

1. 矢量运算基本性质

1. 加法及数乘

$$\vec{x} + \vec{y} = \vec{y} + \vec{x} \quad \alpha(\vec{x} + \vec{y}) = \alpha\vec{x} + \alpha\vec{y} \quad \vec{x} + \vec{y} + \vec{z} = \vec{x} + (\vec{y} + \vec{z}) \quad (\alpha + \beta)\vec{x} = \alpha\vec{x} + \beta\vec{x}$$

2. 内积

$$\vec{x} \cdot \vec{y} = \vec{x}^T \vec{y} = \vec{y}^T \vec{x} = \sum_{i=1}^n x_i y_i \quad \vec{x}, \vec{y} \in R^n$$

$$3. \text{欧氏距离: } ||\vec{x} - \vec{y}|| = \sqrt{\sum_{i=1}^n (x_i - y_i)^2}$$

4. 线性不相关

若 $\alpha_1 \vec{x}_1 + \alpha_2 \vec{x}_2 + \alpha_3 \vec{x}_3 + \dots + \alpha_n \vec{x}_n = \vec{0}$ 仅在 $\alpha_1 = \alpha_2 = \alpha_3 = \dots = \alpha_n = 0$ 时成立, 则 $\vec{x}_1, \vec{x}_2, \vec{x}_3, \dots, \vec{x}_n$ 时不相关

$$5. \text{正交: } \vec{x}^T \vec{y} = 0$$

6. 子空间 (U) 与矢量 (\vec{x}) 正交: 矢量和该子空间中的所有向量正交

立体几何: 向量和平面垂直

2. 矩阵的部分性质

1. 对称阵、对角阵、单位阵

- 对称阵: n阶方阵且元素满足 $a_{ij} = a_{ji}$
- 对角阵: n阶方阵且仅有对角线元素不为0
- 单位阵: n阶方阵, 且对角线元素均为1, 其余元素均为0

2. 特征根

$$\mathbf{A}\vec{x} = \lambda\vec{x} \Rightarrow \det(\mathbf{A} - \lambda\mathbf{I}) = 0 \quad \text{则该矩阵存在特征根和特征向量}$$

3. 矩阵的秩

$$\text{rank}(\mathbf{A}) = \text{rank}(\mathbf{A}^T) = \text{rank}(\mathbf{A}\mathbf{A}^T) = \text{rank}(\mathbf{A}^T\mathbf{A})$$

$$4. \text{奇异阵: } \det(\mathbf{A}) = 0 \Rightarrow \mathbf{A}^{-1} \text{ 不存在}$$

$$5. \text{迹 (trace): } \text{tr}(\mathbf{A}) = \sum_i a_{ii} \quad (\text{对角线元素的和}) \quad \text{tr}(\mathbf{AB}) = \text{tr}(\mathbf{BA})$$

$$6. \text{模: } ||\mathbf{A}|| = \sqrt{\text{tr}(\mathbf{A}^T\mathbf{A})}$$

$$7. (\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$$

3. 矢量及矩阵求导

3.1. 标量对矢量求导

1. 梯度

$$\frac{\partial g(\vec{\omega})}{\partial \vec{\omega}} = \begin{bmatrix} \frac{\partial g}{\partial \omega_1} \\ \frac{\partial g}{\partial \omega_2} \\ \vdots \\ \frac{\partial g}{\partial \omega_n} \end{bmatrix} \quad \frac{\partial g}{\partial \vec{\omega}} \in R^n \text{ 也可记作 } \nabla_{\omega} g$$

2. Hessian 矩阵

$$\frac{\partial^2 g}{\partial \vec{\omega}^2} = \begin{bmatrix} \frac{\partial^2 g}{\partial \omega_1^2} & \frac{\partial^2 g}{\partial \omega_1 \partial \omega_2} & \cdots & \frac{\partial^2 g}{\partial \omega_1 \partial \omega_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 g}{\partial \omega_n \partial \omega_1} & \frac{\partial^2 g}{\partial \omega_n \partial \omega_2} & \cdots & \frac{\partial^2 g}{\partial \omega_n^2} \end{bmatrix} \quad \text{也可记作 } \nabla_{\omega}^2 g$$

$$3. \frac{\partial f(\vec{w})g(\vec{w})}{\partial \vec{w}} = g(\vec{w}) \frac{\partial f(\vec{w})}{\partial \vec{w}} + f(\vec{w}) \frac{\partial g(\vec{w})}{\partial \vec{w}}$$

$$4. \frac{\partial \left(\frac{f(\vec{w})}{g(\vec{w})} \right)}{\partial \vec{w}} = \frac{g(\vec{w}) \frac{\partial f(\vec{w})}{\partial \vec{w}} - f(\vec{w}) \frac{\partial g(\vec{w})}{\partial \vec{w}}}{g^2(\vec{w})}$$

$$5. \frac{\partial (\mathbf{A} \vec{g}(\vec{w}))}{\partial \vec{w}} = \mathbf{A} \frac{\partial \vec{g}(\vec{w})}{\partial \vec{w}}$$

$$6. \frac{\partial [\vec{f}(\vec{w})^\top \vec{g}(\vec{w})]}{\partial \vec{w}} = \left(\frac{\partial \vec{f}(\vec{w})}{\partial \vec{w}} \right)^\top \vec{g}(\vec{w}) + \left(\frac{\partial \vec{g}(\vec{w})}{\partial \vec{w}} \right)^\top \vec{f}(\vec{w})$$

$$\text{例1: } g(\vec{\omega}) = \vec{\omega}^\top [A] \vec{\omega}, \text{ 则可知 } \frac{\partial g(\vec{\omega})}{\partial \vec{\omega}} = ([A] + [A]^\top) \vec{\omega}$$

$$\text{例2: } \frac{\partial \left(\vec{A}^\top \vec{\omega} \right)}{\partial \vec{\omega}} = \left(\vec{A}^\top \right)^\top = \vec{A}$$

3.2. 矩阵求导

$$\frac{\partial g([w])}{\partial [w]} = \begin{pmatrix} \frac{\partial g}{\partial w_{11}} & \cdots & \frac{\partial g}{\partial w_{1n}} \\ \vdots & \ddots & \vdots \\ \frac{\partial g}{\partial w_{m1}} & \cdots & \frac{\partial g}{\partial w_{mn}} \end{pmatrix} \quad [w] = \begin{pmatrix} w_{11} & \cdots & w_{1n} \\ \vdots & \ddots & \vdots \\ w_{m1} & \cdots & w_{mn} \end{pmatrix}$$

3.3. Taylor 展开

1. $g(w)$ 在 w_0 处展开

$$g(w) = g(w_0) + \left. \frac{\partial g}{\partial w} \right|_{w=w_0} (w - w_0) + \frac{1}{2} \left. \frac{\partial^2 g}{\partial w^2} \right|_{w=w_0} (w - w_0)^2 + \cdots$$

2. $g(\vec{w})$ 在 \vec{w}_0 处展开 (存疑)

$$g(\vec{w}) = g(\vec{w}_0) + \left(\frac{\partial g}{\partial \vec{w}} \right)^T \bigg|_{\vec{w}=\vec{w}_0} (\vec{w} - \vec{w}_0) + \frac{1}{2} (\vec{w} - \vec{w}_0)^T \left. \frac{\partial^2 g}{\partial \vec{w}^2} \right|_{\vec{w}=\vec{w}_0} (\vec{w} - \vec{w}_0) + \cdots$$

$$3. g([w]) = g([w_0]) + tr \left[\left(\frac{\partial g}{\partial [w]} \right)^T \bigg|_{[w]=[w_0]} ([w] - [w_0]) \right] + \cdots$$

3.4. 无约束条件的优化算法

训练集 $\vec{x}(t) = \{\vec{x}^1, \dots, \vec{x}^p, \dots, \vec{x}^N\}$, 对应教师 $\vec{d}(t) = \{d^1, \dots, d^p, \dots, d^N\}$

3.4.1. 线性

$$\text{目标函数 } g(\vec{w}) = \frac{1}{2} E \{ (d(t) - y(t))^2 \} = \frac{1}{2} E \{ (d(t) - \vec{w}^T \vec{x}(t))^2 \}$$

则可知

$$g(\vec{w}) = \frac{1}{2} E \{ d^2(t) \} - E \{ d(t)(\vec{x}(t))^T \} \vec{w} + \frac{1}{2} \vec{w}^T E \{ \vec{x}(t)(\vec{x}(t))^T \} \vec{w} = \frac{1}{2} E \{ d^2(t) \} - \vec{R}_{dx}^T \vec{w} + \frac{1}{2} \vec{w}^T [R_{xx}] \vec{w}$$

$$\text{求极值, } \frac{\partial g(\vec{w})}{\partial \vec{w}} = 0 \Rightarrow 0 - \vec{R}_{dx} + [R_{xx}] \vec{w} = 0 \Rightarrow \vec{w} = [R_{xx}]^{-1} \vec{R}_{dx}$$

3.4.2. 梯度法 (非线性, 只能求局部极小值)

$$g(\vec{w}) = \sum_{p=1}^N (d^p - y^p(\vec{w}))^2 \quad \text{使得目标函数最小}$$

$$\text{梯度调整公式: } \Delta \vec{w} = -\eta \frac{\partial g(\vec{w})}{\partial \vec{w}}, \eta > 0$$

将 $g(\vec{w})$ 在 $\vec{w} = \vec{w}(t-1)$ 处 Taylor 展开,

$$\begin{aligned} g(\vec{w}(t)) &= g(\vec{w}(t-1)) + \left(\frac{\partial g(\vec{w})}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)} (\vec{w}(t) - \vec{w}(t-1)) \\ \Rightarrow g(\vec{w}(t)) - g(\vec{w}(t-1)) &= \left(\frac{\partial g(\vec{w})}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)} (\vec{w}(t) - \vec{w}(t-1)) \end{aligned}$$

要求 $g(\vec{w}(t)) < g(\vec{w}(t-1))$ 时, 目标函数不断变小, 即误差不断的变小

$$\begin{aligned} \left(\frac{\partial g(\vec{w})}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)} < 0 &\Leftrightarrow \vec{w}(t) - \vec{w}(t-1) > 0 \\ \left(\frac{\partial g(\vec{w})}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)} > 0 &\Leftrightarrow \vec{w}(t) - \vec{w}(t-1) < 0 \end{aligned}$$

可调整如下, $\Delta \vec{w} = \vec{w}(t) - \vec{w}(t-1) = -\eta \left(\frac{\partial g(\vec{w})}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)}, \eta > 0$

若 $g(\vec{w}^*) = 0$ 时存在, \vec{w}^* 存在, 梯度下降法收敛于局部最小值

3.4.3. 牛顿法

将 $g(\vec{w})$ 在 $\vec{w}(t-1)$ 处 Taylor 展开如下,

$$g(\vec{w}(t)) - g(\vec{w}(t-1)) = \left(\frac{\partial g}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)} (\vec{w}(t) - \vec{w}(t-1)) + \frac{1}{2} (\vec{w}(t) - \vec{w}(t-1))^\top \frac{\partial^2 g}{\partial \vec{w}^2} (\vec{w}(t) - \vec{w}(t-1))$$

令 $\Delta \vec{w} = \vec{w}(t) - \vec{w}(t-1)$, 则可知

$$\Delta g(\vec{w}) = \left(\frac{\partial g}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)} \Delta \vec{w} + \frac{1}{2} \Delta \vec{w}^\top \mathbf{H} \bigg|_{\vec{w}=\vec{w}(t-1)} \Delta \vec{w}$$

要使得 $\Delta g(\vec{w})$ 极小, 则 $\frac{\partial(\Delta g(\vec{w}))}{\partial \vec{w}} = 0$, 可知

$$\left(\frac{\partial g}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)} + \mathbf{H}(t-1) \Delta \vec{w} = 0 \Rightarrow \Delta \vec{w} = -\mathbf{H}^{-1} \left(\frac{\partial g}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)}$$

为防止奇异, 则 $\Delta \vec{w} = -(\mathbf{H} + \delta)^{-1} \left(\frac{\partial g}{\partial \vec{w}} \right)^\top \bigg|_{\vec{w}=\vec{w}(t-1)}$

3.4.4. Gauss - Newton 法

目标函数: $g(\vec{w}) = \frac{1}{2} \sum_{p=1}^N (d^p - y^p(\vec{w}))^2 = \frac{1}{2} \vec{e}(\vec{w})^\top \vec{e}(\vec{w})$

其中 $\vec{e}(\vec{\omega}) = [d^1 - y^1(\vec{\omega}) \quad d^2 - y^2(\vec{\omega}) \quad \dots \quad d^N - y^N(\vec{\omega})]^\top$

将 $\vec{e}(\vec{\omega})$ 在 $\vec{\omega} = \vec{\omega}(t-1)$ 处 Taylor 展开,

$$\vec{e}(\vec{\omega}(t)) = \vec{e}(\vec{\omega}(t-1)) + \left. \frac{\partial \vec{e}(\vec{\omega})}{\partial \vec{\omega}} \right|_{\vec{\omega}=\vec{\omega}(t-1)} (\vec{\omega}(t) - \vec{\omega}(t-1))$$

记 $\mathbf{J}(\vec{\omega}) = \frac{\partial \vec{e}(\vec{\omega})}{\partial \vec{\omega}}$, 则

$$\mathbf{J}(\vec{\omega}(t-1)) = \left. \frac{\partial \vec{e}(\vec{\omega})}{\partial \vec{\omega}} \right|_{\vec{\omega}=\vec{\omega}(t-1)} \Rightarrow \vec{e}(\vec{\omega}(t)) = \vec{e}(\vec{\omega}(t-1)) + \mathbf{J}(\vec{\omega}(t-1))\Delta\vec{\omega}$$

欲求出 $\vec{\omega}^* = \operatorname{argmin}_{\vec{\omega}}(g(\vec{\omega}))$, 将上式代入 $g(\vec{\omega})$ 的表达式中可得

$$g(\vec{\omega}) = \frac{1}{2} \|\vec{e}(\vec{\omega}(t-1))\|^2 + \vec{e}^\top(\vec{\omega}(t-1))[\mathbf{J}(\vec{\omega}(t-1))\Delta\vec{\omega}] + \frac{1}{2}(\Delta\vec{\omega})^\top([\mathbf{J}(\vec{\omega}(t-1))]^\top[\mathbf{J}(\vec{\omega}(t-1))])\Delta\vec{\omega}$$

又 $g(\vec{\omega})$ 取极值时, $\frac{\partial g(\vec{\omega})}{\partial \vec{\omega}} = 0$, 则可知

$$\begin{aligned} & (\vec{e}^\top(\vec{\omega}(t-1))[\mathbf{J}(\vec{\omega}(t-1))]^\top + [\mathbf{J}(\vec{\omega}(t-1))]^\top[\mathbf{J}(\vec{\omega}(t-1))])\Delta\vec{\omega} = 0 \\ \therefore \Delta\vec{\omega} &= -[\mathbf{J}(\vec{\omega}(t-1))]^\top[\mathbf{J}(\vec{\omega}(t-1))]^{-1}[\mathbf{J}(\vec{\omega}(t-1))]^\top\vec{e}(\vec{\omega}(t-1)) \end{aligned}$$

3.4.5. 自然梯度法 (黎曼空间)

$$\Delta\vec{\omega} = -\eta \frac{\partial g(\vec{\omega})}{\partial \vec{\omega}} \vec{\omega}^\top \vec{\omega} \quad \eta > 0$$