

## Practice Problems - Weeks 1 through 4

With Answers. Important Note: Answers are NOT as full as I would expect in an exam.

These are quick answers I wrote for you to check your results. They are not the best answers possible.

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### Question One. Statistical Models

Which of the following describe statistical models?

If it is a statistical model, what are the parameters? What are the parameter spaces?

a)  $X_t \sim P$  where  $P \in \{N(0, 2.3), N(1, 2.3), N(2, 2.3), N(3, 2.3)\}$

**YES;**  $\mu; \{0, 1, 2, 3\}$

b)  $X_t = \beta_0 + \beta_1 X_{t-1} + \epsilon_t$

where  $\epsilon_t \sim N(0, \sigma^2)$

**YES;**  $\beta_0, \beta_1, \sigma^2; R \times R \times R^+$

c)  $E[X_t|Y_t] = 4$

where  $Y_t \sim N(2, 4)$

**NO,** the distribution of the  $X_t$  is not defined, just the first moment.

### Question Two. Weak Stationarity, Autocovariances.

Which of the following processes are Weakly Stationary? Show a proof.

If they are weakly stationary, derive their autocovariance functions.

a)

$$X_t = .9X_{t-2} + \epsilon_t$$

Where  $\epsilon_t \sim N(0, \sigma_\epsilon^2)$  iid  $\forall t$

b)

$$X_t = .9X_{t-2} + \epsilon_t + .5\epsilon_{t-1}$$

Where  $\epsilon_t \sim N(0, \sigma_\epsilon^2)$  iid  $\forall t$

c)

$$X_t = X_{t-2} + \epsilon_t$$

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\*I invented these questions. All faults are my own. If you have any concerns, queries, or find something that is incorrect, please email me (ndm2125@columbia.edu)

Where  $\epsilon_t \sim N(0, \sigma_\epsilon^2)$  iid  $\forall t$

d)

$$X_t = t\mu + \epsilon_t + \theta\epsilon_{t-1}$$

Where  $\epsilon_t \sim N(0, \sigma_\epsilon^2)$  iid  $\forall t$

e)

$$X_t = \theta^t \exp(\epsilon_t)$$

Where  $\epsilon_t \sim N(0, \sigma_\epsilon^2)$  iid  $\forall t$

a) and b) are weakly stationary. The others are not.

**Question Two.** Stationarity.

a) If a time series process is weakly stationary, is it strongly stationary? Show why using a proof, or why not using a counterexample. [This is a harder question]

**No, not necessarily. For example, the third moments could change as t changes.**

b) Take the MA(1) process

$$X_t = \theta_0\epsilon_t + \theta_1\epsilon_{t-1}$$

Where  $\epsilon_t \sim N(0, \sigma^2)$

Under what values of  $\theta_0, \theta_1$ , and  $\sigma^2$  is the process stationary?

**All finite values (i.e.  $|\theta_0| < \infty, |\theta_1| < \infty, |\sigma^2| < \infty$ )**

**Question Three.** Applications of LLN and CLT.

**\* Note: c) and d) were originally much harder than I had wanted to make them. A more reasonable question would have said added the bold statement below.**

Take the time series process

$$X_t = \phi X_{t-1} + \epsilon_t$$

Where  $\epsilon_t \sim N(\mu, \sigma^2)$  iid  $\forall t$

a) Suggest a consistent estimator for  $\mu$ . Which Law of large numbers can you use to prove consistency? Show that its conditions are satisfied.

**Answer:**

There are many possible answers to this questions. One is as follows:

Step One: Define a consistent estimator for  $\phi$  as  $\hat{\phi} = \frac{\text{Cov}(X_t, X_{t-1})}{\text{Cov}(X_t, X_t)}$

Step Two: Define  $Z_t = X_t - \hat{\phi}X_{t-1}$

Step Three: Our suggested estimator is  $\hat{\mu} = \frac{1}{T} \sum Z_t$ . We can apply the LLN to this estimator:

$$\hat{\mu} \rightarrow E[Z_t] = E[X_t - \phi X_{t-1}] = E[e_t] = \mu$$

This shows that the suggested estimator is consistent if the LLN applies.

To show that the LLN applies to  $Z_t$ , show that  $Z_t$  is a weakly stationary process.

b) What will be the asymptotic variance of this estimator?

By iid central limit theorem,

$$\sqrt{T}\hat{\mu} \rightarrow N(\mu, \sigma^2)$$

Therefore, the asymptotic variance of the estimator

$\hat{\mu}$  is  $\frac{1}{T}\sigma^2$

**Note: c) and d) were originally much harder than I had wanted to make them. A more reasonable question would have said added the bold statement below.**

Now take the time series process

$$X_t = \phi X_{t-1} + \eta_t$$

Where

$$\eta_t = \mu + \epsilon_t + \theta\epsilon_{t-1}$$

And  $\epsilon_t \sim N(0, \sigma^2)$  iid  $\forall t$

**Assume you know the values  $\theta = \theta^0$  and  $\phi = \phi^0$ .**

c) Suggest a consistent estimator for  $\mu$ . Which Law of large numbers can you use to prove consistency? Show that its conditions are satisfied.

**Answer:**

$$\frac{1}{T} \sum Z_t$$

where  $Z_t = X_t - \phi X_{t-1}$  d) What will be the asymptotic variance of this estimator?

**Answer:** By time series central limit theorem,

$$\sqrt{T} \frac{1}{T} \sum Z_t \rightarrow N(\mu, \sum_{h=-\infty}^{\infty} \gamma(h))$$

Where  $\gamma(h)$  is the autocovariance function of  $Z_t$ .

$$\gamma(0) = \text{Cov}(Z_t, Z_t) = \text{Cov}(\epsilon_t + \theta\epsilon_{t-1}, \epsilon_t + \theta\epsilon_{t-1}) = (1 + \theta^2)\sigma^2$$

$$\gamma(-1) = \gamma(1) = \text{Cov}(Z_t, Z_{t-1}) = \text{Cov}(\epsilon_t + \theta\epsilon_{t-1}, \epsilon_{t-1} + \theta\epsilon_{t-2}) = (\theta)\sigma^2$$

Therefore, the asymptotic variance of the estimator

$$\begin{aligned} \sum_{h=-\infty}^{\infty} \gamma_Z(h) &= \gamma(-1) + \gamma(0) + \gamma(1) \\ &= (1\theta^2 + 2\theta)\sigma^2 \end{aligned}$$

**Question Four.** Consistency and Unbiasedness.

Suppose that you have an iid sample  $X_1, X_2, \dots, X_N$  from a  $N(\theta, 1)$  model. Consider the following two estimators. The first one is the sample mean:

$$\hat{\theta}_1 = \frac{1}{N} \sum_{i=1}^N X_i$$

The second one is the shrinkage estimator:

$$\hat{\theta}_2 \equiv \frac{1}{2}\hat{\theta}_1$$

For each estimator answer the following questions:

a) Is the estimator consistent? Show this.

**Answer:** Yes for estimator 1. The LLN applies to  $X_i$ , as  $X_i$  are iid draws.

No for estimator 2.  $\hat{\theta}_2 \rightarrow E[\frac{1}{2}\hat{\theta}_1] = \frac{1}{2}E[\hat{\theta}_1] = \frac{1}{2}\theta \neq \theta$

b) Is the estimator unbiased? Show this.

**Answer:** Yes for estimator 1.  $E[\frac{1}{N} \sum_{i=1}^N X_i] = \frac{1}{N} \sum_{i=1}^N E[X_i] = \frac{1}{N} \sum_{i=1}^N \theta = \theta$ .

No for estimator 2.  $E[\hat{\theta}_2] = E[\frac{1}{2}\hat{\theta}_1] = \frac{1}{2}E[\hat{\theta}_1] = \frac{1}{2}\theta \neq \theta$