STAT 323 Assignment 5

1. 1

We set our two tests equal:

$$P(Y_1 > .95 | \theta = 0) = P(Y_1 + Y_2 > c)$$

$$\int_{0.95}^{1} 1 dx = P(Y_1 + Y_2 > c)$$

$$0.05 = P(Y_1 + Y_2 > c).$$

Now from example 1.1.6, $Y_1 + Y_2 = Z$ has the pdf given by:

$$f_Z(z) = \begin{cases} z & 0 \le z \le 1 \\ -z + 2 & 1 < z \le 2 \\ 0 & \text{otherwise} \end{cases}$$

And so we integrate;

$$0.95 = P(Z \le c)$$

$$= \int_0^1 z \, dz + \int_1^c -z + 2 \, dz$$

$$= {}^{wolfram} 1.68377.$$

Technically this gives us two roots, but we can reject the one greater than 2 since it is outside the support.

- 2. 2
- 3. 3

The power of our test $1 - \beta$ is the probability that we reject our null hypothesis correctly;

$$1 - \beta = P(RH_0|H_0 \text{ is false}) = P(x_1x_2 > 0.75|\theta = 2).$$

To find this probability we must find the joint distribution of x_1x_2 . Since our variables are independent, our joint distribution is given by $f_{x_1}f_{x_2} = 4x_1x_2$, and integrating our support,

$$\int_{\frac{3}{4}}^{1} \int_{\frac{3}{4\pi\alpha}}^{1} 4x_1 x_2 dx_1 dx_2 =^{wolfram} 0.113857$$

- 4. 4
- 5. 5

We load our data into R:

```
> june=c(153.3, 155.9, 176.2, 189.9, 200.0, 214.9, 229.9, 231.5, 257.9, 299.9)
> december=c(151.1, 154.2, 169.9, 169.9, 185.9, 199.5, 229.9, 232.9, 279.9, 289.9)
```

And we can use t.test in R since our data is normally distributed. Our null hypothesis is that the mean selling price of a condo is the same in December as in June, and our alternative is that the mean in December is less than June.

Before using t.test we must check the ratio of variances, which we can do using var.test

> var.test(june,december)

And the interval that this gives us does not contain 1, so we cannot conclude equal variance. So our t.test:

> t.test(december, june, alternative="less", var.equal=FALSE)

This gives us a p-value of 0.5841, too high to reject the null hypothesis. Therefore there is not significant evidence to suggest that the mean price of a condo decreased from December to June.

6. 6

7. 7

Let \hat{p}_1 be the proportion of defects while the water was contaminated. We know this to be 16 out of 414 births. Let \hat{p}_2 be the defects after the water was contaminated. We know this to be 2 out of 228 births. Then our hypotheses are $H_0: p_1-p_2 \leq 0$, and $H_a: p_1-p_2 > 0$. Our test statistic:

$$Z_{calc} = \frac{\widehat{p}_1 - \widehat{p}_2}{\sqrt{\widehat{p}(1-\widehat{p})\left(\frac{1}{n_1} + \frac{1}{n_2}\right)}} \sim Normal(0,1)$$

Where \hat{p} is our pooled sample proportion, when simplified with our above values gives us a test statistic of 2.194423. To find our p-value, the probability of observing a sample more extreme than this, is given by:

$$P(Z > \text{Test Stat}) = 1\text{-pnorm(2.194423)} = 0.0141.$$

Which is less than our tolerance for type 1 error, so we conclude that the contaminated water had a significant effect on the proportion of birth defects.

8.8

9. 9 Begin with neyman-pearson, using both sets of RVs,

$$\begin{split} \frac{L(H_0)}{L(H_a)} &= \frac{\prod_{i=1}^n 2^{y_i} e^{-2}/y_i! \prod_{j=1}^m 2^{x_i} e^{-2}/x_i!}{\prod_{i=1}^n 1/2^{y_i} e^{-\frac{1}{2}}/y_i! \prod_{j=1}^m 3^{x_i} e^{-3}/x_i!} \\ &= \frac{\prod_{i=1}^n 2^{y_i} e^{-2} \prod_{j=1}^m 2^{x_i} e^{-2}}{\prod_{i=1}^n 1/2^{y_i} e^{-\frac{1}{2}} \prod_{j=1}^m 3^{x_i} e^{-3}} \\ 2^{\sum y_i} \left(\frac{1}{2}\right)^{-\sum y_i} \left(\frac{2}{3}\right)^{\sum x_i} e^{-2} e^{-\frac{1}{2}} e^{\frac{1}{2}} e^{3} < k \\ \left(\frac{2}{3}\right)^{m\bar{x}} 4^{n\bar{y}} < k'. \end{split}$$

Taking the ln of both sides brings our variables out of the exponent but complicates the expression so I left this the way it was.

Our test statistic is

$$\left(\frac{2}{3}\right)^{m\bar{x}}4^{n\bar{y}}.$$

10. 10

11. 11

We use Neyman-Pearson; beginning with the ratio of our likelihood functions:

$$\frac{L(\sigma^2 = 2)}{L(\sigma^2 = 3)} = \frac{\prod_{i=1}^8 \frac{1}{\sqrt{4\pi}} \exp\left[-\left(\frac{1}{4}(y_i - \mu)^2\right)\right]}{\prod_{i=1}^8 \frac{1}{\sqrt{6\pi}} \exp\left[-\left(\frac{1}{6}(y_i - \mu)^2\right)\right]}$$

$$= \frac{\frac{1}{(4\pi)^4} \prod_{i=1}^8 \exp\left[-\left(\frac{1}{4}(y_i - \mu)^2\right)\right]}{\frac{1}{(6\pi)^4} \prod_{i=1}^8 \exp\left[-\left(\frac{1}{6}(y_i - \mu)^2\right)\right]}$$

$$= \left(\frac{6\pi}{4\pi}\right)^4 \frac{\exp\left[-\frac{1}{4}\sum_{i=1}^8 (y_i - \mu)^2\right]}{\exp\left[-\frac{1}{6}\sum_{i=1}^8 (y_i - \mu)^2\right]} < k$$

$$\frac{\exp\left[-\frac{1}{4}(8-1)S^2\right]}{\exp\left[-\frac{1}{6}(8-1)S^2\right]} < k'$$

$$\exp\left[\frac{7}{6}S^2 - \frac{7}{4}S^2\right] < k'$$

$$\frac{7}{6}S^2 - \frac{7}{4}S^2 < k''$$

$$-\frac{14}{24}S^2 < k''$$

$$S^2 > k'''$$

So we have our decision rule, to reject H_0 if S^2 is greater than our critical value. Now to find our critical value, setting the probability of Type I error to 0.05:

$$0.05 = P(RH_0|H_0 \text{ is true})$$

$$= P(S^2 > CV|\sigma^2 = 2)$$

$$= P(\frac{7S^2}{2} > \frac{7}{2}CV)$$

$$= P(\chi_7^2 > \frac{7}{2}CV)$$

And so in R our CV is given by qchisq(.05,7)*(2/7)=0.6192428. Our uniformly most powerful test is to reject the null hypothesis when our sample variance is greater than this value.