

## APPENDIX A: THE AUTOCORRELATION MATRIX

There are many ways to introduce this theory. We will consider the power output to a *WSS (Wide-Sense Stationary)* input for a *FIR* filter:

$$\underline{w} \equiv [w_0 \ w_1 \ w_2 \ \dots \ w_{N-1}]^T \quad \underline{x}(u) \equiv [x(u) \ x(u-1) \ x(u-2) \ \dots \ x(u-N+1)]^T$$

$$\boxed{y(u) = \underline{w}^H \underline{x}(u)}$$

( $H$  denotes transpose-conjugate for complex data and just only transpose for real data)

The filter output power is then given by:

$$p_y \equiv E[|y(u)|^2] = E[y(u)y^*(u)] = \underline{w}^H E[\underline{x}(u)\underline{x}^H(u)] \underline{w} \equiv \underline{w}^H \underline{\underline{R}} \underline{w}$$

where:

$$\underline{\underline{R}} \equiv E[\underline{x}(u)\underline{x}^H(u)] \quad \text{Autocorrelation Matrix}$$

**Thus:**

$$\underline{\underline{R}} = E \left[ \underline{x}(u) \underline{x}^H(u) \right] = \begin{bmatrix} R_{xx}(0) & R_{xx}(1) & \cdots & R_{xx}(N-1) \\ R_{xx}^*(1) & R_{xx}(0) & \cdots & \vdots \\ R_{xx}^*(2) & \vdots & \ddots & \vdots \\ \vdots & \vdots & \cdots & R_{xx}(1) \\ R_{xx}^*(N-1) & \cdots & \cdots & R_{xx}(0) \end{bmatrix}$$

Toeplitz matrix

**We see also that:**

$$P_y \geq 0 \quad \forall \underline{w} \Rightarrow \underline{w}^H \underline{\underline{R}} \underline{w} \geq 0 \quad (\text{real}) \quad \forall \underline{w} \quad \underline{\underline{R}} \text{ Semi-positive matrix}$$

**and:**

$$\underline{\underline{R}}^H = \underline{\underline{R}} \quad \underline{\underline{R}} \text{ Hermitic matrix} \longrightarrow \text{The eigenstructure is very useful}$$

We now consider:

$$\underline{\underline{R}}\underline{q} = \lambda\underline{q} \quad \text{with} \quad \|\underline{q}\|^2 = \underline{q}^H \underline{q} = 1$$

1.-  $\lambda \geq 0$  and real

$$\text{From } \underline{\underline{R}}\underline{q} = \lambda\underline{q} \Rightarrow \underline{q}^H \underline{\underline{R}}\underline{q} = \lambda \underline{q}^H \underline{q} = \lambda \geq 0 \quad !!$$

Real!!

$$2.- \left. \begin{array}{l} \underline{\underline{R}}\underline{q}_i = \lambda_i \underline{q}_i \\ \underline{\underline{R}}\underline{q}_j = \lambda_j \underline{q}_j \end{array} \right\} \text{with } \lambda_i \neq \lambda_j \Rightarrow \underline{q}_i^H \underline{q}_j = \delta_{ij} \text{ (orthonormal)}$$

$$\left. \begin{array}{l} \underline{q}_j^H \underline{\underline{R}}\underline{q}_i = \lambda_i \underline{q}_j^H \underline{q}_i \\ \underline{q}_i^H \underline{\underline{R}}\underline{q}_j = \lambda_j \underline{q}_i^H \underline{q}_j \end{array} \right\} \quad 0 = (\lambda_i - \lambda_j) \underline{q}_j^H \underline{q}_i$$

$$\lambda_i \neq \lambda_j \Rightarrow \underline{q}_j^H \underline{q}_i = 0$$

↑  
 $\lambda_i, \lambda_j$  real

$$3.- \underline{\underline{Q}} \equiv [\underline{q}_1 \quad \underline{q}_2 \quad \dots \quad \underline{q}_N]$$

$$\underline{\underline{\Lambda}} \equiv \text{diag}[\lambda_1 \quad \lambda_2 \quad \dots \quad \lambda_N]$$

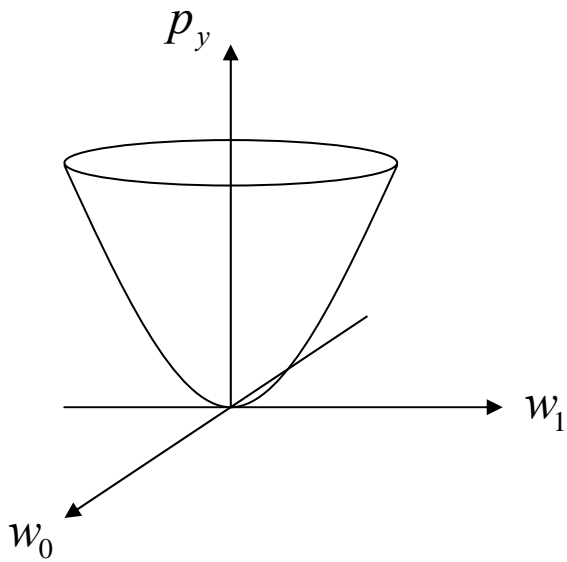
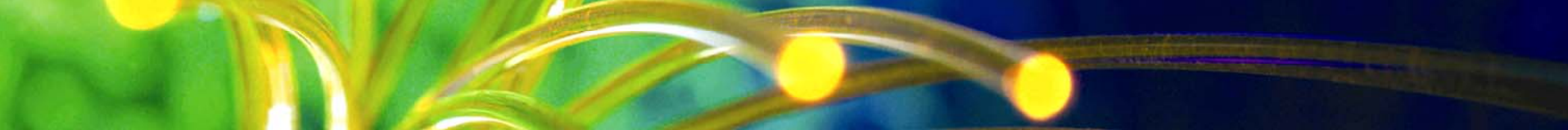
$$\underline{\underline{R}} = \underline{\underline{Q}} \underline{\underline{\Lambda}} \underline{\underline{Q}}^H$$

with  $\underline{\underline{Q}} \underline{\underline{Q}}^H = \underline{\underline{Q}}^H \underline{\underline{Q}} = \underline{\underline{I}}$  Unitary

Or:

$$\underline{\underline{R}} = \sum_{i=1}^N \lambda_i \underline{q}_i \underline{q}_i^H$$

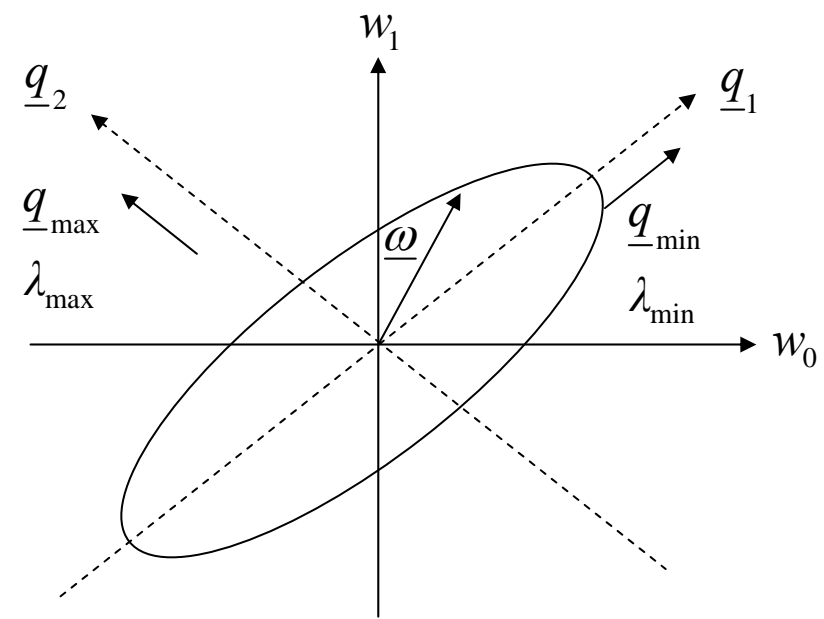
$$4.- \mathbf{P}_y = \underline{w}^H \underline{\underline{R}} \underline{w} = \underline{w}^H \sum_{i=1}^N \lambda_i \underline{q}_i \underline{q}_i^H \underline{w} = \sum_{i=1}^N \lambda_i \underline{w}^H \underline{q}_i \underline{q}_i^H \underline{w} = \sum_{i=1}^N \lambda_i \|\underline{w}^H \underline{q}_i\|^2$$



$$P_y = \sum_{i=1}^N \lambda_i \left\| \underline{w}^H \underline{q}_i \right\|^2$$

Or, in other words:

$$\left. \begin{array}{l} \max \\ \min \end{array} \right\} \underline{w}^H \underline{R} \underline{w}$$



Subject to:

$$\underline{w}^H \underline{w} = 1$$

## APPENDIX B: THE SINGULAR VALUE DECOMPOSITION OF A MATRIX (SVD)

We saw in *appendix A* that for a hermitic  $N \times N$  matrix there exists an orthonormal (unitary) matrix  $\underline{\underline{Q}}$  such that:

$$\underline{\underline{R}} = \underline{\underline{Q}} \underline{\underline{\Lambda}} \underline{\underline{Q}}^H \quad (\text{Schur decomposition}) \quad \text{With} \quad \left| \begin{array}{l} \underline{\underline{Q}}^H \underline{\underline{Q}} = \underline{\underline{Q}} \underline{\underline{Q}}^H = \underline{\underline{I}} \\ \underline{\underline{\Lambda}} = \text{diag}(\lambda_1 \ \lambda_2 \ \dots \ \lambda_N) \quad \lambda_i \geq 0 \end{array} \right.$$

Now, we will consider an  $M \times N$  matrix  $\underline{\underline{H}}$  of rank  $r(\underline{\underline{H}}) = \underline{\underline{R}} > 0$

Then, there exist an  $M \times R$  matrix  $\underline{\underline{S}}$  and an  $N \times R$  matrix  $\underline{\underline{T}}$  such that:

$$\left. \begin{array}{l} \underline{\underline{S}}^H \underline{\underline{S}} = \underline{\underline{I}} \\ \underline{\underline{T}}^H \underline{\underline{T}} = \underline{\underline{I}} \end{array} \right\} \quad \text{with} \quad \underline{\underline{H}} = \underline{\underline{S}} \underline{\underline{\Lambda}} \underline{\underline{T}}^H \quad (\text{SVD})$$
$$\underline{\underline{\Lambda}} = \text{diag}(\lambda_1 \ \lambda_2 \ \dots \ \lambda_N) \quad \lambda_i \geq 0$$

(see ref. [3] for details)