## 1 Snapshot Model

$$x(t) = v(\theta)f(t) + n(t)$$

$$X = VF + N$$

$$X = X_s + N$$

- x (Nx1) Received signal vector at time t
- v (NxD) Matrix of Steering Vectors
- f(t) (Dx1) Zero Mean Random Vector that contains desired and possibly undesired signals (D signals in total)
- n(t) (Nx1) Complex Additive White Gaussian Noise
- X (NxT) Full Signal as received by the Array
- V (NxD) Matrix of Steering Vectors
- F (DxT) Zero Mean Random Vector that contains desired and possibly undesired signals (D signals in total)
- N (NxT) Complex Additive White Gaussian Noise

## 2 Bartlett Beamforming

$$y = w^{\mathsf{H}} X$$
 
$$y = v^{\mathsf{H}}(\theta_s) X$$
 
$$f = y = v^{\mathsf{H}}(\theta_s) X$$
 
$$\mathcal{P} = \frac{1}{T} \sum_{t=0}^{T-1} |v^{\mathsf{H}}(\theta_s) x(t)|^2$$

$$\begin{split} \mathcal{P} &= Var(f) \\ &= E[(f - \mu_f)^2] \\ &= E[f^2] \qquad \text{zero mean signal} \\ &= E[|v^{\mathsf{H}}(\theta_s)X|^2] \qquad \text{ideal Beamformer} \\ &= \frac{1}{T} \sum_{t=0}^{T-1} |v^{\mathsf{H}}(\theta_s)x(t)|^2 \end{split}$$

$$\begin{split} \mathcal{P} &= \frac{1}{T} \sum_{t=0}^{T-1} |v^{\mathsf{H}}(\theta) x(t)|^2 \\ &= \frac{1}{T} \sum_{t=0}^{T-1} (v^{\mathsf{H}} x(t)) (v^{\mathsf{H}} x(t))^{\mathsf{H}} \\ &= \frac{1}{T} \sum_{t=0}^{T-1} v^{\mathsf{H}} x(t) x^{\mathsf{H}}(t) v_s \\ &= v^{\mathsf{H}} \left( \frac{1}{T} \sum_{t=0}^{T-1} x(t) x^{\mathsf{H}}(t) \right) v \\ &= v^{\mathsf{H}} E[XX^{\mathsf{H}}] v \\ \mathcal{P}_B &= v^{\mathsf{H}} R_{XX} v \end{split}$$

## 3 Capon Beamformer

$$\hat{y} = w^{\mathsf{H}} X$$

$$\hat{y} = f$$

$$v^{\mathsf{H}}(\theta_s) w = 1$$

$$f = y = v^{\mathsf{H}}(\theta_s) X$$

$$\mathcal{P} = \frac{1}{T} \sum_{t=0}^{T-1} |v^{\mathsf{H}}(\theta_s) x(t)|^2$$

$$\begin{split} \hat{y} &= w^{\mathsf{H}} X \\ &= w^{\mathsf{H}} (X_s + N) \\ &= w^{\mathsf{H}} X_s + w^{\mathsf{H}} N \\ &= f + w^{\mathsf{H}} N \qquad \text{Distortionaless Response} \\ \hat{y} &= f + Y_n \end{split}$$

$$\begin{split} Var(\hat{y}) &= E[|\hat{y}|^2] + E[\hat{y}]^2 \\ &= E[|\hat{y}|^2] + E[f + Y_n]^2 \\ &= E[|\hat{y}|^2] + (E[f] + E[Y_n])^2 \qquad \text{both } f \text{ and } Y_n \text{ are zero mean} \\ &= E[|\hat{y}|^2] + 0 \\ &= E[|w^{\text{H}}X|^2] \\ &= E[(w^{\text{H}}X)(w^{\text{H}}X)^{\text{H}}] \\ &= E[w^{\text{H}}XX^{\text{H}}w] \\ &= w^{\text{H}}E[XX^{\text{H}}]w \\ &= w^{\text{H}}\left(\frac{1}{T}\sum_{t=0}^{T-1}x(t)x(t)^{\text{H}}\right)w \\ &= w^{\text{H}}R_{XX}w \end{split}$$

Minimize  $w^{\mathsf{H}}R_{XX}w$ subject to :  $w^{\mathsf{H}}v_s=1$ 

first construct the objective function

$$\mathcal{L} = w^{\mathsf{H}} R_{XX} w + \mathsf{Re}[\lambda(w^{\mathsf{H}} v_s - 1)]$$

$$\begin{split} \mathcal{L} &= \boldsymbol{w}^{\mathsf{H}} R_{XX} \boldsymbol{w} + \mathrm{Re}[\lambda(\boldsymbol{w}^{\mathsf{H}} \boldsymbol{v}_s - 1)] \\ &= \boldsymbol{w}^{\mathsf{H}} R_{XX} \boldsymbol{w} + \frac{\lambda}{2} (\boldsymbol{w}^{\mathsf{H}} \boldsymbol{v}_s - 1) + \left[ \frac{\lambda}{2} (\boldsymbol{w}^{\mathsf{H}} \boldsymbol{v}_s - 1) \right]^* \\ &= \boldsymbol{w}^{\mathsf{H}} R_{XX} \boldsymbol{w} + \frac{\lambda}{2} (\boldsymbol{w}^{\mathsf{H}} \boldsymbol{v}_s - 1) + \frac{\lambda^*}{2} (\boldsymbol{w}^{\mathsf{T}} \boldsymbol{v}_s^* - 1) \\ &= \boldsymbol{w}^{\mathsf{H}} R_{XX} \boldsymbol{w} + \frac{\lambda}{2} (\boldsymbol{w}^{\mathsf{H}} \boldsymbol{v}_s - 1) + \frac{\lambda^*}{2} (\boldsymbol{v}_s^{\mathsf{H}} \boldsymbol{w} - 1) \end{split}$$

$$\mathcal{L} = w^{\mathsf{H}} R_{XX} w + \frac{\lambda}{2} (w^{\mathsf{H}} v_s - 1) + \frac{\lambda^*}{2} (v_s^{\mathsf{H}} w - 1)$$
$$\frac{\partial \mathcal{L}}{\partial w^{\mathsf{H}}} = R_{XX} w + \frac{\lambda}{2} v_s = 0$$
$$R_{XX} w = -\frac{\lambda}{2} v_s$$
$$w = -\frac{\lambda}{2} R_{XX}^{-1} v_s$$

$$\begin{split} \boldsymbol{w}^{\mathsf{H}} \boldsymbol{v}_s &= \boldsymbol{v}_s^{\mathsf{H}} \boldsymbol{w} = 1 \\ \boldsymbol{v}_s^{\mathsf{H}} \boldsymbol{w} &= -\frac{\lambda}{2} \boldsymbol{v}_s^{\mathsf{H}} \boldsymbol{R}_{XX}^{-1} \boldsymbol{v}_s = 1 \\ &-\frac{\lambda}{2} = (\boldsymbol{v}_s^{\mathsf{H}} \boldsymbol{R}_{XX}^{-1} \boldsymbol{v}_s)^{-1} \end{split}$$

$$w_{mvdr} = (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} R_{XX}^{-1} v_s$$

**OLD** 

$$\begin{split} w_{mvdr} &= (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} R_{XX}^{-1} v_s \\ \mathcal{P} &= \frac{1}{T} \sum_{t=0}^{T-1} |w^{\mathsf{H}} x(t)|^2 \\ &= \frac{1}{T} \sum_{t=0}^{T-1} |(v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} v_s^{\mathsf{H}} R_{XX}^{-1} x(t)|^2 \\ &= \left| (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} \right|^2 (v_s^{\mathsf{H}} R_{XX}^{-1}) \frac{1}{T} \sum_{t=0}^{T-1} |x(t)|^2 (R_{XX}^{-1} v_s) \\ &= \left| (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} \right|^2 (v_s^{\mathsf{H}} R_{XX}^{-1}) R_{XX} (R_{XX}^{-1} v_s) \\ &= \left| (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} \right|^2 v_s^{\mathsf{H}} R_{XX}^{-1} v_s \\ \mathcal{P}_{mvdr} &= (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} \end{split}$$

NEW

$$\begin{split} w_{mvdr} &= (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} R_{XX}^{-1} v_s \\ \mathcal{P} &= \frac{1}{T} \sum_{t=0}^{T-1} |w^{\mathsf{H}} x(t)|^2 \\ &= \frac{1}{T} \sum_{t=0}^{T-1} |(v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} v_s^{\mathsf{H}} R_{XX}^{-1} x(t)|^2 \\ &= \left| (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} \right|^2 \frac{1}{T} \sum_{t=0}^{T-1} |(v_s^{\mathsf{H}} R_{XX}^{-1} x(t)|^2 \\ &= \left| (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1} \right|^2 (v_s^{\mathsf{H}} R_{XX}^{-1} R_{XX})^2 \end{split}$$

NEW

$$\mathcal{P}_{mvdr} = (v_s^{\mathsf{H}} R_{XX}^{-1} v_s)^{-1}$$