

线性回归闭式解推导

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1 基本概念

1、线性模型

$$f_{\theta}(x) = \theta_0 + \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_n x_n$$

矩阵形式为：

$$f(x) = \sum_{i=0}^n \theta_i x_i = \theta^T x$$

2、目标函数

由于

$$z^T z = \sum_i z_i^2$$

所以损失函数可写为：

$$J(\theta) = \frac{1}{2} \sum_{i=0}^m (f_{\theta}(x^{(i)}) - y^{(i)})^2 = \frac{1}{2} (X\theta - y)^T (X\theta - y)$$

2 闭式解推导

基本性质

$$\text{tr} A = \sum_{i=1}^n A_{ii}$$

$$\text{tr} AB = \text{tr} BA$$

$$\text{tr} A = \text{tr} A^T$$

$$\text{tr}(\text{real}) = \text{real}$$

$$\nabla_A \text{tr} AB = B^T$$

$$\nabla_{A^T} f(A) = (\nabla_A f(A))^T$$

$$\nabla_A \text{tr} ABA^T C = CAB + C^T AB^T$$

$$\nabla_A |A| = |A| (A^{-1})^T$$

$$\nabla_{A^T} \text{tr} ABA^T C = (\nabla_A \text{tr} ABA^T C)^T = (CAB + C^T AB^T)^T = B^T A^T C^T + BA^T C$$

推导过程

$$\begin{aligned}
\nabla_{\theta} J(\theta) &= \nabla_{\theta} \frac{1}{2} (X\theta - y)^T (X\theta - y) \\
&= \frac{1}{2} \nabla_{\theta} (\theta^T X^T X \theta - \theta^T X^T y - y^T X \theta + y^T y) \\
&= \frac{1}{2} \nabla_{\theta} \text{tr}(\theta^T X^T X \theta - \theta^T X^T y - y^T X \theta + y^T y) \\
-- > \text{tr} A^T = \text{tr} A \Rightarrow \text{tr}(\theta^T X^T y) &= \text{tr}(y^T X \theta) \\
&= \frac{1}{2} \nabla_{\theta} (\text{tr} \theta^T X^T X \theta - 2 \text{tr}(y^T X \theta)) \\
-- > (\theta = A^T, B = X^T X, C = I) & \\
&= \frac{1}{2} \nabla_{A^T} (\text{tr}(A B A^T C)) - \frac{1}{2} \nabla_{\theta} 2 \text{tr}(y^T X \theta) \\
&= \frac{1}{2} (B^T A^T C^T + B A^T C) - X^T y \\
&= \frac{1}{2} (X^T X \theta + X^T X \theta) - X^T y \\
&= X^T X \theta - X^T y
\end{aligned} \tag{1}$$

于是令 $\nabla_{\theta} J(\theta) = 0$ 得到 $X^T X \theta = X^T y$, 于是:

$$\theta = (X^T X)^{-1} X^T y$$

上式即为 θ 的闭式解。