

- Tap-weight adjustment of NLMS: $w[n+1] - w[n]$ in the direction of $u[n]$.
- Movement of tap-weight adjustment a function of correlation between $u[n]$ and $u[n-1]$.
- Larger correlation indicates $u[n]$ and $u[n-1]$ produces a smaller tap-weight movement.
- Need to pre-whiten observations to attain a faster uniform rate of convergence.

- Modified Cost function:

$$J(n) = \|w[n+1] - w[n]\|^2 + \sum_{k=0}^{N-1} \lambda_k (d[n-k] - w^T[n+1]u[n-k]).$$

- Some definitions:

$$A^T[n] = [u[n], u[n-1], \dots, u[n-N+1]],$$

$$\Lambda^T[n] = [\lambda_0, \lambda_1, \dots, \lambda_{N-1}],$$

$$d^T[n] = [d[n], d[n-1], \dots, d[n-N+1]].$$

- Matrix form:

$$J(n) = \|w[n+1] - w[n]\|^2 + (d[n] - A[n]w[n+1])^T \Lambda.$$

- Derivative w.r.s.t. Λ :

$$d[n] = A[n]w[n+1].$$

- Derivative w.r.s.t. $w[n+1]$:

$$w[n+1] = w[n] + \frac{1}{2} A^T[n] \Lambda$$

- Tap-weight update:

$$w[n+1] = w[n] + \tilde{\mu} A_R^\dagger[n] (d[n] - A[n]w[n])$$

$$\tilde{\mu} = \frac{\beta}{\|u[n]\|^2 + \delta}$$

- Alternative form of tap-weight update:

$$w[n+1] = (I - \tilde{\mu} A_R^\dagger[n] A[n]) w[n] + \tilde{\mu} A_R^\dagger[n] d[n].$$

- Regularization for pseudo-inverse:

$$\tilde{A}_R^\dagger[n] = A^T[n] (A[n] A^T[n] + \gamma I)^{-1}$$

- Better convergence characteristics comes at the price of computational complexity.