CS-453 Machine Learning

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Automatic Differentiation

Automatic Differentiation (AD)

- Set of techniques to numerically evaluate the derivative of a function specified by a computer program.
- Analytic or symbolic differentiation evaluates the derivative of a function specified by a math expression.
- ▶ AD Also called *algorithmic differentiation* or *computational differentiation*.
- Backpropagation is a special case of AD.

Modern machine learning frameworks (TensorFlow, Theano, PyTorch) employ AD. The programmer only needs to implement the forward pass up to the loss function. Derivatives are handled automatically!

Automatic Differentiation

AD exploits the fact that every computer program, no matter how complicated, executes a sequence of elementary arithmetic operations (addition, subtraction, multiplication, division, etc.) and elementary functions (exp, log, sin, cos, etc.). By applying the chain rule repeatedly to these operations, derivatives of arbitrary order can be computed automatically, accurately to working precision, and using at most a small constant factor more arithmetic operations than the original program.

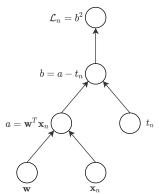
 $https://en.\ wikipedia.\ org/wiki/Automatic_\ differentiation$

Linear Regression via Automatic Differentiation

Consider the squared loss function for linear regression.

$$L_n(\mathsf{w}) = \left(\mathsf{w}^\mathsf{T} \mathsf{x}_n - t_n\right)^2$$

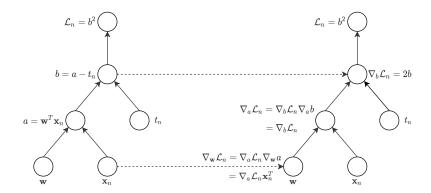
Can be represented as a computational graph consisting of elementary operations.



Linear Regression via Automatic Differentiation

Linear Regression

- ▶ For training, we are interested in the gradient $\nabla_w L_n$.
- \triangleright After the forward pass for a particular w and x_n , gradients can be evaluated numerically.



AD in Python

AD in Python

- ▶ A Python package called *Autograd* implements *reverse mode* automatic differentiation.
- computing their derivates $1, \cos, kx^{k-}$ etc..

 \triangleright Elementary operations such as +, sin, x^k etc. are overloaded by also

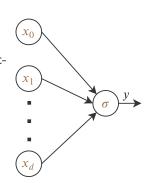
▶ If required, more sophisticated user-defined functions and their derivative implementations can be *registered* with Autograd.

Logistic Regression via Automatic Differentiation Binary classifier with no hidden layer

Just a perceptron with logistic sigmoid activation function. Models probability of class 1 instead of decision.

$$y = p(C_1|x) = \sigma(w^T x)$$

$$1 - y = p(C_2|x) = 1 - p(C_1|x)$$



Binary cross-entropy loss

$$\mathcal{L}(w) = -\sum_{n=1}^{N} t_n \ln y_n + (1 - t_n) \ln (1 - y_n)$$

Logistic Regression via Automatic Differentiation Step 1: Computational Graph for \mathcal{L}_n

Logistic Regression via Automatic Differentiation Step 2: AutoDiff till $\nabla_w \mathcal{L}_n$

Machine Learning