nets on two machine tri ity while being more pan Our model achieves 28 ion task, improving ov 2 BLEU. On the WMT 2 is a new single-model st srs on eight GPUs, a smu e literature. We show th ying it successfully to E aining data.	els are based on compl e an encoder and a deco der and decoder througed twork architecture, th nsing with recurrence a unslation tasks show th rallelizable and requirin .4 BLEU on the WMT or the existing best res 014 English-to-French BLEU sc all fraction of the trainin tat the Transformer gen nglish constituency pa	ex recurrent or oder. The best than attention e Transformer, d convolutions nese models to g significantly 2014 English- ults, including translation task, reo f 41.8 after ng costs of the eralizes well to rsing both with
*Equal contribution. Listing the effort to evaluate this idea has been crucially involved in ev attention and the parameter-free detail. Niki designed, implement tensor2tensor. Llion also experi efficient inference and visualiza implementine tensor2tensor. ren	order is random. Jakob pro Ashish, with Illia, designed ery aspect of this work. No e position representation a tted, tuned and evaluated c mented with novel model v tions. Lukasz and Aidan sp lacine our earlier codebase	posed replacing RNNs wi d and implemented the fir am proposed scaled dot-pin d became the other person ountless model variants ir ariants, was responsible f ent countless long days de: areatly improving results	th self-attention and started st Transformer models and oduct attention, multi-head on involved in nearly every our original codebase and or our initial codebase, and signing various parts of and and massively accelerating

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our research. <sup>†</sup>Work performed while at Google Brain.

Aug 2023

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[cs.CL]

arXiv:1706.03762v7

<sup>‡</sup>Work performed while at Google Brann.

31st Conference on Neural Information Processing Systems (NIPS 2017), Long Beach, CA, USA.

# Large Language Models

# MOAR DATA!!!

For this analysis producing data was expensive

- Time
- Compute time

More data would likely have meant better separation!



# Lawsuits!

A friend of mine who works on large language models at Meta:

"No LLM can be trained without using Copyrighted data. There just isn't enough high quality data to train on."

Right now: No one is making money on LLM's/AI

### The Times Sues OpenAI and Microsoft Over A.I. Use of Copyrighted Work

Millions of articles from The New York Times were used to train chatbots that now compete with it, the lawsuit said.

🛱 Share full article 🔊 🗍 🖓 1.3K



A lawsuit by The New York Times could test the emerging legal contours of generative A.I. technologies. Sasha Maslov for The New York Times



y Michael M. Grynbaum and Ryan Mac

Dec. 27, 2023 Leer en españo

The New York Times sued OpenAI and Microsoft for copyright infringement on Wednesday, opening a new front in the increasingly intense legal battle over the unauthorized use of published work to train artificial intelligence technologies.

# Can LLM's Help In Science?

They already are:

• Writing first draft of code to process our data

- General lookup questions
  - But you sort-of have to know the answers

• Helping to smooth our poor English...

...as assistants

😁 GitHub Copilot

台 🖓 …

Used 1 reference >

You can use the plot\_surface function from the Axes3D module in matplotlib to create a 3D plot. Here's how you can modify your code to create a 3D plot:

from mpl\_toolkits.mplot3d import A>

# Assuming x and y are already defi
b\_range = jnp.linspace(0, 4, 20)
m\_range = jnp.linspace(2, 4, 20)

def loss\_fn(x, y, m, b):
 return jnp.mean((y - m\*x - b)\*\*

# Lets make a grid plot of the loss
losses = jnp.array([[loss\_fn(x, y,

# Create a meshgrid for m and b
m\_mesh, b\_mesh = jnp.meshgrid(m\_rar

#### Create a 3D plot

fig = plt.figure()
ax = fig.add\_subplot(111, projectic
ax.plot\_surface(b\_mesh, m\_mesh, los

ax.set\_xlabel('b') ax.set\_ylabel('m') ax.set\_zlabel('loss')

plt.show()

This will create a 3D plot with **b** and **m** as the x and y axes, and loss as the z axis.

# But much more imaginative uses are coming...

- 1. Physicist must write talk night before conference presentation.
- 2. Physicist is bored...
- 3. Writes a prompt and feeds it and a conference abstract to a LLM
- 4. Discovers it can tell them if they are interested in the conference talk!
- 5. Writes a python script to feed all the conference talks to the LLM and rank them for interest!

This was 3 hours of coding, starting from not really knowing how to do this...

### From github...

- data = load\_indico\_json(event\_url)
- prompt = """I am an expert in experimental particle physics as well as computing for
- particle physics. You are my expert AI assistant who is well versed in particle physics
- and particle physics computing. My interests are in the following areas:
  - 1. Hidden Sector Physics
- 2. Long Lived Particles (Exotics or RPV SUSY)
- 3. Analysis techniques and methods and frameworks, particularly those based around python or ROOT's DataFrame (RDF)
- 4. Machine Learning and AI for particle physics
- 5. Distributed computing for analysis (e.g. Dask, Spark, etc)
- 6. Data Preservation and FAIR principles
- 7. Differentiable Programming
- I'm not very interested in:
  - 1. Quantum Computing
  - 2. Lattice QCD
  - 3. Neutrino Physics

Please summarize this conference abstract so I can quickly judge the abstract and if I want to see the talk it represents.

Your reply should have the following format:

summary: <One line, terse, summary of the abstract that does not repeat the title. It should add extra information beyond experiment: <If you can guess the experiment this abstract is associated with (e.g. ATLAS, CMS, LHCb, etc), place it here keywords: <comma separated list of keywords that match my interest list above. If you can't find any, leave blank.> interest: <If you can guess how interested I am from above, put "low", "medium", or "high" here. Otherwise blank.>

Here is the talk title and Abstract:"""

# LLM for Particle Physics?

Question: Do [large language] models actually understand what they are saying?

"Clearly, some people believe they do, and some people believe they are just stochastic parrots." Geoff Hinton

Wait... what is a LLM? What is it actually trained to do!???

# Training Is... Simple...

Artificial intelligence seems more powerful than ever, with chatbots like Bard and ChatGPT capable of producing uncannily humanlike text.

Source Text (copyrighted?)

Artificial intelligence seems more powerful than ever, with chatbots like Bard and ChatGPT capable of producing uncannily humanlike

Training Input

**Expected Output** 

text

They are trained to predict the next word....

That is all

# Stochastic Parrot

There is some randomness injected into the process

It was trained to complete a sentence! Sheesh!

# For Science!



We could train a NN to know how a detector works

- Train on all our Simulated Data or even Real Data
- It would learn about objects
- How objects occur together

Then we could tune it to answer specific questions...

This effort is just starting (funding!!!)

# Will It Work?

I can't wait to find out!



# But... we have a way to go...

### Ask Copilot (ChatGPT 4 turbo?)...

Could you tell me if this looks like it is a straigth line, and then estimate the slope and intercept?



Certainly! The scatter plot in the image appears to form a **straight ascending line**. Let's estimate the slope and intercept:

- Slope (m): The rise from left to right suggests a positive slope.
- Intercept (b): The line intersects the y-axis slightly above 0.

Given this, we can approximate the equation of the line as:

#### y = mx + b

However, without specific data points or coordinates, I cannot provide precise values for the slope and intercept. If you have the data, I'd be happy to calculate them for you! I had to convince it to try – as if it knew it couldn't do a good job...

#### Copilot

Certainly! Let's analyze the scatter plot to estimate the slope and intercept.

- Slope (m): The line rises from left to right, indicating a positive slope. To estimate the slope, we can visually approximate the change in y-values divided by the change in x-values. Let's select two points on the line:
  - Point A: Approximately (-2, -6)
  - Point B: Approximately (2, 8)

The change in y-values is (8 - (-6) = 14), and the change in x-values is (2 - (-2) = 4). Therefore, the slope (m) can be estimated as: [ m \approx  $\frac{14}{4} = 3.5$ ]

 Intercept (b): The line intersects the y-axis slightly above 0. Let's estimate the y-intercept visually. It appears to be around (y = 4).

Therefore, the equation of the line can be approximated as: [y = 3.5x + 4] Keep in mind that the inspection. If you have y = 3x + 2exact slope and intercept.

3 ♀ □ ↓) ● 2 of 30

### Machine Learning sucks! (compared to humans and animals)

- Supervised learning (SL) requires large numbers of labeled samples.
- Reinforcement learning (RL) requires insane amounts of trials.
- Self-Supervised Learning (SSL) works great but...
  - Generative prediction only works for text and other discrete modalities

### Animals and humans:

- Can learn new tasks very quickly.
- Understand how the world works
- Can reason an plan
- Humans and animals have common sense
- There behavior is driven by objectives (drives)

### Future AI Assistants need Human-Level AI

- Al assistants will require (super-)human-level intelligence
  - Like having a staff of smart "people" working for us

### But, we are nowhere near human-level AI today

- Any 17 year-old can learn to drive in 20 hours of training
- Any 10 year-old can learn to clear the dinner table in one shot
- Any house cat can plan complex actions

### What are we missing?

- Learning how to world works (not just from text)
- World models. Common sense
- Memory, Reasoning, Hierarchical Planning

# Fin!

- The only AI game in town is Machine Learning
- Machine Learning is just a function fit
- It has had a huge impact on Science
  - Which started about a decade ago
- Is now having a huge impact on Society for the same reasons

