#### Theory of Matrices

#### Chapter 4. Vector/Matrix Derivatives and Integrals

4.2. Types of Differentiation—Proofs of Theorems



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Theorem 4.2.1

#### Theorem 4.2.1

**Theorem 4.2.1.** Differentiation of scalar valued function f satisfies the following.

$$(1) \ \frac{\partial f}{\partial X^T} = \left(\frac{\partial f}{\partial X}\right)^T.$$

- (2) For X square and  $f(X) = \operatorname{tr}(X)$ ,  $\frac{\partial f}{\partial X} = \mathcal{I}$ .
- (3) For AX a square matrix where A is constant,  $\frac{\partial [\operatorname{tr}(AX)]}{\partial X} = A^{T}.$
- (4)  $\frac{\partial [\operatorname{tr}(X^TX)]}{\partial X} = 2X.$
- (5) With a and b constant vectors,  $\frac{\partial [a^T X b]}{\partial X} = ab^T$ .
- (6)  $\frac{\partial [\det(X)]}{\partial X} = (\operatorname{adj}(X))^T$ .

### Theorem 4.2.1 (continued 1)

$$(1) \ \frac{\partial f}{\partial X^T} = \left(\frac{\partial f}{\partial X}\right)^T.$$

(2) For X square and  $f(X) = \operatorname{tr}(X)$ ,  $\frac{\partial f}{\partial X} = \mathcal{I}$ .

**Proof.** (1) From the definition, with  $X^T = [x_{ij}^T]$  where  $x_{ij}^T = x_{ji}$ , we have

$$\frac{\partial f}{\partial X^T} = \left[\frac{\partial f}{\partial x_{ij}^T}\right] = \left[\frac{\partial f}{\partial x_{ij}}\right]^T = \left(\frac{\partial f}{\partial X}\right)^T.$$

(2) With X square and  $f(X) = \operatorname{tr}(X) = \sum_{k=1}^{n} x_{kk}$  we have

$$\frac{\partial f}{\partial X} = \left[ \frac{\partial [\sum x_{kk}]}{\partial x_{ij}} \right] = \left[ \sum_{k=1}^{n} \frac{\partial x_{kk}}{\partial x_{ij}} \right] = \mathcal{I}$$

since  $\frac{\partial x_{kk}}{\partial x_{ij}} = \begin{cases} 1 & \text{if } (i,j) = (k,k) \\ 0 & \text{if } (i,j) \neq (k,k) \end{cases}$  for  $k = 1, 2, \dots, n$ ; that is,  $\frac{\partial [\text{tr}(X)]}{\partial X} = \mathcal{I}$ .

### Theorem 4.2.1 (continued 1)

$$(1) \frac{\partial f}{\partial X^T} = \left(\frac{\partial f}{\partial X}\right)^T.$$

(2) For X square and  $f(X) = \operatorname{tr}(X)$ ,  $\frac{\partial f}{\partial X} = \mathcal{I}$ .

**Proof.** (1) From the definition, with  $X^T = [x_{ij}^T]$  where  $x_{ij}^T = x_{ji}$ , we have

$$\frac{\partial f}{\partial X^T} = \left[\frac{\partial f}{\partial x_{ij}^T}\right] = \left[\frac{\partial f}{\partial x_{ij}}\right]^T = \left(\frac{\partial f}{\partial X}\right)^T.$$

(2) With X square and  $f(X) = \operatorname{tr}(X) = \sum_{k=1}^{n} x_{kk}$  we have

$$\frac{\partial f}{\partial X} = \left[ \frac{\partial [\sum x_{kk}]}{\partial x_{ij}} \right] = \left[ \sum_{k=1}^{n} \frac{\partial x_{kk}}{\partial x_{ij}} \right] = \mathcal{I}$$

since 
$$\frac{\partial x_{kk}}{\partial x_{ij}} = \begin{cases} 1 & \text{if } (i,j) = (k,k) \\ 0 & \text{if } (i,j) \neq (k,k) \end{cases}$$
 for  $k = 1, 2, \dots, n$ ; that is,  $\frac{\partial [\operatorname{tr}(X)]}{\partial X} = \mathcal{I}$ .

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## Theorem 4.2.1 (continued 2)

(3) For AX a square matrix where A is constant,  $\frac{\partial [\operatorname{tr}(AX)]}{\partial X} = A^{T}.$ 

**Proof.** (3) For AX a square matrix where A is constant, the diagonal entries are  $\sum_{\ell=1}^{n} a_{k\ell} x_{\ell k}$  for  $k=1,2,\ldots,n$ . So

$$\frac{\partial [\operatorname{tr}(AX)]}{\partial X} = \left[ \frac{\partial}{\partial x_{ij}} \left[ \sum_{k=1}^{n} \sum_{\ell=1}^{n} a_{k\ell} x_{\ell k} \right] \right] = [a_{ji}] = A^{T}$$

since 
$$\frac{\partial}{\partial x_{ij}} \left[ \sum_{k=1}^{n} \sum_{\ell=1}^{n} a_{k\ell} x_{\ell k} \right] = \sum_{k=1}^{n} \sum_{\ell=1}^{n} a_{k\ell} \frac{\partial x_{\ell k}}{\partial x_{ij}}$$
 and  $\frac{\partial x_{\ell k}}{\partial x_{ij}} = \begin{cases} 1 & \text{if } (\ell, k) = (i, j) \\ 0 & \text{if } (\ell, k) \neq (i, j) \end{cases}$ .

## Theorem 4.2.1 (continued 3)

(4) 
$$\frac{\partial [\operatorname{tr}(X^TX)]}{\partial X} = 2X.$$

**Proof.** (4) In  $X^TX$ , the diagonal entries are  $\sum_{\ell=1}^n x_{\ell k} x_{\ell k} = \sum_{\ell=1}^n (x_{\ell k})^2$  and so

$$\frac{\partial [\operatorname{tr}(X^T X)]}{\partial X} = \left[ \frac{\partial}{\partial x_{ij}} \left[ \sum_{k=1}^n \sum_{\ell=1}^n x_{\ell k} \right)^2 \right] = [2x_{ij}] = 2X$$

since 
$$\frac{\partial x_{\ell k}}{\partial x_{ij}} = \begin{cases} 1 & \text{if } (\ell, k) = (i, j) \\ 0 & \text{if } (\ell, k) \neq (i, j) \end{cases}$$
.

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### Theorem 4.2.1 (continued 4)

(5) With a and b constant vectors,  $\frac{\partial [a^T X b]}{\partial X} = ab^T$ .

**Proof.** (5) With  $a = [a_1, a_2, ..., a_n]$  and  $b = [b_1, b_2, ..., b_n]$  constant vectors, the matrix  $a^T X b$  is a  $1 \times 1$  matrix. Now the kth entry of  $a^T X$  is  $\sum_{\ell=1}^n a_\ell x_{\ell k}$  and so

$$a^T X b = \left[ \sum_{k=1}^n \left( \sum_{\ell=1}^n a_\ell x_{\ell k} \right) b_k \right].$$

Therefore  $\frac{\partial [a^T X b]}{\partial x_{ij}} = a_i b_j$  since  $\frac{\partial x_{\ell k}}{\partial x_{ij}} = \begin{cases} 1 & \text{if } (\ell, k) = (i, j) \\ 0 & \text{if } (\ell, k) \neq (i, j) \end{cases}$ . So  $\frac{\partial [a^T X b]}{\partial X} = [a_i b_j] = a b^T$ .

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## Theorem 4.2.1 (continued 4)

(5) With a and b constant vectors,  $\frac{\partial [a^T X b]}{\partial X} = ab^T$ .

**Proof.** (5) With  $a = [a_1, a_2, \dots, a_n]$  and  $b = [b_1, b_2, \dots b_n]$  constant vectors, the matrix  $a^T X b$  is a  $1 \times 1$  matrix. Now the kth entry of  $a^T X$  is  $\sum_{\ell=1}^n a_\ell x_{\ell k}$  and so

$$a^T X b = \left[ \sum_{k=1}^n \left( \sum_{\ell=1}^n a_\ell x_{\ell k} \right) b_k \right].$$

Therefore  $\frac{\partial [a^T X b]}{\partial x_{ij}} = a_i b_j$  since  $\frac{\partial x_{\ell k}}{\partial x_{ij}} = \begin{cases} 1 & \text{if } (\ell, k) = (i, j) \\ 0 & \text{if } (\ell, k) \neq (i, j) \end{cases}$ . So  $\frac{\partial [a^T X b]}{\partial X} = [a_i b_i] = a b^T$ .

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# Theorem 4.2.1 (continued 5)

(6) 
$$\frac{\partial [\det(X)]}{\partial X} = (\operatorname{adj}(X))^T$$
.

**Proof.** (6) By Theorem 3.1.F, we can express  $\det(A)$  by expansion along row k as  $\det(X) = \sum_{k=1}^{n} x_{kj} \chi_{kj}$  where  $\chi_{kj}$  is the cofactor of  $x_{kj}$ . None of the cofactors  $\chi_{kj}$  involve  $x_{ij}$  so the only occurrence of  $x_{ij}$  in this representation of  $\det(X)$  is when k = i in the term  $x_{ij} \chi_{ij}$ . Hence,  $\frac{\partial [\det(X)]}{\partial x_{ij}} = \chi_{ij}$  and  $\frac{\partial [\det(X)]}{\partial X} = [\chi_{ij}]$ . Recall that the adjoint of X is  $\det(X) = [\chi_{ij}]^T$  and so  $\frac{\partial [\det(X)]}{\partial X} = [\chi_{ij}] = (\det(X))^T$ .

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## Theorem 4.2.1 (continued 5)

(6) 
$$\frac{\partial [\det(X)]}{\partial X} = (\operatorname{adj}(X))^T$$
.

**Proof.** (6) By Theorem 3.1.F, we can express  $\det(A)$  by expansion along row k as  $\det(X) = \sum_{k=1}^{n} x_{kj} \chi_{kj}$  where  $\chi_{kj}$  is the cofactor of  $x_{kj}$ . None of the cofactors  $\chi_{kj}$  involve  $x_{ij}$  so the only occurrence of  $x_{ij}$  in this representation of  $\det(X)$  is when k=i in the term  $x_{ij}\chi_{ij}$ . Hence,  $\frac{\partial [\det(X)]}{\partial x_{ij}} = \chi_{ij}$  and  $\frac{\partial [\det(X)]}{\partial X} = [\chi_{ij}]$ . Recall that the adjoint of X is

$$\operatorname{adj}(X) = [\chi_{ij}]^T \text{ and so } \frac{\partial [\det(X)]}{\partial X} = [\chi_{ij}] = (\operatorname{adj}(X))^T.$$

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