

# EE 503

## Homework 12 Solution

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**Problem 1.** Let  $S$  be a set and  $n$  a natural number. Recall

$$\limsup_{n \rightarrow \infty} A_n = \bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k.$$

Show

$$\limsup_{n \rightarrow \infty} A_n = \{x \in S : x \in A_n \text{ for infinitely many } n\}.$$

**Solution:**

We want to show the two sets are identical:

$$\bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k = \{x \in S : x \in A_n \text{ for infinitely many } n\}.$$

Thus, we pick an arbitrary  $x \in \bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k$ . This implies that  $\forall n \in \mathbb{N}$ ,  $x \in \bigcup_{k=n}^{\infty} A_k$  which tells us that  $\forall n \in \mathbb{N}$ ,  $\exists k \geq n$ , s.t.  $x \in A_k$ , or equivalently,  $x \in A_n$  for infinitely many  $n$ .

Conversely, we pick an arbitrary  $x \in A_n$  for infinitely many  $n$ , then  $\forall n \in \mathbb{N}$ ,  $\exists k \geq n$ , s.t.  $x \in A_k$ , or equivalently,  $\forall n \in \mathbb{N}$ ,  $x \in \bigcup_{k=n}^{\infty} A_k$ . Thus,  $x \in \bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k$ .

Since we show that  $\bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k \subseteq \{x \in S : x \in A_n \text{ for infinitely many } n\}$  and  $\bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k \supseteq \{x \in S : x \in A_n \text{ for infinitely many } n\}$ , we get

$$\bigcap_{n=1}^{\infty} \bigcup_{k=n}^{\infty} A_k = \{x \in S : x \in A_n \text{ for infinitely many } n\}.$$

**Problem 2.** Let  $X_k$ ,  $k = 1, 2, \dots$  be a sequence of independent normal (Gaussian) random variables with mean 0 and variance  $\sigma^2$ . Define

$$Y_n = \frac{1}{N} \sum_{k=n-N+1}^n X_k, \quad n = 0, 1, \dots$$

where we take  $Y_n = 0$  for  $n < 0$ .

This kind of operation of averaging is often used in signal processing. It is a type of low pass filter (LPF) since if the  $X_k$  had sinusoidal components they would average out (that is, be filtered out). We will refer to this type of filter as an I&D (integrate and dump) filter.

Now for implementation purposes we can also construct a LPF to approximate the above averaging filter using an IIR (infinite impulse response) filter as

$$\tilde{Y}_n = (1 - \alpha)X_n + \alpha\tilde{Y}_{n-1}, \quad n = 0, 1, \dots$$

where we take  $\tilde{Y}_{-1} = 0$  and  $\alpha \in (0, 1)$ .

- a. Show that  $E[\tilde{Y}_n] = E[Y_n] = 0$  for any  $\alpha \in (0, 1)$ .
- b. Determine the value of  $\alpha = \alpha(N)$  so that the variance of the IIR filter output ( $\tilde{Y}_n$ ) matches the variance of the I&D filter output ( $Y_n$ ) as  $n \rightarrow \infty$ .

**Solution:**

a.

$$\mathbf{E}[Y_n] = \mathbf{E}\left[\frac{1}{N} \sum_{k=n-N+1}^n X_k\right] = 0,$$

$$\begin{aligned} \mathbf{E}[\tilde{Y}_n] &= \mathbf{E}\left[(1 - \alpha)X_n + \alpha(1 - \alpha)X_{n-1} + \alpha^2(1 - \alpha)X_{n-2} + \dots + \alpha^n(1 - \alpha)X_0 + \alpha^{n+1}\tilde{Y}_{-1}\right] \\ &= 0. \end{aligned}$$

b.

$$\text{var}(Y_n) = \text{var}\left(\frac{1}{N} \sum_{k=n-N+1}^n X_k\right) = \frac{1}{N^2} \cdot N\sigma^2 = \frac{\sigma^2}{N},$$

$$\begin{aligned} \text{var}(\tilde{Y}_n) &= \sigma^2(1 - \alpha)^2 (1 + \alpha^2 + \alpha^4 + \dots + \alpha^{2n}) \\ &= \sigma^2(1 - \alpha)^2 \cdot \frac{1 - \alpha^{2(n+1)}}{1 - \alpha^2} \\ &= \sigma^2 \cdot \frac{(1 - \alpha)(1 - \alpha^{2(n+1)})}{1 + \alpha} \rightarrow \sigma^2 \cdot \frac{1 - \alpha}{1 + \alpha} \text{ as } n \rightarrow \infty \text{ if } \alpha \in (0, 1). \end{aligned}$$

$$\text{Hence, } \frac{1}{N} = \frac{1 - \alpha}{1 + \alpha} \Rightarrow \alpha = \frac{N - 1}{N + 1}.$$