

# Assignment 2

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October 7, 2023

1  $\frac{\partial \hat{y}}{\partial \hat{x}}$ , given that  $\hat{x}$  is a function of  $\hat{z}$

*Proof.* Let  $x = (k \times 1)$  column vector such that  $\frac{\partial}{\partial x} = \begin{pmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \vdots \\ \frac{\partial}{\partial x_k} \end{pmatrix}$ .

$$\frac{\partial \hat{y}}{\partial \hat{x}} = \begin{pmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \dots \\ \frac{\partial}{\partial x_k} \end{pmatrix}_{1 \times m} \begin{pmatrix} y_{x_n} & y_{x_n} & \vdots & y_{x_n} \end{pmatrix}_{n \times 1}.$$

Then by properties of the Kronocker product, denoted:

$$\begin{pmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \vdots \\ \frac{\partial}{\partial x_k} \end{pmatrix} \otimes \begin{pmatrix} y_{x_n} & y_{x_n} & \vdots & y_{x_n} \end{pmatrix}.$$

which results in a  $(n \times m)$  jacobian matrix:

$$\begin{pmatrix} x_1 y_1' & x_2 y_1' & \dots & x_n y_1' \\ x_1 y_2' & x_2 y_2' & \dots & x_n y_2' \\ \vdots & & \ddots & \\ x_1 y_n' & x_2 y_n' & \dots & x_n y_n' \end{pmatrix}$$

Let this  $(m \times n)$  matrix be  $A$ .

Then, because  $\hat{x}$  was a function of  $\hat{z}$ , by properties of the chain rule wrt to matrix operations, we multiply  $A$  by the derivative of  $\hat{x}$  wrt  $\hat{z}$

$$\therefore \frac{\partial \hat{y}}{\partial \hat{x}} = A \frac{\partial \hat{x}}{\partial \hat{z}}$$

□

## 2 $\frac{\partial \alpha}{\partial \hat{x}}$ given the scalar matrix $\alpha = \hat{y}^T A \hat{x}$ .

It is sufficient to show the case of  $\frac{\partial \alpha}{\partial \hat{x}}$  as the case for  $\frac{\partial \alpha}{\partial \hat{y}}$  follows closely:

*Proof.* Replace  $\alpha = \hat{y}^T A \hat{x} \implies \frac{\partial (\hat{y}^T A \hat{x})}{\partial \hat{x}}$ , and let A be a constant (mxn) constant matrix:

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & \ddots & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix}$$

let  $\hat{y}^T$  be a (1xm) row vector,  $\begin{pmatrix} y_1 & y_2 & \dots & y_n \end{pmatrix}$  and let  $\hat{x}$  be (kx1) column vector, such that  $\frac{\partial}{\partial \hat{x}} = \begin{pmatrix} \frac{\partial}{\partial x_1} & \frac{\partial}{\partial x_2} & \dots & \frac{\partial}{\partial x_k} \end{pmatrix}$

Then by the product rule wrt matrix operations,

$$\begin{aligned} & \frac{\partial (\hat{y}^T A \hat{x})}{\partial \hat{x}} \\ &= (y^T \frac{\partial \hat{x}}{\partial \hat{x}} + x^T \frac{\partial \hat{y}^T}{\partial \hat{x}}) A \end{aligned}$$

$\frac{\partial \hat{x}}{\partial \hat{x}}$  results in the Identity Matrix,  $I$  and  $\frac{\partial \hat{y}^T}{\partial \hat{x}} = 0$  given by,

$$\begin{pmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \vdots \\ \frac{\partial}{\partial x_k} \end{pmatrix} \begin{pmatrix} y_1 & y_2 & \dots & y_n \end{pmatrix} = 0$$

Therefore our result gives,

$$\begin{aligned} & \frac{\partial \alpha}{\partial \hat{x}} = y^T A \\ &= \begin{pmatrix} y_1 & y_2 & \dots & y_n \end{pmatrix} \otimes \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & \ddots & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \end{aligned}$$

□

### 3 $\frac{\partial \alpha}{\partial \hat{x}}$ given $\alpha = \hat{x}^T A \hat{x}$

*Proof.* Given  $\alpha = \hat{x}^T A \hat{x}$ , A is a constant (mxn) matrix and x is a (kx1) column vector such that

$\frac{\partial}{\partial x} = \left( \frac{\partial}{\partial x_1} \quad \frac{\partial}{\partial x_2} \quad \dots \quad \frac{\partial}{\partial x_k} \right)$  and  $x^T$  is a (1xk) row vector.

Then by properties of the product rule wrt to matrix operations, we have:  $\frac{\partial x^T}{\partial x} A + x^T A \frac{\partial \hat{x}}{\partial \hat{x}}$

$$\begin{aligned} & \left( \begin{pmatrix} x_1 & x_2 & \dots & x_k \end{pmatrix} \begin{pmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \dots \\ \frac{\partial}{\partial x_k} \end{pmatrix} \right) \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & \ddots & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \\ & + \begin{pmatrix} x_1 & x_2 & \dots & x_k \end{pmatrix} \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & & \ddots & \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix} \left( \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_k \end{pmatrix} \begin{pmatrix} \frac{\partial}{\partial x_1} \\ \frac{\partial}{\partial x_2} \\ \vdots \\ \frac{\partial}{\partial x_k} \end{pmatrix} \right) \end{aligned}$$

gives,

$$\begin{aligned} & \begin{pmatrix} x'_1 & 0 & 0 & \dots \\ 0 & x'_2 & 0 & \dots \\ \vdots & & \ddots & \\ 0 & 0 & \dots & x'_k \end{pmatrix} \otimes \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & & \ddots & \\ a_{m1} & a_{m2} & \dots & a_{mn} \end{pmatrix} \\ & \oplus \begin{pmatrix} x_1 & x_2 & \dots & x_k \end{pmatrix} \otimes \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1m} \\ a_{21} & a_{22} & \dots & a_{2m} \\ \vdots & & \ddots & \\ a_{n1} & a_{n2} & \dots & a_{nm} \end{pmatrix} \begin{pmatrix} I \end{pmatrix} \end{aligned}$$

without loss of generality, factoring  $x^T$  gives  $x^T(A + A^T)$

Therefore  $\frac{\partial \alpha}{\partial \hat{x}} = x^T$  gives  $x^T(A + A^T)$

In the case that  $A=A^T$  we see that  $x^T(A + A^T)$  gives  $2x^T A$ . □