

① $X = X_c + Z$ X, Y zero-mean

$\hat{X} = K_0 Y$ where $K_0 R_Y = R_{XY}$

$$R_{XY} = E[X Y^H] = E[(X_c + Z) Y^H] = E[X_c Y^H] + E[Z Y^H]$$

$$= E[X_c Y^H] = R_{X_c Y}$$

Therefore, $K_0 R_Y = R_{X_c Y}$.

$\hat{X}_c = K_0 Y = \hat{X}$.

[Even if Y was not zero mean, we will get the same result.]

$\hat{X} = K_0 (Y - E[Y])$

where $K_0 R_Y = R_{X_c Y}$ $R_Y = E[(Y - E[Y])(Y - E[Y])^H]$

$R_{X_c Y} = E[X_c (Y - E[Y])^H] = R_{X_c Y}$

$\Rightarrow \hat{X}_c = \hat{X}$

② $\hat{X} = K_0 Y$ where K_0 is any solution to $K_0 R_Y = R_{XY}$.

We want to minimize $E[\tilde{X}^H W \tilde{X}]$ for some $W \geq 0$. ($\tilde{X}' = X - \hat{X}'$)
 $\hat{X}' = K Y$.

$$E[\tilde{X}^H W \tilde{X}] = E[(X - KY)^H W (X - KY)]$$

$$= E[(X - K_0 Y + K_0 Y - KY)^H W (X - K_0 Y + K_0 Y - KY)]$$

$$= E[(X - K_0 Y)^H W (X - K_0 Y)] + E[(X - K_0 Y)^H W (K_0 Y - KY)]$$

$$+ E[(K_0 Y - KY)^H W (X - K_0 Y)] + E[(K_0 Y - KY)^H W (K_0 Y - KY)]$$

We know that $K_0 Y$ is the linear MMSE estimate of X from Y .

Therefore, we know $(X - K_0 Y) \perp Y$, ie, $E[(X - K_0 Y) Y^H] = 0$
error

Let $\tilde{X}_e = X - K_0 Y$. $E[\tilde{X}_e Y^H] = 0 \Rightarrow E[\tilde{X}_e Y_j^*] = 0 \quad \forall i, j$.

