

# EECS 16B    Designing Information Devices and Systems II

## Spring 2021    Discussion Worksheet

# Discussion 13A

This discussion will recap a lot of the key concepts covered in [lecture last week](#).

### 1. Linear Approximation

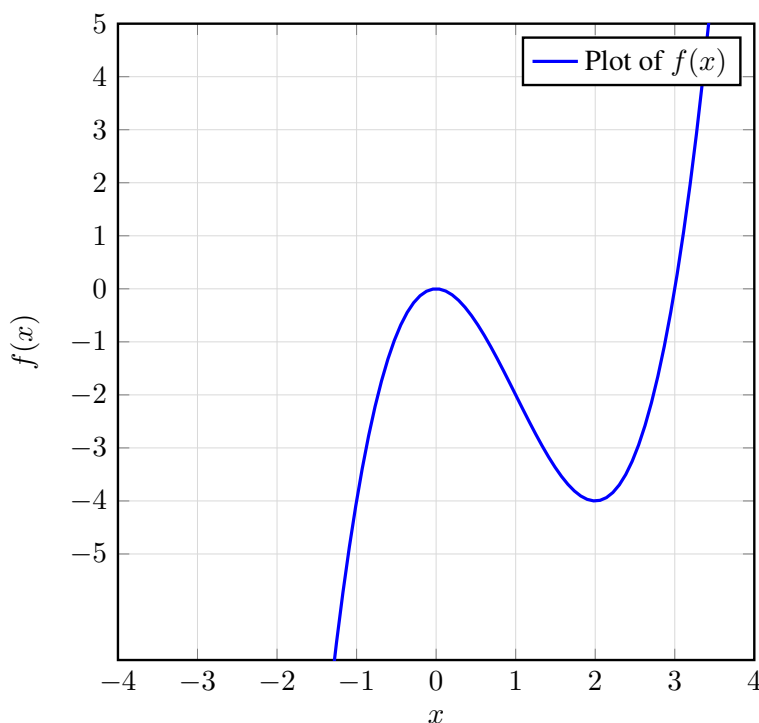
A common way to approximate a nonlinear function is to perform linearization near a point. In the case of a one-dimensional function  $f(x)$ , the linear approximation of  $f(x)$  at a point  $x_*$  is given by

$$f(x) \approx f(x_*) + f'(x_*) \cdot (x - x_*), \quad (1)$$

where  $f'(x_*) := \left. \frac{df(x)}{dx} \right|_{x=x_*}$  is the derivative of  $f(x)$  at  $x = x_*$ .

Keep in mind that wherever we see  $x_*$ , this denotes a *constant value* or operating point.

- (a) Suppose we have the single-variable function  $f(x) = x^3 - 3x^2$ . We can plot the function  $f(x)$  as follows:



- i. Write the linear approximation of the function around an arbitrary point  $x_*$ .

**Answer:**

$$f(x) \approx f(x_*) + f'(x_*) \cdot (x - x_*) \quad (2)$$

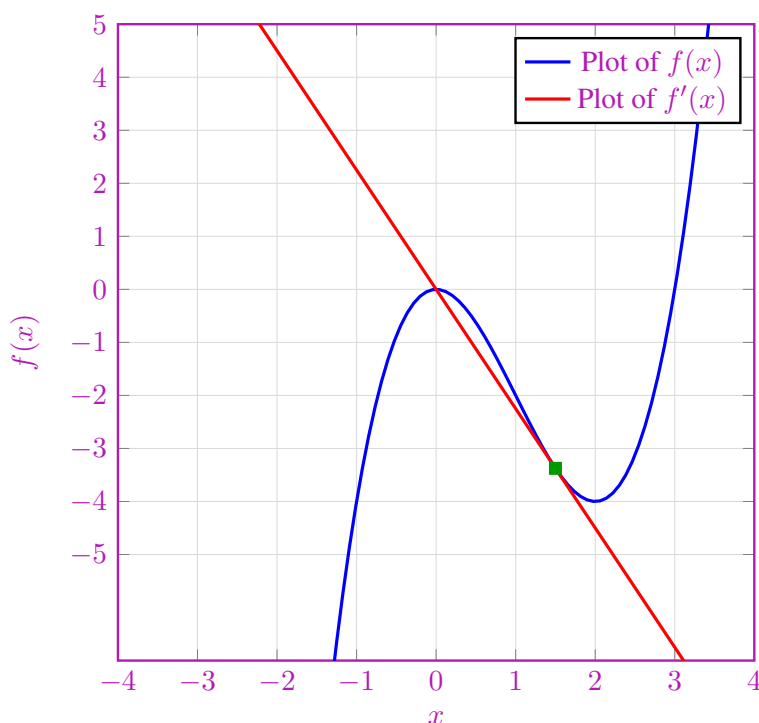
$$\approx f(x_*) + (3 \cdot x_*^2 - 6x_*) \cdot (x - x_*) \quad (3)$$

- ii. Use the expression above to linearize the function around the point  $x = 1.5$ . Draw the linearization into the plot of part i).

**Answer:**

$$f(x) \approx f(1.5) + (3 \cdot 1.5^2 - 6 \cdot 1.5) \cdot (x - 1.5) \quad (4)$$

$$\approx -3.375 + (-2.25) \cdot (x - 1.5) \quad (5)$$



Now that we have this specific point's linearization, we understand how the function behaves around the point. Let's use this linearization to evaluate the function's approximation at  $x = 1.7$  (based on our approximation at  $x = 1.5$ , we want to see how a  $\delta = +0.2$  shift in the  $x$  value changes the corresponding  $f(x)$  value). How does this approximation compare to the exact value of the function at  $x = 1.7$ ?

$$f(1.7) \approx -3.375 + (-2.25) \cdot (1.7 - 1.5) \quad (6)$$

$$\approx -3.375 - 0.45 \quad (7)$$

$$\approx -3.825 \quad (8)$$

Comparing to the exact value  $f(1.7) = 1.7^3 - 3 \cdot 1.7^2 = -3.757$ , we find that the difference is 0.068. Not too bad! What if we repeat with  $\delta = 1$ ? To do so, we must use the approximation around  $x = 1.5$  to compute  $x = 2.5$ , and compare to the exact value  $f(2.5)$ . How does our new approximation compare to the exact result?

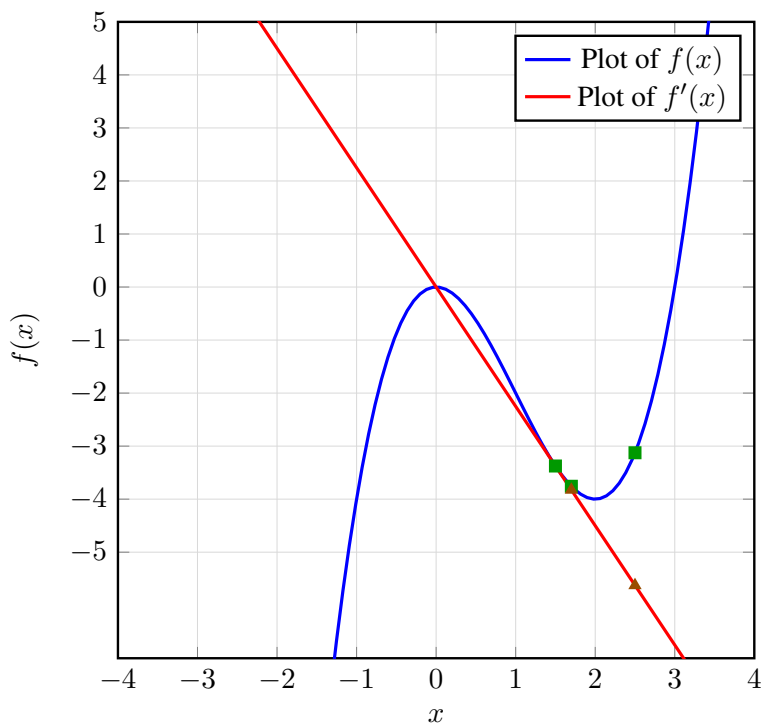
$$f(2.5) \approx -3.375 + (-2.25) \cdot (2.5 - 1.5) \quad (9)$$

$$\approx -3.375 - 2.25 \quad (10)$$

$$\approx -5.625 \quad (11)$$

Comparing to the exact value  $f(2.5) = 2.5^3 - 3 \cdot 2.5^2 = -3.125$ , we find that the difference is much larger; the error jumped to 2.5! This is an error multiplication of  $\frac{2.5}{0.068} \approx 37$ , even though our  $\delta$  only multiplied by 5. What happened?

Looking at the actual function, we see that the function has a significant curvature between our "anchor point" of  $x_* = 1.5$  and  $x = 2.5$ . Our linear model is unable to capture this curvature, and so we estimated  $f(2.5)$  as if the function kept decreasing, as it did around  $x = 1.5$  (where the slope was  $-2.25$ ).



Now, we can extend this to higher dimensional functions. In the case of a two-dimensional function  $f(x, y)$ , the linear approximation of  $f(x, y)$  at a point  $(x_*, y_*)$  is given by

$$f(x, y) \approx f(x_*, y_*) + f_x(x_*, y_*) \cdot (x - x_*) + f_y(x_*, y_*) \cdot (y - y_*). \quad (12)$$

where  $f_x(x_*, y_*)$  is the partial derivative of  $f(x, y)$  with respect to  $x$  at the point  $(x_*, y_*)$ :

$$f_x(x_*, y_*) = \left. \frac{\partial f(x, y)}{\partial x} \right|_{(x_*, y_*)} \quad (13)$$

and  $f_y(x_*, y_*)$  is the partial derivative of  $f(x, y)$  with respect to  $y$  at the point  $(x_*, y_*)$ .

- (b) Now, let's see how we can derive partial derivatives. When we are given a function  $f(x, y)$ , we calculate the partial derivative of  $f$  with respect to  $x$  by fixing  $y$  and taking the derivative with respect to  $x$ . Given the function  $f(x, y) = x^2y$ , find the partial derivatives  $f_y(x, y)$  and  $f_x(x, y)$ .

**Answer:** We have

$$f_y(x, y) = x^2 \quad (14)$$

and

$$f_x(x, y) = 2xy. \quad (15)$$

(c) Write out the linear approximation of  $f$  near  $(x_*, y_*)$ .

**Answer:** Based on the formula in eq. (12), we can write that:

$$f(x, y) \approx f(x_*, y_*) + 2x_*y_* \cdot (x - x_*) + x_*^2 \cdot (y - y_*). \quad (16)$$

(d) We want to see if the approximation arising from linearization of this function is reasonable for a point close to our point of evaluation. First, approximate  $f(x, y)$  at the point  $(2.01, 3.01)$  using  $(x_*, y_*) = (2, 3)$ , and compare the result to  $f(2.01, 3.01)$ .

**Answer:** Let  $\delta = 0.01$ . Then, the true value of  $f(2.01, 3.01)$  is

$$f(2.01, 3.01) = (2 + \delta)^2(3 + \delta) = (4 + 4\delta + \delta^2)(3 + \delta) = 12 + 16\delta + 7\delta^2 + \delta^3. \quad (17)$$

On the other hand, our approximation is

$$f(2.01, 3.01) \approx f(2, 3) + 2 \cdot 2 \cdot 3 \cdot \delta + 2^2 \cdot \delta = 12 + 16\delta. \quad (18)$$

As we can see, our approximation removes the terms with  $\delta^2$  and  $\delta^3$ . When  $\delta$  is sufficiently small, these terms become very small, and hence our approximation is reasonable.

The actual numerical values are:

$$\begin{aligned} f(2, 3) &= 12 \\ f(2.01, 3.01) &\approx 12.16 \quad (\text{using linearization}) \\ f(2.01, 3.01) &= 12.160701 \quad (\text{exact evaluation of } f) \end{aligned}$$

- (e) Suppose we have now a vector-valued function  $f(\vec{x}, \vec{y})$ , which takes in vectors  $\vec{x} \in \mathbb{R}^n$ ,  $\vec{y} \in \mathbb{R}^k$  and outputs a scalar  $\in \mathbb{R}$ . That is,  $f(\vec{x}, \vec{y})$  is  $\mathbb{R}^n \times \mathbb{R}^k \rightarrow \mathbb{R}$ . With this new model, how can we adapt our previous linearization method?

One way to linearize the function  $f$  is to do it for every single element in  $\vec{x} = [x_1 \ x_2 \ \dots \ x_n]^\top$  and  $\vec{y} = [y_1 \ y_2 \ \dots \ y_k]^\top$ . Then, when we are looking at  $x_i$  or  $y_j$ , we fix everything else as constant. This would give us the linear approximation

$$f(\vec{x}, \vec{y}) \approx f(\vec{x}_*, \vec{y}_*) + \sum_{i=1}^n \frac{\partial f(\vec{x}, \vec{y})}{\partial x_i} (x_i - x_{*,i}) + \sum_{j=1}^k \frac{\partial f(\vec{x}, \vec{y})}{\partial y_j} (y_j - y_{*,j}). \quad (19)$$

In order to simplify this equation, we can define the rows  $D_{\vec{x}}$  and  $D_{\vec{y}}$  as

$$D_{\vec{x}}f = \left[ \frac{\partial f}{\partial x_1} \quad \dots \quad \frac{\partial f}{\partial x_n} \right], \quad (20)$$

$$D_{\vec{y}}f = \left[ \frac{\partial f}{\partial y_1} \quad \dots \quad \frac{\partial f}{\partial y_k} \right]. \quad (21)$$

Then, Equation (19) can be rewritten as

$$f(\vec{x}, \vec{y}) \approx f(\vec{x}_*, \vec{y}_*) + (D_{\vec{x}}f) \Big|_{(\vec{x}_*, \vec{y}_*)} \cdot (\vec{x} - \vec{x}_*) + (D_{\vec{y}}f) \Big|_{(\vec{x}_*, \vec{y}_*)} \cdot (\vec{y} - \vec{y}_*). \quad (22)$$

Assume that  $n = k$  and we define the function  $f(\vec{x}, \vec{y}) = \vec{x}^\top \vec{y} = \sum_{i=1}^k x_i y_i$ . Find  $D_{\vec{x}}f$  and  $D_{\vec{y}}f$ .

**Answer:** The derivative is a row vector (as denoted above), so if we apply the definition (and write out the given function explicitly as  $x_1 y_1 + x_2 y_2 + \dots + x_k y_k$ ), we have:

$$D_{\vec{x}}f = \vec{y}^\top \quad (23)$$

and

$$D_{\vec{y}}f = \vec{x}^\top. \quad (24)$$

(f) Following the above part, find the linear approximation of  $f(\vec{x}, \vec{y})$  near  $\vec{x}_* = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  and  $\vec{y}_* = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$ .

Recall that  $f(\vec{x}, \vec{y}) = \vec{x}^\top \vec{y} = \sum_{i=1}^k x_i y_i$ .

**Answer:** From the solution in the previous part, we can write

$$f(\vec{x}, \vec{y}) \approx f(\vec{x}_*, \vec{y}_*) + (D_{\vec{x}} f) \Big|_{(\vec{x}_*, \vec{y}_*)} \cdot (\vec{x} - \vec{x}_*) + (D_{\vec{y}} f) \Big|_{(\vec{x}_*, \vec{y}_*)} \cdot (\vec{y} - \vec{y}_*) \quad (25)$$

$$= \vec{x}_*^\top \vec{y}_* + \vec{y}_*^\top (\vec{x} - \vec{x}_*) + \vec{x}_*^\top (\vec{y} - \vec{y}_*). \quad (26)$$

Putting in  $\vec{x}_* = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$  and  $\vec{y}_* = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$ , and let's find the approximation of  $f \left( \begin{bmatrix} 1 + \delta_1 \\ 2 + \delta_2 \end{bmatrix}, \begin{bmatrix} -1 + \delta_3 \\ 2 + \delta_4 \end{bmatrix} \right)$ ,

we have

$$f \left( \begin{bmatrix} 1 + \delta_1 \\ 2 + \delta_2 \end{bmatrix}, \begin{bmatrix} -1 + \delta_3 \\ 2 + \delta_4 \end{bmatrix} \right) \approx \vec{x}_*^\top \vec{y}_* + \vec{y}_*^\top (\vec{x} - \vec{x}_*) + \vec{x}_*^\top (\vec{y} - \vec{y}_*) \quad (27)$$

$$= 3 + \begin{bmatrix} -1 & 2 \end{bmatrix} \begin{bmatrix} \delta_1 \\ \delta_2 \end{bmatrix} + \begin{bmatrix} 1 & 2 \end{bmatrix} \begin{bmatrix} \delta_3 \\ \delta_4 \end{bmatrix} \quad (28)$$

$$= 3 - \delta_1 + 2\delta_2 + \delta_3 + 2\delta_4. \quad (29)$$

Let's compare this with the true value  $f \left( \begin{bmatrix} 1 + \delta_1 \\ 2 + \delta_2 \end{bmatrix}, \begin{bmatrix} -1 + \delta_3 \\ 2 + \delta_4 \end{bmatrix} \right)$  We have:

$$f \left( \begin{bmatrix} 1 + \delta_1 \\ 2 + \delta_2 \end{bmatrix}, \begin{bmatrix} -1 + \delta_3 \\ 2 + \delta_4 \end{bmatrix} \right) = (1 + \delta_1)(-1 + \delta_3) + (2 + \delta_2)(2 + \delta_4) \quad (30)$$

$$= 3 - \delta_1 + 2\delta_2 + \delta_3 + 2\delta_4 + \delta_1\delta_3 + \delta_2\delta_4. \quad (31)$$

As we can see, our approximation removes the second order  $\delta$  terms  $\delta_1\delta_3$  and  $\delta_2\delta_4$ , which is valid for small  $\delta_i$ .

- (g) When the function  $\vec{f}(\vec{x}, \vec{y}) : \mathbb{R}^n \times \mathbb{R}^k \rightarrow \mathbb{R}^m$  takes in vectors and outputs a vector (rather than a scalar), we can view each dimension in  $\vec{f}$  independently as a separate function  $f_i$ , and linearize each of them:

$$\vec{f}(\vec{x}, \vec{y}) = \begin{bmatrix} f_1(\vec{x}, \vec{y}) \\ f_2(\vec{x}, \vec{y}) \\ \vdots \\ f_m(\vec{x}, \vec{y}) \end{bmatrix} \approx \begin{bmatrix} f_1(\vec{x}_*, \vec{y}_*) + D_{\vec{x}}f_1 \cdot (\vec{x} - \vec{x}_*) + D_{\vec{y}}f_1 \cdot (\vec{y} - \vec{y}_*) \\ f_2(\vec{x}_*, \vec{y}_*) + D_{\vec{x}}f_2 \cdot (\vec{x} - \vec{x}_*) + D_{\vec{y}}f_2 \cdot (\vec{y} - \vec{y}_*) \\ \vdots \\ f_m(\vec{x}_*, \vec{y}_*) + D_{\vec{x}}f_m \cdot (\vec{x} - \vec{x}_*) + D_{\vec{y}}f_m \cdot (\vec{y} - \vec{y}_*) \end{bmatrix} \quad (32)$$

We can rewrite this in a clean way with the *Jacobian*:

$$D_{\vec{x}}\vec{f} = \begin{bmatrix} D_{\vec{x}}f_1 \\ D_{\vec{x}}f_2 \\ \vdots \\ D_{\vec{x}}f_m \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix}, \quad (33)$$

and similarly

$$D_{\vec{y}}\vec{f} = \begin{bmatrix} \frac{\partial f_1}{\partial y_1} & \cdots & \frac{\partial f_1}{\partial y_k} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial y_1} & \cdots & \frac{\partial f_m}{\partial y_k} \end{bmatrix}. \quad (34)$$

Then, the linearization becomes

$$\vec{f}(\vec{x}, \vec{y}) \approx \vec{f}(\vec{x}_*, \vec{y}_*) + (D_{\vec{x}}\vec{f})\Big|_{(\vec{x}_*, \vec{y}_*)} \cdot (\vec{x} - \vec{x}_*) + (D_{\vec{y}}\vec{f})\Big|_{(\vec{x}_*, \vec{y}_*)} \cdot (\vec{y} - \vec{y}_*). \quad (35)$$

Let  $\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  and  $\vec{f}(\vec{x}) = \begin{bmatrix} x_1^2 x_2 \\ x_1 x_2^2 \end{bmatrix}$ . Find  $D_{\vec{x}}\vec{f}$ , applying the definition above.

**Answer:** Here, we have

$$D_{\vec{x}}\vec{f} = \begin{bmatrix} 2x_1 x_2 & x_1^2 \\ x_2^2 & 2x_1 x_2 \end{bmatrix}. \quad (36)$$

- (h) Compare the approximation of  $\vec{f}$  at the point  $\begin{bmatrix} 2.01 \\ 3.01 \end{bmatrix}$  using  $\vec{x}_* = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$  versus  $\vec{f}\left(\begin{bmatrix} 2.01 \\ 3.01 \end{bmatrix}\right)$ . Recall the definition that  $\vec{f}(\vec{x}) = \begin{bmatrix} x_1^2 x_2 \\ x_1 x_2^2 \end{bmatrix}$ .

**Answer:** Let  $\delta = 0.01$ . The true value is

$$\vec{f}\left(\begin{bmatrix} 2.01 \\ 3.01 \end{bmatrix}\right) = \begin{bmatrix} (2 + \delta)^2(3 + \delta) \\ (2 + \delta)(3 + \delta)^2 \end{bmatrix} = \begin{bmatrix} 12 + 16\delta + 7\delta^2 + \delta^3 \\ 18 + 21\delta + 8\delta^2 + \delta^3 \end{bmatrix}. \quad (37)$$

On the other hand, our approximation is

$$\vec{f}\left(\begin{bmatrix} 2.01 \\ 3.01 \end{bmatrix}\right) \approx \vec{f}\left(\begin{bmatrix} 2 \\ 3 \end{bmatrix}\right) + \begin{bmatrix} 12 & 4 \\ 9 & 12 \end{bmatrix} \cdot \begin{bmatrix} \delta \\ \delta \end{bmatrix} = \begin{bmatrix} 12 + 16\delta \\ 18 + 21\delta \end{bmatrix}. \quad (38)$$

Again, our approximation essentially removes the higher order terms of  $\delta$ .

When we plug in  $\delta = 0.01$ , we have

$$\vec{f}\left(\begin{bmatrix} 2.01 \\ 3.01 \end{bmatrix}\right) = \begin{bmatrix} 12.160701 \\ 18.210801 \end{bmatrix} \quad (39)$$

and our approximation is

$$\vec{f}\left(\begin{bmatrix} 2.01 \\ 3.01 \end{bmatrix}\right) \approx \begin{bmatrix} 12.16 \\ 18.21 \end{bmatrix}. \quad (40)$$

- (i) **Practice:** Let  $\vec{x}$  and  $\vec{y}$  be vectors with 2 rows, and let  $\vec{w}$  be another vector with 2 rows. Let  $\vec{f}(\vec{x}, \vec{y}) = \vec{x}\vec{y}^\top\vec{w}$ . Find  $D_{\vec{x}}\vec{f}$  and  $D_{\vec{y}}\vec{f}$ .

**Answer:** Here, recall that

$$\vec{f} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} \cdot \begin{bmatrix} y_1 & y_2 \end{bmatrix} \cdot \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} x_1y_1 & x_1y_2 \\ x_2y_1 & x_2y_2 \end{bmatrix} \cdot \begin{bmatrix} w_1 \\ w_2 \end{bmatrix} = \begin{bmatrix} x_1y_1w_1 + x_1y_2w_2 \\ x_2y_1w_1 + x_2y_2w_2 \end{bmatrix}. \quad (41)$$

Then,

$$D_{\vec{x}}\vec{f} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} \end{bmatrix} = \begin{bmatrix} y_1w_1 + y_2w_2 & 0 \\ 0 & y_1w_1 + y_2w_2 \end{bmatrix} \quad (42)$$

and

$$D_{\vec{y}}\vec{f} = \begin{bmatrix} \frac{\partial f_1}{\partial y_1} & \frac{\partial f_1}{\partial y_2} \\ \frac{\partial f_2}{\partial y_1} & \frac{\partial f_2}{\partial y_2} \end{bmatrix} = \begin{bmatrix} x_1w_1 & x_1w_2 \\ x_2w_1 & x_2w_2 \end{bmatrix}. \quad (43)$$

We can also write

$$D_{\vec{x}}\vec{f} = \vec{y}^\top\vec{w} \cdot I \quad (44)$$

and

$$D_{\vec{y}}\vec{f} = \vec{x}\vec{w}^\top, \quad (45)$$

which can be derived by noticing that  $\vec{y}^\top\vec{w} = \vec{w}^\top\vec{y}$ .

- (j) **Practice:** Continuing the above part, find the linear approximation of  $\vec{f}$  near  $\vec{x} = \vec{y} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$  and with

$$\vec{w} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}.$$

**Answer:** We have

$$\vec{f}(\vec{x}, \vec{y}) \approx \vec{f}(\vec{x}_*, \vec{y}_*) + D_{\vec{x}}\vec{f} \cdot (\vec{x} - \vec{x}_*) + D_{\vec{y}}\vec{f} \cdot (\vec{y} - \vec{y}_*) \quad (46)$$

$$= \begin{bmatrix} 3 \\ 3 \end{bmatrix} + \begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} \cdot \begin{bmatrix} x_1 - 1 \\ x_2 - 1 \end{bmatrix} + \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} \cdot \begin{bmatrix} y_1 - 1 \\ y_2 - 1 \end{bmatrix} \quad (47)$$



(48)

Let's do an approximation of  $\vec{f}\left(\begin{bmatrix} 1 + \delta_1 \\ 1 + \delta_2 \end{bmatrix}, \begin{bmatrix} 1 + \delta_3 \\ 1 + \delta_4 \end{bmatrix}\right)$ , then,

$$\vec{f}\left(\begin{bmatrix} 1 + \delta_1 \\ 1 + \delta_2 \end{bmatrix}, \begin{bmatrix} 1 + \delta_3 \\ 1 + \delta_4 \end{bmatrix}\right) \approx \begin{bmatrix} 3 \\ 3 \end{bmatrix} + \begin{bmatrix} 3 & 0 \\ 0 & 3 \end{bmatrix} \cdot \begin{bmatrix} \delta_1 \\ \delta_2 \end{bmatrix} + \begin{bmatrix} 2 & 1 \\ 2 & 1 \end{bmatrix} \cdot \begin{bmatrix} \delta_3 \\ \delta_4 \end{bmatrix} = \begin{bmatrix} 3 + 3\delta_1 + 2\delta_3 + \delta_4 \\ 3 + 3\delta_2 + 2\delta_3 + \delta_4 \end{bmatrix}.$$

We can compare with the true value

$$\begin{aligned} \vec{f}\left(\begin{bmatrix} 1 + \delta_1 \\ 1 + \delta_2 \end{bmatrix}, \begin{bmatrix} 1 + \delta_3 \\ 1 + \delta_4 \end{bmatrix}\right) &= \begin{bmatrix} 1 + \delta_1 \\ 1 + \delta_2 \end{bmatrix} \begin{bmatrix} 1 + \delta_3 & 1 + \delta_4 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} 1 + \delta_1 + \delta_3 + \delta_1\delta_3 & 1 + \delta_1 + \delta_4 + \delta_1\delta_4 \\ 1 + \delta_2 + \delta_3 + \delta_2\delta_3 & 1 + \delta_2 + \delta_4 + \delta_2\delta_4 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} \\ &= \begin{bmatrix} 3 + 3\delta_1 + 2\delta_3 + \delta_4 + 2\delta_1\delta_3 + \delta_1\delta_4 \\ 3 + 3\delta_2 + 2\delta_3 + \delta_4 + 2\delta_2\delta_3 + \delta_2\delta_4 \end{bmatrix}, \end{aligned} \quad (49)$$

and we see that our approximation removes the second order  $\delta$  terms  $\delta_1\delta_3$ ,  $\delta_1\delta_4$ ,  $\delta_2\delta_3$  and  $\delta_2\delta_4$ .

These linearizations are important for us because we can do many easy computations using linear functions.

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