Discussion 11B

EECS 16B

The following note is useful for this discussion: Note 18.

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

$$\widehat{f}(x; x_{\star}) = f(x_{\star}) + f'(x_{\star}) \cdot (x - x_{\star}), \tag{1}$$

where $f'(x_*) := \frac{\mathrm{d}f}{\mathrm{d}x}(x_*)$ is the derivative of f(x) at $x = x_*$.

Keep in mind that wherever we see x_{\star} , this denotes a *constant value* or operating point.

We can evaluate the accuracy of our approximation by calculating the approximation error, namely $|f(x) - \widehat{f}(x; x_{\star})|.$

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

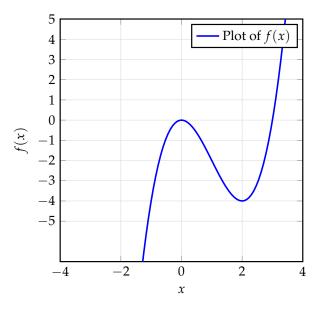


Figure 1: Plot of $f(x) = x^3 - 3x^2$

(a) Write the linear approximation of the function around an arbitrary point x_{\star} .

(b) Using the expression above, linearize the function around the point $x_* = 1.5$. Draw the linearization into the plot in fig. 1. Then evaluate the accuracy of the linear approximation at x = 1.7 and x = 2.5. Does the difference in accuracy make sense, based on the plot?

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

$$\widehat{f}(x,y;x_{\star},y_{\star}) = f(x_{\star},y_{\star}) + \frac{\partial f}{\partial x}(x_{\star},y_{\star}) \cdot (x - x_{\star}) + \frac{\partial f}{\partial y}(x_{\star},y_{\star}) \cdot (y - y_{\star}). \tag{2}$$

where $\frac{\partial f}{\partial x}(x_{\star},y_{\star})$ is the partial derivative of f(x,y) with respect to x at the point (x_{\star},y_{\star}) , and similarly for $\frac{\partial f}{\partial y}(x_{\star},y_{\star})$

- (c) Now, let's see how we can find 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 $\frac{\partial f(x,y)}{\partial x}$ and $\frac{\partial f(x,y)}{\partial y}$.
- (d) Write out the linear approximation of f near (x_*, y_*) .

(e) We want to see if the approximation arising from linearization of this function is reasonable for a point close to our point of evaluation. Suppose we want to evaluate the accuracy of our approximation at some point $(x_* + \delta, y_* + \delta)$, where $x_* = 2$ and $y_* = 3$. Find the accuracy of this approximation in terms of δ . What if $\delta = 0.01$?

(f) Suppose we have now a scalar-valued function $f(\vec{x}, \vec{y})$, which takes in vector-valued arguments $\vec{x} \in \mathbb{R}^n$, $\vec{y} \in \mathbb{R}^k$ and outputs a scalar $\in \mathbb{R}$. That is, $f(\vec{x}, \vec{y})$ is $\mathbb{R}^n \times \mathbb{R}^k \to \mathbb{R}$.

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

$$f(\vec{x}, \vec{y}) \approx f(\vec{x}_{\star}, \vec{y}_{\star}) + \sum_{i=1}^{n} \frac{\partial f(\vec{x}, \vec{y})}{\partial x_{i}} \Big|_{(\vec{x}_{\star}, \vec{y}_{\star})} (x_{i} - x_{i, \star}) + \sum_{i=1}^{k} \frac{\partial f(\vec{x}, \vec{y})}{\partial y_{j}} \Big|_{(\vec{x}_{\star}, \vec{y}_{\star})} (y_{j} - y_{j, \star}).$$
(3)

In order to simplify this equation, we can define the following two vector quantities:

$$J_{\vec{x}}f = \begin{bmatrix} \frac{\partial f}{\partial x_1} & \cdots & \frac{\partial f}{\partial x_n} \end{bmatrix} \tag{4}$$

$$J_{\vec{y}}f = \begin{bmatrix} \frac{\partial f}{\partial y_1} & \cdots & \frac{\partial f}{\partial y_k} \end{bmatrix} \tag{5}$$

First, how can we "vectorize" eq. (3) using $J_{\vec{x}}f$ and $J_{\vec{y}}f$? Next, assume that n=k and we define the function $f(\vec{x},\vec{y})=\vec{x}^{\top}\vec{y}=\sum_{i=1}^k x_iy_i$. Find $J_{\vec{x}}f$ and $J_{\vec{y}}f$ for this specific f.

(HINT: For vectorizing, think about replacing the summations as the multiplication of a row and column vector. What would these vectors be?)

(g) Following the above part, find the linear approximation of $f(\vec{x}, \vec{y})$ near $\vec{x}_{\star} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $\vec{y}_{\star} = \begin{bmatrix} -1 \\ 2 \end{bmatrix}$. Recall that $f(\vec{x}, \vec{y}) = \vec{x}^{\top} \vec{y} = \sum_{i=1}^{k} x_i y_i$.

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

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