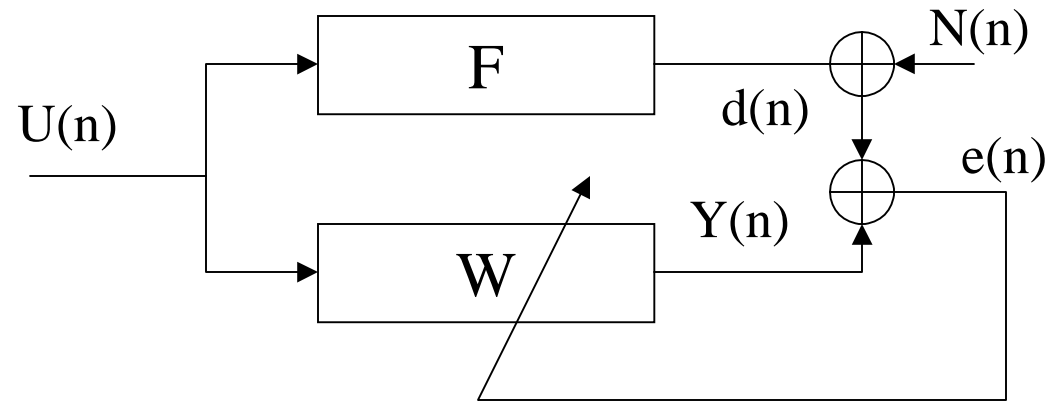


# Optimal Filtering



Focus on linear, discrete time filters using mean square error as the statistical criterion for optimization

The above model covers equalizer, echo canceller, and linear predictors

# Optimal Filter Theory

- Wiener Filtering, spectral factorization:

Estimation error

$$e(n) = d(n) - \sum w_k^* u(n-k)$$

Cost function:

$$\varepsilon = E\{e(n)e^*(n)\} = E\{|d(n)|^2\} - \sum_{k=0}^{M-1} w_k^* \rho(-k) - \sum_{k=0}^{M-1} w_k \rho^*(-k) + \sum_{k=0}^{M-1} \sum_{i=0}^{M-1} w_k^* w_i r(i-k)$$

$$r(i-k) = E\{u(n-k)u^*(n-i)\}$$

$$\rho(i) = E\{u(n+k)d^*(n)\}$$

Minimizing the cost function:

$$\frac{\partial \varepsilon}{\partial w_k} = \rho(-k) - \sum_{i=0}^{M-1} w_{oi} r(i-k) = 0 \quad k = 0, 1, \dots, M-1$$

Optimum coefficients:

In matrix form:

$$\sum_{i=0}^{M-1} w_{oi} r(i-k) = \rho(-k) \quad k = 0, \dots, M-1$$

$$RW_o = P$$

$$R = \begin{pmatrix} r(0) & r(1) & \cdots & r(M-1) \\ r^*(1) & r(0) & & r(M-2) \\ \vdots & & \ddots & \\ r^*(M-1) & r^*(M-2) & & r(0) \end{pmatrix}$$

$$P = [\rho(0), \rho(-1), \dots, \rho(1-M)]^T$$

$$W = [w_{o0}, w_{o1}, \dots, w_{oM-1}]$$

$$\varepsilon = \sigma_d^2 - W^H P - P^H W + W^H R W$$

The optimum solution is

$$W_o = R^{-1} P$$

$$\varepsilon_{\min} = E\{|d(n)|^2\} - P^H W_o$$

$$\boldsymbol{\varepsilon} = \boldsymbol{\varepsilon}_{\min} + (\boldsymbol{w} - \boldsymbol{w}_o)^H \boldsymbol{R}(\boldsymbol{w} - \boldsymbol{w}_o)$$

Canonical form

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

$$\boldsymbol{v} = \boldsymbol{Q}^H(\boldsymbol{w} - \boldsymbol{w}_o)$$

$$\boldsymbol{\varepsilon} = \boldsymbol{\varepsilon}_{\min} + \sum_{k=1}^M \lambda_k \boldsymbol{v}_k \boldsymbol{v}_k^* = \boldsymbol{\varepsilon}_{\min} + \sum_{k=1}^M \lambda_k |\boldsymbol{v}_k|^2$$

The new coefficient vector contains the principle axes of error-performance surface.

- Steepest descent Algorithm: Iterative solution to Wiener equation

$$w(n+1) = w(n) + \mu[-\nabla(\epsilon)]$$

$$w(n+1) = w(n) + \mu[P - RW(n)]$$

- Gradient operator

$$\begin{aligned} \nabla \epsilon(W) &= \left[ \frac{\partial \epsilon(W)}{\partial w_0} \dots \frac{\partial \epsilon(W)}{\partial w_{M-1}} \right] \\ &= -2P + 2RW(n) \end{aligned}$$

- The correction is the expectation of inner product of input vector and estimation error

$$\nabla(\epsilon(W)) = 2E[u(n-k)e^*(n)]$$

Stability analysis:

$$c(n) = w(n) - w_0(n)$$

Then:

$$c(n+1) = (I - \mu R)c(n)$$

$$R = Q\Lambda Q^H$$

$$c(n+1) = (I - \mu Q\Lambda Q^H)c(n)$$

Define:  $v(n) = Q^H c(n)$

Then

$$v(n+1) = (I - \mu\Lambda)v(n)$$

For the kth natural mode:

$$v_k(n) = (1 - \mu\lambda_k)^n v_k(0)$$

Stability requirement:

$$-1 < 1 - \mu\lambda_k < 1 \Rightarrow 0 < \mu < \frac{2}{\lambda_{\max}}$$