

Optimization of deep networks

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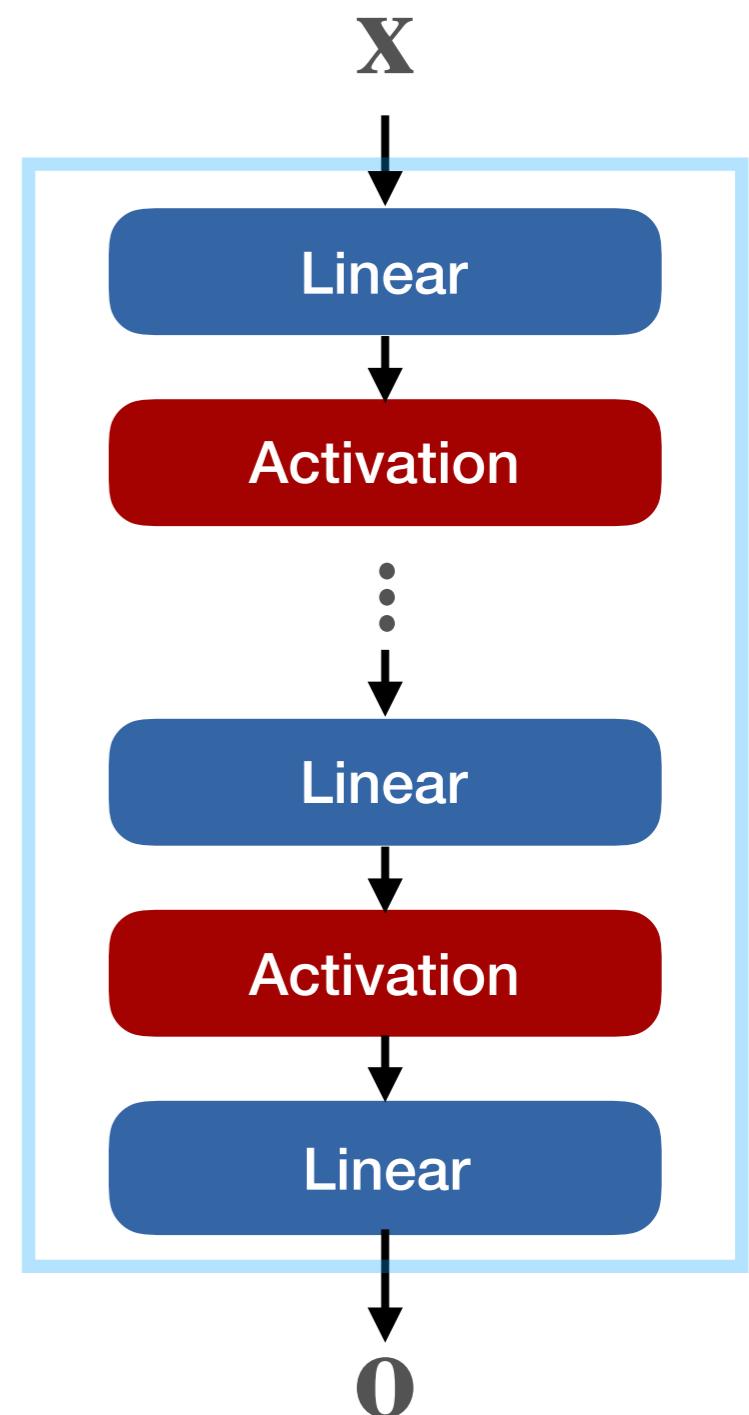
Data

- Input: $\{\mathbf{x}_0, \dots, \mathbf{x}_{N-1}\}$
- Label: $\{\mathbf{y}_0, \dots, \mathbf{y}_{N-1}\}$
- Dataset: $D = \{(\mathbf{x}_0, \mathbf{y}_0), \dots, (\mathbf{x}_{N-1}, \mathbf{y}_{N-1})\}$

- ( , dog)
- ( , dog)
- ( , cat)
- ( , cat)
- ⋮

Model

- Deep network $f: (\mathbf{x}, \theta) \rightarrow \mathbf{o}$
 - Layers of computation
 - Parameters θ
 - Differentiable computation graph



LOSS

- Differentiable $\ell(\mathbf{o}, \mathbf{y})$

- Regression
 - Distance norm

$$\ell(\mathbf{o}, \mathbf{y}) = \|\mathbf{o} - \mathbf{y}\|$$

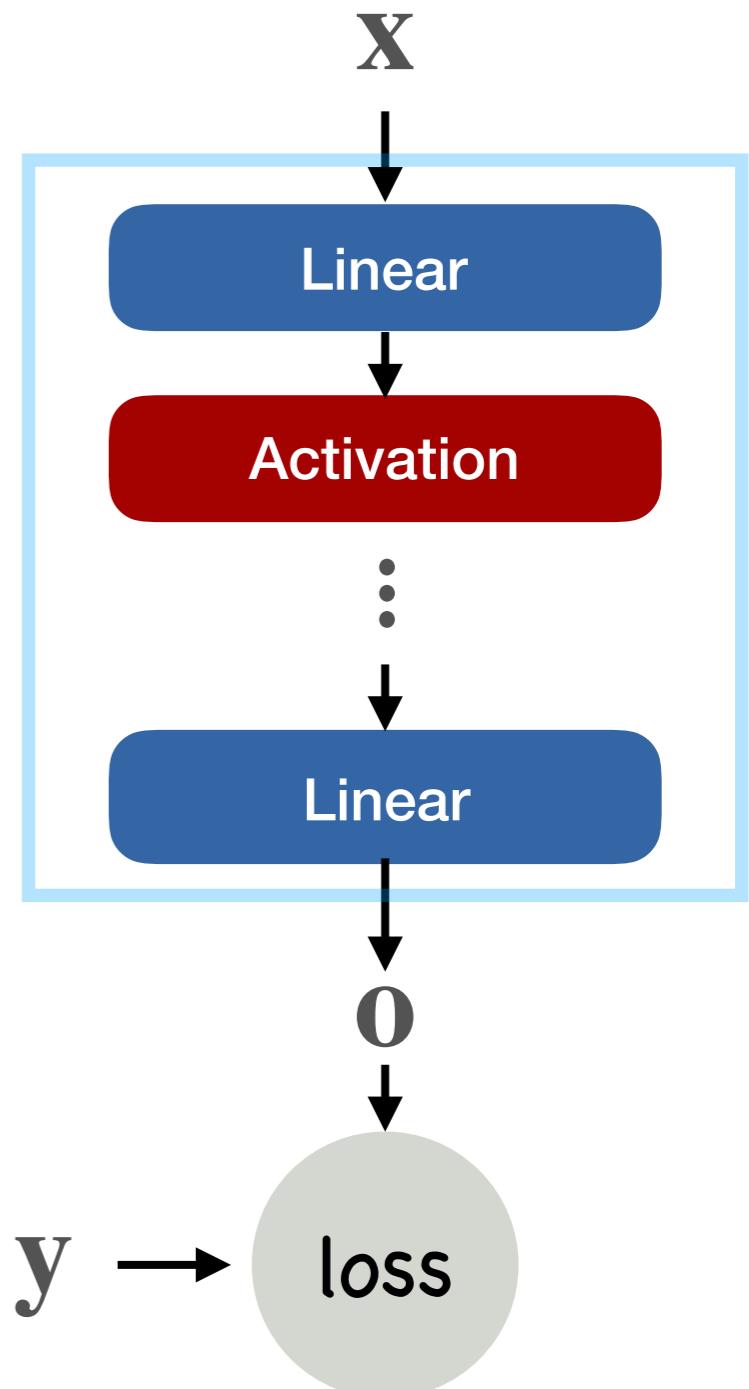
- Classification

- Cross Entropy

$$\ell(\mathbf{o}, \mathbf{y}) = -\log p(y)$$

- Over training dataset

- $L(\theta) = \mathbb{E}_{\mathbf{x}, \mathbf{y} \sim D} [\ell(f(\mathbf{x}, \theta), \mathbf{y})]$

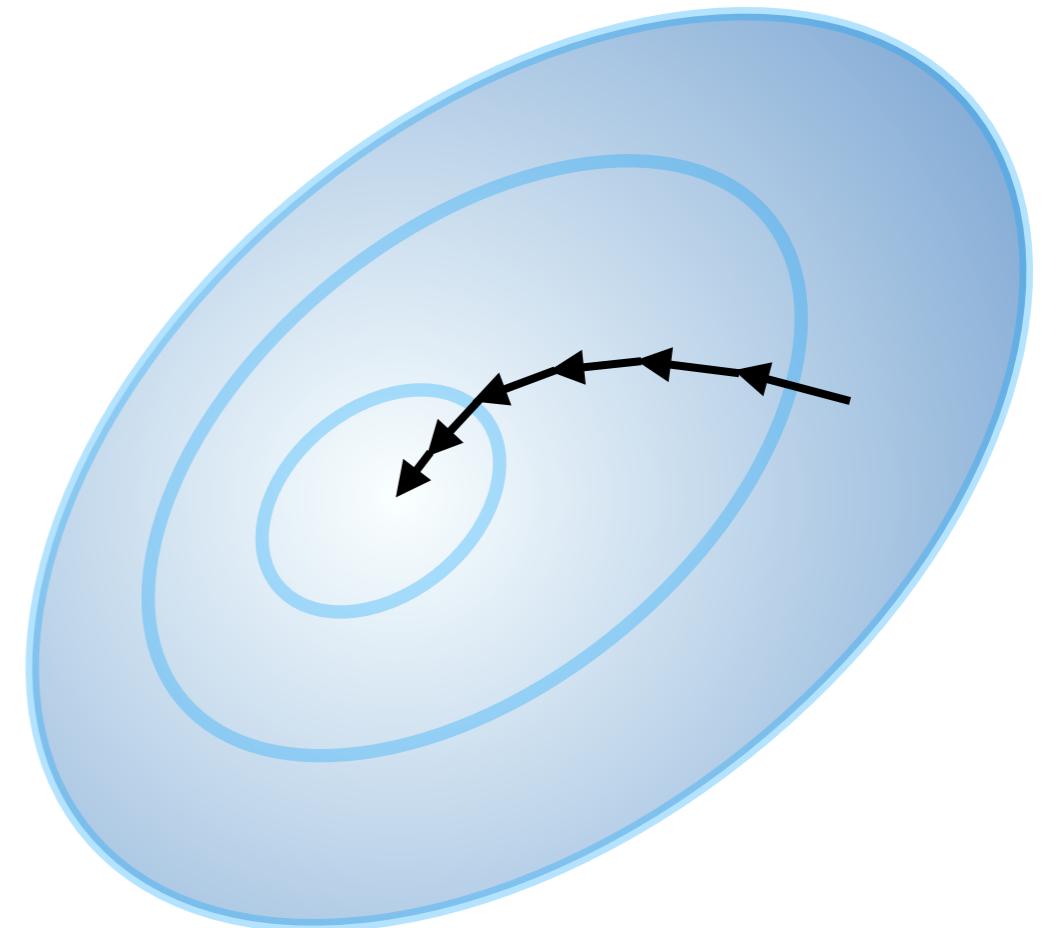


Optimization

- Minimize $L(\theta)$

Gradient Descent

- Repeat until convergence:
 - $\theta := \theta - \epsilon \frac{dL(\theta)}{d\theta}$



Issue with Gradient Descent

- Slow to compute gradient
- $\frac{dL(\theta)}{d\theta} = \mathbb{E}_{\mathbf{x}, \mathbf{y} \in D} \left[\frac{d\ell(f(\mathbf{x}, \theta), \mathbf{y})}{d\theta} \right]$