

Assignment #2: FIR causal Wiener solution

Consider the two following setups:

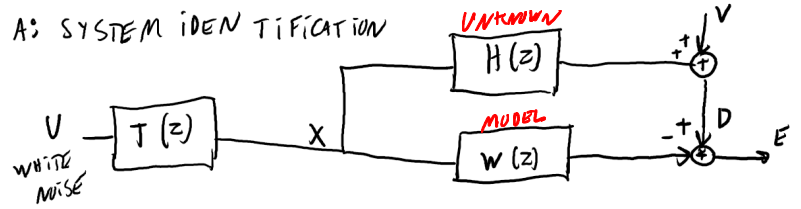
Setup A System identification

$$x(n) = u(n) * t(n)$$

$$d(n) = x(n) * h(n) + v(n)$$

$$e(n) = d(n) - y(n) = d(n) - \sum_{k=0}^{N-1} w_k x(n-k) = d(n) - \mathbf{w}^H \mathbf{x}(n) \text{ with } \mathbf{w} = [w_0^* \ w_1^* \ \dots \ w_{N-1}^*]^T.$$

SETUP A: SYSTEM IDENTIFICATION



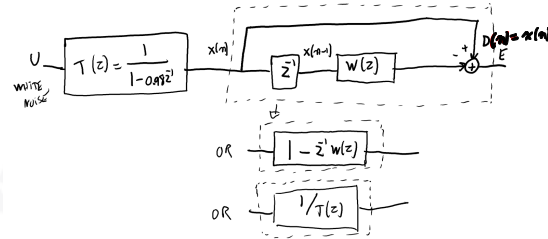
Setup B Linear prediction

$$x(n) = u(n) * t(n)$$

$$d(n) = x(n)$$

$$e(n) = d(n) - y(n) = d(n) - \sum_{k=0}^{N-1} w_k x(n-k-1) = d(n) - \mathbf{w}^H \mathbf{x}(n-1) \text{ with } \mathbf{w} = [w_0^* \ w_1^* \ \dots \ w_{N-1}^*]^T.$$

SETUP B: LINEAR PREDICTION



Use $T(z) = \frac{1}{1-0.98z^{-1}}$ $|z| > 0.98$ ($t(n) = 0.98^n u(n)$), $H(z) = \frac{1}{1-0.95z^{-1}}$ $|z| > 0.95$ ($h(n) = 0.95^n u(n)$), and $u(n), v(n)$ from uncorrelated zero-mean Gaussian white noise stochastic processes with power (and variance) $\sigma^2 = 1$.

1) Assuming a very large value for the number of coefficients N , find an intuitive solution for \mathbf{w}_{opt} in the two setups, just by simple observation (e.g. from the z-transform of the systems).

2) For both setups, using causal FIR Wiener filters \mathbf{w}_{opt} with $N = 10$ coefficients:

- Compute (by coding) the Wiener solution \mathbf{w}_{opt} and the normalized MMSE, based on the theoretical correlation functions $\phi_{dx}(l)$, $\phi_{xx}(l)$, $\phi_{dd}(l)$ (required for \mathbf{R} and \mathbf{p} in $\mathbf{w}_{opt} = \mathbf{R}^{-1} \mathbf{p}$ and for the normalized MMSE $\frac{\xi(\mathbf{w}_{opt})}{\sigma_d^2} = \frac{\sigma_d^2 - \mathbf{p}^H \mathbf{R}^{-1} \mathbf{p}}{\sigma_d^2}$). Plot the resulting \mathbf{w}_{opt} solution for each setup.

$$\phi_d = \phi_{dd}(0)$$

- Note: for Setup A, when $x(n)$ is not white noise (as here) and when the length N of \mathbf{w}_{opt} is shorter than the impulse response $h(n)$ (as here), you will find that \mathbf{w}_{opt} does more than just modeling $h(n)$, and the resulting shape for the coefficients in \mathbf{w}_{opt} may not be fully as expected (i.e., as in part 1)). Explain why this is the case.

3) For both setups, estimate (by coding) the Wiener solution \mathbf{w}_{opt} and the normalized MMSE based on $L = 10,000,000$ samples of the signals $x(n)$, $d(n)$, by estimating the $\phi_{dx}(l)$ and $\phi_{xx}(l)$ functions to compute \mathbf{w}_{opt} , and by direct estimation of the normalized MMSE $\frac{\sigma_{e_{opt}}^2}{\sigma_d^2}$ from the error signal $e_{opt}(n)$ produced with the estimated \mathbf{w}_{opt} . Plot the resulting \mathbf{w}_{opt} solution for each setup.

Use the following code to generate the signals $x(n)$ and $d(n)$ for Setup A and Setup B in Q3:

```
L=10000000;
u=randn(L+200,1); % white Gaussian, variance of one
x=filter(1,[1 -0.98],u); % t(n) filter
x=x(101:end); % discard first 100 samples to remove transients
               % in simulated signals
v=randn(L+100,1); % white Gaussian, variance of one

% setup A
dA=filter(1,[1 -0.95],x) + v; % h(n) filter plus additive noise
dA=dA(101:end); % discard first 100 samples to remove transients
               % in simulated signals
xA=x(101:end); % discard first 100 samples to remove transients
               % in simulated signals

% setup B
dB=x(101:end); % discard first 100 samples to remove transients
               % in simulated signals
xB=x(100:end-1); % x(n-1), delayed by one sample
```

Appendix Some z-transform relations for correlation functions

$$a^n u(n) \xleftrightarrow{z\text{-transf.}} \frac{1}{1-az^{-1}} \quad |z| > |a| \qquad -a^n u(-n-1) \xleftrightarrow{z\text{-transf.}} \frac{1}{1-az^{-1}} \quad |z| < |a|$$

$$a^{-n} u(-n) \xleftrightarrow{z\text{-transf.}} \frac{-a^{-1}z^{-1}}{1-a^{-1}z^{-1}} \quad |z| < |a^{-1}|$$

$$a^n u(n) * b^{-n} u(-n) \xleftrightarrow{z\text{-transf.}} \frac{-b^{-1}z^{-1}}{(1-az^{-1})(1-b^{-1}z^{-1})} = \frac{A}{1-az^{-1}} + \frac{B}{1-b^{-1}z^{-1}} \quad |a| < |z| < |b^{-1}|$$

$$A = \frac{-b^{-1}a^{-1}}{1-b^{-1}a^{-1}} = \frac{1}{1-ab}$$

$$B = \frac{-b^{-1}b}{1-ab} = \frac{-1}{1-ab}$$

$$a^n u(n) * b^{-n} u(-n) = Aa^n u(n) - Bb^{-n} u(-n-1)$$

$$a^n u(n) * a^{-n} u(-n) * b^n u(n) \xleftrightarrow{z\text{-transf.}} \frac{-a^{-1}z^{-1}}{(1-az^{-1})(1-a^{-1}z^{-1})(1-bz^{-1})} = \frac{A}{1-az^{-1}} + \frac{B}{1-a^{-1}z^{-1}} + \frac{C}{1-bz^{-1}} \quad \max(|a|, |b|) < |z| < |a^{-1}|$$

$$A = \frac{-a^{-1}a^{-1}}{(1-a^{-1}a^{-1})(1-ba^{-1})} = \frac{a}{(1-a^2)(a-b)}$$

$$B = \frac{-a^{-1}a}{(1-aa)(1-ba)} = \frac{-1}{(1-a^2)(1-ab)}$$

$$C = \frac{-a^{-1}b^{-1}}{(1-ab^{-1})(1-a^{-1}b^{-1})} = \frac{-b}{(a-b)(1-ab)}$$

$$a^n u(n) * a^{-n} u(-n) * b^n u(n) = Aa^n u(n) - Ba^{-n} u(-n-1) + Cb^n u(n)$$

$$a^n u(n) * a^{-n} u(-n) * b^n u(n) * b^{-n} u(-n) \xleftrightarrow{z\text{-transf.}} \frac{a^{-1}b^{-1}z^{-2}}{(1-az^{-1})(1-a^{-1}z^{-1})(1-bz^{-1})(1-b^{-1}z^{-1})} \\ = \frac{A}{1-az^{-1}} + \frac{B}{1-a^{-1}z^{-1}} + \frac{C}{1-bz^{-1}} + \frac{D}{1-b^{-1}z^{-1}} \quad \max(|a|, |b|) < |z| < \min(|a^{-1}|, |b^{-1}|)$$

$$A = \frac{a^{-1}b^{-1}a^{-2}}{(1-a^{-1}a^{-1})(1-ba^{-1})(1-b^{-1}a^{-1})} = \frac{a}{(1-a^2)(a-b)(1-ab)}$$

$$B = \frac{a^{-1}b^{-1}a^2}{(1-aa)(1-ba)(1-b^{-1}a)} = \frac{-a}{(1-a^2)(1-ab)(a-b)}$$

$$C = \frac{a^{-1}b^{-1}b^{-2}}{(1-ab^{-1})(1-a^{-1}b^{-1})(1-b^{-1}b^{-1})} = \frac{-b}{(a-b)(1-ab)(1-b^2)}$$

$$D = \frac{a^{-1}b^{-1}b^2}{(1-ab)(1-a^{-1}b)(1-bb)} = \frac{b}{(1-ab)(a-b)(1-b^2)}$$

$$a^n u(n) * a^{-n} u(-n) * b^n u(n) * b^{-n} u(-n) = Aa^n u(n) - Ba^{-n} u(-n-1) + Cb^n u(n) - Db^{-n} u(-n-1)$$