Naive Bayes from Scratch with the Titanic Data

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We will use the same data and data cleaning as in the first notebook in this chapter, so we repeate those steps first with no commentary.

Load the data

```
df <- read.csv("data/titanic3.csv", header=TRUE, stringsAsFactors = FALSE)</pre>
# subset to just columns survived, pclass, sex, and age
df \leftarrow df[,c(1,2,4,5)]
# pclass and survived and sex should be factors
df$pclass <- factor(df$pclass)</pre>
df$survived <- factor(df$survived)</pre>
df$sex <- factor(df$sex, levels=c("female", "male"))</pre>
# remove NAs
df <- df[!is.na(df$pclass),]</pre>
df <- df[!is.na(df$survived),]</pre>
df$age[is.na(df$age)] <- median(df$age,na.rm=T)</pre>
# divide into train and test
set.seed(1234)
i <- sample(1:nrow(df), 0.75*nrow(df), replace=FALSE)
train <- df[i,]
test <- df[-i,]
# perform Naive Bayes
library(e1071)
nb1 <- naiveBayes(df[,-2], df[,2], data=train)</pre>
pred <- predict(nb1, newdata=test[,-2], type="raw")</pre>
# look at first 5 (actual: 0 1 1 1 0)
pred[1:5,]
                  0
## [1,] 0.07032876 0.9296712
## [2,] 0.13421950 0.8657805
## [3,] 0.64596090 0.3540391
## [4,] 0.11063661 0.8893634
```

Calculate priors

[5,] 0.11954430 0.8804557

Using the training data we calculate prior probability of survived/perished as the percentage for each category.

```
apriori <- c(
    nrow(df[df$survived=="0",])/nrow(df),
    nrow(df[df$survived=="1",])/nrow(df)
    )
print("Prior probability, survived=no, survived=yes:")</pre>
```

```
## [1] "Prior probability, survived=no, survived=yes:"
apriori
```

```
## [1] 0.618029 0.381971
```

Calculate likelihoods for qualitative data

The likelihood for qualitative data is calculated as follows:

- for each class
- for each factor level i
- likelihood (class=i|survived=yes) = count(factor = i and survived=yes) / count(survived=yes)
- likelihood (class=i|survived=no) = count(factor = i and survived=no) / count(survived=np)

we will use nrow() to get N

```
# get survived counts for no and yes
count_survived <- c(</pre>
  length(df$survived[df$survived=="0"]),
  length(df$survived[df$survived=="1"])
# likelihood for pclass
lh_pclass <- matrix(rep(0,6), ncol=3)</pre>
for (sv in c("0", "1")){
  for (pc in c("1","2","3")) {
    lh_pclass[as.integer(sv)+1, as.integer(pc)] <-</pre>
      nrow(df[df$pclass==pc & df$survived==sv,]) / count_survived[as.integer(sv)+1]
  }
}
# likelihood for sex
lh_sex <- matrix(rep(0,4), ncol=2)</pre>
for (sv in c("0", "1")){
  for (sx in c(1, 2)) {
    lh_sex[as.integer(sv)+1, sx] <-</pre>
      nrow(df[as.integer(df$sex)==sx & df$survived==sv,]) /
       count_survived[as.integer(sv)+1]
  }
}
```

likelihood p(survived|pclass)

print("Likelihood values for p(sex|survived):")

```
print("Likelihood values for p(pclass|survived):")
## [1] "Likelihood values for p(pclass|survived):"
lh_pclass
## [,1] [,2] [,3]
## [1,] 0.1520396 0.1953028 0.6526576
## [2,] 0.4000000 0.2380000 0.3620000
likelihood p(survived|sex)
```

```
## [1] "Likelihood values for p(sex|survived):"
lh_sex

## [,1] [,2]
## [1,] 0.1569839 0.8430161
## [2,] 0.6780000 0.3220000
```

Calculate likelihoods for quantitative data

Age is quantitative. We need to compute the mean and variance.

```
age_mean <- c(0, 0)
age_var <- c(0, 0)
for (sv in c("0", "1")){
    age_mean[as.integer(sv)+1] <-
        mean(df$age[df$survived==sv])
    age_var[as.integer(sv)+1] <-
        var(df$age[df$survived==sv])
}</pre>
```

Probability density for quantitative data

For the qualitative variable we can calculate probabilities by dividing but for the age variable we need a function that will calculate its probability.

```
calc_age_lh <- function(v, mean_v, var_v){
    # run like this: calc_age_lh(6, 25.9, 138)
    1 / sqrt(2 * pi * var_v) * exp(-((v-mean_v)^2)/(2 * var_v))
}</pre>
```

Function for scratch model

Write a function to calculate raw probabilities given pclass, sex, and age.

Apply to the first 5 test observations

Let's look at just the first 5 test observations.

```
for (i in 1:5){
   raw <- calc_raw_prob(test[i,1], as.integer(test[i,3]), test[i,4])
   print(paste(raw[2], raw[1]))
}
## [1] "0.0703287599020377 0.929671240097962"
## [1] "0.134219499226771 0.865780500773229"
## [1] "0.64596090198233 0.35403909801767"</pre>
```

```
## [1] "0.110636612039489 0.889363387960511"
```

[1] "0.119544295476936 0.880455704523064"

pred[1:5,]

```
## 0 1
## [1,] 0.07032876 0.9296712
## [2,] 0.13421950 0.8657805
## [3,] 0.64596090 0.3540391
## [4,] 0.11063661 0.8893634
## [5,] 0.11954430 0.8804557
```