

# Matrix Differentiation [Detailed Derivation]

## Variables

- $A$  - Input Matrix

$$A = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} \\ a_{21} & a_{22} & \dots & a_{2j} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ij} \end{bmatrix}$$

- $W$  - Weight Matrix

$$W = \begin{bmatrix} w_{11} \\ w_{21} \\ \vdots \\ w_{j1} \end{bmatrix}$$

- $\hat{Y}$  -  $Y$  predict

$$\hat{Y} = A \cdot W$$

$$\hat{Y} = \begin{bmatrix} \hat{y}_{11} \\ \hat{y}_{21} \\ \vdots \\ \hat{y}_{i1} \end{bmatrix}$$

$$\hat{Y} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} \\ a_{21} & a_{22} & \dots & a_{2j} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ij} \end{bmatrix} \cdot \begin{bmatrix} w_{11} \\ w_{21} \\ \vdots \\ w_{j1} \end{bmatrix}$$

- $Y$  -  $Y$  true

$$Y = \begin{bmatrix} y_{11} \\ y_{21} \\ \vdots \\ y_{i1} \end{bmatrix}$$

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Expanding  $\hat{Y}$

$$\hat{Y} = \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} \\ a_{21} & a_{22} & \dots & a_{2j} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ij} \end{bmatrix} \cdot \begin{bmatrix} w_{11} \\ w_{21} \\ \vdots \\ w_{j1} \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}w_{11} + a_{12}w_{21} + \dots + a_{1j}w_{j1} \\ a_{21}w_{11} + a_{22}w_{21} + \dots + a_{2j}w_{j1} \\ \vdots \\ a_{i1}w_{11} + a_{i2}w_{21} + \dots + a_{ij}w_{j1} \end{bmatrix}$$


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loss

$$loss = \frac{1}{N} \sum (Y - \hat{Y})^2$$

$$= \frac{1}{N} \begin{pmatrix} (y_{11} - a_{11}w_{11} - a_{12}w_{21} - \dots - a_{1j}w_{j1})^2 \\ + (y_{21} - a_{21}w_{11} - a_{22}w_{21} - \dots - a_{2j}w_{j1})^2 \\ \vdots \\ + (y_{i1} - a_{i1}w_{11} - a_{i2}w_{21} - \dots - a_{ij}w_{j1})^2 \end{pmatrix}$$

Here  $N$  is number of rows in input matrix  $A$

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Finding  $\delta W$

$$\frac{d(loss)}{d(w_{j1})} = \frac{1}{N} \frac{d}{d(w_{j1})} \begin{pmatrix} (y_{11} - a_{11}w_{11} - a_{12}w_{21} - \dots - a_{1j}w_{j1})^2 \\ + (y_{21} - a_{21}w_{11} - a_{22}w_{21} - \dots - a_{2j}w_{j1})^2 \\ \vdots \\ + (y_{i1} - a_{i1}w_{11} - a_{i2}w_{21} - \dots - a_{ij}w_{j1})^2 \end{pmatrix}$$

$$\frac{d(loss)}{d(w_{j1})} = \frac{1}{N} \begin{pmatrix} \frac{d}{d(w_{j1})} (y_{11} - a_{11}w_{11} - a_{12}w_{21} - \dots - a_{1j}w_{j1})^2 \\ + \frac{d}{d(w_{j1})} (y_{21} - a_{21}w_{11} - a_{22}w_{21} - \dots - a_{2j}w_{j1})^2 \\ \vdots \\ + \frac{d}{d(w_{j1})} (y_{i1} - a_{i1}w_{11} - a_{i2}w_{21} - \dots - a_{ij}w_{j1})^2 \end{pmatrix}$$

$$\frac{d(loss)}{d(w_{j1})} = \frac{1}{N} \begin{pmatrix} 2(y_{11} - a_{11}w_{11} - a_{12}w_{21} - \dots - a_{1j}w_{j1})(-a_{1j}) \\ + 2(y_{21} - a_{21}w_{11} - a_{22}w_{21} - \dots - a_{2j}w_{j1})(-a_{2j}) \\ \vdots \\ + 2(y_{i1} - a_{i1}w_{11} - a_{i2}w_{21} - \dots - a_{ij}w_{j1})(-a_{ij}) \end{pmatrix}$$

$$\frac{d(loss)}{d(w_{j1})} = -\frac{2}{N} \begin{pmatrix} (y_{11} - a_{11}w_{11} - a_{12}w_{21} - \dots - a_{1j}w_{j1})(a_{1j}) \\ +(y_{21} - a_{21}w_{11} - a_{22}w_{21} - \dots - a_{2j}w_{j1})(a_{2j}) \\ \vdots \\ +(y_{i1} - a_{i1}w_{11} - a_{i2}w_{21} - \dots - a_{ij}w_{j1})(a_{ij}) \end{pmatrix}$$

$$\frac{d(loss)}{d(w_{j1})} = -\frac{2}{N} [a_{1j} \quad a_{2j} \quad \dots \quad a_{ij}] \cdot \begin{bmatrix} (y_{11} - a_{11}w_{11} - a_{12}w_{21} - \dots - a_{1j}w_{j1}) \\ (y_{21} - a_{21}w_{11} - a_{22}w_{21} - \dots - a_{2j}w_{j1}) \\ \vdots \\ (y_{i1} - a_{i1}w_{11} - a_{i2}w_{21} - \dots - a_{ij}w_{j1}) \end{bmatrix}$$

$$\frac{d(loss)}{d(w_{j1})} = -\frac{2}{N} \begin{bmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{ij} \end{bmatrix}^T \cdot \begin{bmatrix} (y_{11} - a_{11}w_{11} - a_{12}w_{21} - \dots - a_{1j}w_{j1}) \\ (y_{21} - a_{21}w_{11} - a_{22}w_{21} - \dots - a_{2j}w_{j1}) \\ \vdots \\ (y_{i1} - a_{i1}w_{11} - a_{i2}w_{21} - \dots - a_{ij}w_{j1}) \end{bmatrix}$$

$$\frac{d(loss)}{d(w_{j1})} = -\frac{2}{N} \begin{bmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{ij} \end{bmatrix}^T \cdot \left[ \begin{bmatrix} y_{11} \\ y_{21} \\ \vdots \\ y_{i1} \end{bmatrix} - \begin{bmatrix} a_{11}w_{11} + a_{12}w_{21} + \dots + a_{1j}w_{j1} \\ a_{21}w_{11} + a_{22}w_{21} + \dots + a_{2j}w_{j1} \\ \vdots \\ a_{i1}w_{11} + a_{i2}w_{21} + \dots + a_{ij}w_{j1} \end{bmatrix} \right]$$

$$\frac{d(loss)}{d(w_{j1})} = -\frac{2}{N} \begin{bmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{ij} \end{bmatrix}^T \cdot \left[ \begin{bmatrix} y_{11} \\ y_{21} \\ \vdots \\ y_{i1} \end{bmatrix} - \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} \\ a_{21} & a_{22} & \dots & a_{2j} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ij} \end{bmatrix} \cdot \begin{bmatrix} w_{11} \\ w_{21} \\ \vdots \\ w_{j1} \end{bmatrix} \right]$$

$$\delta W = -\frac{2}{N} \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} \\ a_{21} & a_{22} & \dots & a_{2j} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ij} \end{bmatrix}^T \cdot \left[ \begin{bmatrix} y_{11} \\ y_{21} \\ \vdots \\ y_{i1} \end{bmatrix} - \begin{bmatrix} a_{11} & a_{12} & \dots & a_{1j} \\ a_{21} & a_{22} & \dots & a_{2j} \\ \vdots & \vdots & \ddots & \vdots \\ a_{i1} & a_{i2} & \dots & a_{ij} \end{bmatrix} \cdot \begin{bmatrix} w_{11} \\ w_{21} \\ \vdots \\ w_{j1} \end{bmatrix} \right]$$

$$\mathbf{grad:} \delta W = -\frac{2}{N} A^T \cdot (Y - A \cdot W)$$


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