EE 416 - Image Processing and Computer Vision

(UH Manoa, Fall 2020)

Homework Assignment #6

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Problem 1, a)

$$\begin{split} E[||x - x^*||^2] = & E[(x - y^T \theta)^2] \\ = & E[x^2 - 2\theta^T y x + \theta^T y y^T \theta] \\ = & E[x^2] - 2\theta^T E[y x] + \theta^T E[y y^t] \theta \\ = & \sigma^2 - 2\theta^T r_{y,x} + \theta^T C_y \theta \end{split}$$

Problem 1, b)

$$\underset{\theta}{\operatorname{argmin}} E[||x - x^*||^2] = \frac{\partial}{\partial \theta} E[||x - x^*||^2]$$

$$= \frac{\partial}{\partial \theta} [\sigma^2 - 2\theta^T r_{y,x} + \theta^T C_y \theta]$$

$$= -2r_{y,x} + 2C_y \theta$$

$$= 0$$

Therefore

$$\theta = C_y^{-1} r_{y,x}$$

Problem 1, c)

$$C_y = E[yy^T]$$

$$= E[(ax + n)(ax + n)^T]$$

$$= E[x^2aa^T + xan^T + xna^T + nn^T]$$

$$= E[x^2]aa^T + E[xn^T]a + E[xn]a^T + E[nn^T]$$
Since n is independent. $E[xn] = E[xn^T] = 0$

$$= \sigma^2aa^T + 0 + 0 + C_n$$

$$= \sigma^2aa^T + C_n$$

Problem 1, d)

$$r_{y,x} = E[yx]$$

$$= E[(ax + n)(x)]$$

$$= E[ax^{2} + nx]$$

$$= aE[x^{2}] + E[n]E[x]$$

$$= \sigma^{2}a + 0$$

$$= \sigma^{2}a$$

Problem 1, e)

$$x^* = y^T \theta$$

$$= y^T C_y^{-1} r_{y,x}$$

$$= (ax + n)^T (\sigma^2 a a^T + C_n)^{-1} \sigma^2 a$$

Problem 2, a)

$$E[MSE_{2}(\hat{\theta})] = E\left[\frac{1}{M} \sum_{n \in S_{2}} ||x_{n} - f(y_{n}, \hat{\theta})||_{2}^{2}\right]$$

$$= \frac{1}{M} E\left[\sum_{n \in S_{2}} ||x_{n} - f(y_{n}, \hat{\theta})||_{2}^{2}\right]$$

$$= \frac{1}{M} \sum_{n \in S_{2}} E[||x_{n} - f(y_{n}, \hat{\theta})||_{2}^{2}]$$

Since when we know that $\hat{\theta} > \theta^*$.

$$\frac{1}{M} \sum_{n \in S_2} E[||x_n - f(y_n, \hat{\theta})||_2^2] \ge \frac{1}{M} \sum_{n \in S_2} E[||x_n - f(y_n, \theta^*)||_2^2]$$
$$E[MSE_2(\hat{\theta})] \ge E[MSE_2\theta^*]$$

Thus, $MSE_2(\theta^*)$ is smaller.

Problem 2, b)

We will need to know the distribution of x_n and y_n . This values are usually not given/available.

Problem 2, c)

I think L needs to be at least larger than p, where p is the dimension of $\theta(\theta \in R^p)$. Is also important to note that if $L \Rightarrow \infty$, the better $\hat{\theta}$ becomes.

Problem 2, d)

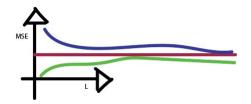


Figure 1: Problem 2-d

Blue Line (Top line) is $MSE_2\hat{\theta}$ Purple Line (Middle Line) is $MSE_2(\theta^*)$ Green Line (Bottom Line) is $MSE_1(\hat{\theta})$

Problem 3, e)

The PSNR for the moving average filter (22.8478) is worst than the learned filter (24.9928).