

# Homework 6

1.  $p(t) = x_0 + x_1 t + x_2 t^2$

$t$	-2	-1	0	1	2
$y$	2	-10	0	2	1

let  $A$  be a  $5 \times 3$  matrix and  $y$  be  $5 \times 1$  vector

$$A = \begin{pmatrix} 1 & -2 & 4 \\ 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 2 & 4 \end{pmatrix} \quad y = \begin{pmatrix} 2 \\ -10 \\ 0 \\ 2 \\ 1 \end{pmatrix}$$

$$A^T A = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ -2 & -1 & 0 & 1 & 2 \\ 4 & 1 & 0 & 1 & 4 \end{pmatrix} \begin{pmatrix} 1 & -2 & 4 \\ 1 & -1 & 1 \\ 1 & 0 & 0 \\ 1 & 1 & 1 \\ 1 & 2 & 4 \end{pmatrix} = \begin{pmatrix} 5 & 0 & 10 \\ 0 & 10 & 0 \\ 10 & 0 & 34 \end{pmatrix}$$

$$A^T y = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 \\ -2 & -1 & 0 & 1 & 2 \\ 4 & 1 & 0 & 1 & 4 \end{pmatrix} \begin{pmatrix} 2 \\ -10 \\ 0 \\ 2 \\ 1 \end{pmatrix} = \begin{pmatrix} -5 \\ 10 \\ 4 \end{pmatrix}$$

$$\det(A^T A) = 5 \times 10 \times 34 + 0 + 0 - 10 \times 10 \times 10 - 0 - 0 = 700$$

$$M_{11} = 10 \times 34 - 0 = 340 ; M_{12} = -(0 - 0) = 0 ; M_{13} = 0 - 10 \times 10 = -100$$

$$M_{21} = -(0 - 0) = 0 ; M_{22} = 5 \times 34 - 10 \times 10 = 70 ; M_{23} = -(0 - 0) = 0$$

$$M_{31} = 0 - 10 \times 10 = -100 ; M_{32} = -(0 - 0) = 0 ; M_{33} = 5 \times 10 - 0 = 50$$

$$(A^T A)^{-1} = \frac{1}{\det(A^T A)} \text{adj}(A^T A) = \frac{1}{700} \begin{pmatrix} 340 & 0 & -100 \\ 0 & 70 & 0 \\ -100 & 0 & 50 \end{pmatrix}$$

$$\text{Thus, } A^T A x = A^T y$$

$$\begin{aligned} x = (A^T A)^{-1} A^T y &= \frac{1}{700} \begin{pmatrix} 340 & 0 & -100 \\ 0 & 70 & 0 \\ -100 & 0 & 50 \end{pmatrix} \begin{pmatrix} -5 \\ 10 \\ 4 \end{pmatrix} \\ &= \frac{1}{700} \begin{pmatrix} -2100 \\ 700 \\ 50 \end{pmatrix} = \begin{pmatrix} -3 \\ 1 \\ 1 \end{pmatrix} \end{aligned}$$

$$\text{Therefore, } p(t) = -3t + t + t^2$$

2.  $\begin{pmatrix} I & A \\ A^T & 0 \end{pmatrix} \begin{pmatrix} r \\ x \end{pmatrix} = \begin{pmatrix} b \\ 0 \end{pmatrix}$

$$\Rightarrow \begin{cases} Ir + Ax = b \\ A^T r + 0 \cdot x = 0 \end{cases} \Rightarrow \begin{cases} Ir = b - Ax \\ A^T r = 0 \end{cases}$$

Since  $r = b - Ax$ , we have  $A^T(b - Ax) = 0$

$A^TAx = A^Tb$ . This is the solution to the least squares

optimization problem of  $\min \|Ax - b\|^2$

Therefore,  $\min \|Ax - b\|^2$  can be obtained from solving  $\begin{pmatrix} \frac{1}{2}A \\ A^T \\ 0 \end{pmatrix} \begin{pmatrix} r \\ x \\ 0 \end{pmatrix} = \begin{pmatrix} b \\ 0 \\ 0 \end{pmatrix}$

$$3. f(x) = \frac{1}{2} x^T Q x - c^T x$$

$$(a) \nabla f(x) = \frac{\partial}{\partial x} \left( \frac{1}{2} x^T Q x - c^T x \right) = Qx - c$$

$$Hf(x) = Q^T = Q$$

$$(b) f(x) \approx f(a) + \nabla f(a)^T(x-a) + \frac{1}{2}(x-a)^T \nabla^2 f(a)(x-a)$$

$$\approx \frac{1}{2} a^T Q a - c^T a + (Qa - c)^T(x-a) + \frac{1}{2}(x-a)^T Q(x-a)$$

$$(c) \text{ Since } r^* \text{ is the optimal step size, we have } r^* = \arg \min_r f(x - r \nabla f(x))$$

$$f(x - r \nabla f(x)) = \frac{1}{2} (x - r \nabla f(x))^T Q (x - r \nabla f(x)) - c^T (x - r \nabla f(x))$$

$$= \frac{1}{2} (x^T Q - r \nabla f(x)^T Q)(x - r \nabla f(x)) - c^T x + r c^T \nabla f(x)$$

$$= \frac{1}{2} (x^T Q x - r x^T Q \nabla f(x) - r \nabla f(x)^T Q x + r^2 \nabla f(x)^T Q \nabla f(x)) - c^T x + r c^T \nabla f(x)$$

$$= \left( \frac{1}{2} x^T Q x - c^T x \right) + \frac{1}{2} r^2 \nabla f(x)^T Q \nabla f(x) - r x^T Q \nabla f(x) + r c^T \nabla f(x)$$

$$= f(x) + \frac{1}{2} r^2 (\nabla f(x) - c)^T Q (\nabla f(x) - c) - r (\nabla f(x)^T Q (\nabla f(x) - c) + c^T Q (\nabla f(x) - c))$$

$$\frac{\partial}{\partial r} f(x - r \nabla f(x)) = r (\nabla f(x) - c)^T Q (\nabla f(x) - c) - (x^T Q Q x - x^T Q c + c^T Q x - c^T c) = 0$$

$$\Rightarrow r^* = \frac{x^T Q \nabla f(x) - c^T c}{(\nabla f(x) - c)^T Q (\nabla f(x) - c)} = \frac{(\nabla f(x) - c)^T (\nabla f(x) - c)}{(\nabla f(x) - c)^T Q (\nabla f(x) - c)}$$

non zero vectors

4. Suppose  $d_1, \dots, d_n \in \mathbb{R}^n$  are not linearly independent, then

$\exists \alpha_1, \dots, \alpha_n \text{ not all zero, such that } \alpha_1 d_1 + \alpha_2 d_2 + \dots + \alpha_n d_n = \sum_{j=1}^n \alpha_j d_j = 0$

Thus, we have  $d_i^T Q \left( \sum_{j=1}^n \alpha_j d_j \right) = 0$

Since  $d_i^T Q d_j = 0$  for all  $i, j$  with  $i \neq j$ , we can get

$$\sum_{j=1}^n \alpha_j d_i^T Q d_j = \alpha_i d_i^T Q d_i = 0$$

Because  $Q$  is a  $n \times n$  symmetric and positive definite matrix

$$x^T Q x > 0, \forall x \in \mathbb{R}^n, d_i^T Q d_i > 0$$

This means  $\alpha_i = 0$  for each  $i$

This result contradicts the assumption that  $\alpha_1, \dots, \alpha_n$  are not all zero

Therefore,  $d_1, \dots, d_n$  are linearly independent in the standard sense

$$5. (a) x^* - x^0 = \alpha_0 d_0 + \dots + \alpha_{n-1} d_{n-1} = \sum_{i=0}^{n-1} \alpha_i d_i$$

$$d_k^T Q (x^* - x^0) = \alpha_0 d_k^T Q d_0 + \dots + \alpha_{n-1} d_k^T Q d_{n-1}$$

Since  $d_i^T Q d_j = 0$  for  $i \neq j$ , then we have

$$d_k^T Q (x^* - x^0) = \alpha_k d_k^T Q d_k$$

$$\text{Thus, } \alpha_k = \frac{d_k^T Q (x^* - x^0)}{d_k^T Q d_k}$$

$$(b) x^{k+1} = x^k + \alpha_k d_k \Rightarrow x^k = x^0 + \alpha_0 d_0 + \dots + \alpha_{k-1} d_{k-1}$$

Substituting  $k=n$ , we have  $x^n = x^0 + \alpha_0 d_0 + \dots + \alpha_{n-1} d_{n-1}$

$$\text{Since } x^* - x^0 = \sum_{i=0}^{n-1} \alpha_i d_i \quad = x^0 + \sum_{i=0}^{n-1} \alpha_i d_i$$

it follows that  $x^* = x^n$

For  $x^k = x^0 + \sum_{i=0}^{k-1} \alpha_i d_i$ , apply  $d_k^T Q$

$$\begin{aligned} d_k^T Q x^k &= d_k^T Q x^0 + d_k^T Q \sum_{i=0}^{k-1} \alpha_i d_i \\ &= d_k^T Q x^0 + \alpha_0 d_k^T Q d_0 + \dots + \alpha_{k-1} d_k^T Q d_{k-1} \\ &= d_k^T Q x^0 \end{aligned}$$

$$\text{Thus, } d_k^T Q x^k = d_k^T Q x^0$$

$$(c) d_k^T Q x^0 = d_k^T Q x^k$$

$$d_k^T Q (x^* - x^0) = d_k^T Q (x^* - x^k)$$

$$= d_k^T (\alpha x^k - \alpha x^k)$$

$$= d_k^T (b - \alpha x^k) \quad \text{since } g_k = \alpha x^k - b$$

$$= d_k^T (-g_k) = -d_k^T g_k = -g_k^T d_k$$

$$\text{Thus, } d_k^T \alpha (x^k - x^*) = -g_k^T d_k$$

$$\text{since } \alpha_k = \frac{d_k^T \alpha (x^k - x^*)}{d_k^T \alpha d_k} \quad \text{obtained in part (a)}$$

$$\text{we can get } \alpha_k = \frac{-g_k^T d_k}{d_k^T \alpha d_k}$$

$$6. (a) \quad Q = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}, \quad d_0 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \quad d_1 = \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix}, \quad d_2 = \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix}$$

$$d_0^T Q d_1 = (1 \ 0 \ 0) \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = (1 \ 1 \ 1) \begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} = 0$$

$$d_0^T Q d_2 = (1 \ 0 \ 0) \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} = (1 \ 1 \ 1) \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} = 0$$

$$d_1^T Q d_2 = (1 \ 0 \ -1) \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} = (1 \ 1 \ 0) \begin{pmatrix} 1 \\ -1 \\ 0 \end{pmatrix} = 0$$

Thus,  $d_0, d_1, d_2$  are  $Q$ -conjugate

$$(b) \quad f(x) = \frac{1}{2} x^T Q x - b^T x, \quad Q = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}, \quad b = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}, \quad x^* = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$x^{k+1} = x^k + \alpha_k d_k \quad \text{with } \alpha_k = -\frac{g_k^T d_k}{d_k^T \alpha d_k} \text{ where } g_k = Q x^k - b$$

$$g_0 = Q x^* - b = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \\ 0 \end{pmatrix}$$

$$\alpha_0 = -\frac{g_0^T d_0}{d_0^T \alpha d_0} = -\frac{(1 \ -1 \ 0) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}}{(1 \ 0 \ 0) \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}} = -\frac{-1}{1} = 1$$

$$x' = x^* + \alpha_0 d_0 = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} + 1 \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix}$$

$$g_1 = Q x' - b = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} - \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\alpha_1 = -\frac{g_1^T d_1}{d_1^T \alpha d_1} = -\frac{(0 \ -1 \ 0) \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}}{(1 \ 0 \ -1) \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}} = 0$$

$$x'' = x' + \alpha_1 d_1 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + 0 \cdot \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$

$$g_2 = Q x'' - b = \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} - \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}$$

$$\alpha_2 = -\frac{g_2^T d_2}{d_2^T \alpha d_2} = -\frac{(0 \ 0 \ 1) \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}}{(1 \ 0 \ 0) \begin{pmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}} = \frac{-1}{(1 \ 0 \ 0) \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix}} = \frac{-1}{1} = -1$$

$$x''' = x'' + \alpha_2 d_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} + (-1) \begin{pmatrix} 0 \\ 0 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix}$$