

**1. Rounding Error.** Your friend has a list of ten real numbers, whose values are unknown to you:  $x_1, \dots, x_{10} \in \mathbb{R}$ . Your friend rounds each number to the nearest integer and sends you the results:  $X_1, X_2, \dots, X_{10} \in \mathbb{Z}$ . We will assume that  $X_i = x_i + U_i$ , where each  $U_i$  is a uniform random variable on the interval  $[-1/2, 1/2]$ .

- Compute  $E[U_i]$  and  $\text{Var}(U_i)$ .
- Consider the sum of the rounded numbers  $X = X_1 + \dots + X_{10}$  and the sum of the unrounded numbers  $x = x_1 + \dots + x_{10}$ . Prove that  $E[X] = x$ .
- Assuming that the random variables  $U_i$  are independent, use the CLT to estimate the probability that  $|X - x| > 1/2$ . [Hint: The CLT says that  $X - x = U_1 + \dots + U_{10}$  is approximately normal. You just need to compute  $E[X - x]$  and  $\text{Var}(X - x)$ .]

**2. Tail Probabilities.** Consider a standard normal variable  $Z \sim N(0, 1)$ . Solve for  $a$ .

- $P(Z > a) = 93\%$
- $P(Z < a) = 35\%$
- $P(|Z| > a) = 2\%$
- $P(|Z| < a) = 80\%$

**3. A Bernoulli Hypothesis Test.** A six-sided die has sides labeled  $\{1, 2, 3, 4, 5, 6\}$ . Let  $p$  be the probability of getting a 6. Before performing any experiments we will assume that  $H_0 = "p = 1/6"$  is true. Now suppose that you roll the die 600 times and let  $Y$  be the number of times you get 6. Which values of  $Y$  would cause you to reject  $H_0$  in favor of  $H_1 = "p > 1/6"$  at the 99% level of confidence? That is, what is the rejection region?

**4. Confidence Intervals for a Proportion.** Let  $p$  be the proportion of Americans who are left-handed. In order to estimate  $p$ , we randomly selected  $n = 1000$  Americans and we found that  $Y = 125$  of them are left-handed. Use this information to compute two-sided, symmetric  $(1 - \alpha)100\%$  confidence intervals for  $p$  when  $\alpha = 5\%$ ,  $2.5\%$  and  $1\%$ .

**5. Sample Variance.** Consider an iid sample  $X_1, \dots, X_n$  with unknown mean  $\mu$  and unknown variance  $\sigma^2$ . In order to estimate  $\sigma^2$  we define the *sample variance* as follows:

$$S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2,$$

where  $\bar{X} = (\sum_{i=1}^n X_i)/n$  is the usual sample mean.

- Show that  $\sum_{i=1}^n (X_i - \bar{X})^2 = (\sum_{i=1}^n X_i^2) - n\bar{X}^2$ . [Hint:  $n\bar{X} = \sum_{i=1}^n X_i$ .]
- Show that  $E[X_i^2] = \mu^2 + \sigma^2$  and  $E[\bar{X}^2] = \mu^2 + \sigma^2/n$ . [Hint: By definition we have  $E[X_i] = \mu$  and  $\text{Var}(X_i) = \sigma^2$ , which implies that  $E[\bar{X}] = \mu$  and  $\text{Var}(\bar{X}) = \sigma^2/n$ .]
- Combine (a) and (b) to show that  $E[S^2] = \sigma^2$ . This is why the definition of  $S^2$  has  $n - 1$  in the denominator instead of  $n$ .

**6. A Small Sample.** The label weight of a Cadbury Creme Egg is 1.2oz. In order to test this you weighed 10 eggs and obtained the following values (in ounces):

1.12	1.01	1.04	1.10	1.00	1.04	1.28	1.17	1.19	1.24
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Let  $X$  represent the underlying distribution with unknown mean  $\mu = E[X]$ . For simplicity we assume that  $X$  is normal.

- (a) Compute the sample mean  $\bar{X}$  and the sample variance  $S^2$ .
- (b) Look up the  $t$ -tail probabilities  $t_{5\%}(9)$  and  $t_{2.5\%}(9)$ .
- (c) Test the hypothesis  $H_0 = \mu = 1.2$  against the one-sided alternative  $H_1 = \mu < 1.2$  at the 5% level of significance.
- (d) Compute a two-sided symmetric 95% confidence interval for the unknown  $\mu$ .