

## 414HW4

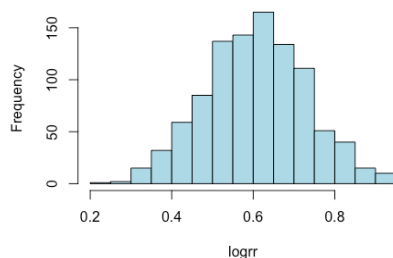
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### 1

- a. The histogram of the log relative risk was shown below, and the estimated standard error is 0.1236. 95% CI for the RR is (1.4268, 2.3160)

```
set.seed(456)
dat = c(189, 104, 10845, 10933)
myoasp = matrix(dat, nrow = 2, ncol = 2)
colnames(myoasp) = c("Yes", "No")
rownames(myoasp) = c("Placebo", "Aspirin")
simdat = rmultinom(1000, size = 22071,
                  prob = c(189, 104, 10845, 10933)/22071)
logthfun = function(thta){
  log((thta[1]/(thta[1]+thta[3])) / (thta[2]/(thta[2]+thta[4])))}
logrr = apply(simdat, 2, logthfun)
hist(logrr, col="lightblue", breaks = 15, main = "")
```



```
round(sd(logrr), 4)
## [1] 0.1236
# 95% CI
dat_log = log((189/(189+10845)) / (104/(104+10933)))#0.6071993
exp(dat_log - 1.96*sd(logrr))
## [1] 1.426781
exp(dat_log + 1.96*sd(logrr))
## [1] 2.315986
```

- b. The estimated standard error is 0.1213, which is close to the result in part a

```

pihat = cbind(c(189, 10845, 104, 10933)/22071)
sigma = diag(c(pihat)) - pihat %*% t(pihat)
temp = c(-1/(pihat[1] + pihat[2]), 1/(pihat[3] + pihat[4]))
B = rbind(c(temp[1]+(1/pihat[1]), temp[1], temp[2]-(1/pihat[3]), temp[2]))
sqrt(B %*% sigma %*% t(B)/22071)

##           [,1]
## [1,] 0.1213473

```

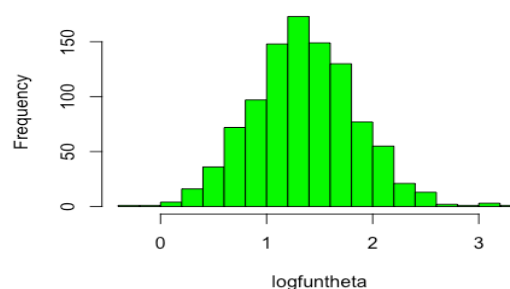
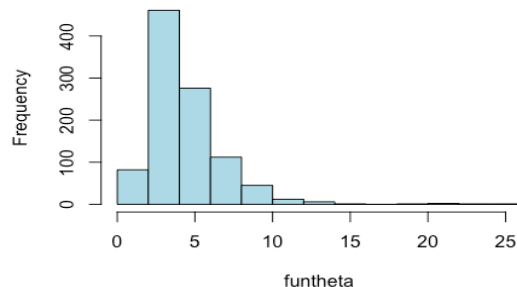
## 2

- The histogram of the odds ratio is shown below, and the standard error of the sample is 2.4569.
- The standard error of the log odds ratio samples is 0.4893. The log odds ratio follows a normal distribution. The log odds ratio is 1.3218 (please check the hand-written part below) and 95% CI of theta is (1.4391, 9.7726)
- The estimated standard error of log odds ratio calculated by hand is 0.4726, and the 95% CI is (1.4815, 9.4697), which is pretty close to the result in part b. Please check the hand-written part below. The small difference is probably because of the sample size.

```

set.seed(456)
dat2 = c(30, 10, 20, 25)
myoasp2 = matrix(dat2, nrow = 2, ncol = 2)
colnames(myoasp2) = c("Yes", "No")
rownames(myoasp2) = c("Yes", "No")
simdat2 = rmultinom(1000, size = 85,
                    prob = c(30, 20, 10, 25)/85) # row-based
fun = function(thta){
  (thta[1] * thta[4]) / (thta[2] * thta[3])
}
funtheta = apply(simdat2, 2, fun)
hist(funtheta, col="lightblue", breaks = 15, main = "")

```



```

round(sd(funtheta), 4)

```

```

## [1] 2.4569

```

```
#b.  
logfun = function(thta){  
  log((thta[1] * thta[4]) / (thta[2]*(thta[3])))}  
logfuntheta = apply(simdat2, 2, logfun)  
hist(logfuntheta, col="green", breaks = 15, main = "")  
  
round(sd(logfuntheta), 4)  
## [1] 0.4893  
  
dat_log2 = 1.3218  
exp(dat_log2 - 1.96*sd(logfuntheta))  
## [1] 1.437172  
  
exp(dat_log2 + 1.96*sd(logfuntheta))  
## [1] 9.785703
```

$$2b. \log \hat{\theta} = \log \left( \frac{n_{11} n_{22}}{n_{12} n_{21}} \right) = \log \left( \frac{30 \cdot 25}{20 \cdot 10} \right) = \log(3.75) \\ = 1.3218$$

$$2c. \hat{se}(\log \hat{\theta}) = \sqrt{\frac{1}{30} + \frac{1}{20} + \frac{1}{10} + \frac{1}{25}} \\ = 0.4726$$

95% CI:

$$e^{1.3218 \pm 1.96 \cdot 0.4726} = e^{(0.3955, 2.2481)} \\ = (1.4851, 9.4697)$$

$$3a. \sum_{i=1}^2 \sum_{j=1}^2 (n_{ij} - \hat{n}_{ij}) = (n_{11} - \hat{n}_{11}) + (n_{12} - \hat{n}_{12}) + \\ (n_{21} - \hat{n}_{21}) + (n_{22} - \hat{n}_{22}) \\ = \underbrace{n_{11}} - \frac{n_{1+} n_{+1}}{n} + \underbrace{n_{12}} - \frac{n_{1+} n_{+2}}{n} + \underbrace{n_{21}} - \frac{n_{2+} n_{+1}}{n} + \underbrace{n_{22}} - \frac{n_{2+} n_{+2}}{n} \\ = \left( n_{1+} - \frac{n_{1+} n_{+1}}{n} - \frac{n_{1+} n_{+2}}{n} \right) + \left( n_{2+} - \frac{n_{2+} n_{+1}}{n} - \frac{n_{2+} n_{+2}}{n} \right) \\ = n_{1+} \left[ 1 - \frac{n_{+1} + n_{+2}}{n} \right] + n_{2+} \left[ 1 - \frac{n_{+1} + n_{+2}}{n} \right] \\ = 0$$

$$3b. \sum_{i=1}^2 \sum_{j=1}^2 f(n_{ij}) \approx \sum_{i=1}^2 \sum_{j=1}^2 f(\hat{n}_{ij}) + (n_{ij} - \hat{n}_{ij}) f'(\hat{n}_{ij}) + \\ \frac{1}{2} (n_{ij} - \hat{n}_{ij})^2 f''(\hat{n}_{ij})$$

find out  $f(\hat{n}_{ij})$ ,  $f'(\hat{n}_{ij})$ , and  $f''(\hat{n}_{ij})$ :

since  $f(n_{ij}) = n_{ij} \log(n_{ij} / \hat{n}_{ij}) = n_{ij} \log(n_{ij}) - n_{ij} \log(\hat{n}_{ij})$

$$f'(n_{ij}) = \log(n_{ij}) + 1 - \log(\hat{n}_{ij})$$

$$f''(n_{ij}) = \frac{1}{n_{ij}}$$

So,  $f(\hat{n}_{ij}) = 0$

$$f'(\hat{n}_{ij}) = 1$$

$$f''(\hat{n}_{ij}) = \frac{1}{\hat{n}_{ij}}$$

$$= 2 \sum_{i=1}^2 \sum_{j=1}^2 \left[ 0 + (n_{ij} - \hat{n}_{ij}) + \frac{1}{2} \frac{(n_{ij} - \hat{n}_{ij})^2}{\hat{n}_{ij}} \right]$$

$$= \sum_{i=1}^2 \sum_{j=1}^2 \frac{(n_{ij} - \hat{n}_{ij})^2}{\hat{n}_{ij}} = \chi^2$$

↑ Pearson

chi-squared