MSI exercise 3

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1 Task 1

1.

$$\nabla f(x) = 2Qx + c, \nabla^2 f(x) = 2Q \tag{1}$$

2. Q should be invertible and Positive Semi-Definite

3.

$$2Qx^* + c = 0$$

$$x^* = -\frac{1}{2}Q^{-1}C$$
(2)

$$f(x^*) = \frac{1}{4}c^T Q^{-1} Q Q^{-1} c - \frac{1}{2}c^T Q^{-1} c$$

$$f(x^*) = -\frac{1}{4}c^T Q^{-1} c$$
(3)

2 Task 2

$$\nabla f(x) = 2 \begin{bmatrix} 4 & 1 \\ 1 & 2 \end{bmatrix} x + \begin{bmatrix} 3 \\ 4 \end{bmatrix} = 0 \tag{4}$$

$$x = -\frac{1}{2} \begin{bmatrix} 4 & 1 \\ 1 & 2 \end{bmatrix}^{-1} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} -0.1429 \\ -0.9286 \end{bmatrix}$$
 (5)

There exists a single value for x that satisfies the First order Necessary Condition. This is the Global Minimizer as well because the objective function is convex.

3 Task 3

$$E[S^{2}] = E\left[\frac{1}{N-1} \sum_{n=1}^{N} (Y(n) - M(Y_{N}))^{2}\right]$$

$$= \frac{1}{N-1} \sum_{n=1}^{N} E\left[Y(n)^{2} - 2 * Y(n) * M(Y_{N}) + M(Y_{N})^{2}\right]$$

$$= \frac{N}{N-1} \left(E[Y(n)^{2}] - 2E[Y(n) * M(Y_{N})] + E[M(Y_{N})^{2}]\right)$$

$$= \frac{N}{N-1} \left(E[Y(n)^{2}] - 2E\left[Y(n) * \frac{(Y(1) + \dots + Y(n) + \dots + Y(N))}{N}\right] + E[M(Y_{N})^{2}]\right)$$

$$= \frac{N}{N-1} \left(\sigma_{Y}^{2} + \mu_{y}^{2} - 2\left(\frac{\sigma_{Y}^{2}}{N} + \mu_{y}^{2}\right) + \frac{\sigma_{Y}^{2}}{N} + \mu_{y}^{2}\right)$$

$$= \frac{N}{N-1} \left(\sigma_{Y}^{2} - \frac{\sigma_{Y}^{2}}{N}\right) = \sigma_{Y}^{2}$$
(6)

If we use the factor (N - 1) for the variance, we end up with an unbiased estimator, as shown.

4 Task 4

• Part b

$$\Phi = \begin{bmatrix} 1 & i(1) \\ 1 & i(2) \\ \vdots & \vdots \\ 1 & i(N) \end{bmatrix}, y = \begin{bmatrix} u(1) \\ u(2) \\ \vdots \\ u(N) \end{bmatrix}, \theta = \begin{bmatrix} E \\ R \end{bmatrix}$$
(7)

$$m = \Phi\theta \tag{8}$$

• Part e The second dataset is noisier. The first dataset assumes Gaussian Noise. The second dataset assumes Uniform Noise.