

NASA-STIG — AD / PyTorch / JAX

Accelerated Computation with Auto-differentiation

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Outline:

- Autodiff Motivation: Why do we need anything beyond NumPy & SciPy?
- PyTorch (Dec. 15th)
 - Model Building
 - Training
- JAX + JAX Ecosystem (Dec. 22nd)

NumPy & SciPy



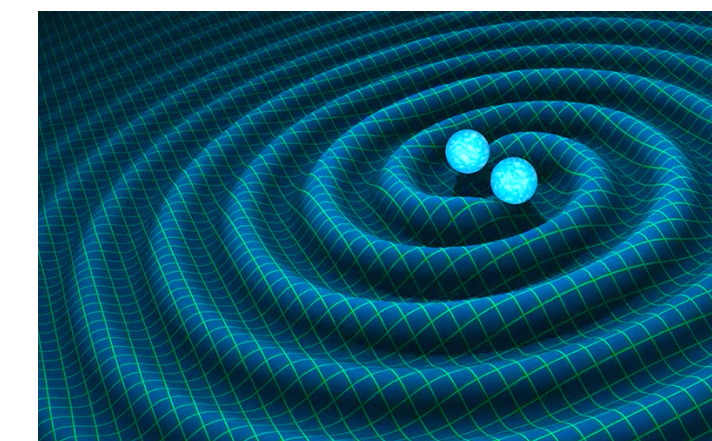
Tasks:

- ◆ Efficient N-D Array Math
- ◆ Vectorized Operations
- ◆ Broadcasting
- ◆ Statistical Functions
- ◆ Linear Algebra
- ◆ FFT
- ◆ Random Numbers



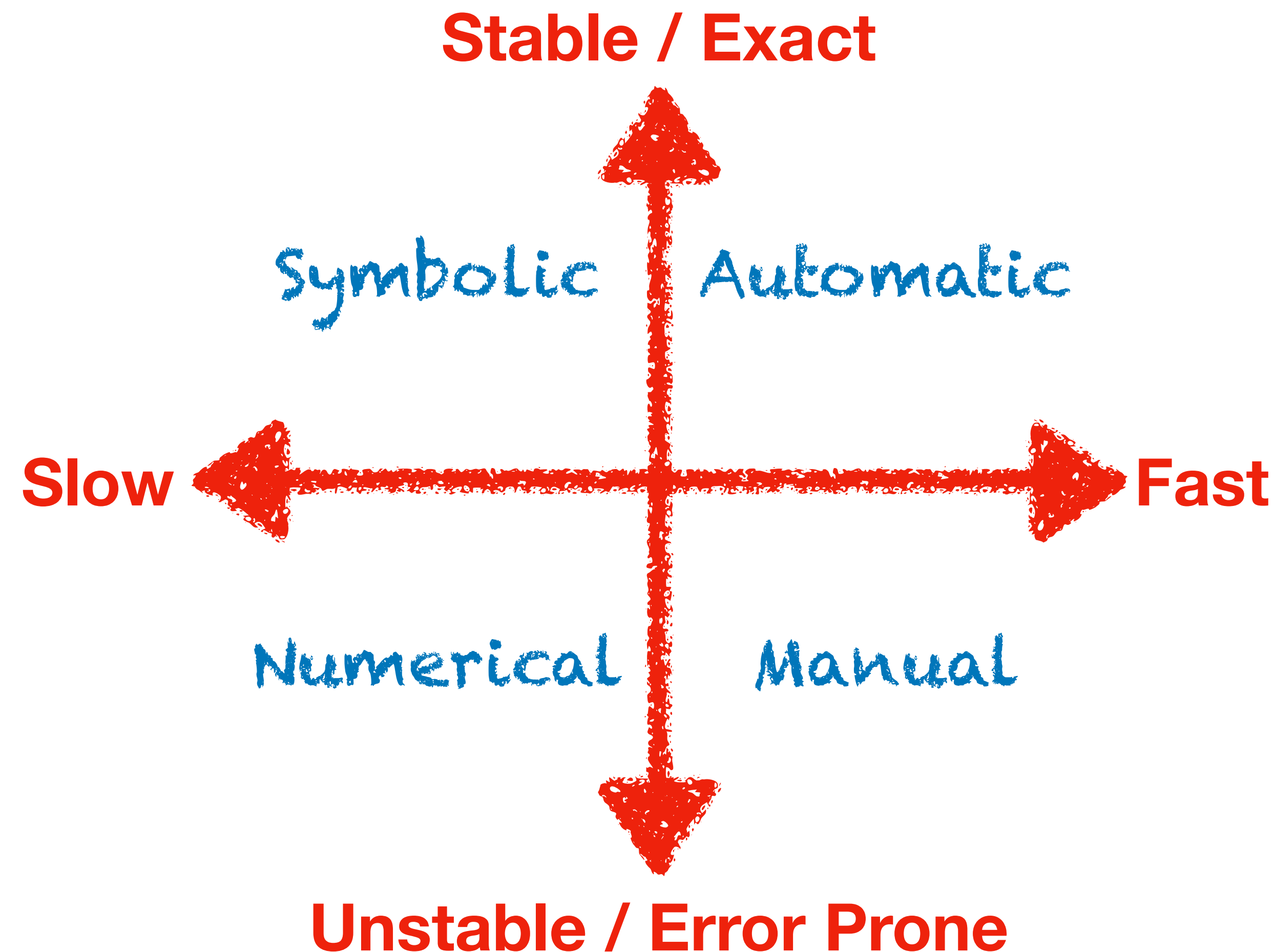
Tasks:

- ◆ Regression / Optimization
- ◆ Signal Processing
- ◆ Interpolation
- ◆ Integration



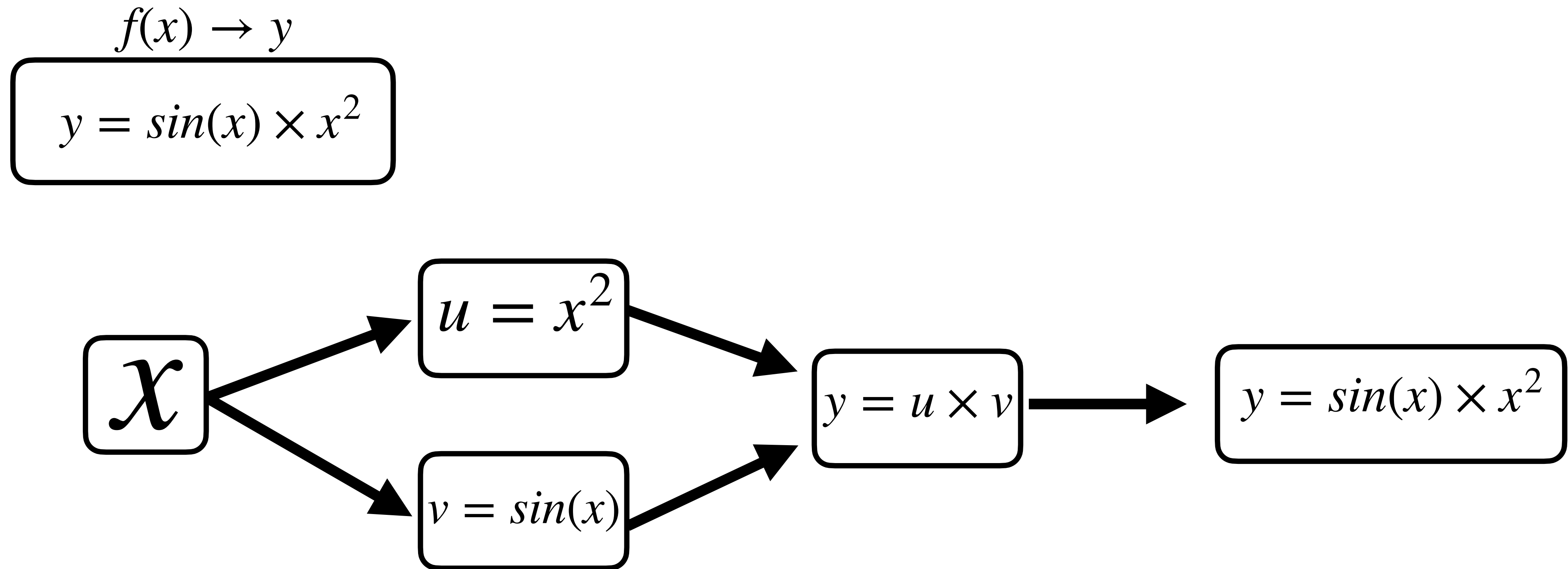
What is the partial derivative of a Python function?

Ways to compute a derivative:



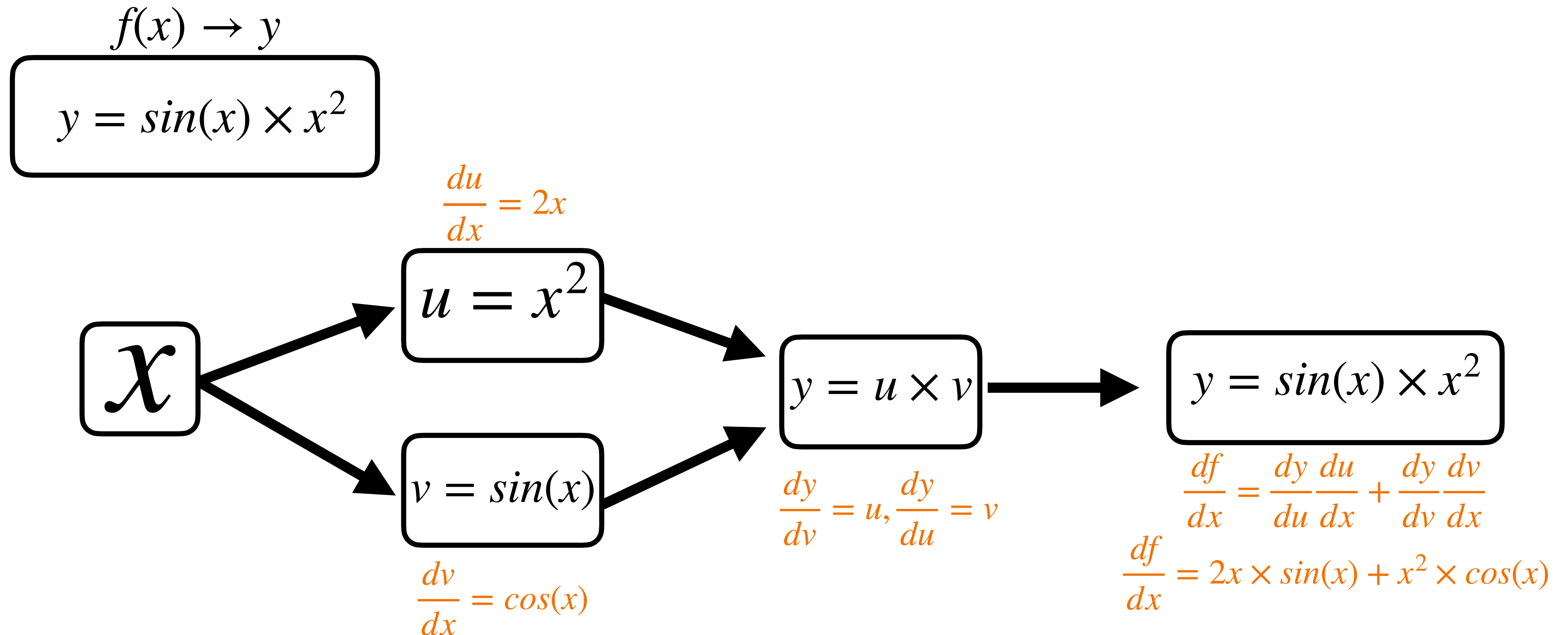
Automatic Differentiation (Autodiff, AD, etc.)

i.e., the chain-rule!



Automatic Differentiation (Autodiff, AD, etc.)

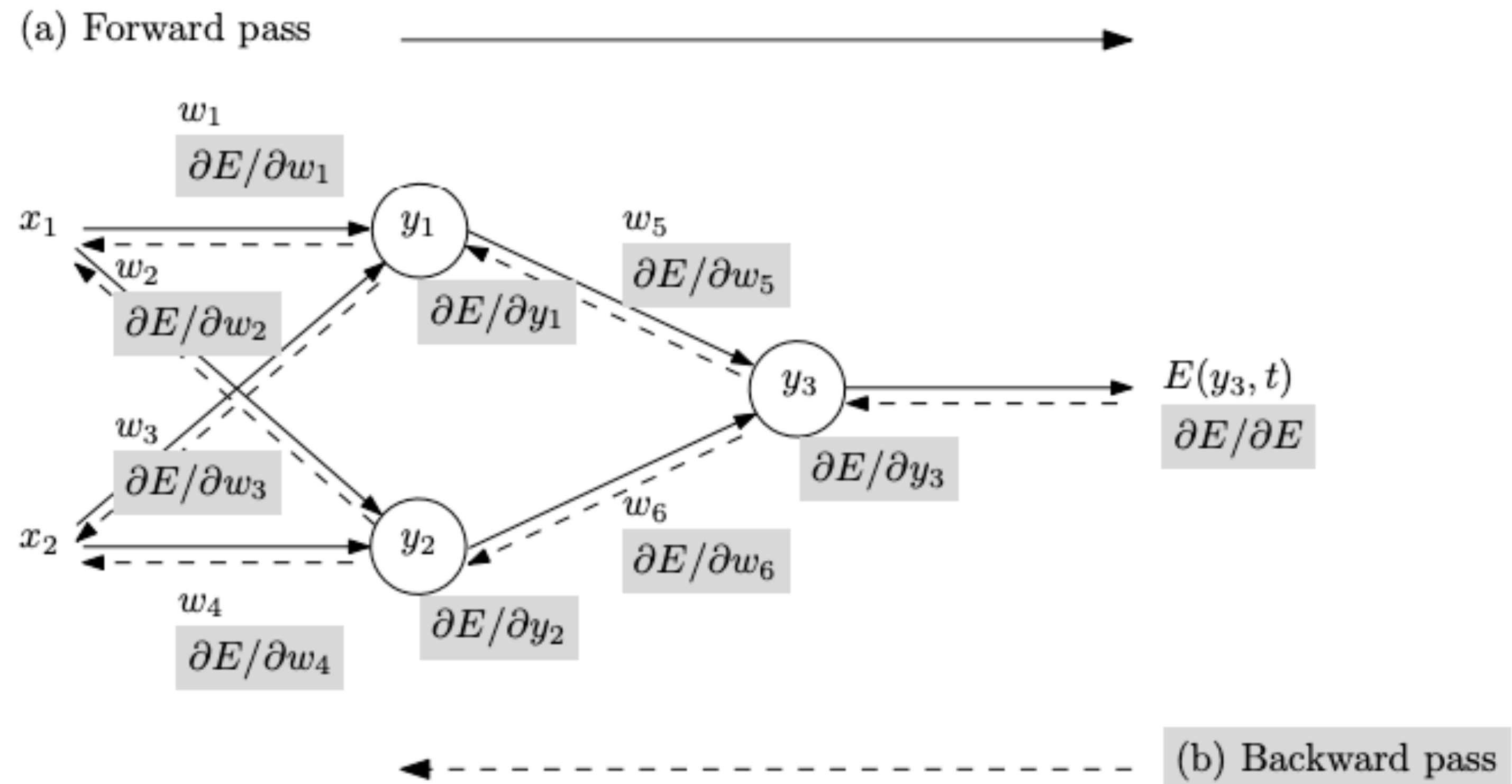
i.e., the chain-rule!



Now scale this to millions of operations!

Two Flavors of AD:

$$f(x_1, x_2) = E$$



Rule of Thumb:

Forwards: $f: R^n \rightarrow R^m, n \ll m$

Backwards: $f: R^n \rightarrow R^m, n \gg m$

Where can I use AD in astronomy?

- Physics & Motion Simulations (e.g., N-body)
- Solving complex non-linear PDEs (e.g., Fluid Dynamics)
- Verification of code correctness through derivatives
- Information Theory (e.g., Fisher information of system)
- Probabilistic & Bayesian Modeling (see you next week)
- Complex, high-order, non-linear optimization & regression (AI / ML)

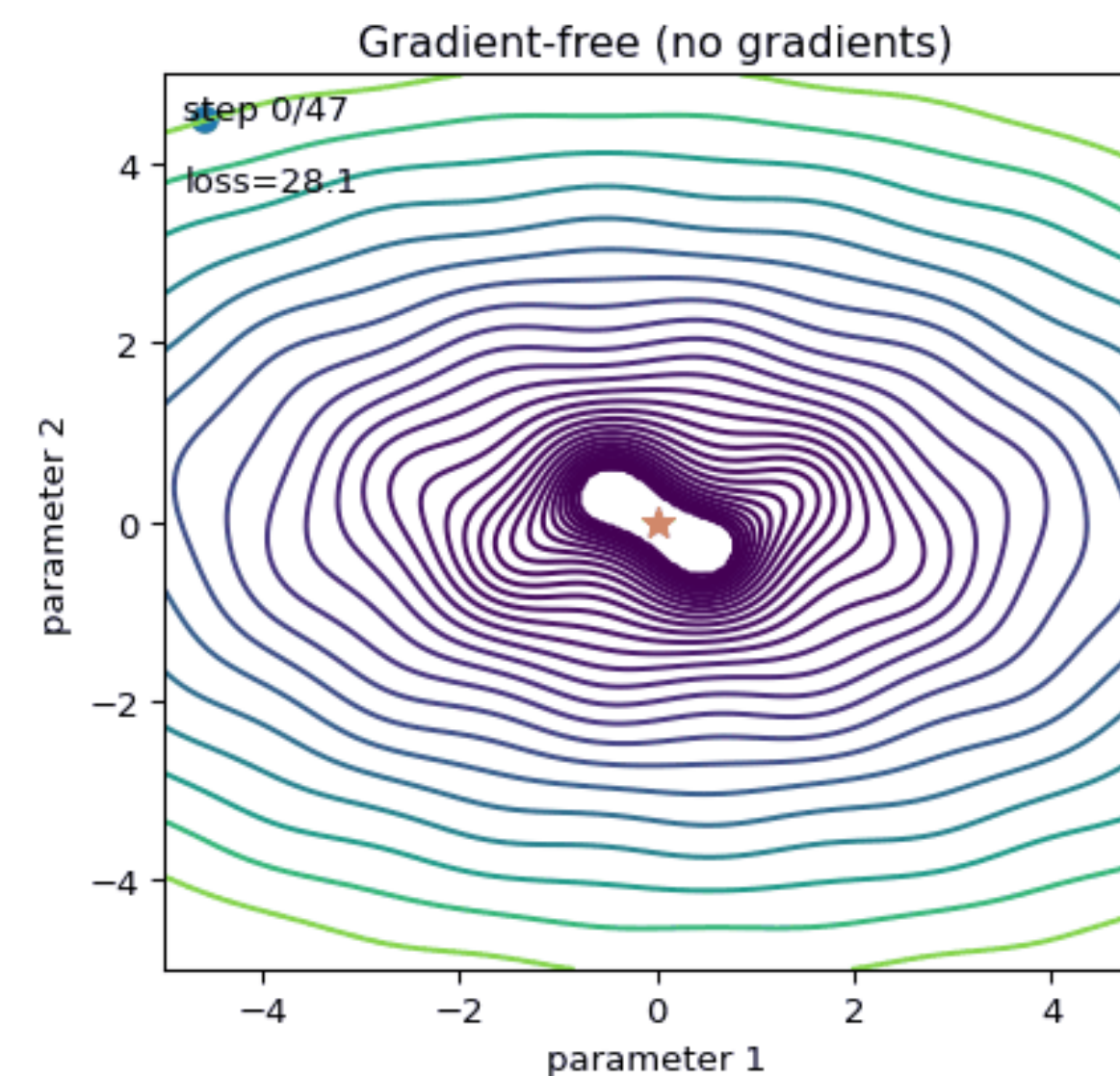
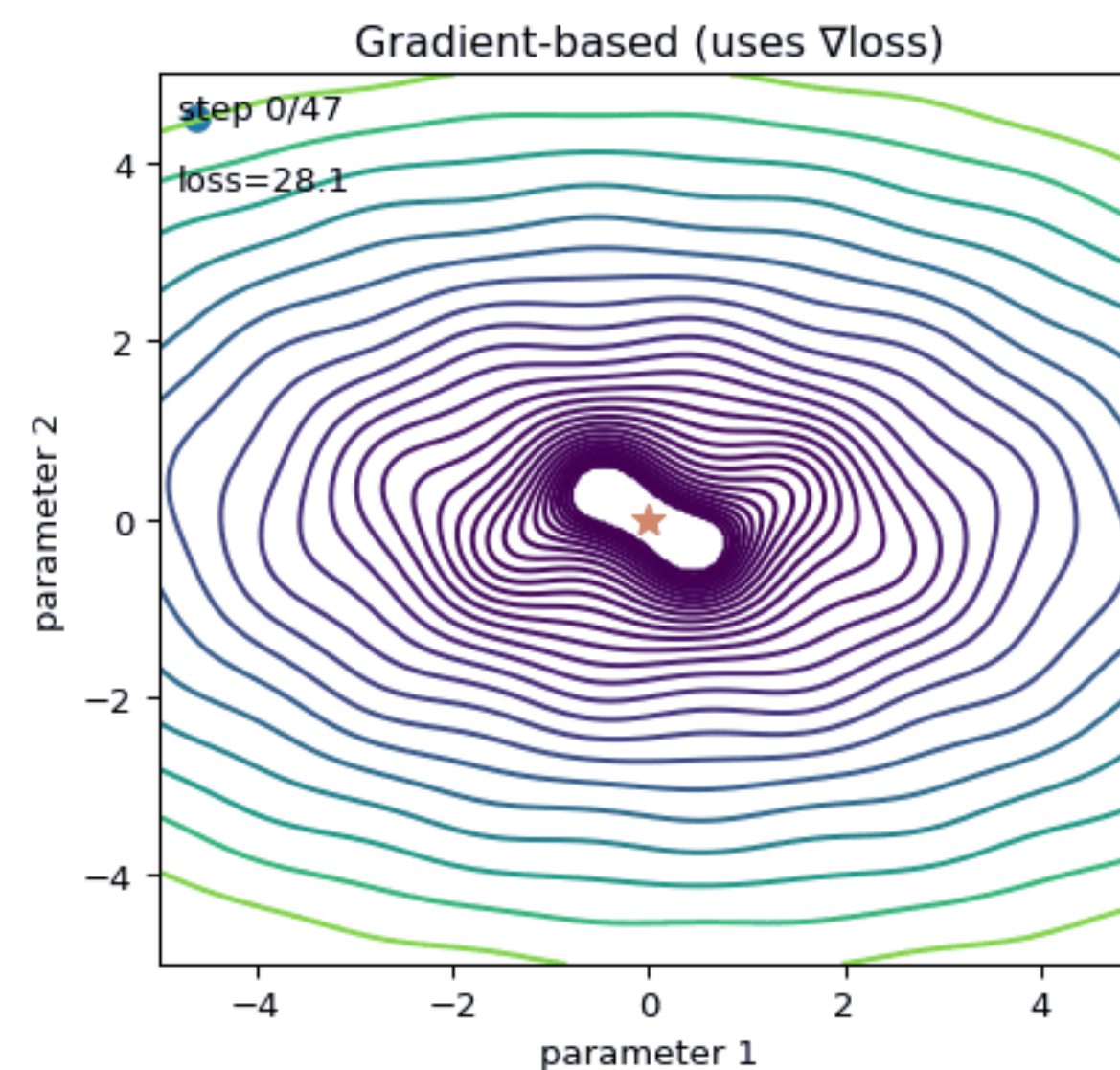
Example: Optimization of large models

Fit a complex model with lots of parameters to some data!

Loss Function: what you are trying to optimize in regression (e.g., mean-squared error)

Regression Approaches:

- ◆ Gradient-free Methods (e.g., Nelder-Mead, Simulated Annealing, LM)
- ◆ Gradient-based Methods (e.g., SGD, L-BFGS, Adam)



Example: Optimization of large models

Gradient-free Methods

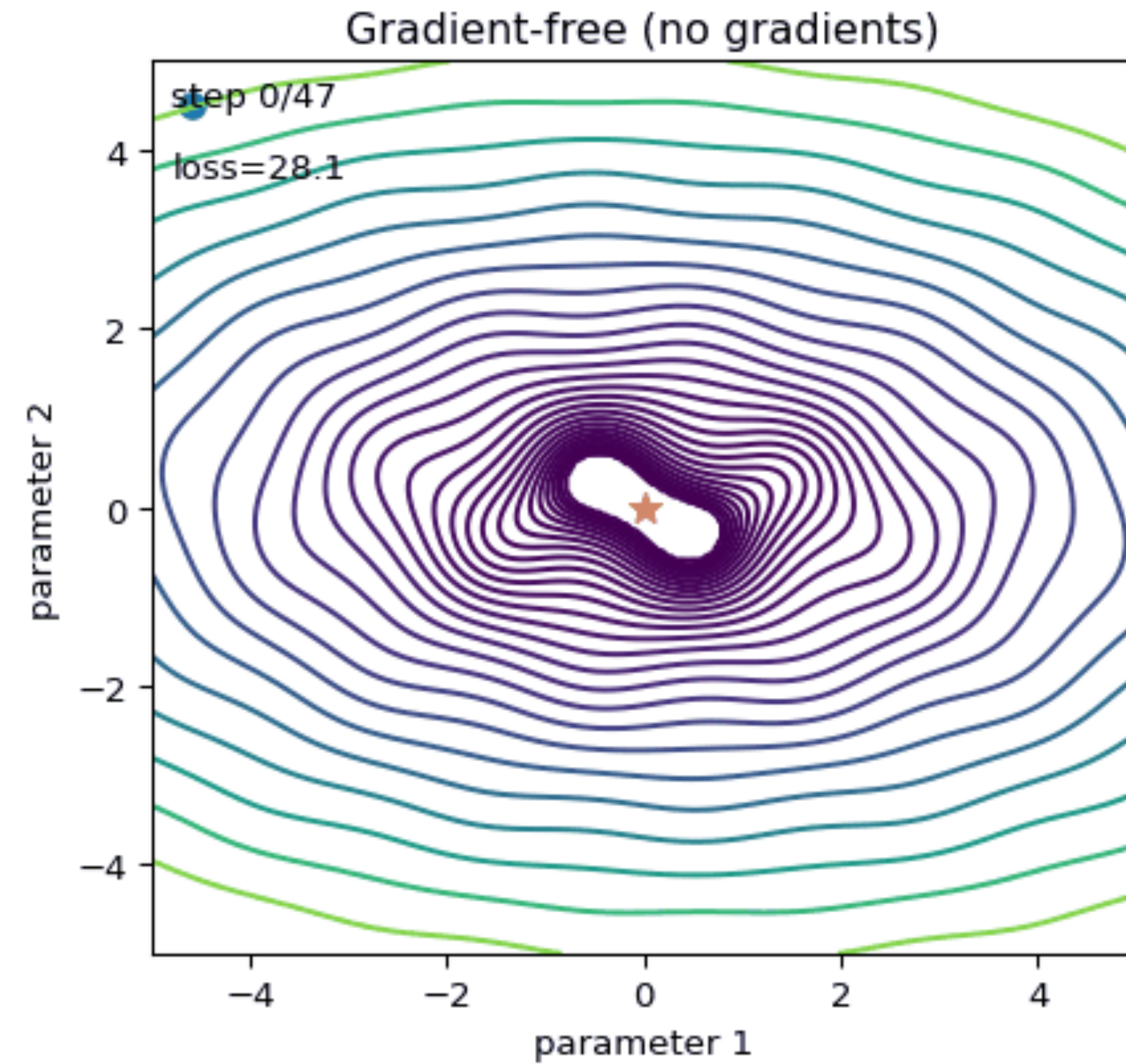
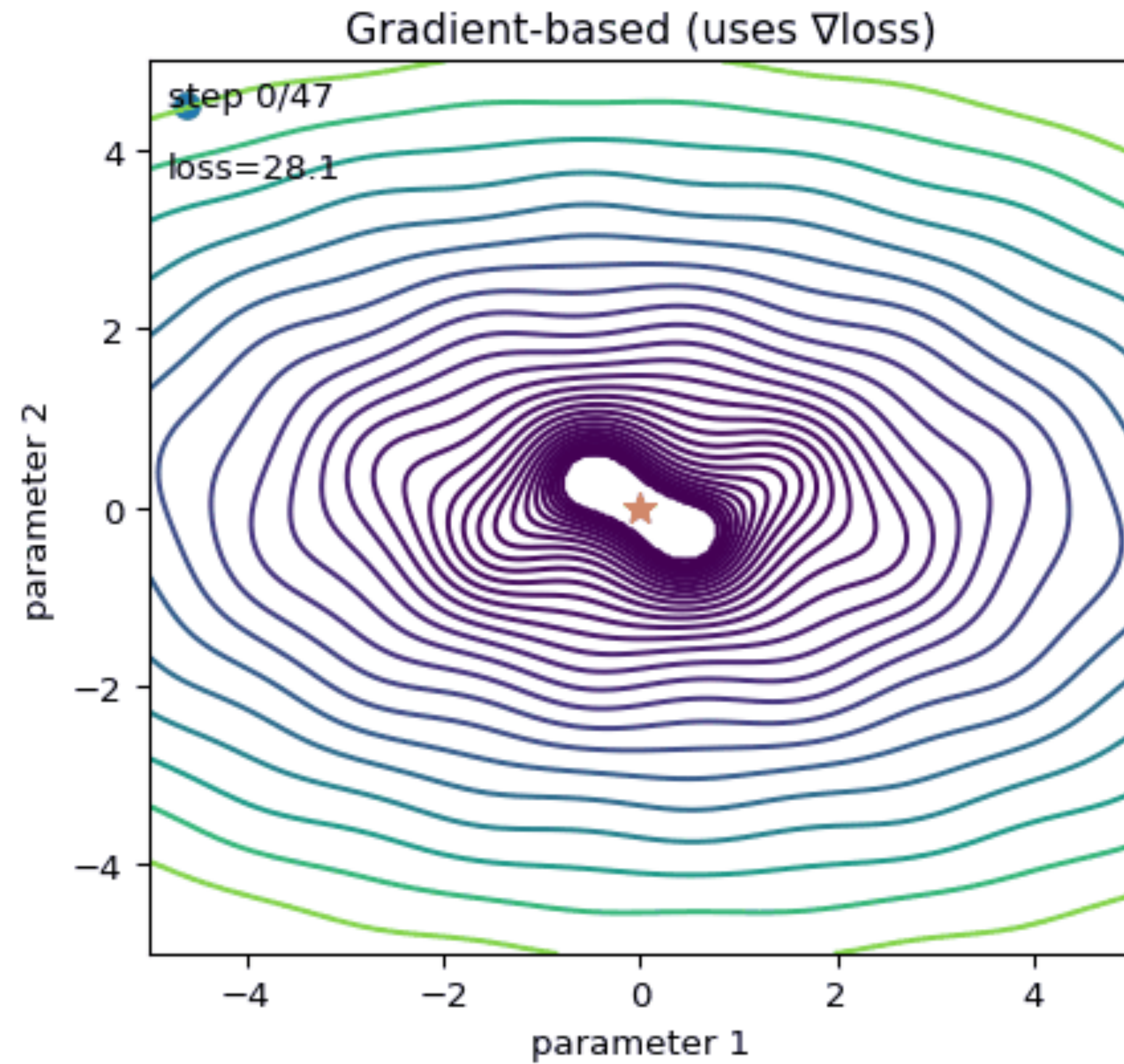
- Explores Loss Function without a priori knowledge to find optimal solution
- Optimization Scales $\sim O(N^2)$

Gradient-based Methods

- Uses partial derivatives of loss to decide how to navigate to the optimal solution
- Optimization Scales $\sim O(N)$

State-of-the-art LLM: 100's billions – trillions of free parameters

Any questions about AD?



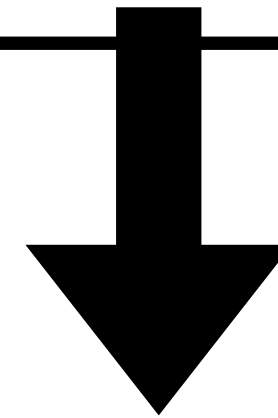


`torch.autograd`
backwards pass

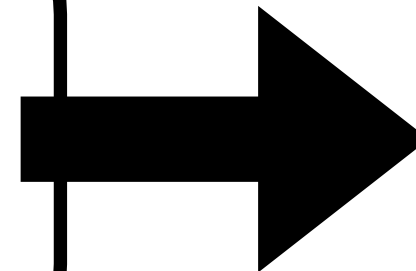


PyTorch Ecosystem

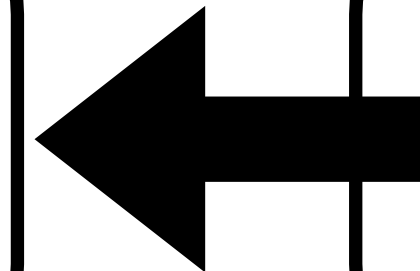
Model Building API



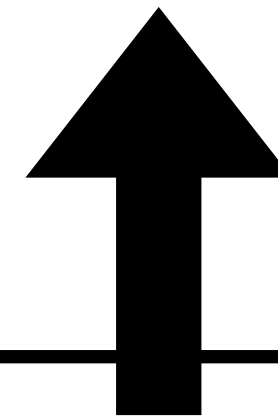
Flexible Training



`torch.autograd`
backwards pass



Dataloading



GPU Offloading