# Dynamic Programming with Google JAX CEF 2024

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### **Topics**

- Types of programming languages
- Vectorization vs JIT compilers
- Motivation for JAX
- Intro to JAX
- JAX for DP

Target audience: people who are new to / curious about JAX

Please feel free to question / debate / share your experiences

Slides, code:

https://github.com/QuantEcon/cef\_2024\_singapore

### Quick poll:

- Python programmers?
  - NumPy? Numba? PyTorch? JAX?
- Julia?
- MATLAB?
- C?
- Fortran?

### Regular GPU users?

# Old school: static types & AOT compilers

Example. Consider the Solow model

$$k_{t+1} = s k_t^\alpha + (1-\delta) k_t \qquad \text{with } k_0 \text{ given}$$

Fortran code:

```
program main
 implicit none
 integer, parameter :: dp=kind(0.d0)
 integer :: n=1000
 real(dp) :: s=0.3 dp
 real(dp) :: a=1.0 dp
 real(dp) :: delta=0.1 dp
 real(dp) :: alpha=0.4_dp
 real(dp) :: k=0.2 dp
 integer :: i
 do i = 1, n - 1
  k = a * s * k**alpha + (1 - delta) * k
 end do
 print *,'k = ', k
end program main
```

Relative merits of Fortran / C / other static type AOT compiled languages?

#### **Pros**

fast loops / arithmetic

#### Cons

- low interactivity
- time consuming to write / read / debug
- hard to parallelize

For comparison, the same operation in Python:

```
α = 0.4
s = 0.3
δ = 0.1
n = 1_000
k = 0.2

for i in range(n):
    k = s * k**α + (1 - δ) * k

print(k)
```

#### Pros

- high interactivity
- easy to write / read / debug

#### Cons

• slow loops / arithmetic

Why is pure Python slow?

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Why is pure Python slow?

# Problem 1: Type checking

How does Python know which operation to perform?

### Answer: Python checks the type of the objects first

```
>> x = 1
>> type(x)
int
```

```
>> x = 'foo'
>> type(x)
str
```

In a large loop, this type checking generates massive overhead

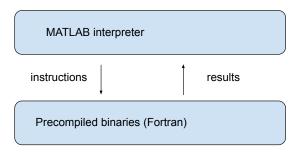
# Problem 2: Memory management

```
>>> import sys
>>> x = [1.0, 2.0]
>>> sys.getsizeof(x) * 8  # number of bits
576  # whaaaat???
```

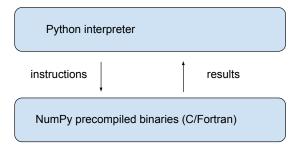
So how can we get

good execution speeds and high productivity / interactivity?

### MATLAB's vectorization trick



# Python + NumPy - stealing MATLAB's idea



### Vectorization: pros and cons

#### **Pros**

high interactivity

#### Cons

- some tasks cannot be efficiently vectorized
- cannot adapt flexibly to function arguments / hardware

# Julia — rise of the JIT compilers

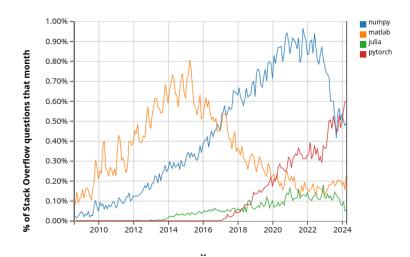
Function solow is efficiently JIT compiled after the first call

# Python + Numba copy Julia

```
from numba import jit
@jit
def solow(k0, \alpha=0.4, \delta=0.1, n=1_000):
    k = k0
    for i in range(n-1):
         k = s * k**\alpha + (1 - \delta) * k
     return k
solow(0.2)
```

Runs at same speed as Julia / C / Fortran

#### Some trends:



So where does JAX fit in?

Let's start with some motivation and background

# Al-driven scientific computing

### Al is changing the world

- image processing / computer vision
- speech recognition, translation
- scientific knowledge discovery
- forecasting and prediction
- generative AI

Plus killer drones, skynet, etc....

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### Projected spending on AI in 2024:

• Google: \$48 billion

• Microsoft: \$60 billion

• Meta: \$40 billion

• etc.

# Deep learning in two slides

Supervised deep learning: find a good approximation to an unknown functional relationship

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

#### Examples.

- x = sequence of words, y = next word
- $ullet x = {\sf weather \ sensor \ data}, \ y = {\sf max \ temp \ tomorrow}$

#### Problem:

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

Training: minimize the empirical loss

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

But what is  $\{f_{\theta}\}_{\theta\in\Theta}$ ?

In the case of ANNs, we consider all  $f_{ heta}$  having the form

$$f_\theta = \sigma \circ A_1 \circ \cdots \circ \sigma \circ A_{k-1} \circ \sigma \circ A_k$$

where

- $A_i x = W_i x + b_i$  is an affine map
- $\sigma$  is a nonlinear "activation" function

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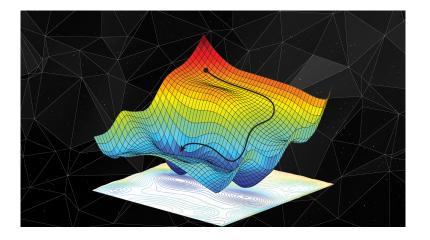
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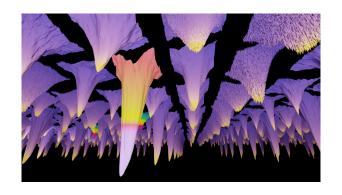
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### Minimizing a smooth loss functions – what algorithm?



Source: https://danielkhv.com/

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



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

# Hardware



"NVIDIA supercomputers are the factories of the AI industrial revolution." – Jensen Huang

### Software

#### Core elements

- automatic differentiation (for gradient descent)
- parallelization (GPUs! how many?)
- Compilers / JIT-compilers

Crucially, these components must be well integrated



```
import jax.numpy as jnp
from jax import grad, jit
def f(\theta, x):
  for W, b in \theta:
    W = W \otimes X + b
    x = jnp.tanh(w)
  return x
def loss(\theta, x, y):
  return jnp.sum((y - f(\theta, x))**2)
grad loss = jit(grad(loss)) # Now use gradient descent
```

Source: JAX readthedocs

### JAX for economists

I do mathematical modeling / optimization / simulation

#### My wishlist:

- <u>automated</u> parallelization
- JIT compiler
- integrated autodiff
- automatically / transparently supports CPUs / GPUs / TPUs

JAX ticks these boxes