

YOUR NAME  
YOUR EMAIL  
November 18, 2025

## Homework 8

1. (1 point) Suppose a turtle lays a random number of eggs  $Y$ , and the probability that a hatchling from any given egg survives is  $p$ . If  $Y$  is modeled using a Poisson( $\lambda$ ) distribution, and the  $i$ th egg survives following an independent model  $X_i \sim \text{Bernoulli}(p)$  what is the expected value and variance of  $X$ , the total number of hatchlings that survive?
2. (1 point) Show that  $E[\text{Var}(Y|X)] \leq \text{Var}(Y)$ .
3. (1 point) Let  $T \sim \text{Exp}(\lambda)$ , and let  $U$  be uniform on the interval  $[0, T]$  (that is, we have a hierarchical model where  $U$  depends on  $T$ ). Find  $E[U]$  and  $\text{Var}(U)$ .
4. Use the definition of Covariance to show the following:
  - (a) (0.5 points)  $\text{Cov}(aX, bY) = ab \text{Cov}(X, Y)$  .
  - (b) (0.5 points)  $\text{Cov}(X, Y + Z) = \text{Cov}(X, Y) + \text{Cov}(X, Z)$  .
  - (c) (1 point) Now combine the results from parts 4a and 4b to show the following:

$$\text{Cov}(aW + bX, cY + dZ) = ac \text{Cov}(W, Y) + bc \text{Cov}(X, Y) + ad \text{Cov}(W, Z) + bd \text{Cov}(X, Z).$$

5. (1 point) Consider a symmetric random walk, starting at the position  $S_0 = 0$ . At time  $n$ , we write the position  $S_n$  as

$$S_n = S_{n-1} + X_n,$$

where the  $X_n$  are mean zero, independent steps with variance  $\sigma^2$ . For any  $n, m > 0$ , find an expression for  $\text{Cov}(S_n, S_m)$ .

6. (3 points) In time series analysis, we are interested in the behaviour of a sequence of random variables  $X_1, X_2, \dots$ , where the order of the random variables is an important feature of their behavior. An important concept in time series is *stationarity*. A times series is said to be (*weakly*) *stationary* if  $\text{Var}(X_n) < \infty$  for all  $n$ , and

- $E[X_n] = \mu < \infty$ , for all  $n \in \{1, 2, \dots\}$  (the mean doesn't depend on time).
- $\text{Cov}(X_i, X_j) = \gamma_{|i-j|}$ , or that the covariance only depends on the distance between  $i$  and  $j$ . In other words, for all time points  $n$  and values  $h \in \{0, 1, 2, \dots\}$ , there exists some  $\gamma_h$  such that

$$\text{Cov}(X_n, X_{n+h}) = \text{Cov}(X_{n+h}, X_n) = \gamma_h. \quad (1)$$

For this problem, we will assume that  $X_n$  comes from an Auto-regressive (1) model. That is, for some  $-1 < \phi < 1$ , we assume that

$$X_n = \phi X_{n-1} + \epsilon_n, \quad (2)$$

where  $\epsilon_n$  are iid  $N(0, \sigma^2)$  random variables. We will now assume that  $X_0$  is sampled from the *stationary distribution*, or that  $E[X_0] = 0$ , and  $\text{Var}(X_0) = \gamma_0$ . **Your task is to use the information above to find an expression for  $\gamma_h$  in terms of  $h, \sigma^2$  and  $\phi$** . The steps below will help you through this process.

- (a) (1 point) Step (1): Use properties of the covariance, Equations 1 and 2 to express  $\gamma_h$  in terms of  $\gamma_{h-1}$ :

$$\begin{aligned} \gamma_h &= \text{Cov}(X_n, X_{n+h}) \quad (\text{By Eq. 1}) \\ &= \text{Cov}(X_n, \phi X_{n+h-1} + \epsilon_{n+h}) \quad (\text{By Eq. 2}) \\ &= \dots \text{ You continue from here } \dots \end{aligned}$$

- (b) (1 point) Step (2): Using your result in the previous step, get an expression for  $\gamma_h$  in terms of  $\gamma_0$ .  
(c) (1 point) Step (3): Now our goal is to get an expression for  $\gamma_0$ , which is  $\text{Cov}(X_n, X_n)$  for all  $n$ :

$$\begin{aligned}\gamma_0 &= \text{Cov}(X_n, X_n) \\ &= \text{Cov}(\phi X_{n-1} + \epsilon_{n-1}, \phi X_{n-1} + \epsilon_{n-1}) \quad (\text{By Eq. 2}) \\ &= \dots \text{ You continue from here } \dots\end{aligned}$$

- (d) (0 points, trivial) Step (4): Combine your results from Step 2 and Step 3 to get an expression of  $\gamma_h$  in terms of  $h, \sigma^2$  and  $\phi$ .  
(e) (BONUS, +0.5): How does your result change if  $X_0$  is not from the stationary distribution? Is the process weakly stationary? For instance, suppose  $X_0 = x_0$ , so that  $X_0$  is not random but some constant.