

## Homework 7

Due: Thursday, March 18 – 10:00 EST

### Problem 1: Orthogonal projections [10 Points]

1. Show that  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \in \mathbb{M}(2 \times 2, \mathbb{R})$  is invertible iff  $ad - bc \neq 0$ . Find  $A^{-1}$ .
2. In  $\mathbb{R}^3$ , compute the orthogonal projection  $P$  to  $S = \{[x, y, z]^T \in \mathbb{R}^3 \mid x - y - 2z = 0\}$ .
3. Compute the orthogonal projection  $Q$  to  $S^\perp$  and show that  $P + Q = I$ .
4. **Math 513:** Formulate the following as orthogonal projection and solve it: Find  $f \in \text{Span}_{\mathbb{R}}\{1, \sin(x), \cos(x)\}$  which minimizes  $\int_0^{2\pi} (\sin(2x) - f(x))^2$ .

### Problem 2: Least square approximation [10 Points]

We compute the parabola  $P(D, E, F) = \left\{ \begin{bmatrix} t \\ D + Et + Ft^2 \end{bmatrix} \mid t \in \mathbb{R} \right\}$  closest to

$$\vec{b}_1 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}, \quad \vec{b}_2 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, \quad \vec{b}_3 = \begin{bmatrix} 2 \\ 3 \end{bmatrix}, \quad \vec{b}_4 = \begin{bmatrix} 3 \\ 5 \end{bmatrix}. \quad (1)$$

1. First consider a general  $A \in \mathbb{M}(n \times m, \mathbb{R})$  and  $\vec{b} \in \mathbb{R}^n$ .
  - By repeating the steps in the lecture, argue that the orthogonal projection  $A\vec{x}$  of  $\vec{b}$  to  $C(A)$  is defined by the demand that  $A^T A\vec{x} = A^T \vec{b}$ .
  - Consider the length of the error vector  $l_e: \mathbb{R}^3 \rightarrow \mathbb{R}$ ,  $\vec{x} \mapsto l_e(\vec{x}) = \left\langle A\vec{x} - \vec{b}, A\vec{x} - \vec{b} \right\rangle_{\text{Std}}$ .  
Prove that the Jacobian matrix of  $l_e$  vanishes at  $\vec{x} \in \mathbb{R}^n$  iff  $A^T A\vec{x} = A^T \vec{b}$ .
  - Explain the significance of the relation among  $A^T A\vec{x} = A^T \vec{b}$  and  $l_e$ .
2. Now focus on the  $\vec{b}_i$  in eq. (1) and the parabola  $P(D, E, F)$ . Find  $A \in \mathbb{M}(4 \times 3, \mathbb{R})$ ,  $\vec{b} \in \mathbb{R}^4$  such that
 
$$\{\vec{b}_1, \vec{b}_2, \vec{b}_3, \vec{b}_4\} \subset P(D, E, F) \Leftrightarrow \vec{x} = [D \ E \ F]^T \text{ satisfies } A\vec{x} = \vec{b}. \quad (2)$$
3. Show that no parabola contains  $\vec{b}_1, \vec{b}_2, \vec{b}_3$  and  $\vec{b}_4$ .
4. Find the best approximation parabola. You may assume that the Hessian matrix of  $l_e$  is positive definite.
5. Plot the best approximation parabola and the points  $\vec{b}_i$ .

### Problem 3: Least square approximation in Python [10 Points]

1. Write a function `LineFit` which accepts  $[(t_1, b_1), (t_2, b_2), \dots, (t_n, b_n)]$  and fits a line to this data by the method of least square. You may assume that the Hessian matrix is positive definite. Plot the line and the data points  $(t_i, b_i)$ .
2. Similarly, write a function `ParaFit`, which fits a parabola  $C + Dt + Et^2$  to  $[(t_1, b_1), (t_2, b_2), \dots, (t_n, b_n)]$ , plots the parabola and the data points.
3. Apply `LineFit` and `ParaFit` to  $[(1, 2), (2, 2), (3, 5), (4, 3), (4.5, 8)]$ . By looking at the plots, does the line or the parabola describe the data better?
4. Expand your functions by a criterion for the quality of the fit. Justify your criterion and use it to tell if the line or parabola fits the data better.
5. **Math 513:** Compare your line fit with the linear regression fit in *scikit-learn*.

### Problem 4: Gram-Schmidt procedure [10 Points]

1. Be  $\{\vec{a}, \vec{b}, \vec{c}\}$  a family of linearly independent vectors in an inner product space  $(V, \langle \cdot, \cdot \rangle)$ . Show that  $\vec{U}, \vec{V}, \vec{W}$  are orthogonal:

$$\vec{U} = \vec{a}, \quad \vec{V} = \vec{b} - \frac{\langle \vec{U}, \vec{b} \rangle}{\langle \vec{U}, \vec{U} \rangle} \cdot \vec{U}, \quad \vec{W} = \vec{c} - \frac{\langle \vec{U}, \vec{c} \rangle}{\langle \vec{U}, \vec{U} \rangle} \cdot \vec{U} - \frac{\langle \vec{V}, \vec{c} \rangle}{\langle \vec{V}, \vec{V} \rangle} \cdot \vec{V}. \quad (3)$$

2. In  $\mathbb{R}^4$ , let  $W = \text{Span}_{\mathbf{R}}(\vec{v}_1, \vec{v}_2, \vec{v}_3)$  with

$$\vec{v}_1 = 2\vec{e}_1 + \vec{e}_3, \quad \vec{v}_2 = -\vec{e}_1 + 3\vec{e}_2, \quad \vec{v}_3 = 2\vec{e}_1 - \vec{e}_2 + 3\vec{e}_3, \quad (4)$$

where  $(\vec{e}_1, \vec{e}_2, \vec{e}_3, \vec{e}_4)$  is the standard basis of  $\mathbb{R}^4$ .

- Find an orthonormal basis of  $W$ .
- Find an orthonormal basis of  $N(A^T)$  where  $A = \begin{bmatrix} \vec{v}_1 & \vec{v}_2 & \vec{v}_3 \end{bmatrix}$ .
- Use these results to find an orthonormal basis of  $\mathbb{R}^4$ .
- Compute the projection of  $\vec{b} = \begin{bmatrix} 1 & 0 & 0 & 1 \end{bmatrix}^T \in \mathbb{R}^4$  to  $W$ .