## What's JAX

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# **Topics**

- Foo
- Bar

### Focus on JAX

https://jax.readthedocs.io/en/latest/

- Just-in-time compilation
- Automatic differentiation
- Xccelerated linear algebra

```
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
```

Example. AlphaFold3 (Google JAX)

### Highly accurate protein structure prediction with AlphaFold

John Jumper, Richard Evans, Alexander Pritzel, Tim Green, Michael Figurnov, Olaf Ronneberger, Kathryn Tunyasuvunakool,...

Nature Vol. 596 (2021)

- Citation count = 30K
- Nobel Prize in Chemistry 2024

## Functional Programming

JAX adopts a functional programming style

Key feature: Functions are pure

- Deterministic: same input ⇒ same output
- Have no side effects (don't modify state outside their scope)

#### A non-pure function

```
tax_rate = 0.1 # Global
price = 10.0 # Global

def add_tax_non_pure():
    global price
    # The next line both accesses and modifies global state
    price = price * (1 + tax_rate)
    return price
```

#### A pure function

```
def add_tax_non_pure(price, tax_rate=0.1):
    price = price * (1 + tax_rate)
    return price
```

### General advantages:

- Helps testing: each function can operate in isolation
- Data dependencies are explicit, which helps with understanding and optimizing complex computations
- Promotes deterministic behavior and hence reproducibility
- Prevents subtle bugs that arise from mutating shared state

#### Advantages for JAX:

- Functional programming facilitates autodiff because pure functions are more straightforward to differentiate (don't mod external state
- Pure functions are easier to parallelize and optimize for hardware accelerators like GPUs (don't depend on shared mutable state, more independence)
- Transformations can be composed cleanly with multiple transformations yielding predictable results
- Portability across hardware: The functional approach helps JAX create code that can run efficiently across different hardware accelerators without requiring hardware-specific implementations.

# JAX PyTrees

A PyTree is a concept in the JAX library that refers to a tree-like data structure built from Python containers.

#### Examples.

- A dictionary of lists of parameters
- A list of dictionaries of parameters, etc.

#### JAX can

- apply functions to all leaves in a PyTree structure
- differentiate functions with respect to the leaves of PyTrees
- etc.

#### JAX PyTree Structure

```
pytree = {
    "a": [1, 2, 3],
    "b": {"c": jnp.array([4, 5]), "d": jnp.array([[6, 7], [8, 9]])}
                                        dict
                "a"
                                                                  "h"
                list
                                                                 dict
                                                                              "d"
                                                    [4, 5]
                                                                        [[6, 7], [8, 9]]
   Container nodes (dict, list, tuple)
   Leaf nodes (arrays, scalars)
```

```
# Apply gradient updates to all parameters
def sqd update(params, grads, learning rate):
    return jax.tree map(
        lambda p, g: p - learning rate * g,
        params,
        grads
# Calculate gradients (PyTree with same structure as params)
grads = jax.grad(loss fn)(params, inputs, targets)
# Update all parameters at once
updated params = sgd update(params, grads, 0.01)
```