

Gradients of Matrices

■ Useful identities:

$$\frac{\partial}{\partial \mathbf{X}} f(\mathbf{X})^\top = \left(\frac{\partial f(\mathbf{X})}{\partial \mathbf{X}} \right)^\top \quad (5.99)$$

$$\frac{\partial}{\partial \mathbf{X}} \text{tr}(f(\mathbf{X})) = \text{tr} \left(\frac{\partial f(\mathbf{X})}{\partial \mathbf{X}} \right) \quad (5.100)$$

$$\frac{\partial}{\partial \mathbf{X}} \det(f(\mathbf{X})) = \det(f(\mathbf{X})) \text{tr} \left(f(\mathbf{X})^{-1} \frac{\partial f(\mathbf{X})}{\partial \mathbf{X}} \right) \quad (5.101)$$

$$\frac{\partial}{\partial \mathbf{X}} f(\mathbf{X})^{-1} = -f(\mathbf{X})^{-1} \frac{\partial f(\mathbf{X})}{\partial \mathbf{X}} f(\mathbf{X})^{-1} \quad (5.102)$$

$$\frac{\partial a^\top \mathbf{X}^{-1} b}{\partial \mathbf{X}} = -(\mathbf{X}^{-1})^\top a b^\top (\mathbf{X}^{-1})^\top \quad (5.103)$$

$$\frac{\partial x^\top a}{\partial x} = a^\top \quad (5.104)$$

$$\frac{\partial a^\top x}{\partial x} = a^\top \quad (5.105)$$

$$\frac{\partial a^\top \mathbf{X} b}{\partial \mathbf{X}} = a b^\top \quad (5.106)$$

$$\frac{\partial x^\top B x}{\partial x} = x^\top (B + B^\top) \quad (5.107)$$

$$\frac{\partial}{\partial \mathbf{s}} (x - \mathbf{A} \mathbf{s})^\top W (x - \mathbf{A} \mathbf{s}) = -2(x - \mathbf{A} \mathbf{s})^\top W \mathbf{A} \quad \text{for symmetric } W \quad (5.108)$$

➔ **Note:**

Trace of a $D \times D \times E \times F$ tensor $\in \mathbb{R}^{E \times F}$

➔ $\mathbf{X} \in \mathbb{R}^{m \times n}$, $\mathbf{f}(\mathbf{X}) \in \mathbb{R}^{p \times p}$

Left term: $\frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X})^{-1} \in \mathbb{R}^{p \times p \times m \times n}$

Right term:

$$\mathbf{f}(\mathbf{X})^{-1} \in \mathbb{R}^{p \times p}$$

$$\frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} \in \mathbb{R}^{p \times p \times m \times n}$$

For x_{ij} , $-\mathbf{f}(\mathbf{X})^{-1} \frac{\partial \mathbf{f}(\mathbf{X})}{\partial x_{ij}} \mathbf{f}(\mathbf{X})^{-1} \in \mathbb{R}^{p \times p}$

➔ $-\mathbf{f}(\mathbf{X})^{-1} \frac{\partial \mathbf{f}(\mathbf{X})}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X})^{-1} \in \mathbb{R}^{p \times p \times m \times n}$

Gradients of Matrices

- Clarification of some identities:

- $$\frac{\partial}{\partial \mathbf{X}} f(\mathbf{X})^{-1} = -f(\mathbf{X})^{-1} \frac{\partial f(\mathbf{X})}{\partial \mathbf{X}} f(\mathbf{X})^{-1} \quad (5.102)$$

- $$\mathbf{0} = \frac{\partial \mathbf{I}}{\partial \mathbf{X}} = \frac{\partial}{\partial \mathbf{X}} (\mathbf{f}(\mathbf{X})^{-1} \mathbf{f}(\mathbf{X})) = \left(\frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X})^{-1} \right) \mathbf{f}(\mathbf{X}) + \mathbf{f}(\mathbf{X})^{-1} \left(\frac{\partial}{\partial \mathbf{X}} \mathbf{f}(\mathbf{X}) \right)$$

- $$\frac{\partial}{\partial s} (x - \mathbf{A}\mathbf{s})^T \mathbf{W} (x - \mathbf{A}\mathbf{s}) = -2(x - \mathbf{A}\mathbf{s})^T \mathbf{W} \mathbf{A} \quad \text{for symmetric } \mathbf{W} \quad (5.108)$$

- Let $\mathbf{z} = \mathbf{x} - \mathbf{A}\mathbf{s}$, and use $\frac{\partial x^T \mathbf{B} x}{\partial x} = x^T (\mathbf{B} + \mathbf{B}^T)$:

- $\frac{\partial \mathbf{z}^T \mathbf{W} \mathbf{z}}{\partial \mathbf{z}} = \mathbf{z}^T (\mathbf{W} + \mathbf{W}^T) = 2(\mathbf{x} - \mathbf{A}\mathbf{s})^T \mathbf{W}$

- Then compute $\frac{\partial \mathbf{z}^T \mathbf{W} \mathbf{z}}{\partial \mathbf{z}} \frac{\partial \mathbf{z}}{\partial s} = 2(\mathbf{x} - \mathbf{A}\mathbf{s})^T \mathbf{W} \cdot (-\mathbf{A})$