

Practical Deep Neural Networks

GPU computing perspective

Softmax Regression

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Outline

- 1 Introduction
- 2 Logistic Regression
- 3 Softmax Regression
- 4 Stochastic Gradient Descent

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Assumed Prerequisites

- ★ Basic Linear Algebra (DL book chapter 2)
- ★ Basic Probability and Information Theory (DL book chapter 3)
- ★ Basic Numerical Computation (DL book chapter 4)
- ★ Machine Learning Basics (DL book chapter 5)

Suggested Readings

- 📖 UFLDL Tutorial: Logistic Regression, Softmax Regression and Stochastic Gradient Descent.
- 📖 CS231n: Linear classification: Support Vector Machine, Softmax and Optimization: Stochastic Gradient Descent.
- 📖 DL Book Chapter 4 Numerical Computation 4.3 DL Book Chapter 8 Numerical Optimization 8.3

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Logistic Function

$$f(\mathbf{x}) = \frac{1}{1 + \exp(-\mathbf{W}^\top \mathbf{x})}$$

$$p(y = 1|\mathbf{x}) = f(\mathbf{x})$$

$$p(y = 0|\mathbf{x}) = 1 - f(\mathbf{x})$$

Logistic Regression — Binary Classifier

$$L(X, \mathbf{y} | \mathbf{W}) = -\frac{1}{N} \sum_i (y^i \log(P(y = 1 | \mathbf{x}^i)) + (1 - y^i) \log(p(y = 0 | \mathbf{x}^i)))$$
$$W^* = \arg \min_W L(X, \mathbf{y})$$

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Extending Logistic Function — Softmax Function

$$P(y = k | \mathbf{x}; \mathbf{W}) = \frac{\exp(\mathbf{W}^{(k)\top} \mathbf{x})}{\sum_{j=1}^K \exp(\mathbf{W}^{(j)\top} \mathbf{x})}$$

Softmax Regression — Multi-classes Classifier

$$L(X, \mathbf{y} | \mathbf{W}) = -\frac{1}{N} \sum_i \sum_{k=1}^K \mathbf{1}\{y^i = k\} \log P(y^i = k | \mathbf{x}^i, \mathbf{W})$$
$$\mathbf{W}^* = \arg \min_{\mathbf{W}} L(X, \mathbf{y})$$

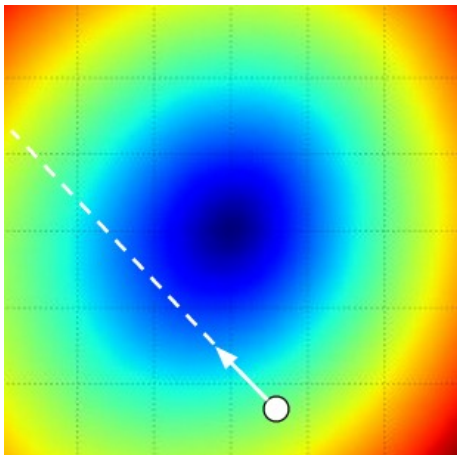
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Cost

- Target: to learn some optimal parameters θ
- Strategy: to minimize some cost L respect to θ
- Cost choice: cross-entropy cost, mean-squared error cost
- Solution: Gradient Descent!

Gradient optimization



Demo 1; Demo 2

Stochastic Gradient Descent (SGD)

$$\theta^* = \theta - \alpha \frac{\partial}{\partial \theta} L(\theta)$$

Variants: momentum SGD

$$V^* = \mu V - \alpha \nabla L(\theta)$$
$$\theta^* = \theta + V^*$$

Variants: Nesterov's Accelerated Gradient (NAG)

$$V^* = \mu V - \alpha \nabla L(\theta + \mu V)$$
$$\theta^* = \theta + V^*$$

Q&A

