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A. DERIVING LINEAR CONSTRAINED MINIMUM VARIANCE DECTECTOR (LCMV)

Let $\mathbf{w}^* = \underset{\mathbf{w}}{\operatorname{argmin}} \{ \mathbf{w}^T \mathbf{R} \mathbf{w} \}$ such that $\mathbf{D}^T \mathbf{w} = \mathbf{c}$

 $\text{where} \ \ \boldsymbol{w} \in \mathfrak{R}^{Lx1} \text{ , } \ \boldsymbol{R} \in \mathfrak{R}^{LxL} \text{ , } \ \boldsymbol{D} = [\boldsymbol{d}_1 | \boldsymbol{d}_2 | ... | \boldsymbol{d}_p] \in \mathfrak{R}^{Lxp} \quad \text{and} \quad \boldsymbol{c} \in \mathfrak{R}^{px1}$

B. LAGRANGE MULTIPLER FORM

$$J(\mathbf{w}) = \mathbf{w}^{T} \mathbf{R} \mathbf{w} + \boldsymbol{\lambda}^{T} (\mathbf{D}^{T} \mathbf{w} - \mathbf{c}) \quad \text{such that} \quad \boldsymbol{\lambda} \in \mathfrak{R}^{px1}$$

C. DERIVATIVES OF COST EQUATION

$$\frac{\partial J(\mathbf{w})}{\partial \mathbf{w}} \Big|_{\mathbf{w}^*} = \mathbf{0} \text{ and } \frac{\partial J(\mathbf{w})}{\partial \lambda} \Big|_{\mathbf{\lambda}^*} = \mathbf{0}$$

$$2\mathbf{w}^{\mathrm{T}}\mathbf{R} + \mathbf{\lambda}^{\mathrm{T}}\mathbf{D}^{\mathrm{T}} = \mathbf{0}$$

$$\mathbf{D}^T\mathbf{w} - \mathbf{c} = \mathbf{0}$$

D. SOLVE FOR **w**

$$\mathbf{w}^T \mathbf{R} = -\frac{1}{2} \boldsymbol{\lambda}^T \mathbf{D}^T$$

 $\mathbf{R}\mathbf{w} = -\frac{1}{2}\mathbf{D}\lambda$, note $\mathbf{R} = \mathbf{R}^T$ is symmetric and invertible

$$\mathbf{w} = -\frac{1}{2}\mathbf{R}^{-1}\mathbf{D}\boldsymbol{\lambda}$$

E. Solve for λ by substituting in w

$$\mathbf{D}^{\mathrm{T}}\mathbf{w}-\mathbf{c}=\mathbf{0}$$

 $-\frac{1}{2}D^TR^{-1}D\lambda = c$, note that $D^TR^{-1}D$ is square and full rank and invertible

$$\lambda = -2(\mathbf{D}^{\mathrm{T}}\mathbf{R}^{-1}\mathbf{D})^{-1}\mathbf{c}$$

F. Solve for \boldsymbol{w} by substituting in expression for λ

$$\mathbf{w} = -\frac{1}{2}\mathbf{R}^{-1}\mathbf{D}\boldsymbol{\lambda}$$

$$\mathbf{w}^{\text{LCMV}} = \mathbf{R}^{-1} \mathbf{D} (\mathbf{D}^{\text{T}} \mathbf{R}^{-1} \mathbf{D})^{-1} \mathbf{c}$$

A. DERIVING LINEAR CONSTRAINED MINIMUM VARIANCE CLASSIFIER (LCMVC)

$$\text{Let} \quad \boldsymbol{w}_{j}^{*} = \underset{\boldsymbol{w}_{j}}{\text{argmin}} \, \{\boldsymbol{w}_{j}^{T}\boldsymbol{R}\boldsymbol{w}_{j}\} \quad \text{ such that } \quad \boldsymbol{M}^{T}\boldsymbol{w}_{j} = \boldsymbol{c}_{j} \text{, where } 1 \leq j \leq p$$

 $\text{where} \quad \mathbf{w}_j \in \mathfrak{R}^{Lx1} \;, \quad \mathbf{R} \in \mathfrak{R}^{LxL} \;, \quad \mathbf{M} = [\mathbf{m}_1 | \mathbf{m}_2 | ... | \mathbf{m}_p] \in \mathfrak{R}^{Lxp} \quad \text{ and } \quad \mathbf{c} = (0,...,c_j,...,0)^T \in \mathfrak{R}^{px1} \;.$

B. LAGRANGE MULTIPLER FORM

$$J(\boldsymbol{w}_j) = \boldsymbol{w}_j^T \boldsymbol{R} \boldsymbol{w}_j + \boldsymbol{\lambda}^T \, (\boldsymbol{M}^T \boldsymbol{w}_j - \boldsymbol{c}_j) \quad \text{ such that } \quad \boldsymbol{\lambda} \in \mathfrak{R}^{px1}$$

C. DERIVATIVES OF COST EQUATION

$$\frac{\partial J(\mathbf{w}_j)}{\partial \mathbf{w}_j} \Big|_{\mathbf{w}_i^*} = \mathbf{0} \text{ and } \frac{\partial J(\mathbf{w}_j)}{\partial \lambda} \Big|_{\mathbf{\lambda}^*} = \mathbf{0}$$

$$2\boldsymbol{w}_{j}^{T}\boldsymbol{R}+\boldsymbol{\lambda}^{T}\boldsymbol{M}^{T}=\boldsymbol{0}$$

$$\boldsymbol{M}^T\boldsymbol{w}_j-\boldsymbol{c}_j=\boldsymbol{0}$$

D. SOLVE FOR **w**

$$\mathbf{w}_{i}^{T}\mathbf{R} = -\frac{1}{2}\boldsymbol{\lambda}^{T}\mathbf{M}^{T}$$

 $\mathbf{R}\mathbf{w}_{i} = -\frac{1}{2}\mathbf{M}\lambda$, note $\mathbf{R} = \mathbf{R}^{T}$ is symmetric and invertible

$$\mathbf{w}_{j} = -\frac{1}{2}\mathbf{R}^{-1}\mathbf{M}\boldsymbol{\lambda}$$

E. Solve for λ by substituting in w

$$\boldsymbol{M}^T\boldsymbol{w}_i - \boldsymbol{c}_i = \boldsymbol{0}$$

 $-\frac{1}{2}\mathbf{M}^T\mathbf{R}^{-1}\mathbf{M}\boldsymbol{\lambda} = \mathbf{c}_j$, note that $\mathbf{M}^T\mathbf{R}^{-1}\mathbf{M}$ is square and full rank and invertible

$$\boldsymbol{\lambda} = -2(\mathbf{M}^{\mathrm{T}}\mathbf{R}^{-1}\mathbf{M})^{-1}\mathbf{c}_{j}$$

F. Solve for \boldsymbol{w} by substituting in expression for λ

$$\mathbf{w}_{i} = -\frac{1}{2}\mathbf{R}^{-1}\mathbf{M}\boldsymbol{\lambda}$$

$$\boldsymbol{w}_{j}^{LCMVC} = \boldsymbol{R}^{-1}\boldsymbol{M}(\boldsymbol{M}^{T}\boldsymbol{R}^{-1}\boldsymbol{M})^{-1}\boldsymbol{c}_{j}$$

A. DERIVING TARGET-CONSTRAINED INTERFERENCE-MINIMIZED FILTER (TCIMF)

$$\text{Let} \quad \boldsymbol{w}^* = \underset{\boldsymbol{w}}{\text{argmin}} \{ \boldsymbol{w}^T \boldsymbol{R} \boldsymbol{w} \} \quad \text{ such that } \quad [\boldsymbol{D} \boldsymbol{U}]^T] \boldsymbol{w} = {[\boldsymbol{1}_{px1}^T \boldsymbol{0}_{qx1}^T]}^T$$

where
$$\mathbf{w} \in \mathfrak{R}^{Lx1}$$
, $\mathbf{R} \in \mathfrak{R}^{LxL}$, $\mathbf{D} = [\mathbf{d}_1 | \mathbf{d}_2 | ... | \mathbf{d}_p]$ and $\mathbf{U} = [\mathbf{u}_1 | \mathbf{u}_2 | ... | \mathbf{u}_q]$

B. LAGRANGE MULTIPLIER FORM

$$J(\boldsymbol{w}) = \boldsymbol{w}^T \boldsymbol{R} \boldsymbol{w} + \boldsymbol{\lambda}^T \left([\boldsymbol{D} \boldsymbol{U}]^T \boldsymbol{w} - [\boldsymbol{1}_{px1}^T \boldsymbol{0}_{qx1}^T]^T \right) \quad \text{such that} \quad \boldsymbol{\lambda} \in \mathfrak{R}^{(p+q)x1}$$

C. DERIVATIVES OF COST EQUATION

$$\frac{\partial J(\mathbf{w})}{\partial \mathbf{w}} \Big|_{\mathbf{w}^*} = \mathbf{0} \text{ and } \frac{\partial J(\mathbf{w})}{\partial \lambda} \Big|_{\mathbf{\lambda}^*} = \mathbf{0}$$

$$2\mathbf{w}^{\mathrm{T}}\mathbf{R} + \mathbf{\lambda}^{\mathrm{T}} [\mathbf{D}\mathbf{U}]^{\mathrm{T}} = \mathbf{0}$$

$$[\mathbf{D}\mathbf{U}]^T\mathbf{w} - [\mathbf{1}_{px1}^T\mathbf{0}_{qx1}^T]^T = \mathbf{0}$$

D. SOLVE FOR **w**

$$\mathbf{w}^T \mathbf{R} = -\frac{1}{2} \boldsymbol{\lambda}^T [\mathbf{D} \mathbf{U}]^T$$

 $\mathbf{R}\mathbf{w} = -\frac{1}{2} [\mathbf{D}\mathbf{U}] \lambda$, note $\mathbf{R} = \mathbf{R}^{\mathrm{T}}$ is symmetric and invertible

$$\mathbf{w} = -\frac{1}{2}\mathbf{R}^{-1} \left[\mathbf{D} \mathbf{U} \right] \boldsymbol{\lambda}$$

E. Solve for λ by substituting in w

$$[\boldsymbol{D}\boldsymbol{U}]^T\boldsymbol{w}\!=\![\boldsymbol{1}_{px1}^T\boldsymbol{0}_{qx1}^T]^T$$

$$-\frac{1}{2}\left[\mathbf{D}\mathbf{U}\right]^{T}\mathbf{R}^{-1}\left[\mathbf{D}\mathbf{U}\right]\boldsymbol{\lambda}=\left[\mathbf{1}_{nx1}^{T}\mathbf{0}_{\alpha x1}^{T}\right]^{T}\text{, note that }\left[\mathbf{D}\mathbf{U}\right]^{T}\mathbf{R}^{-1}\left[\mathbf{D}\mathbf{U}\right]\text{ is square and full rank} => invertible$$

$$\lambda = -2([\mathbf{D}\mathbf{U}]^{\mathrm{T}}\mathbf{R}^{-1}[\mathbf{D}\mathbf{U}])^{-1}[\mathbf{1}_{\mathrm{px1}}^{\mathrm{T}}\mathbf{0}_{\mathrm{qx1}}^{\mathrm{T}}]^{\mathrm{T}}$$

F. Solve for \boldsymbol{w} by substituting in expression for λ

$$\mathbf{w} = -\frac{1}{2}\mathbf{R}^{-1} [\mathbf{D}\mathbf{U}] \lambda$$

$$\mathbf{w}^{\mathrm{TCIMF}} = \mathbf{R}^{-1} \left[\mathbf{D} \mathbf{U} \right] \left(\left[\mathbf{D} \mathbf{U} \right]^{\mathrm{T}} \mathbf{R}^{-1} \left[\mathbf{D} \mathbf{U} \right] \right)^{-1} \left[\mathbf{1}_{\mathrm{px1}}^{\mathrm{T}} \mathbf{0}_{\mathrm{qx1}}^{\mathrm{T}} \right]^{\mathrm{T}}$$

REFERENCES

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