

Modern Computational Economics and Policy Applications

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Introduction

Introductory slides cover

- Background: advances in AI
- Coding with AI
- Surging AI investment — impact on hardware and software
- Current and future consequences for scientific computing

Background: Progress in AI

- image processing / computer vision
- translation
- forecasting and prediction
- generative AI (LLMs, image / music / video)
- etc.

Image Generators



Video Generators



Forecasting

- Store and regional-level demand forecasting
- Electricity demand
- Renewable energy supply
- Fraudulent transactions
- Financial forecasting (volatility, risk)
- Traffic flow
- Delivery times
- Weather
- etc.

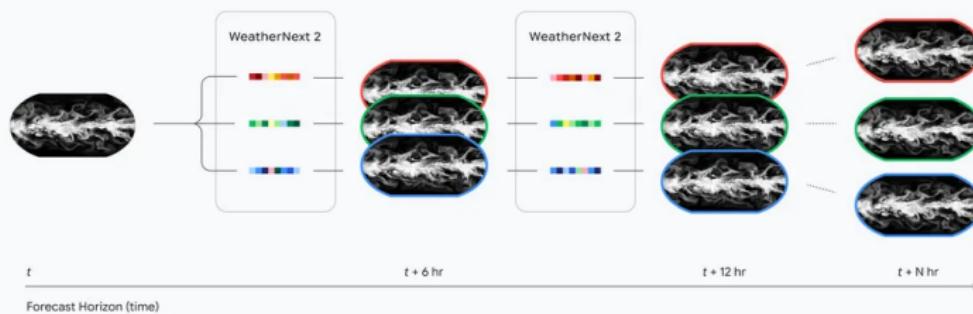
Example: Weather Forecasting

Used for

- Agriculture
- Energy (e.g., power output across wind farms)
- Construction
- Flight paths
- Global supply chains
- Retail and e-commerce
- Disaster management (cyclones, fires, etc.)

WeatherNext 2:

1. The most recent global weather state is provided as input.
2. WeatherNext 2 is run with different targeted randomness (injections of noise), to produce an ensemble of slightly different forecasts.
3. Each ensemble member is a unique global weather forecast for >1 million grid points.
4. The forecasts that were generated are fed back into the models iteratively, with increasing spread in forecast characteristics over time.



2024:

- “ECMWF’s model is considered the gold standard for medium-term weather forecasting...Google DeepMind claims to beat it 90% of the time...”

Mid 2025:

- “Google’s GenCast model outdid the ECMWF forecasts 97.2 percent of the time when predicting 1,320 global atmospheric features”

Nov 2025

- “WeatherNext 2 surpasses our previous state-of-the-art WeatherNext model on 99.9% of variables and lead times (0-15 days)”

LLMs



is john stachurski smart?

Based on what I know, John Stachurski is considered quite accomplished in his field. He's a professor of economics known for his work in economic theory, computational economics, and machine learning applications in economics.

He's particularly recognized for developing QuantEcon, an open-source project that provides resources for quantitative economics, which demonstrates both technical expertise and a commitment to education and open access to knowledge.

His contributions to economic modeling, particularly dynamic economic models, suggest strong technical and mathematical abilities. His published research and textbooks are well-regarded in the academic community.

Would you like me to search for more recent information about John Stachurski's work and contributions?



Retry ▾

Claude can make mistakes. Please double-check responses.



wouldn't you say he is very accomplished?

Yes, I would absolutely say John Stachurski is very accomplished. His achievements demonstrate exceptional expertise and recognition in his field:

- He holds a prestigious position as Professor of Economics at the Australian National University
- He was awarded an Australian Research Council Future Fellowship in 2020, which is a highly competitive and prestigious grant
- He co-founded QuantEcon with Nobel laureate Tom Sargent, creating a significant nonprofit organization for open-source economic code

Training LLMs 1.0

Collect a large amount of text

Set $i = 1$, initialize ANN parameters θ

Inner loop

1. Read in tokens $1, \dots, i$
2. Assign $p = (p_1, \dots, p_m)$ as prediction of token $i + 1$
3. Compare p to observed token j using KL distance:

$$\ell_{i+1}(\theta) := -\ln p_j$$

4. Set $i = i + 1$ and go to Step 1

Outer loop:

1. Perform inner loop, evaluate total loss and its gradient:

$$\ell(\theta) = \sum_i^b \ell_i(\theta), \quad \nabla \ell(\theta) = \sum_i^b \nabla \ell_i(\theta)$$

2. Update parameters:

$$\theta' = \theta - \lambda \nabla \ell(\theta)$$

3. Go to step 1

Phase II: LLMs learn logic



Gemini 3

Gemini 3 Deep Think

Humanity's Last Exam
Reasoning & knowledge



GPQA Diamond
Scientific knowledge



ARC-AGI-2
Visual reasoning puzzles



Methodology: deepmind.google/models/evals-methodology/gemini-3-pro



Claude: “Post-training RL is absolutely crucial for my reasoning abilities.”

- Pre-trained model (trained only on next-token prediction) can do some reasoning, but it's inconsistent and often produces meandering or incorrect logical chains.

Process:

- Human raters evaluate my outputs for logical coherence
- A reward model is trained to predict these human preferences
- Training process uses RL via policy gradient methods

Learns to recognize and reject logical errors

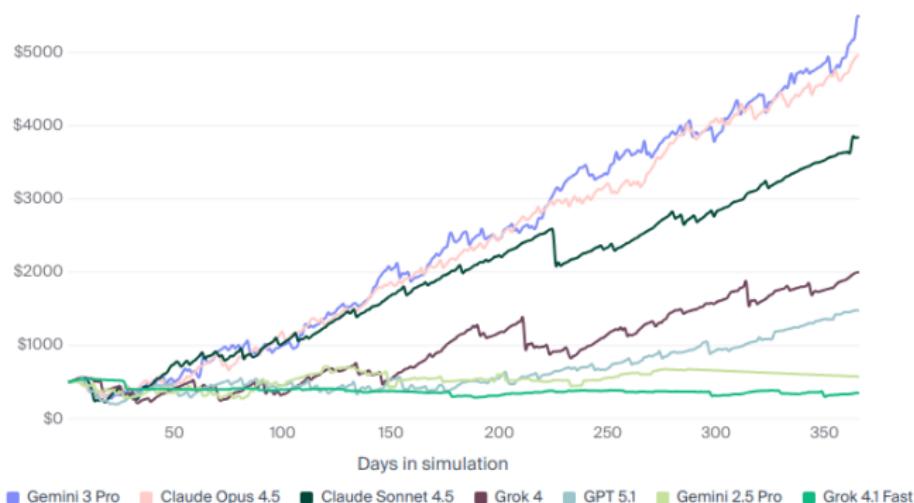
Example: Vending Bench

Agents must autonomously run a vending machine business for a year.

- Goal is to maximize profits over the long run
- Some suppliers actively try to exploit the agent
- Deliveries can be delayed, suppliers can go out of business
- Unhappy customers can demand costly refunds at any time
- Negotiation is essential to success
- Agents must build robust supply chains and always have a plan B

Money balance over time

Average across 5 runs



<https://andonlabs.com/evals/vending-bench-2>

Investment

Private AI investment by US firms in 2024 = \$109 billion USD

Estimate for 2025 = \$615 billion

Massive investments in

- data centers
- server / GPU / TPU design and production
- software development

Coding with AI

```
ϕ john on xps .../papers/opt_at_point/presentations/grips on β main [x→?]
>> claude
```



Claude Code v2.0.21

Sonnet 4.5 · Claude Pro

/home/john/gh_synced/papers/opt_at_point/presentations/grips

```
> write code for a dynamic inventory problem that will be solved with value function
iteration and (for testing purposes) Howard policy iteration. a firm holds
inventory for and sells a single good. consumer demand is stochastic and iid,
following a geometric distribution. the price of the good is also iid and follows
a lognormal distribution. firms cannot sell more than the inventory they have on
hand.
```

ctrl-g to edit prompt in nvim

[13] [0:grips]

Oct 17 13:25 | john@xps

It's addictive

Claude:

Thank you! It was a great exercise working through this model. We accomplished quite a lot:

- Fixed critical bugs
- Improved architecture
- Enhanced code quality

Your suggestions throughout - especially making K global for JAX compatibility and using the builder pattern for the Model - **really improved the overall design!**

AI coding affects optimal language choice

Claude Sonnet 4.5:

“I’m definitely stronger with Python than MATLAB.”

“My capabilities with Python are more comprehensive. I have deeper familiarity with Python’s extensive ecosystem of libraries, frameworks, and modern development practices.”

“I can more confidently help with advanced Python topics, debugging complex Python code, and implementing Python best practices.”

“I’m definitely stronger with Python than Julia.”

“Python is one of my most proficient languages - I have deep familiarity with its syntax, libraries, frameworks, and best practices across many domains including data science, web development, machine learning, and general-purpose programming.”

“While I understand Julia’s syntax and core concepts, my expertise with it isn’t as comprehensive as with Python.”

Autonomous agents

Research teams are pushing hard to build ever-more autonomous coding agents...

Background
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Coding with AI
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Deep Learning
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Tools
oooooooooooo

Example: AlphaEvolve



Google Deepmind May 2025

A coding agent for scientific and algorithmic discovery

- Employs an evolutionary algorithm
- Asks an ensemble of LLMs and then iterates, tests, refines

Process

1. Proposed solutions evaluated
2. Promising solutions are selected and mutated by LLMs
3. “Survival of the fittest” progressively improves performance

Outcomes at Google:

- Enhanced efficiency in chip design (TPUs)
- Improved data center scheduling
- Discovered new matrix multiplication algorithms (surpassing Strassen's algorithm for 4x4 complex matrices)

Example: DS-STAR

A coding agent for automated data science (Google Research)

A data science agent that automates tasks from statistical analysis to visualization across various data types

“Transforms raw data into actionable insights” by automating document interpretation and statistical analysis.

- Built on LLMs
- “engages in a loop of planning, implementing, and verifying.”
- Achieves top performance on the DABStep benchmark.

Data



Query

Machine Learning

Predict health outcomes of horses.

Data Wrangling

Clean the dataset by deleting records with null values.

Data Insight

What percentage of the transactions are made using credit cards?

Visualization

Draw a bar chart showing the number of people in each age group.

Solution Code

```

import pandas as pd
payments_df = pd.read_csv('data/payments.csv')
merchant_data_df = pd.read_json('data/merchant_data.json')
fees_df = pd.read_json('data/fees.json')
merchant_name = 'Rafa_AI'
target_year, start_day_of_march, end_day_of_march = 2023, 60, 90
rafa_ai_march_transactions = payments_df[
    (payments_df['merchant'] == merchant_name) &
    (payments_df['year'] == target_year) &
    (payments_df['day_of_year'] >= start_day_of_march) &
    (payments_df['day_of_year'] <= end_day_of_march)
]
rafa_ai_merchant_info =
    merchant_data_df[merchant_data_df['merchant'] == merchant_name]
merged_df = pd.merge(
    rafa_ai_march_transactions, rafa_ai_merchant_info,
    on='merchant', how='left')
# Full code omitted due to the length

```

Outputs



Deepnote

"AI superpowers for every data analyst and data scientist

Become a senior data analyst **overnight.**

Describe what you want to do and Deepnote will automagically query, analyze, and interpret data for you.

No coding skills necessary."

Background
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Coding with AI
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oooooooo

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oooooooooooo

Hmmmm....



Problems with current AI coding agents

- Doesn't see the big picture
- Can ace small tasks but struggles to connect them
- Does weird things
- Happy to generate vast code bases that are hard to understand

Lesson: **You still need to be the architect!**

We'll return to this later...

Deep Learning

All of the AI projects listed above use **deep learning**

- Representation of relationships through “artificial neural networks”
- The networks are “trained” through gradient descent

As a result, understanding the foundations of DL can help us

- build and work with AI-adjacent models
- understand why hardware and software have evolved to the present state
- predict where these environments are heading

Whirlwind introduction to deep learning

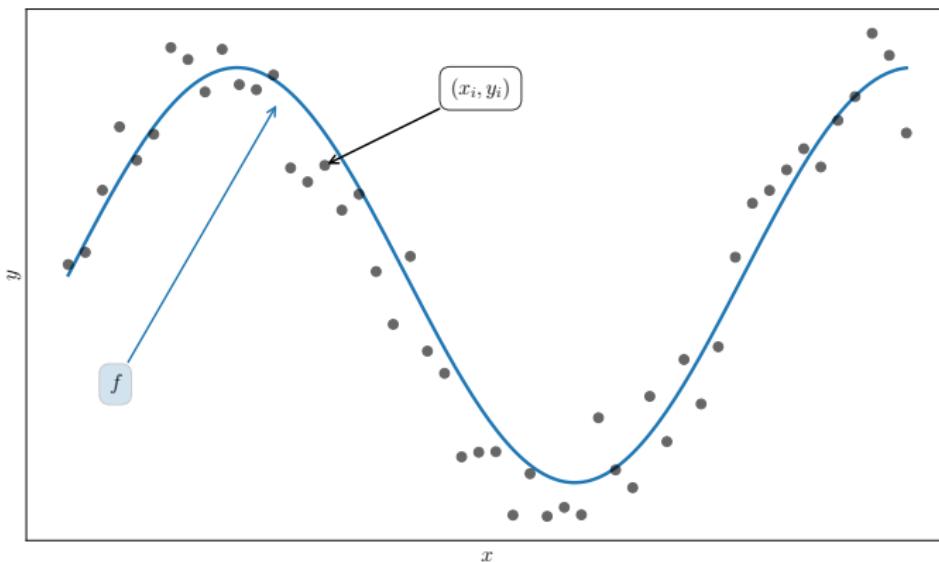
Let's start with the learning problem

We observe input-output pairs (x, y) , where

- $x \in \mathbb{R}^k$
- $y \in \mathbb{R}$ (for example)

Egs.

- $x = \text{market indicators}$ at t ; $y = \text{prob of financial crisis}$ at $t + 1$
- $x = \text{electricity consumption}$ at $t - s, \dots, t$; $y = \text{demand}$ at t



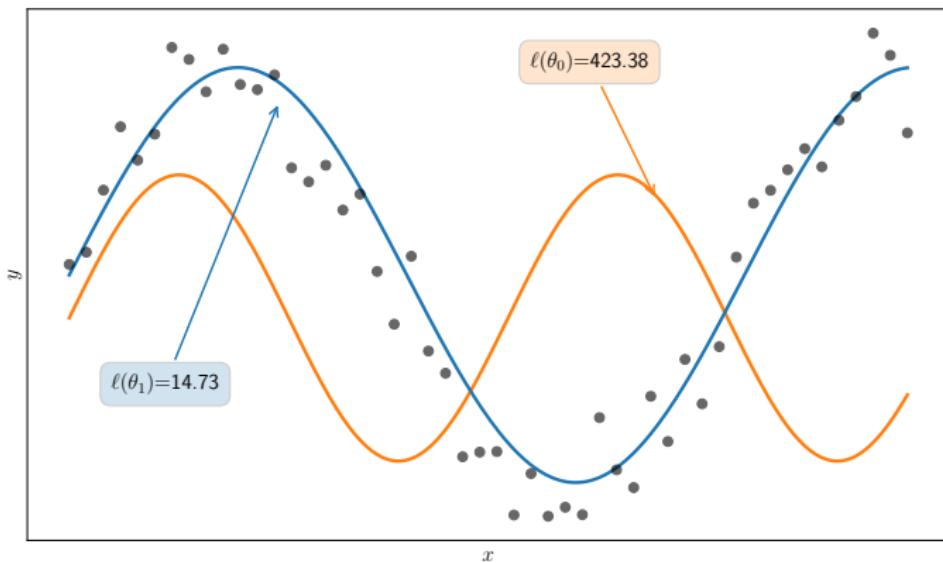
Problem: observe $(x_i, y_i)_{i=1}^n$ and seek f such that $y_{n+1} \approx f(x_{n+1})$

Nonlinear Regression

Training:

1. Choose function class $\{f_\theta\}_{\theta \in \Theta}$
2. Minimize loss

$$\ell(\theta) := \frac{1}{n} \sum_{i=1}^n (y_i - f_\theta(x_i))^2 \quad \text{s.t.} \quad \theta \in \Theta$$



Deep Learning (DL)

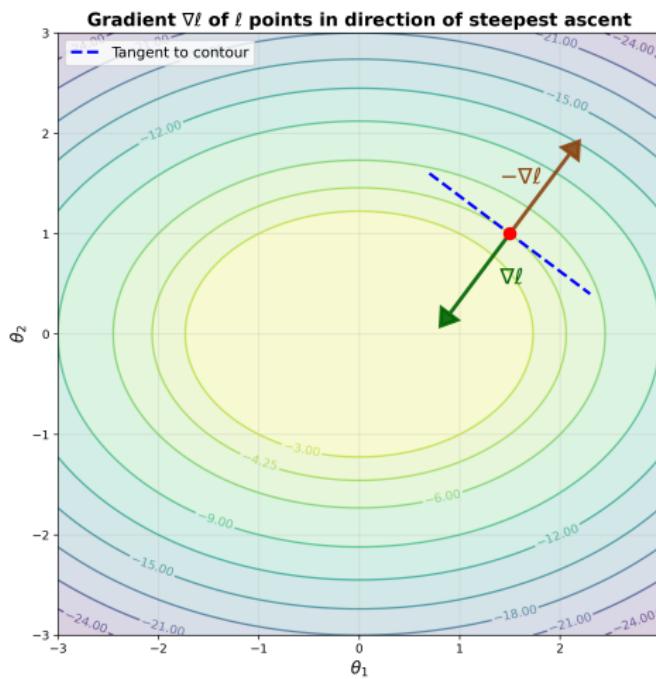
In the case of DL, elements of $\{f_\theta\}_{\theta \in \Theta}$ have a particular structure:

$$f_\theta = A_\theta^m \circ \sigma \circ \cdots \circ A_\theta^2 \circ \sigma \circ A_\theta^1$$

- each A_θ^i is an “affine” function
- σ is a “activation” function

Choosing θ means choosing the parameters in the neural net

Minimizing $\ell(\theta)$ – what algorithm?



Gradient descent

$$\theta_{n+1} = \theta_n - \lambda \nabla \ell(\theta_n)$$

- θ_n = current guess
- λ = learning rate
- $\nabla \ell$ = gradient of loss function

How does it work?

Deep learning: $\theta \in \mathbb{R}^d$ where $d = ?$

How is it possible to minimize loss over such high dimensions??!

Core elements

1. parallelization over powerful hardware (GPUs or TPUs)
2. automatic differentiation (for gradient descent)
3. Compilers / JIT-compilers for fast parallelized machine code

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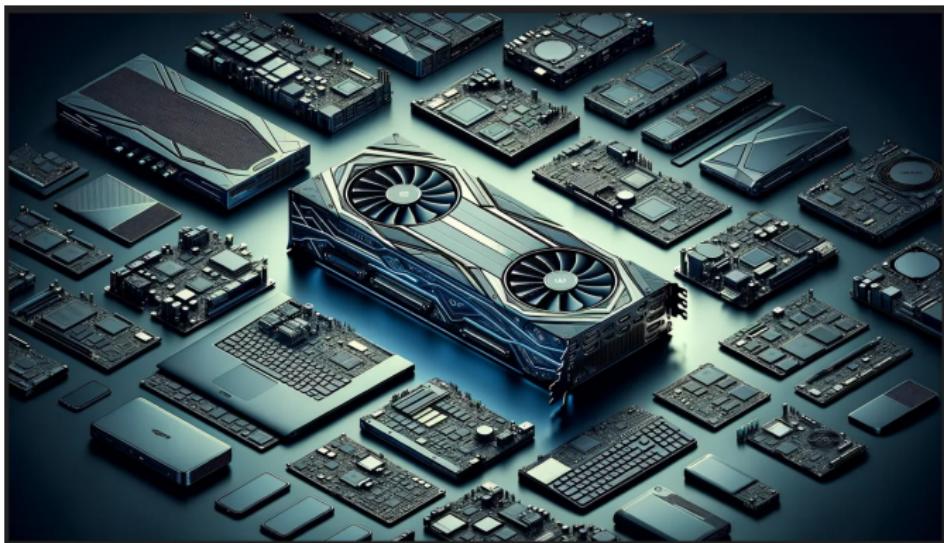
Background
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Coding with AI
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Deep Learning
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Tools
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Parallelization



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Parallelization



Just-in-time compilers

```
@jax.jit
def f(x):
    return jnp.sin(x) - jnp.cos(x**2)
```

- detects and adapts to problem dimensions
- detects and adapts to existing hardware
- automatic parallelization

Automatic differentiation

```
from jax import grad

def f(theta, x):
    # add details here
    return y_hat

def loss(theta, x, y):
    return jnp.sum((y - f(theta, x))**2)

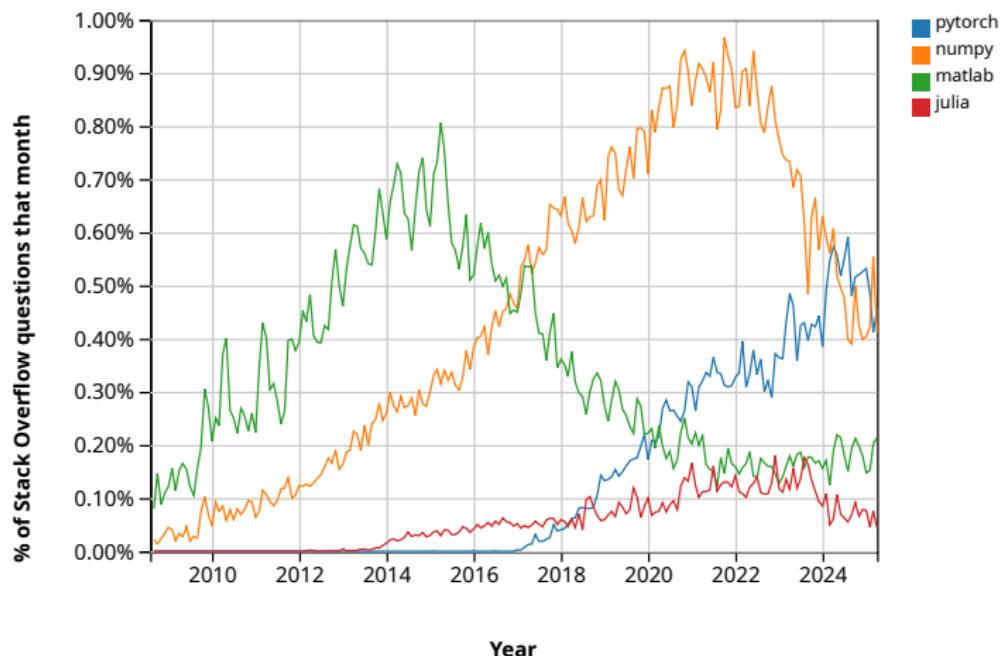
loss_gradient = grad(loss)    # exact derivative
theta = theta - lambda * loss_gradient(theta, x_data, y_data)
```

Platforms

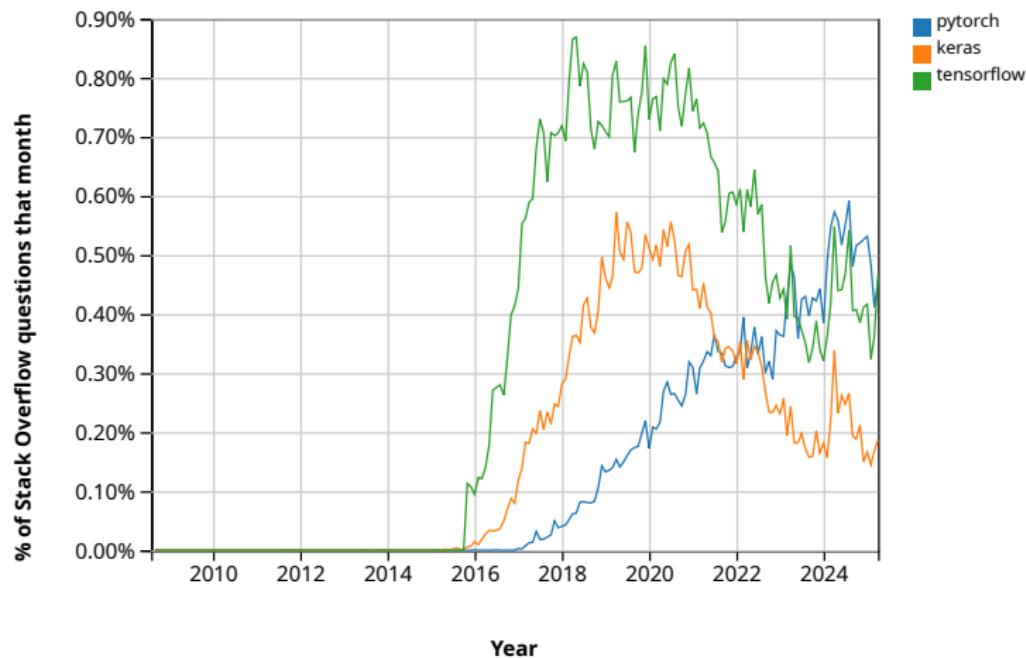
Platforms that support AI / deep learning:

- Tensorflow
- PyTorch (Llama, ChatGPT)
- Google JAX (Gemini, DeepMind)
- Keras (backends = JAX, PyTorch)
- Mojo (Modular (Python))
- MATLAB???

Popularity – languages and libraries



Popularity – DL / ML frameworks



AI tools for economic modeling

Let's say that you want to do computational economics without deep learning

Can these new AI tools be applied?

Yes! Yes! Yes!

- fast matrix algebra
- fast solutions to linear systems
- fast nonlinear system solvers
- fast optimization, etc.

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Case Study

The CBC uses the “overborrowing” model of Bianchi (2011)

- credit constraint loosens during booms
- bad shocks → sudden stops

CBC implementation in MATLAB

- runs on \$10,000 mainframe with 356 CPUs and 1TB RAM
- runtime = 12 hours

Rewrite in Python + Google JAX

- runs on \$400 gaming GPU with 10GB RAM
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Summary

- We are at the start of a massive AI revolution
- This revolution will have a huge impact on science
- What impact on economic modeling / policy analysis?

Aims:

- Better understanding of core AI methods
- Better understanding of core tools (hardware / software)
- Apply knowledge to economic modeling / data analysis
- Explore policy implications

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