

Please provide complete and well-written solutions to the following exercises.

Due September 1, 9AM, to be submitted in blackboard, under the Assignments tab.

## Homework 1

**Exercise 1.** Estimate the probability that 1000000 coin flips of fair coins will result in more than 501,000 heads, using the Central Limit Theorem. (Some of the following integrals may be relevant:  $\int_{-\infty}^0 e^{-t^2/2} dt / \sqrt{2\pi} = 1/2$ ,  $\int_{-\infty}^1 e^{-t^2/2} dt / \sqrt{2\pi} \approx .8413$ ,  $\int_{-\infty}^2 e^{-t^2/2} dt / \sqrt{2\pi} \approx .9772$ ,  $\int_{-\infty}^3 e^{-t^2/2} dt / \sqrt{2\pi} \approx .9987$ .) (Hint: use Bernoulli random variables.)

**Exercise 2** (Numerical Integration). In computer graphics in video games, etc., various integrations are performed in order to simulate lighting effects. Here is a way to use random sampling to integrate a function in order to quickly and accurately render lighting effects. Let  $\Omega = [0, 1]$ , and let  $\mathbf{P}$  be the uniform probability law on  $\Omega$ , so that if  $0 \leq a < b \leq 1$ , we have  $\mathbf{P}([a, b]) = b - a$ . Let  $X_1, \dots, X_n$  be independent random variables such that  $\mathbf{P}(X_i \in [a, b]) = b - a$  for all  $0 \leq a < b \leq 1$ , for all  $i \in \{1, \dots, n\}$ . Let  $f: [0, 1] \rightarrow \mathbf{R}$  be a continuous function we would like to integrate. Instead of integrating  $f$  directly, we instead compute the quantity

$$\frac{1}{n} \sum_{i=1}^n f(X_i).$$

Show that

$$\lim_{n \rightarrow \infty} \mathbf{E} \left( \frac{1}{n} \sum_{i=1}^n f(X_i) \right) = \int_0^1 f(t) dt.$$

$$\lim_{n \rightarrow \infty} \text{var} \left( \frac{1}{n} \sum_{i=1}^n f(X_i) \right) = 0.$$

That is, as  $n$  becomes large,  $\frac{1}{n} \sum_{i=1}^n f(X_i)$  is a good estimate for  $\int_0^1 f(t) dt$ .

**Exercise 3.** Let  $X := (X_1, \dots, X_n)$  be a random sample of size  $n$  from a binomial distribution with parameters  $n$  and  $p$ . Here  $n$  is a positive (known) integer and  $0 < p < 1$  is unknown. (That is,  $X_1, \dots, X_n$  are i.i.d. and  $X_1$  is a binomial random variable with parameters  $n$  and  $p$ , so that  $\mathbf{P}(X_1 = k) = \binom{n}{k} p^k (1-p)^{n-k}$  for all integers  $0 \leq k \leq n$ .)

You can freely use that  $\mathbf{E}X_1 = np$  and  $\text{Var}X_1 = np(1-p)$ .

- Compute the Fisher information  $I_X(p)$  for any  $0 < p < 1$ . (Consider  $n$  to be fixed.)
- Let  $Z$  be an unbiased estimator of  $p^2$  (assume that  $Z$  is a function of  $X_1, \dots, X_n$ ). State the Cramér-Rao inequality for  $Z$ .
- Let  $W$  be an unbiased estimator of  $1/p$  (assume that  $W$  is a function of  $X_1, \dots, X_n$ ). State the Cramér-Rao inequality for  $W$ .

**Exercise 4.** Let  $X_1, \dots, X_n$  be a random sample of size  $n$  from a Poisson distribution with unknown parameter  $\lambda > 0$ . (So,  $\mathbf{P}(X_1 = k) = e^{-\lambda} \lambda^k / k!$  for all integers  $k \geq 0$ .)

- Find an MLE (maximum likelihood estimator) for  $\lambda$ .
- Is the MLE you found unique? That is, could there be more than one MLE for this problem?

**Exercise 5.** Suppose  $X$  is a Gaussian distributed random variable with known variance  $\sigma^2 > 0$  but unknown mean. Fix  $\mu_0, \mu_1 \in \mathbf{R}$ . Assume that  $\mu_0 - \mu_1 > 0$ . We want to test the hypothesis  $H_0$  that  $\mu = \mu_0$  versus the hypothesis  $H_1$  that  $\mu = \mu_1$ . Fix  $\alpha \in (0, 1)$ . Explicitly describe the UMP test for the class of tests whose significance level is at most  $\alpha$ .

Your description of the test should use the function  $\Phi(t) := \int_{-\infty}^t e^{-x^2/2} dx / \sqrt{2\pi}$ ,  $\Phi: \mathbf{R} \rightarrow (0, 1)$ , and/or the function  $\Phi^{-1}: (0, 1) \rightarrow \mathbf{R}$ . (Recall that  $\Phi(\Phi^{-1}(s)) = s$  for all  $s \in (0, 1)$  and  $\Phi^{-1}(\Phi(t)) = t$  for all  $t \in \mathbf{R}$ .)

**Exercise 6.** This exercise demonstrates that a UMP might not always exist.

Let  $X_1, \dots, X_n$  be i.i.d. Gaussian random variables with known variance and unknown mean  $\mu \in \mathbf{R}$ . Fix  $\mu_0 \in \mathbf{R}$ . Let  $H_0$  denote the hypothesis  $\{\mu = \mu_0\}$  and let  $H_1$  denote the hypothesis  $\mu \neq \mu_0$ . Fix  $0 < \alpha < 1$ . Let  $\mathcal{T}$  denote the set of hypothesis tests with significance level at most  $\alpha$ . Show that no UMP class  $\mathcal{T}$  test exists, using the following strategy.

- Let  $\mu_1 < \mu_0$ . You may take as given the following fact (that follows from the Karlin-Rubin Theorem): the power at  $\mu_1$  is maximized among class  $\mathcal{T}$  tests by the hypothesis test  $\phi$  that rejects  $H_0$  when the sample mean satisfies  $\bar{X} < c$  for an appropriate choice of  $c \in \mathbf{R}$ . Assume for the sake of contradiction that a UMP class  $\mathcal{T}$  test  $\phi'$  exists. Then, using the necessity part of the Neyman-Pearson Lemma (i.e. consider testing  $\mu = \mu_0$  versus  $\mu = \mu_1$ ), conclude that  $\phi'$  must have the same rejection region as  $\phi$  (just by examining the power of the tests at  $\mu_1$ .)
- Consider now a test in  $\mathcal{T}$  that rejects  $H_0$  when the sample mean satisfies  $\bar{X} > c'$  for an appropriate choice of  $c' \in \mathbf{R}$ . Repeating the previous argument, conclude that  $\phi'$  must reject when  $\bar{X} > c'$ , leading to a contradiction.

That is, let  $\mu_2 > \mu_0$ . You may take as given the following fact (that follows from the Karlin-Rubin Theorem): the power at  $\mu_2$  is maximized among class  $\mathcal{T}$  tests by the hypothesis test  $\phi''$  that rejects  $H_0$  when the sample mean satisfies  $\bar{X} > c'$  for an appropriate choice of  $c' \in \mathbf{R}$ . Then, using the necessity part of the Neyman-Pearson Lemma (i.e. consider testing  $\mu = \mu_0$  versus  $\mu = \mu_2$ ), conclude that  $\phi'$  must have the same rejection region as  $\phi''$ .

**Exercise 7.** The rejection regions  $C_\alpha$  for UMP hypothesis tests of significance level at most  $\alpha \in (0, 1)$  are often nested in the sense that  $C_\alpha \subseteq C_{\alpha'}$  for all  $0 < \alpha < \alpha' < 1$ . This exercise demonstrates an example of UMP tests where this nesting behavior does not occur.

Let  $\theta_0, \theta_1 \in \mathbf{R}$  be unequal parameters. Let  $H_0$  denote the hypothesis  $\{\theta = \theta_0\}$  and let  $H_1$  denote the hypothesis  $\{\theta = \theta_1\}$ . Suppose  $X \in \{1, 2, 3\}$  is a random variable. If  $\theta = \theta_0$ , assume that  $X$  takes the values 1, 2, 3 with probabilities .85, .1, .05, respectively. If  $\theta = \theta_1$ , assume that  $X$  takes the values 1, 2, 3 with probabilities .7, .2, .1, respectively. Let  $\mathcal{T}$  denote the set of hypothesis tests with significance level at most  $\alpha$ .

- Let  $0 < \alpha < .15$ . Show that a UMP class  $\mathcal{T}$  test is not unique.
- When  $\alpha = .05$ , show there is a unique nonrandomized hypothesis UMP class  $\mathcal{T}$  test.
- When  $\alpha = .1$ , show there is a unique nonrandomized hypothesis UMP class  $\mathcal{T}$  test.
- Show that the  $\alpha = .05$  and  $\alpha' = .1$  UMP nonrandomized tests from above do not have nested rejection regions.
- However, when  $\alpha = .05$  and  $\alpha' = .1$ , there are randomized UMP tests  $\phi, \phi': \mathbf{R}^n \rightarrow [0, 1]$  respectively, that are nested in the sense that  $\phi \leq \phi'$ .