

矩阵微积分

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定义1: 若 $f: R \rightarrow R^m$, 即, $f(x) = \begin{bmatrix} f_1(x) \\ f_2(x) \\ \vdots \\ f_m(x) \end{bmatrix}$, 则 $\frac{\partial f}{\partial x} = \begin{bmatrix} \frac{\partial f_1}{\partial x} \\ \frac{\partial f_2}{\partial x} \\ \vdots \\ \frac{\partial f_m}{\partial x} \end{bmatrix}$

定义2: 若 $f: R^n \rightarrow R$, 即, $f(x) = f(x_1, x_2, \dots, x_n)$, 则 $\frac{\partial f}{\partial x} = \begin{bmatrix} \frac{\partial f}{\partial x_1} & \frac{\partial f}{\partial x_2} & \dots & \frac{\partial f}{\partial x_n} \end{bmatrix}$

定义3: 若 $f: R^n \rightarrow R^m$, 即, $f(x) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \vdots \\ f_m(x_1, x_2, \dots, x_n) \end{bmatrix}$, 则 $\frac{\partial f}{\partial x} =$

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \dots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}$$

一些常见的例子:

例1: 若 $z = Wx, W \in R^{m \times n}$, 则:

$$\begin{aligned} \frac{\partial z}{\partial x} &= \begin{bmatrix} \frac{\partial z_1}{\partial x_1} & \frac{\partial z_1}{\partial x_2} & \dots & \frac{\partial z_1}{\partial x_n} \\ \frac{\partial z_2}{\partial x_1} & \frac{\partial z_2}{\partial x_2} & \dots & \frac{\partial z_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial z_m}{\partial x_1} & \frac{\partial z_m}{\partial x_2} & \dots & \frac{\partial z_m}{\partial x_n} \end{bmatrix} & \text{(根据定义3)} \\ &= W & (\frac{\partial z_i}{\partial x_j} = W_{ij}) \end{aligned}$$

例2: 若 $z = f(x), x \in R^n$, 其中 $z_i = f(x_i)$, 则:

$$\frac{\partial z}{\partial x} = \begin{bmatrix} f'(x_1) & 0 & \dots & 0 \\ 0 & f'(x_2) & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & f'(x_n) \end{bmatrix} \quad \text{(根据定义3)}$$

例3: 若 $z = g(y), y = f(x), x \in R^n, y \in R^p, z \in R^m$, 则:

$$\frac{\partial z}{\partial x} = \begin{bmatrix} \frac{\partial z_1}{\partial y} \frac{\partial y}{\partial x_1} & \frac{\partial z_1}{\partial y} \frac{\partial y}{\partial x_2} & \cdots & \frac{\partial z_1}{\partial y} \frac{\partial y}{\partial x_n} \\ \frac{\partial z_2}{\partial y} \frac{\partial y}{\partial x_1} & \frac{\partial z_2}{\partial y} \frac{\partial y}{\partial x_2} & \cdots & \frac{\partial z_2}{\partial y} \frac{\partial y}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial z_m}{\partial y} \frac{\partial y}{\partial x_1} & \frac{\partial z_m}{\partial y} \frac{\partial y}{\partial x_2} & \cdots & \frac{\partial z_m}{\partial y} \frac{\partial y}{\partial x_n} \end{bmatrix} \quad (\text{根据定义3})$$

$$= \frac{\partial z}{\partial y} \frac{\partial y}{\partial x}$$

例4: 若 $\frac{\partial J}{\partial z} = \delta, z = Wx, W \in R^{m \times n}, J \in R$, 则:

$$\frac{\partial J}{\partial W} = \begin{bmatrix} \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{11}} & \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{12}} & \cdots & \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{1n}} \\ \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{21}} & \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{22}} & \cdots & \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{2n}} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{m1}} & \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{m2}} & \cdots & \frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{mn}} \end{bmatrix}$$

其中,

$$\frac{\partial z_k}{\partial W_{ij}} = \begin{cases} x_j, & \text{若 } k = i \\ 0, & \text{其它} \end{cases}$$

于是,

$$\frac{\partial J}{\partial z} \frac{\partial z}{\partial W_{ij}} = \delta_i x_j$$

综上, $\frac{\partial J}{\partial W} = \delta^T x^T$

例5, 人工神经网络:

$$\begin{aligned} x & \\ z &= Wx + b_1 \\ h &= ReLU(z) \\ r &= Uh + b_2 \\ \hat{y} &= Softmax(r) \\ J &= CE(y - \hat{y}) \end{aligned}$$

已知 $\frac{\partial J}{\partial r} = \delta_1$, 则:

$$\frac{\partial J}{\partial U} = \delta_1^T h^T$$

$$\frac{\partial J}{\partial h} = \frac{\partial J}{\partial r} \frac{\partial r}{\partial h} = \delta_1 U$$

$$\frac{\partial J}{\partial b_2} = \frac{\partial J}{\partial r} \frac{\partial r}{\partial b_2} = \delta_1$$

$$\frac{\partial J}{\partial z} = \frac{\partial J}{\partial h} \frac{\partial h}{\partial z} = \delta_1 U \circ ReLU'(z) = \delta_2$$

$$\frac{\partial J}{\partial W} = \delta_2^T x^T$$

$$\frac{\partial J}{\partial b_1} = \frac{\partial J}{\partial z} \frac{\partial z}{\partial b_1} = \delta_2$$

(根据例4)