

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.

```
LungCapData<-read.csv("/Users/akamau/Desktop/Stats forum/Stat training/LungCapData.csv", header=TRUE)
```

Part 2: Explore the dataset

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

Explore the relationship between various variables

Load the R packages 'ggplot2', 'lme4', 'descr'.

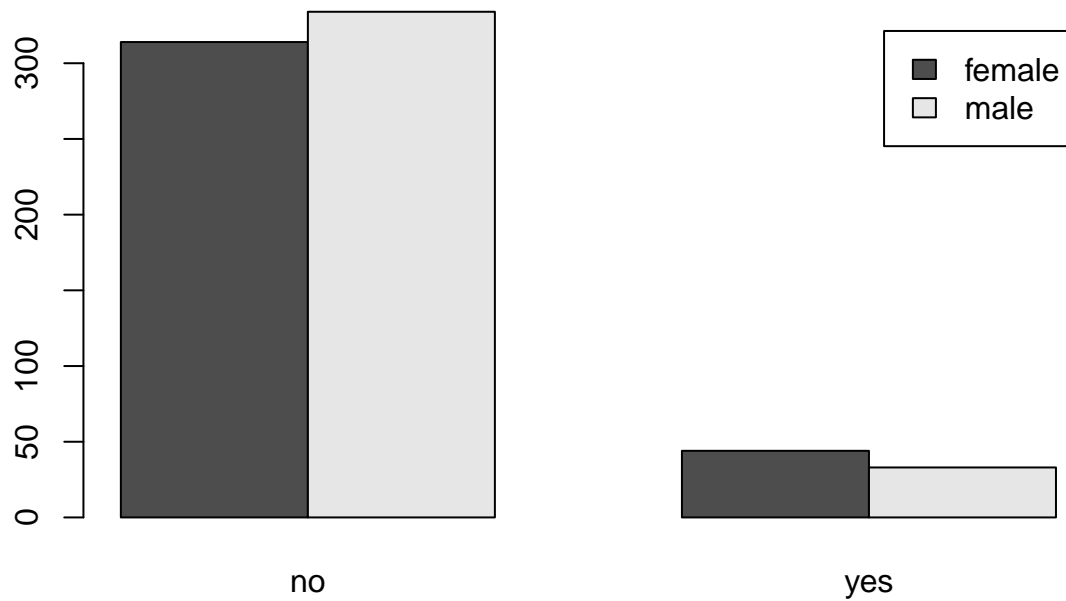
You can use the help function to determine what each package does e.g. ?ggplot

```
library(ggplot2)
library(lme4)
library(descr)
```

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

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

```
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
```

```
fisher.test(tab, conf.int=T, conf.level=0.95)
```

```
##
## Fisher's Exact Test for Count Data
##
## data:  tab
## p-value = 0.1845
## alternative hypothesis: true odds ratio is not equal to 1
## 95 percent confidence interval:
##  0.4233701 1.1659426
## sample estimates:
## odds ratio
##  0.7054345
```

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

```
library(epitools)
library(epiR)
library(abd)
library(epiDisplay)
```

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
```

```
fisher.test(tab2, conf.int=T, conf.level=0.95)
```

```
##
## Fisher's Exact Test for Count Data
##
## data:  tab2
## p-value = 0.1845
## alternative hypothesis: true odds ratio is not equal to 1
## 95 percent confidence interval:
##  0.8576752 2.3619999
## sample estimates:
## odds ratio
##  1.417566
```

Using epitools package

```
epitab(tab2, method="oddsratio")
```

```
## $tab
##           yes      p0  no      p1 oddsratio      lower      upper      p.value
## female   44 0.5714286 314 0.4845679  1.000000         NA         NA         NA
## male     33 0.4285714 334 0.5154321  1.418259 0.8803057 2.284955 0.1845426
##
## $measure
## [1] "wald"
##
## $conf.level
## [1] 0.95
##
## $pvalue
## [1] "fisher.exact"
```

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 (W)                1.37 (0.89, 2.10)
## Odds ratio (W)                   1.42 (0.88, 2.28)
## Attrib risk (W) *                 3.30 (-1.19, 7.79)
## Attrib risk in population (W) *   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)
## -----
## X2 test statistic: 2.077 p-value: 0.15
## W: Wald confidence limits
## * Cases per 100 population units
```

Part 5: Interpretation

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 (W)                2.50 (2.19, 2.85)
## Odds ratio (W)                   9.99 (7.67, 13.01)
## Attrib risk (W) *                49.98 (45.01, 54.96)
## Attrib risk in population (W) *  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)
## -----
## X2 test statistic: 332.057 p-value: < 0.001
## W: Wald confidence limits
## * Cases 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 gestwks into quartiles and analyse as groups.
3. Check the Odds ratio for the association between LBW and ethnicity

4. Finally look at LBW and sex, maternal age, height.
5. Make Conclusion