Homework 8

Problem 1. Which of the following hypotheses are simple, and which are composite?

- (a) X follows a uniform distribution on [0, 1].
- (b) A die is unbiased.
- (c) X follows a normal distribution with mean 0 and variance $\sigma^2 > 10$.
- (d) X follows a normal distribution with mean $\mu = 0$.
 - (a) Simple because the distribution is completely determined.
- (b) Simple, assuming the die has six sides. Otherwise composite because we can't determine the distribution completely from the hypothesis.
 - (c) Simple because the distribution is given by the hypothesis.
 - (d) Composite because the distribution isn't determined by the hypothesis. The variance is unknown.

Problem 2. Suppose that $X \sim \text{bin}(100, p)$. Consider the test that rejects $H_0: p = .5$ in favor of $H_A: p \neq .5$ for |X - 50| > 10. Use the normal approximation to the binomial distribution to answer the following:

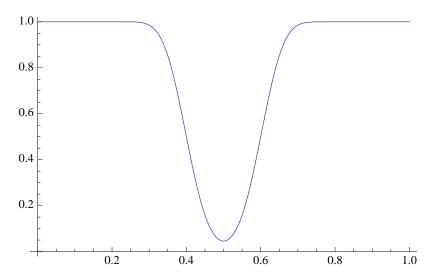
- (a) What is α ?
- (b) Graph the power as a function of p.
- (a) We know α is $P(\text{reject } H_0 \mid H_0)$ which is $P(|X 50| > 10 \mid p = .5)$. The normal approximation to this distribution is N(np, np(1-p)) = N(50, 25) in this case. Then

$$\begin{split} \alpha &= P(|X - 50| > 10 \mid p = .5) \\ &= P(X < 40) + P(X > 60) \\ &= P\left(\frac{X - 50}{5} > \frac{40 - 50}{5}\right) + P\left(\frac{X - 50}{5} > \frac{60 - 50}{5}\right) \\ &= \Phi(-2) + (1 - \Phi(2)) \\ &\approx .0456. \end{split}$$

(b) We use the same normal approximation N(np, np(1-p)) = N(100p, 100p(1-p)) to get

$$\begin{split} 1-\beta &= P(X<40) + P(X>60) \\ &= P\left(\frac{X-100p}{\sqrt{100p(1-p)}} < \frac{40-100p}{\sqrt{100p(1-p)}}\right) + P\left(\frac{X-100p}{\sqrt{100p(1-p)}} < \frac{60-100p}{\sqrt{100p(1-p)}}\right) \\ &= \Phi\left(\frac{40-100p}{\sqrt{100p(1-p)}}\right) + 1 - \Phi\left(\frac{60-100p}{\sqrt{100p(1-p)}}\right). \end{split}$$

The graph of this for $0 \le p \le 1$ is



Problem 3. Consider the coin tossing example of Section 9.1. Suppose that instead of tossing the coin 10 times, the coin was tossed until a head came up and the total number of tosses, X, was recorded.

- (a) If the prior probabilities are equal, which outcomes favor H_0 and which favor H_1 ?
- (b) Suppose $P(H_0)/P(H_1) = 10$. What outcomes favor H_0 ?
- (c) What is the significance level of a test that rejects H_0 if $X \geq 8$?
- (d) What is the power of this test?
 - (a) The posterior probabilities are

$$P(H_0 \mid X) = \frac{(.5)P(X \mid H_0)}{(.5)P(X \mid H_0) + (.5)P(X \mid H_1)} = \frac{(1 - .5)^{x - 1}(.5)}{(1 - .5)^{x - 1}(.5) + (1 - .7)^{x - 1}(.7)}$$

$$P(H_1 \mid X) = \frac{(.5)P(X \mid H_1)}{(.5)P(X \mid H_0) + (.5)P(X \mid H_1)} = \frac{(1 - .7)^{x-1}(.7)}{(1 - .5)^{x-1}(.5) + (1 - .7)^{x-1}(.7)}.$$

Then

$$\frac{P(H_0 \mid X)}{P(H_1 \mid X)} = \frac{(.5)^x}{(.3)^{x-1}(.7)} > 1$$

when

$$x > \frac{\log(7) - \log(3)}{\log(5) - \log(3)} \approx 1.66.$$

(b) Now we have

$$\frac{P(H_0 \mid X)}{P(H_1 \mid X)} = 10 \frac{(.5)^x}{(.3)^{x-1}(.7)} > 1$$

when

$$x > \frac{\log(30) - \log(7)}{\log(5) - \log(3)} \approx -2.85$$

so all realistic outcomes.

(c) This is

$$\alpha = P(\text{reject } H_0 \mid H_0) = P(X \ge 8 \mid H_0) = 1 - P(X < 7 \mid H_0) = 1 - \sum_{i=1}^{6} (.5)^i = .015625$$

(d) The power is

$$1 - \beta = 1 - P(X < 7 \mid H_1) = 1 - \sum_{i=1}^{6} (.3)^{i-1} (.7) = .000729.$$

Problem 4. Show that the test of Problem 7 is uniformly most powerful for testing $H_0: \lambda = \lambda_0$ versus $H_A: \lambda > \lambda_0$.

Proof. In Exercise 7 the likelihood ratio will be of the form

$$\frac{\lambda_0^{\sum_{i=1}^n X_i} e^{-n\lambda_0}}{\lambda_1^{\sum_{i=1}^n X_i} e^{-n\lambda_1}} = \lambda_0^{\sum_{i=1}^n X_i} \lambda_1^{-\sum_{i=1}^n X_i} e^{n(\lambda_1 - \lambda_0)} = \exp\left(n(\log(\lambda_0) - \log(\lambda_1))\overline{X} + n(\lambda_1 - \lambda_0)\right).$$

The likelihood ratio is then small if \overline{X} is large since $\lambda_1 > \lambda_0$. So the most powerful test rejects for $\overline{X} > x_0$ for some x_0 . The null distribution given H_0 in this case is Poisson with parameter $(1/n)n\lambda_0 = \lambda_0$. Thus we can choose x_0 , specified for a significance α , using this distribution. But note that this is independent of λ_1 so it must hold true for all $\lambda_1 > \lambda_0$. Therefore this test is uniformly most powerful.

Problem 5. Let X_1, \ldots, X_n be a random sample from an exponential distribution with the density function $f(x \mid \theta) = \theta \exp[-\theta x]$. Derive a likelihood ratio test of $H_0: \theta = \theta_0$ versus $H_A: \theta \neq \theta_0$, and show that the rejection region is of the form $\{\overline{X} \exp[-\theta_0 \overline{X}] \leq c\}$.

Proof. First we find the mle for this distribution. The log likelihood is

$$l(\theta) = n\log(\theta) - \theta \sum_{i=1}^{n} X_i$$

so

$$l'(\theta) = \frac{n}{\theta} - \sum_{i=1}^{n} X_i$$

and

$$\hat{\theta} = 1/\overline{X}$$
.

Now we form the likelihood ratio as

$$\frac{\theta_0^n \exp\left(-\theta_0 \sum_{i=1}^n X_i\right)}{\hat{\theta}^n \exp\left(-\hat{\theta} \sum_{i=1}^n X_i\right)} = \left(\frac{\theta_0}{\hat{\theta}}\right)^n \exp\left(-n(\theta_0 - \hat{\theta})\overline{X}\right) = (\theta_0 \overline{X} \exp(-\theta_0 \overline{X} + 1))^n.$$

This gives a rejection region of the form

$$(\theta_0 \overline{X} \exp(-\theta_0 \overline{X} + 1))^n \le c$$

$$\theta_0 e^{-1} \overline{X} \exp(-\theta_0 \overline{X}) \le c^{1/n}$$

$$\overline{X} \exp(-\theta_0 \overline{X}) \le \theta_0^{-1} e c^{1/n}.$$

Now note that θ_0 is simply a constant and n is fixed, so we can redefine c to be the constant on the right and the desired result follows.

Problem 6. Let X_1, X_2, \ldots, X_n be i.i.d. random variables from a double exponential distribution with density $f(x) = \frac{1}{2}\lambda \exp(-\lambda|x|)$. Derive a likelihood ratio test of the hypothesis $H_0: \lambda = \lambda_0$ versus $H_1: \lambda = \lambda_1$, where λ_0 and $\lambda_1 > \lambda_0$ are specified numbers. Is the test uniformly most powerful against the alternative $H_1: \lambda > \lambda_0$?

The likelihood ratio will be of the form

$$\begin{split} \frac{\lambda_0^n \exp\left(-\lambda_0 \sum_{i=1}^n |X_i|\right)}{\lambda_1^n \exp\left(-\lambda_1 \sum_{i=1}^n |X_i|\right)} &= \left(\frac{\lambda_0}{\lambda_1}\right)^n \exp\left(-\left(\lambda_0 - \lambda_1\right) \sum_{i=1}^n |X_i|\right) \\ &= \exp\left(n\left(\log(\lambda_0) - \log(\lambda_1)\right) - \left(\lambda_0 - \lambda_1\right) \sum_{i=1}^n |X_i|\right). \end{split}$$

Since $\lambda_1 > \lambda_0$, this ratio is small when $\sum_{i=1}^n |X_i|$ is small. Note that $n|\overline{X}| = |\sum_{i=1}^n X_i| \leq \sum_{i=1}^n |X_i|$ so when the right hand side is small, so is the left. Now note that the null hypothesis has a double exponential distribution with parameter λ_0 . Thus $n|\overline{X}|$ is a known distribution with a parameter depending only on λ_0 . It's independence of λ_1 shows that it's uniformly most powerful.

Problem 7. Consider two probability density functions on [0,1]: $f_0(x) = 1$, and $f_1(x) = 2x$. Among all tests of the null hypothesis $H_0: X \sim f_0(x)$ versus the alternative $X \sim f_1(x)$, with significance level $\alpha = 0.10$, how large can the power possibly be?

By the Lemma, we know the likelihood ratio test will give the maximum power for a given significance level α . The likelihood ratio is of the form

$$\frac{f_0(x)}{f_1(x)} = \frac{1}{2x} > c$$

where $c \in (0,1)$. We are interested in

$$\alpha = .1 = P(\text{reject } H_0 \mid H_0) = P\left(\frac{1}{2x} \le c \mid H_0\right) = P\left(\frac{1}{2c} \le X \mid H_0\right) = 1 - P\left(X \le \frac{1}{2c} \mid X_0\right).$$

So 1 - 1/2c = 1/10 and c = 5/9. Then

$$1 - \beta = 1 - P(\text{accept } H_0 \mid H_1) = 1 - P\left(\frac{1}{2x} > c \mid H_1\right) = 1 - P\left(\frac{1}{2c} > X \mid H_1\right) = P\left(X \le \frac{1}{2c} \mid H_1\right).$$

The cdf for $f_1(x)$ is $F_X(x) = x^2$ so

$$1 - \beta = \left(\frac{1}{2c}\right)^2 = \frac{1}{4c^2} = .81.$$