# 2019R1 Discrete Data Analysis (STAT5107) Assignment

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set.seed(5107);

1.

- a. nominal
- b. ordinal
- c. interval
- d. nominal
- e. ordinal
- c
- f. nominal
- g. ordinal

2.

Variance of binomial distribution is

$$\sigma^2 = n\pi(1-\pi)$$

- When  $\pi$  is close to zero, everything is multiplied by a value close to zero, hence  $\sigma^2$  is small.
- When  $\pi$  is close to one, everything is multiplied by (1 somehting close to one), which is close to zero. Hence  $\sigma^2$  is also small.
- When  $\pi$  is close to 0.5, n is multiplied by 0.5 \* (1 0.5), which is close to 0.25, which is bigger than the values above.

A smaller variance means more precise estimate of  $\pi$  and vice versa.

### 3.

 $H_0$ :  $\pi = 0.5 \ H_1$ :  $\pi \neq 0.5$ 

```
yes_count <- 842;
no_count <- 982;
n <- yes_count + no_count;
H0 <- 0.5;
alpha <- .05
pie_hat <- yes_count / n;
z_s <- (pie_hat - H0) / sqrt(H0*(1-H0)/n);
p_value <- pnorm(q = z_s)</pre>
CI <- pie_hat + c(1, -1) * qnorm(alpha/2) * sqrt(pie_hat*(1-pie_hat)/n)
```

The p-value is  $5.2263376 \times 10^{-4}$  which is below the two tale cut off at 0.05. There is enough evidence to reject  $H_0$  and favor  $H_1$ . 95% confidence interval: 0.4387446, 0.484501

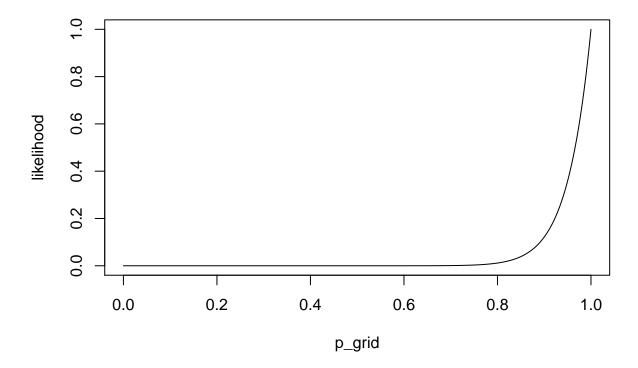
#### 4a.

```
n_better <- 20;
n_trial <- 20;
n_simulate <- 1e4;
H0 <- 0.5;
pie_hat <- n_better/n_trial;

p_grid <- seq(from=0 , to=1 , length.out=n_simulate);

likelihood <- dbinom(x = n_better, size = n_trial, prob=p_grid);
plot(p_grid, likelihood, main = "Likelihood Function: 20 success out of 20 trials", type = 'l');</pre>
```

## Likelihood Function: 20 success out of 20 trials



The maximum likelihood estimate of  $\pi$  is

success/no.trial = 20/20 = 1

**4b.** 

```
se <- sqrt(pie_hat * (1 - pie_hat) / n_trial)
CI <- pie_hat + c(1, -1) * qt(alpha/2, df = n_trial - 1) * se;
z_W <- (pie_hat - HO) / se;</pre>
```

Wald test statistic is inf. The 95% Wald confidence interval for  $\pi$  is between 1 and 1.

4c.

```
se <- sqrt(H0 * (1 - H0) / n_trial);
z_S <- (pie_hat - H0) / se;
CI <- pie_hat + c(1, -1) * qt(alpha/2, df = n_trial - 1) * se;
cutoff <- qt(alpha/2, df = n_trial - 1, lower.tail = F);</pre>
```

Score test statistic is 4.472136, which is greater than the cutoff at 2.0930241. There is evidence to rehject  $H_0$ . The 95% Score confidence interval for  $\pi$  is between 0.7659928 and 1.2340072.

### 4d.

```
L_H0 <- dbinom(n_better, n_trial, prob = H0, log = TRUE);
L_pie_hat <- dbinom(n_better, n_trial, prob = n_better/n_trial, log = TRUE);
z_L <- -2 * (L_H0 - L_pie_hat);
cutoff <- qchisq(.95, 1, lower.tail = T);
upper_bound <- 1 - exp(cutoff/(2*L_H0));</pre>
```

The confidence interval is between zero and 0.1293814; The log likelihood statistic is 27.7258872, which is greater than the cutoff at 3.8414588. At 0.05  $\alpha$  level, we conclude that there is evidence to suggest the data DO NOT follow the null hypothesis.

### 6.

```
deaths <- c(109, 65, 22, 3, 1, 0);
count <- 0:5;
lamb <- mean(rep(count, times=deaths));
expected <- dpois(count, lambda = lamb) * sum(deaths);

#merge cases fewer than 5 observations
deaths_merged <- deaths[1:3];
deaths_merged[3] <- sum(deaths[-c(1, 2)]);
expected_merged <- expected[1:3];
expected_merged[3] <- sum(expected[-c(1, 2)]);
count <- 0:2;

chi_squared <- sum( ((deaths_merged - expected_merged)^2)/expected_merged );

df <- length(count) - 1 - 1;

p_value <- pchisq(chi_squared, df = df, lower.tail = F);</pre>
```

Average death is 0.61 deaths per year. The chi-square statistic is 0.0634512; the p-value of the alternative hypothesis of the data fitting to a Poission distribution is 0.8011221. At 0.05  $\alpha$  level, we conclude that there is no real evidence to suggest the data DO NOT follow a Poisson distribution.