Computational Economics and the Al Revolution

John Stachurski

2025

Topics

- Introduction
- ANNs and deep learning
- Stochastic approximation
- Dynamic programming
- Reinforcement learning
- Economic applications

Al-driven scientific computing

Al is changing the world

- generative AI (LLMs, image / music / video generators)
- image processing / computer vision
- speech recognition, translation
- scientific knowledge discovery
- forecasting and prediction
- etc





AlphaFold

Accelerating breakthroughs in biology with Al

AlphaFold

- Al system by DeepMind that predicts protein structures
- Open-sourced code and database for global access
- Accelerating drug discovery and design
- Enabling research on diseases like cancer and Alzheimer's
- Supporting enzyme engineering for sustainability

2024 Nobel Prize in Chemistry awarded to Demis Hassabis and John Jumper for development of AlphaFold

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?





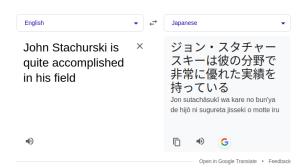
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Image Generators



Translation Engines



Killer drones, Skynet, etc.



Investment

Private Al investment in 2024:

- U.S. = \$109 billion
- China \$9.3 billion
- UK \$4.5 billion

Massive investments in

- data centers
- GPU design and production
- software development

What kinds of problems are they trying to solve?

Deep learning in two slides

Aim: approximate an unknown functional relationship

$$y = f(x)$$
 $(x \in \mathbb{R}^k, y \in \mathbb{R})$

Examples.

- x= cross section of returns today, y= return on oil futures tomorrow
- x = weather sensor data today, y = max temp tomorrow

Problem:

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

Deep learning is nonlinear regression:

- 1. Choose function class $\{f_{\theta}\}_{\theta \in \Theta}$
- 2. Minimize loss over the parameter space, where

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

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

- We discuss this structure soon
- \bullet Typically, $\theta \mapsto f_{\theta}(x)$ is smooth for all x
- MSE is a popular loss function but others are also used

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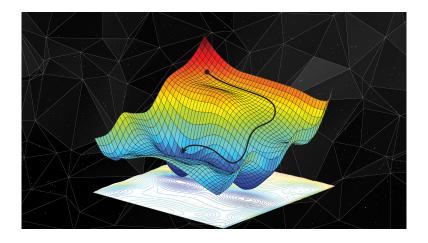
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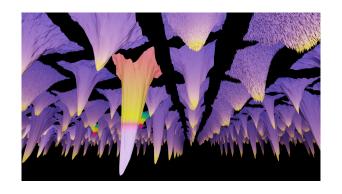
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Minimizing a smooth loss functions – what algorithm?



Source: https://danielkhv.com/

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



Source: https://losslandscape.com/gallery/

How does it work?

Why is it possible to minimize over $\theta \in \mathbb{R}^d$ when $d=10^{12}$?!?

Core elements

- automatic differentiation (for gradient descent)
- parallelization (GPUs or TPUs)
- Compilers / JIT-compilers

How does it work?

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Automatic differentiation

"Exact numerical" differentiation

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

loss_gradient = grad(loss)
dθ = loss_gradient(θ, x_data, y_data)
θ = θ - λ * dθ
```

Parallelization



```
outputs = pmap(f, data)
```

- multithreading over GPU cores (how many?)
- multiprocessing over accelerators in a GPU farm / supercomputing cluster (how many?)

Just-in-time compilers

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

Advantages over AOT compilers:

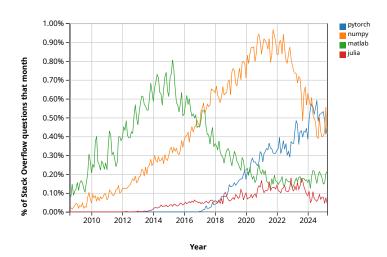
- cleaner code
- more portable
- automatic parallelization (same code for CPUs / GPUs)

Platforms

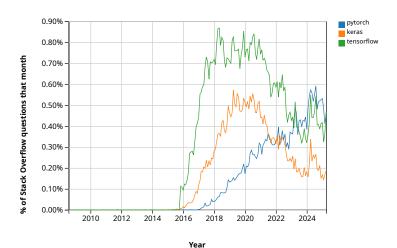
Platforms that support Al / deep learning:

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

Popularity – languages



Popularity - deep learning frameworks



Al for coding

```
please write a function for gradient descent in 6 lines or less
Here's a gradient descent function in 6 lines:
def gradient descent(grad func, init params, lr=0.01, iters=100):
    params = init params.copy()
         gradient = grad func(params)
     return params
```

Affects language choice

"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."

Thoughts from professional developer Lonely-Public2655

- Al doesn't see the big picture
- Can ace small tasks but struggles to connect them meaningfully
- You still need to be the architect
- Context is fragile: Al forgets. A lot.
- Once things get weird, Al starts guessing
- Sometimes AI gets really weird

Al 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
- runtime = 7 seconds

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