

Homework Set #1

Due: Friday, November 6, 2020.

1. Sampling and Matrix Notation

Let's represent a finite duration discrete time sequence with $\mathbf{x} \in \Re^n$, i.e., $\mathbf{x} = \begin{bmatrix} x_1 & x_2 & \dots & x_n \end{bmatrix}^T$. Represent the following linear operations in the form $\mathbf{y} = \mathbf{A}\mathbf{x}$:

(a) Let $\mathbf{y} \in \Re^{2n-1}$ be the output of the upsampling system defined by

$$y_k = \begin{cases} x_{\frac{k+1}{2}} & k \text{ is odd} \\ 0 & k \text{ is even} \end{cases} \quad (1)$$

for $k = 1, \dots, 2n-1$ and $\mathbf{y} = \begin{bmatrix} y_1 & y_2 & \dots & y_{2n-1} \end{bmatrix}^T$.

(b) Assume n is even and $\mathbf{y} \in \Re^{\frac{n}{2}}$. We define $y_k = x_{2k}$ for $k = 1, \dots, \frac{n}{2}$.

(c) Assume n is even and $\mathbf{y} \in \Re^{\frac{n}{2}}$. We define $y_k = \frac{x_{2k} + x_{2k-1}}{2}$ for $k = 1, \dots, \frac{n}{2}$.

2. Vector Space of Polynomials

Let \mathcal{P}_{n-1} represent the vectors space of polynomials with degree less than or equal to $n-1$. Therefore, each element in this vector space can be written as

$$p(x) = c_0 + c_1x + c_2x^2 + \dots + c_{n-1}x^{n-1}. \quad (2)$$

As we discussed in class, we can form a one to one correspondence between \mathcal{P}_{n-1} and \Re^n . Let $\mathbf{a} = \begin{bmatrix} c_0 & c_1 & \dots & c_{n-1} \end{bmatrix}^T$ be the vector in \Re^n corresponding to $p(x)$ above, and let \mathbf{b} be the vector corresponding to $\frac{dp(x)}{dx}$. Show that the relation between \mathbf{b} and \mathbf{a} can be written as

$$\mathbf{b} = \mathbf{H}\mathbf{a} \quad (3)$$

by explicitly writing \mathbf{H} .

3. Affine functions

A function $f : \Re^n \rightarrow \Re^m$ is called affine if and only if for any $\mathbf{x}, \mathbf{y} \in \Re^n$ and any $\alpha, \beta \in \Re$ satisfying $\alpha + \beta = 1$, we have

$$f(\alpha\mathbf{x} + \beta\mathbf{y}) = \alpha f(\mathbf{x}) + \beta f(\mathbf{y}) \quad (4)$$

(Note that, if we didn't restrict $\alpha + \beta = 1$, this would be the definition of a linear function)

(a) Suppose that $\mathbf{A} \in \Re^{m \times n}$ and $\mathbf{b} \in \Re^m$. Show that $f(\mathbf{x}) = \mathbf{A}\mathbf{x} + \mathbf{b}$ is affine.

- (b) Now the converse: Show that any affine function f can be represented as $f(\mathbf{x}) = \mathbf{Ax} + \mathbf{b}$ for some $\mathbf{A} \in \mathbb{R}^{m \times n}$ and $\mathbf{b} \in \mathbb{R}^m$. (Hint: Show that $g(\mathbf{x}) = f(\mathbf{x}) - f(\mathbf{0})$ is linear).

Therefore, an affine function can be considered as a linear function shifted by an offset.

4. *Yet Another Proof of Cauchy-Schwarz...*

- (a) Suppose $a \geq 0$, $c \geq 0$ and for all $\lambda \in \mathbb{R}$, $a + 2b\lambda + c\lambda^2 \geq 0$. Show that $|b| \leq \sqrt{ac}$.
 (b) Given $\mathbf{u}, \mathbf{w} \in \mathbb{R}^n$, explain why $(\mathbf{u} + \lambda\mathbf{w})^T(\mathbf{u} + \lambda\mathbf{w}) \geq 0$ for all $\lambda \in \mathbb{R}$.
 (c) Apply a to the quadratic resulting from the expansion of the expression in (b), to get the Cauchy-Schwarz inequality:

$$|\mathbf{v}^T \mathbf{w}| \leq \sqrt{\mathbf{v}^T \mathbf{v}} \sqrt{\mathbf{w}^T \mathbf{w}}$$

5. *Inequality Proof*

For $\mathbf{a}, \mathbf{b} \in \mathbb{R}^n$, show that $\|\mathbf{a} + \mathbf{b}\| \geq \|\mathbf{a}\| - \|\mathbf{b}\|$, where $\|\cdot\|$ is the Euclidian norm.

6. *First steps in multivariable quadratic functions*

Please write the multivariable quadratic function

$$f(x_1, x_2, x_3) = 2x_1^2 + 3x_1x_3 - 2x_3x_2 - x_2^2 + 4x_3^2 \quad (5)$$

in the form $f(\mathbf{x}) = \mathbf{x}^T \mathbf{Ax}$ where $\mathbf{x} \in \mathbb{R}^3$. Either prove that \mathbf{A} is unique and provide the corresponding \mathbf{A} , or show that it is non-unique by providing two alternative matrices for \mathbf{A} .

7. *On Gradients....*

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be a differentiable multivariate function. Its gradient (at point \mathbf{x}) is defined as the vector

$$\nabla f(\mathbf{x}) = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix}. \quad (6)$$

where the partial derivatives are evaluated at \mathbf{x} .

Find the explicit gradients of the following functions (using matrix notation only)

- (a) $f(\mathbf{x}) = \mathbf{a}^T \mathbf{x} + b$ for some $\mathbf{a} \in \mathbb{R}^n$ and $b \in \mathbb{R}$.
 (b) $f(\mathbf{x}) = \mathbf{x}^T \mathbf{Ax}$ for some $\mathbf{A} \in \mathbb{R}^{n \times n}$.

8. *What is a homework without a computational part?....*

Given a differentiable function $f(\mathbf{x})$, where $\mathbf{x} \in \mathbb{R}^n$,

$$a(\mathbf{x}) = f(\mathbf{x}_0) + \nabla f(\mathbf{x}_0)^T (\mathbf{x} - \mathbf{x}_0) \quad (7)$$

is an affine approximation of $f(\mathbf{x})$ around \mathbf{x}_0 .

Note that this affine function defines an hyperplane in \mathbb{R}^{n+1} in the standard form

$$\begin{bmatrix} \nabla f(\mathbf{x}_0) \\ -1 \end{bmatrix}^T \left(\begin{bmatrix} \mathbf{x} \\ a(\mathbf{x}) \end{bmatrix} - \begin{bmatrix} \mathbf{x}_0 \\ f(\mathbf{x}_0) \end{bmatrix} \right) = 0. \quad (8)$$

Now consider the function

$$f(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x} \quad (9)$$

defined over \mathbb{R}^2 , where

$$\mathbf{A} = \begin{bmatrix} 0.01 & 0.001 \\ 0.001 & 0.01 \end{bmatrix}. \quad (10)$$

Please plot the surfaces corresponding to this function and its affine approximation at the point $\mathbf{x}_0 = \begin{bmatrix} 25 & 0 \end{bmatrix}^T$ as show in Figure below:

Please submit your Matlab/Python/Julia code and your plot (Hint: Yes, you can shamelessly reuse your *drawhyperplane3D* function from your previous homework.) In case of Python/Julia notebooks, please submit a print out of your executed notebook.

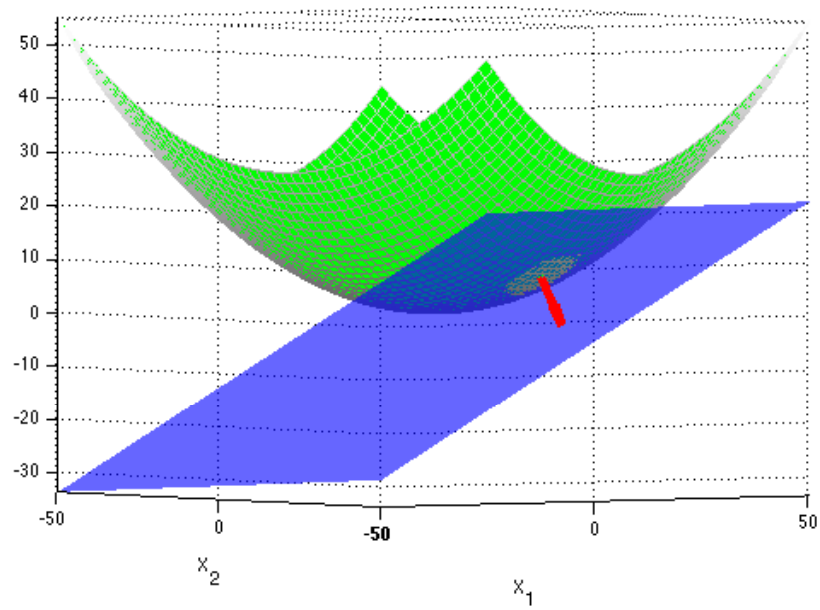


Figure 1: Affine approximation of a quadratic function.