

Practical: Effect Estimates Binary Data

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We will use various R packages to compute these measures such as epicalc, Epi, epiR, and epitools to obtain effect estimates for binary data

Part 1: Load the dataset

We will use the lung capacity dataset to compute effect estimates for binary data (Risk Ratio and Odds Ratio)

Load the *LungCapData.csv* dataset into memory.

```
setwd("/Users/akamau/Work/OneDrive - Kemri Wellcome Trust/Stats forum/Stat training")
LungCapData <- read.csv("Data/LungCapData.csv", header=TRUE)
```

Part 2: Explore the dataset

```
names(LungCapData)
class(LungCapData$Gender)
levels(LungCapData$Gender)
class(LungCapData$Smoke)
levels(LungCapData$Smoke)
```

Explore the relationship between various variables

Load the R packages 'descr'.

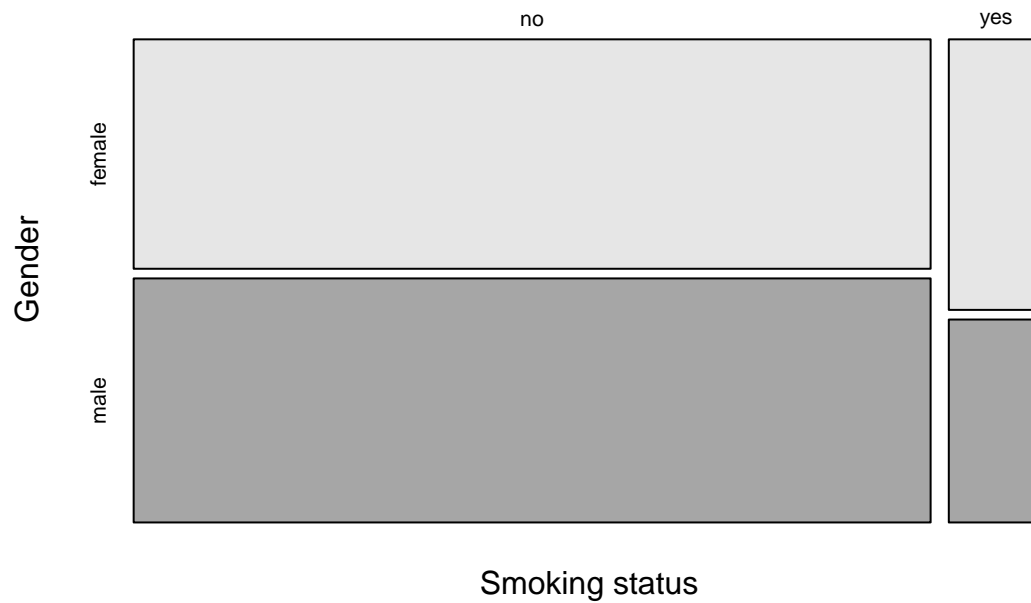
You can use the help function to determine what each package does e.g. ?descr

```
if(!require(descr)) install.packages("descr"); library(descr)
```

```
tab<-table(LungCapData$Gender, LungCapData$Smoke)
tab
```

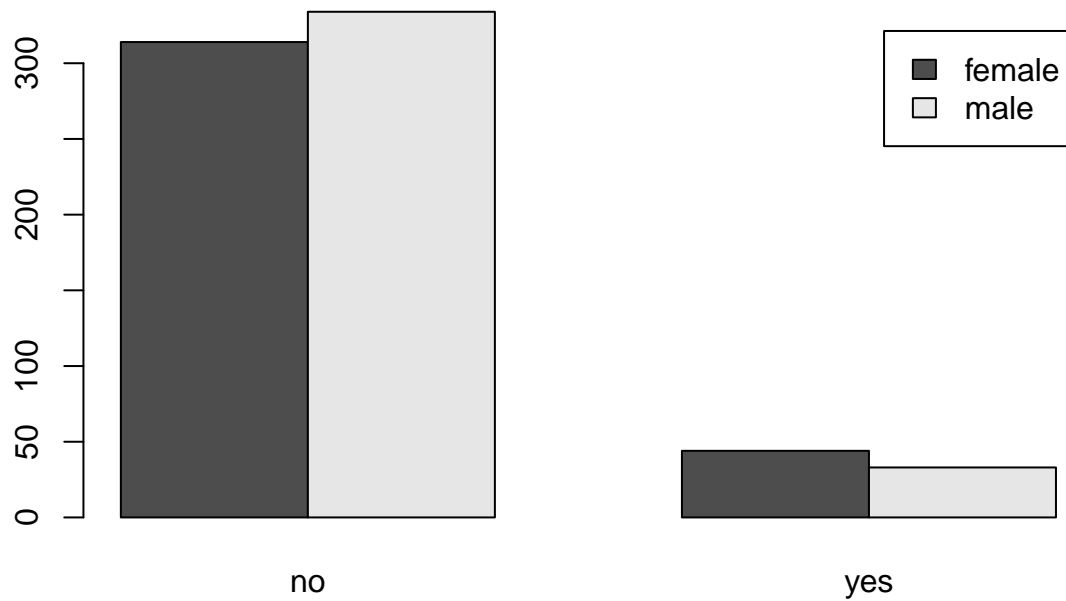
```
##
##          no yes
##  female 314  44
##   male  334  33
```

```
crosstab(LungCapData$Gender,LungCapData$Smoke, prop.c=T, xlab="Smoking status",ylab="Gender")
```



```
##      Cell Contents
## |-----|
## |              Count |
## |      Column Percent |
## |-----|
##
## =====
##              LungCapData$Smoke
## LungCapData$Gender      no      yes  Total
## -----
## female              314      44     358
##                   48.5%   57.1%
## -----
## male                334      33     367
##                   51.5%   42.9%
## -----
## Total               648      77     725
##                   89.4%   10.6%
## =====
```

```
barplot(tab, beside=T, legend=T)
```



```
chisq.test(tab, correct = F)
```

```
##
## Pearson's Chi-squared test
##
## data: tab
## X-squared = 2.0773, df = 1, p-value = 0.1495
```

Part 3: Load the R package ‘epitools’, ‘epiR’ to calculate the RR and OR

```
if(!require(epitools)) install.packages("epitools"); library(epitools)
if(!require(epiR)) install.packages("epiR"); library(epiR)
```

Part 4: Calculate the Risk Ratio and Odds Ratios

Recall the table ‘tab’. We need to set it in the standard abcd format. The standard format dictates that the exposure be presented in rows while the outcome be presented in columns

```
tab2<-cbind(tab[,2],tab[,1])
colnames(tab2)<-c("yes","no")
tab2
```

```
##      yes no
## female 44 314
## male   33 334
```

Using epitools package

```
epitab(tab2, method="oddsratio", rev="both", pvalue = "chi2")
```

```
## $tab
##      no      p0 yes      p1 oddsratio      lower      upper      p.value
## male  334 0.5154321  33 0.4285714  1.000000      NA      NA      NA
## female 314 0.4845679  44 0.5714286  1.418259 0.8803057 2.284955 0.1495036
##
```

```
## $measure
## [1] "wald"
##
## $conf.level
## [1] 0.95
##
## $pvalue
## [1] "chi2"
```

Using epiR package

```
epi.2by2(tab2, method="cohort.count", conf.level=0.95)
```

```
##           Outcome +      Outcome -      Total      Inc risk *
## Exposed +           44           314          358           12.29
## Exposed -           33           334          367           8.99
## Total              77           648          725          10.62
##
##           Odds
## Exposed +       0.1401
## Exposed -       0.0988
## Total          0.1188
##
## Point estimates and 95% CIs:
## -----
## Inc risk ratio                1.37 (0.89, 2.10)
## Odds ratio                    1.42 (0.88, 2.28)
## Attrib risk *                 3.30 (-1.19, 7.79)
## Attrib risk in population *   1.63 (-2.06, 5.32)
## Attrib fraction in exposed (%) 26.84 (-12.15, 52.28)
## Attrib fraction in population (%) 15.34 (-8.10, 33.69)
## -----
## Test that odds ratio = 1: chi2(1) = 2.077 Pr>chi2 = 0.15
## Wald confidence limits
## CI: confidence interval
## * Outcomes per 100 population units
```

Part 5: Interpretation

Females are 1.37 times more likely to be smokers than males

The odds of smoking among female is 1.42 times higher than the odds of smoking among male

##Refer to the example on the relationship between lung cancer and smoking **

Using the example used in the slides to compute effect estimates using R **

```
smoking <- matrix(c(709, 154, 142, 308), nrow = 2)
rownames(smoking) <- c("smokers", "non-smokers")
colnames(smoking) <- c("lungcancer", "no-lung cancer")
```

```
smoking
```

```
##           lungcancer no-lung cancer
## smokers           709           142
## non-smokers        154           308
```

```
epi.2by2(smoking, method="cohort.count", conf.level=0.95)
```

```
##           Outcome +      Outcome -      Total      Inc risk *
## Exposed +           709          142        851          83.3
## Exposed -           154          308        462          33.3
## Total              863          450       1313          65.7
```

```
##           Odds
```

```
## Exposed +           4.99
```

```
## Exposed -           0.50
```

```
## Total              1.92
```

```
##
```

```
## Point estimates and 95% CIs:
```

```
## -----
```

```
## Inc risk ratio                2.50 (2.19, 2.85)
```

```
## Odds ratio                    9.99 (7.67, 13.01)
```

```
## Attrib risk *                 49.98 (45.01, 54.96)
```

```
## Attrib risk in population *   32.39 (27.39, 37.40)
```

```
## Attrib fraction in exposed (%) 59.99 (54.33, 64.95)
```

```
## Attrib fraction in population (%) 49.29 (43.36, 54.59)
```

```
## -----
```

```
## Test that odds ratio = 1: chi2(1) = 332.057 Pr>chi2 = < 0.001
```

```
## Wald confidence limits
```

```
## CI: confidence interval
```

```
## * Outcomes per 100 population units
```

```
epitab(smoking, method="oddsratio", rev="both")
```

```
## $tab
```

```
##           no-lung cancer      p0 lungcancer      p1 oddsratio
```

```
## non-smokers           308 0.6844444         154 0.1784473  1.000000
```

```
## smokers              142 0.3155556         709 0.8215527  9.985915
```

```
##           lower      upper      p.value
```

```
## non-smokers      NA      NA      NA
```

```
## smokers       7.66614 13.00766 4.826448e-74
```

```
##
```

```
## $measure
```

```
## [1] "wald"
```

```
##
```

```
## $conf.level
```

```
## [1] 0.95
```

```
##
```

```
## $pvalue  
## [1] "fisher.exact"
```

Part 6: Assignment

1. Evaluate the birthweight2 data
2. Look at the association between LBW and gestational weeks. Divide gestational week in two groups i.e. group1 < 35 weeks and group2 >= 35 weeks. Then calculate the odds ratio between LBW and gestational weeks.
3. Check the Odds ratio for the association between LBW and gender
4. Make Conclusion