

# assignment\_3

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#summary Summary: Created a dummy variable named "INJURY" based on the "MAX\_SEV\_IR>0" variable. If "MAX\_SEV\_IR" is 1 or 2, If no further information is available for a recently reported accident, the initial prediction should be based on the overall proportion of accidents resulting in injury (value "Yes" in the "INJURY" variable) in the dataset. Utilized the first 24 records in the dataset and focus on "INJURY," "WEATHER\_R," and "TRAF\_CON\_R." Create a pivot table that examines "INJURY" concerning the two predictors for these 24 records. Computed the exact Bayes conditional probabilities of an injury ("INJURY" = Yes) for each of the six possible combinations of the predictors. Classified the 24 accidents using these probabilities with a cutoff of 0.5. Computed the Naive Bayes conditional probability of an injury given "WEATHER\_R" = 1 and "TRAF\_CON\_R" = 1 manually. as we got the same values for the manual and navie bayes as 0. Runs a Naive Bayes classifier on the 24 records and two predictors. model output to obtain probabilities and classifications for all 24 records. Split the dataset into training (60%) and validation (40%) sets. Apply a Naive Bayes classifier on the complete training set with the relevant predictors, including "INJURY" as the response. All predictors in this case are categorical. as required we got the confusion matrix as below predicted no yes no 11 7 yes 0 0 and got the over\_error rate is 0.3888889. By performing these tasks, we can gain insights into the predictive capabilities of the Naive Bayes classifier for accident injury prediction.

#loading the required libraries and reading the accident\_data set

```
library(e1071)
library(caret)

## Loading required package: ggplot2

## Loading required package: lattice

accidents_data = read.csv("C:/Users/varshitha/Downloads/accidentsFull.csv")
accidents_data$INJURY = ifelse(accidents_data$MAX_SEV_IR>0,"yes","no")

# Convert variables to factor
for (i in c(1:dim(accidents_data)[2])){
  accidents_data[,i] <- as.factor(accidents_data[,i])
}
head(accidents_data,n=24)

##      HOUR_I_R ALCHL_I ALIGN_I STRATUM_R WRK_ZONE WKDY_I_R INT_HWY LGTCON_I_R
## 1          0       2       2        1        0        1        0        3
## 2          1       2       1        0        0        1        1        3
## 3          1       2       1        0        0        1        0        3
## 4          1       2       1        1        0        0        0        3
```

## 5	1	1	1	0	0	1	0	3
## 6	1	2	1	1	0	1	0	3
## 7	1	2	1	0	0	1	1	3
## 8	1	2	1	1	0	1	0	3
## 9	1	2	1	1	0	1	0	3
## 10	0	2	1	0	0	0	0	3
## 11	1	2	1	0	0	1	0	3
## 12	1	2	1	1	0	1	0	3
## 13	1	2	1	1	0	1	0	3
## 14	1	2	2	0	0	1	0	3
## 15	1	2	2	1	0	1	0	3
## 16	1	2	2	1	0	1	0	3
## 17	1	2	1	1	0	1	0	3
## 18	1	2	1	1	0	0	0	3
## 19	1	2	1	1	0	1	0	3
## 20	1	2	1	0	0	1	0	3
## 21	1	2	1	1	0	1	0	3
## 22	1	2	2	0	0	1	0	3
## 23	1	2	1	0	0	1	0	3
## 24	1	2	1	1	0	1	9	3
##	MANCOL_I_R	PED_ACC_R	RELJCT_I_R	REL_RWY_R	PROFIL_I_R	SPD_LIM	SUR_COND	
## 1	0	0	1	0	1	40	4	
## 2	2	0	1	1	1	70	4	
## 3	2	0	1	1	1	35	4	
## 4	2	0	1	1	1	35	4	
## 5	2	0	0	1	1	25	4	
## 6	0	0	1	0	1	70	4	
## 7	0	0	0	0	1	70	4	
## 8	0	0	0	0	1	35	4	
## 9	0	0	1	0	1	30	4	
## 10	0	0	1	0	1	25	4	
## 11	0	0	0	0	1	55	4	
## 12	2	0	0	1	1	40	4	
## 13	1	0	0	1	1	40	4	
## 14	0	0	0	0	1	25	4	
## 15	0	0	0	0	1	35	4	
## 16	0	0	0	0	1	45	4	
## 17	0	0	0	0	1	20	4	
## 18	0	0	0	0	1	50	4	
## 19	0	0	0	0	1	55	4	
## 20	0	0	1	1	1	55	4	
## 21	0	0	1	0	0	45	4	
## 22	0	0	1	0	0	65	4	
## 23	0	0	0	0	0	65	4	
## 24	2	0	1	1	0	55	4	
##	TRAF_CON_R	TRAF_WAY	VEH_INVL	WEATHER_R	INJURY_CRASH	NO_INJ_I	PRPTYDMG_C	
RASH								
## 1	0	3	1	1	1	1	1	
0								
## 2	0	3	2	2	0	0	0	

```

1
## 3      1      2      2      2      0      0
1
## 4      1      2      2      1      0      0
1
## 5      0      2      3      1      0      0
1
## 6      0      2      1      2      1      1
0
## 7      0      2      1      2      0      0
1
## 8      0      1      1      1      1      1
0
## 9      0      1      1      2      0      0
1
## 10     0      1      1      2      0      0
1
## 11     0      1      1      2      0      0
1
## 12     2      1      2      1      0      0
1
## 13     0      1      4      1      1      2
0
## 14     0      1      1      1      0      0
1
## 15     0      1      1      1      1      1
0
## 16     0      1      1      1      1      1
0
## 17     0      1      1      2      0      0
1
## 18     0      1      1      2      0      0
1
## 19     0      1      1      2      0      0
1
## 20     0      1      1      2      0      0
1
## 21     0      3      1      1      1      1
0
## 22     0      3      1      1      0      0
1
## 23     2      2      1      2      1      2
0
## 24     0      2      2      2      1      1
0
##      FATALITIES MAX_SEV_IR INJURY
## 1      0          1      yes
## 2      0          0      no
## 3      0          0      no
## 4      0          0      no

```

```

## 5      0      0    no
## 6      0      1   yes
## 7      0      0    no
## 8      0      1   yes
## 9      0      0    no
## 10     0      0    no
## 11     0      0    no
## 12     0      0    no
## 13     0      1   yes
## 14     0      0    no
## 15     0      1   yes
## 16     0      1   yes
## 17     0      0    no
## 18     0      0    no
## 19     0      0    no
## 20     0      0    no
## 21     0      1   yes
## 22     0      0    no
## 23     0      1   yes
## 24     0      1   yes

```

#selecting the first 24 rows of the data and selecting the required variables

```

data2 = accidents_data[1:24,c("INJURY","WEATHER_R","TRAF_CON_R")]
head(data2)

##   INJURY WEATHER_R TRAF_CON_R
## 1   yes      1        0
## 2   no       2        0
## 3   no       2        1
## 4   no       1        1
## 5   no       1        0
## 6   yes      2        0

```

#making the data into a table

```

pivot_Table1 <- ftable(data2)
pivot_Table2 <- ftable(data2[, -1]) # print table only for conditions
pivot_Table1

##                                     TRAF_CON_R 0 1 2
## INJURY WEATHER_R
## no      1                      3 1 1
##         2                      9 1 0
## yes     1                      6 0 0
##         2                      2 0 1

pivot_Table2

##             TRAF_CON_R 0 1 2
## WEATHER_R

```

```
## 1          9  1  1
## 2         11  1  1
```

#presenting the possibiltes of 6 conditional probabilties

```
# Injury = yes
y1 = pivot_Table1[3,1] / pivot_Table2[1,1] # Injury, Weather=1 and Traf=0
y2 = pivot_Table1[4,1] / pivot_Table2[2,1] # Injury, Weather=2, Traf=0
y3 = pivot_Table1[3,2] / pivot_Table2[1,2] # Injury, W=1, T=1
y4 = pivot_Table1[4,2] / pivot_Table2[2,2] # I, W=2,T=1
y5 = pivot_Table1[3,3] / pivot_Table2[1,3] # I, W=1,T=2
y6 = pivot_Table1[4,3]/ pivot_Table2[2,3] #I,W=2,T=2

# Injury = no
n1 = pivot_Table1[1,1] / pivot_Table2[1,1] # Weather=1 and Traf=0
n2 = pivot_Table1[2,1] / pivot_Table2[2,1] # Weather=2, Traf=0
n3 = pivot_Table1[1,2] / pivot_Table2[1,2] # W=1, T=1
n4 = pivot_Table1[2,2] / pivot_Table2[2,2] # W=2,T=1
n5 = pivot_Table1[1,3] / pivot_Table2[1,3] # W=1,T=2
n6 = pivot_Table1[2,3] / pivot_Table2[2,3] # W=2,T=2
print(c(y1,y2,y3,y4,y5,y6))

## [1] 0.6666667 0.1818182 0.0000000 0.0000000 0.0000000 1.0000000

print(c(n1,n2,n3,n4,n5,n6))

## [1] 0.3333333 0.8181818 1.0000000 1.0000000 1.0000000 0.0000000
```

#computing the values for 24 accidents applying the conditon of cutoff probability = 0.5

```
prob.inj <- rep(0,24)

for (i in 1:24) {
  print(c(data2$WEATHER_R[i],data2$TRAF_CON_R[i]))
  if (data2$WEATHER_R[i] == "1") {
    if (data2$TRAF_CON_R[i]=="0"){
      prob.inj[i] = y1
    }
    else if (data2$TRAF_CON_R[i]=="1") {
      prob.inj[i] = y3
    }
    else if (data2$TRAF_CON_R[i]=="2") {
      prob.inj[i] = y5
    }
  }
  else {
    if (data2$TRAF_CON_R[i]=="0"){
      prob.inj[i] = y2
    }
    else if (data2$TRAF_CON_R[i]=="1") {
      prob.inj[i] = y4
    }
  }
}
```



```

## Levels: 1 2 0
## [1] 2 2
## Levels: 1 2 0
## [1] 2 0
## Levels: 1 2 0

data2$prob.inj <- prob.inj

data2$pred.prob <- ifelse(data2$prob.inj>0.5, "yes", "no")

#computing the manually naive Bayes conditional probability of an injury given
WEATHER_R = 1 and TRAF_CON_R = 1.

p(I=Y|W=1,T=1) = P(I=Y | W=1,T=1)/ P(W=1,T=1)

=(0/24)/(1/24) = 0/1 =0

#2. Run a naive Bayes classifier on the 24 records and two predictors. Check the model
output to obtain probabilities and classifications for all 24 records. Compare this to the
exact Bayes classification. Are the resulting classifications equivalent? Is the ranking (= ordering)
of observations equivalent?

nb <- naiveBayes(INJURY ~ TRAF_CON_R + WEATHER_R,
                  data = data2)

nbt <- predict(nb, newdata = data2,type = "raw")
data2$nbpred.prob <- nbt[,2] # Transfer the "Yes" nb prediction

library(klaR)

## Loading required package: MASS

nb2 <- train(INJURY ~ TRAF_CON_R + WEATHER_R,
              data = data2, method = "nb")

## Warning: model fit failed for Resample01: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample02: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

## Warning: model fit failed for Resample03: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample04: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

```

```
## Warning: model fit failed for Resample05: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample06: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

## Warning: model fit failed for Resample07: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample08: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample09: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample10: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample11: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

## Warning: model fit failed for Resample12: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample13: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample14: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

## Warning: model fit failed for Resample15: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2
```

```
## Warning: model fit failed for Resample16: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

## Warning: model fit failed for Resample17: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample18: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample19: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample20: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

## Warning: model fit failed for Resample21: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample22: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1

## Warning: model fit failed for Resample23: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample24: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning: model fit failed for Resample25: usekernel=FALSE, fL=0, adjust=1
Error in NaiveBayes.default(x, y, usekernel = FALSE, fL = param$fL, ...) :
##   Zero variances for at least one class in variables: TRAF_CON_R1, TRAF_CO
N_R2

## Warning in nominalTrainWorkflow(x = x, y = y, wts = weights, info = trainI
nfo,
## : There were missing values in resampled performance measures.
```

```

## Warning in train.default(x, y, weights = w, ...): missing values found in
## aggregated results

predict(nb2, newdata = data2[,c("INJURY", "WEATHER_R", "TRAF_CON_R")])

## [1] no no
## Levels: no yes

predict(nb2, newdata = data2[,c("INJURY", "WEATHER_R", "TRAF_CON_R")],
       type = "raw")

## [1] no no
## Levels: no yes

#Partitioning the data into training and validation sets

set.seed(2808) # For reproducibility
train_indices <- createDataPartition(data2, p=0.60, list = FALSE)

## Warning in createDataPartition(data2, p = 0.6, list = FALSE): Some classes
## have no records ( ) and these will be ignored

## Warning in createDataPartition(data2, p = 0.6, list = FALSE): Some classes
## have a single record ( ) and these will be selected for the sample

train_data <- data2[train_indices, ]
validation_data <- data2[-train_indices, ]

#Running Naive Bayes classifier on the complete training set

nb_model <- naiveBayes(INJURY ~ WEATHER_R + TRAF_CON_R, data = train_data)
# Predict on the validation set
predicted <- predict(nb_model, newdata = validation_data)

```

## Computing the confusion matrix

```
conf_matrix <- table(predicted, validation_data$INJURY)
```

```
#Computing the overall error of the validation set
```

```
overall_error <- mean(predicted != validation_data$INJURY)
```

*#Print the confusion matrix and overall error*

```
print(conf_matrix)
```

```
##
```

```
## predicted no yes
```

```
##      no 11 7
```

```
##      yes 0 0
```

```
print(overall_error)
```

```
## [1] 0.3888889
```