# DSC 462: Assignment 4

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## $\mathbf{Q}\mathbf{1}$

(a)

$$n = 2030, \quad \hat{p} = 1050/n$$
  $CI_{95\%} = \hat{p} \pm z_{0.025} \sqrt{\frac{\hat{p}(1-\hat{p})}{n}}$   $\approx 0.52 \pm 0.02$ 

```
n = 2030

p = 1050/n

p
```

## [1] 0.5172414

```
1.96* sqrt(p*(1-p)/n)
```

## [1] 0.02173801

(b) Note that

$$\alpha = 0.05$$

$$Z_{\alpha/2} = 1.96$$

$$E_o = 0.01$$

So we have  $n = (\frac{Z_{\alpha/2}}{2E_o})^2 = 9604$ .

#### (1.96/2/0.01)^2

## [1] 9604

### $\mathbf{Q2}$

(a)

```
n = 56
pHat = 20.5/n #for continuity correction
p0 = 0.25
Zobs = (pHat-p0)/sqrt(p0*(1-p0)/n)
alpha_obs = 1-pnorm(Zobs)
alpha_obs
```

## [1] 0.02243114

So p-value = 0.0224 and we can reject  $H_o$  at a significance level  $\alpha = 0.05$ .

(b) We can check the previous result by using R.

```
prop.test(x = 21, n = 56, p = .25, alternative = "greater", correct = TRUE)
```

```
##
## 1-sample proportions test with continuity correction
##
## data: 21 out of 56, null probability 0.25
## X-squared = 4.0238, df = 1, p-value = 0.02243
## alternative hypothesis: true p is greater than 0.25
## 95 percent confidence interval:
## 0.268643 1.000000
## sample estimates:
## p
## 0.375
```

#### Q3

We can do this problem by using R.

```
n1 = 250
n2 = 500
x1 = 65
x2 = 100
prop.test(x =c(x2,x1), n = c(n2,n1), correct = F)
```

```
##
## 2-sample test for equality of proportions without continuity
## correction
##
## data: c(x2, x1) out of c(n2, n1)
## X-squared = 3.4965, df = 1, p-value = 0.0615
## alternative hypothesis: two.sided
## 95 percent confidence interval:
## -0.124696627 0.004696627
## sample estimates:
## prop 1 prop 2
## 0.20 0.26
```

- (1) So the 0.95 confidence interval for  $p_2 p_1$  is (-0.12469, 0.00469).
- (2) Since the P-value is 0.0615, the null hypothesis cannot be rejected.

### $\mathbf{Q4}$

```
n11 = 3
n12 = 31
n21 = 12
n22 = 45
OR = (n11*n22)/(n12*n21)
SE.log.OR = sqrt(1/n11+1/n12+1/n21+1/n22)
log(OR)
```

## [1] -1.013619

```
1.96*SE.log.OR
```

## [1] 1.345347

So 95% CI for the log odds ratis is  $-1.10136 \pm 1.3453$ . This also means that we cannot reject the null hypothesis, since 0 is contained in the interval.

## $Q_5$

(a) To construct a 90% upper confidence bound for  $\sigma$ , we need

$$\chi_{n-1,1-\alpha}^2 = \chi_{5,0.90}^2 = 1.61$$

which gives

$$\sigma < \frac{S_n}{\sqrt{\chi_{n-1,1-\alpha}^2/(n-1)}} = \frac{0.088}{\sqrt{1.61/5}} = 0.1550796$$

```
x = c(9.43, 9.85, 10.12, 9.89, 9.81, 10.3)

Sn = var(x)

Sn
```

## [1] 0.088

```
Sn/sqrt(1.61/5)
```

- ## [1] 0.1550796
- (b) Using this upper bound for sample size estimation we get n=10

```
alpha = 0.05
z.alpha.half = qnorm(1-alpha/2)
E.o = 0.1
sigma = 0.1550796
(z.alpha.half*sigma/E.o)^2
```

## [1] 9.238586

#### Q6

```
n1 = 14

n2 = 18

S1.sq = 645.16

S2.sq = 1413.76

alpha = 0.1

F = S1.sq/S2.sq

crit.up = qf(1-0.05, 13,17)

crit.low = qf(0.05, 13,17)

F
```

## [1] 0.4563434

crit.low

## [1] 0.4002126

crit.up

## [1] 2.353063

So the rejection region is  $F \le 0.4$  or  $F \ge 2.35$  So we cannot reject the null hypothesis. Also, we can get our p-value = 0.1569.

```
2*min(pf(F, 13,17), 1- pf(F,13,17))
```

## [1] 0.156946

#### $\mathbf{Q7}$

First calculate the variance for null hypothesis.

$$\sigma_0^2 = (1/2^2 + 1/6^2 + 1/3^2)\sigma_X^2 = 9.722$$

Also the variance for the alternative hypothesis,

$$\sigma_1^2 = \sigma_0^2 + 2/25 * (0.56/12 + 0.18/6 + 0.21/18) = 14.13889$$

So our hypothesis would be

$$H_o: \sigma^2 = \sigma_0^2, \quad H_a: \sigma^2 = \sigma_1^2$$

Now our test statistic

$$W = \frac{(n-1)S_n^2}{\sigma_0^2}$$

Now, if  $H_o$  is true then  $W \sim \chi_{n-1}^2$ . If  $H_a$  is true then  $\frac{\sigma_0^2}{\sigma_1^2}W \sim \chi_{n-1}^2$ . So our power would be

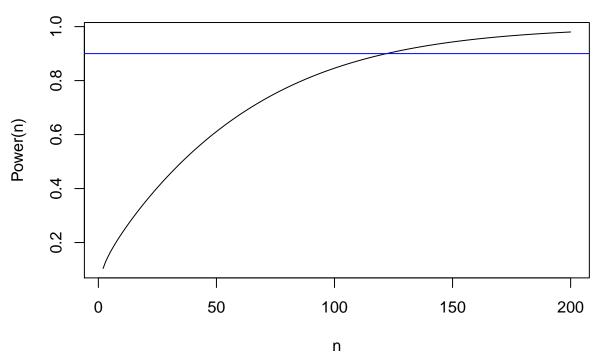
$$Power(n) = P(reject H_0|H_a)$$

$$=P(W>\chi^2_{n-1,0.025})=P(\frac{\sigma_0^2}{\sigma_1^2}\;W>\frac{\sigma_0^2}{\sigma_1^2}\;\chi^2_{n-1,0.025})$$

where  $\frac{\sigma_0^2}{\sigma_1^2}W \sim \chi_{n-1}^2$ .

```
sigma0.sq = (1/4+1/36+1/9)*25
sigma1.sq = sigma0.sq + 2*25*(0.56/12+0.18/6+0.21/18)
Power = function(n){
    crit = sigma0.sq/sigma1.sq*qchisq(p = 0.05, df = n-1, lower.tail = FALSE)
    pchisq(q = crit, df = n-1, lower.tail = FALSE)
}
n = 2:200
plot(n, Power(n), type = 'l', main = "Power vs n")
abline(h = 0.9, col = 'blue')
```

### Power vs n



```
n[Power(n) \ge 0.9][1]
```

## [1] 123

We can see that 123 is the minimum number of sample required to have 90% power.