

## 线性回归

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线性回归基本要素



#### 线性回归基本要素

线性回归输出是一个连续值,因此适用于回归问题。回归问题在实际中很常见,如预测房屋价格、气温、销售额等连续值的问题。与回归问题不同,分类问题中模型的最终输出是一个离散值。我们所说的图像分类、垃圾邮件识别、疾病检测等输出为离散值的问题都属于分类问题的范畴。softmax回归则适用于分类问题。

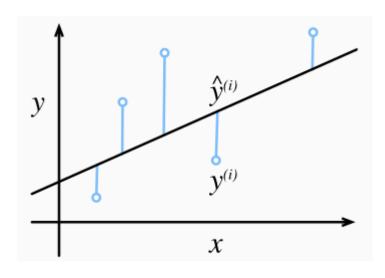
由于线性回归和softmax回归都是单层神经网络,它们涉及的概念和技术同样适用于大多数的深度学习模型。

#### 训练数据

Typically, we will use n to denote the number of samples in our dataset. We index the samples by I, denoting each input data point as  $x^{(i)} = [x_1^{(i)}, x_2^{(i)}]$  and the corresponding label as  $y^{(i)}$ 

#### 损失函数

$$l^{(i)}(\mathbf{w}, b) = \frac{1}{2} (\hat{y}^{(i)} - y^{(i)})^2,$$



$$L(\mathbf{w},b) = \frac{1}{n} \sum_{i=1}^n l^{(i)}(\mathbf{w},b) = \frac{1}{n} \sum_{i=1}^n \frac{1}{2} \Big( \mathbf{w}^\top \mathbf{x}^{(i)} + b - y^{(i)} \Big)^2.$$

$$\mathbf{w}^*, b^* = \underset{\mathbf{w}, b}{\operatorname{argmin}} \ L(\mathbf{w}, b).$$

#### 优化算法

$$(\mathbf{w},b) \leftarrow (\mathbf{w},b) - \frac{\eta}{|\mathcal{B}|} \sum_{i \in \mathcal{B}} \partial_{(\mathbf{w},b)} l^{(i)}(\mathbf{w},b)$$

$$\mathbf{w} \leftarrow \mathbf{w} - \frac{\eta}{|\mathcal{B}|} \sum_{i \in \mathcal{B}} \partial_{\mathbf{w}} l^{(i)}(\mathbf{w}, b) = w - \frac{\eta}{|\mathcal{B}|} \sum_{i \in \mathcal{B}} \mathbf{x}^{(i)} \left( \mathbf{w}^{\top} \mathbf{x}^{(i)} + b - y^{(i)} \right),$$

$$b \leftarrow b - \frac{\eta}{|\mathcal{B}|} \sum_{i \in \mathcal{B}} \partial_{b} l^{(i)}(\mathbf{w}, b) = b - \frac{\eta}{|\mathcal{B}|} \sum_{i \in \mathcal{B}} \left( \mathbf{w}^{\top} \mathbf{x}^{(i)} + b - y^{(i)} \right).$$

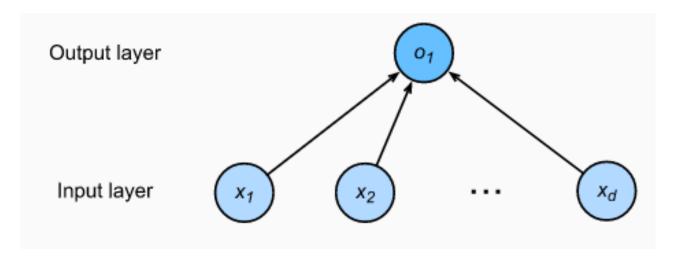
#### 模型预测

$$|\hat{\mathbf{w}}^{\top}x + \hat{b}|$$

模型预测



### 神经网络图



Linear regression is a single-later neural network.

# 谢谢!

#### Does anyone have any questions?

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**Blog**: https://ceteongvanness.wordpress.com

## 资源

Dive into Deep Learning