

Complex value derivative identities

$$\frac{\partial \mathbf{a}^H \mathbf{x}}{\partial \mathbf{x}} = \mathbf{a}^* \quad (1)$$

$$\frac{\partial \mathbf{x}^H \mathbf{a}}{\partial \mathbf{x}} = \mathbf{0} \quad (2)$$

$$\frac{\partial \mathbf{x}^H \mathbf{A} \mathbf{x}}{\partial \mathbf{x}} = \mathbf{A}^T \mathbf{x}^* \quad (3)$$

$$\frac{\partial \mathbf{y}^H \mathbf{X} \mathbf{a}}{\partial \mathbf{X}} = \mathbf{y}^* \mathbf{a}^T \quad (4)$$

Minimum Mean Square Error (MMSE)

The system model:

$$\mathbf{y} = \mathbf{H} \mathbf{x} + \mathbf{n} \quad (5)$$

The MMSE estimator of \mathbf{x} is:

$$\arg \min_{\mathbf{W}} \mathbb{E} \{ \|\mathbf{W} \mathbf{y} - \mathbf{x}\|^2 \} \quad (6)$$

where the expectation is with respect to \mathbf{x} . To find it:

$$\begin{aligned} \frac{\partial \mathbb{E} \{ \|\mathbf{W}^H \mathbf{y} - \mathbf{x}\|^2 \}}{\partial \mathbf{W}} = \\ \mathbb{E} \left\{ \frac{\partial (\mathbf{W}^H \mathbf{y} \mathbf{y}^H \mathbf{W} - \mathbf{y}^H \mathbf{W} \mathbf{x} - \mathbf{x}^H \mathbf{W}^H \mathbf{y} + \|\mathbf{x}\|^2)}{\partial \mathbf{W}} \right\} = 0 \end{aligned} \quad (7)$$

$$(\mathbf{y} \mathbf{y}^H)^T \mathbf{W}^* - \mathbf{y}^* \mathbf{x}^T = 0 \quad (8)$$

which results in:

$$\mathbf{W} = \mathbb{E} \{ (\mathbf{y} \mathbf{y}^H)^{-1} \mathbf{y} \mathbf{x}^H \} \quad (9)$$

Substitute \mathbf{y} with $\mathbf{H} \mathbf{x} + \mathbf{n}$ and we get:

$$\mathbf{W} = (\mathbf{H} \mathbf{R}_{\mathbf{x}\mathbf{x}} \mathbf{H}^H + \mathbf{R}_{\mathbf{nn}})^{-1} \mathbf{H} \mathbf{R}_{\mathbf{x}\mathbf{x}} \quad (10)$$

Therefore, to estimate \mathbf{x} :

$$\hat{\mathbf{x}} = \mathbf{R}_{\mathbf{x}\mathbf{x}} \mathbf{H}^H (\mathbf{H} \mathbf{R}_{\mathbf{x}\mathbf{x}} \mathbf{H}^H + \mathbf{R}_{\mathbf{nn}})^{-1} \mathbf{y} \quad (11)$$