

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 naive 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	LGTCN_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		
0								
## 2	0	3	2	2	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 possibilities of 6 conditional probabilities

```
# 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 condition 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
    }
  }
}
```

```

    }
    else if (data2$TRAF_CON_R[i]=="2") {
      prob.inj[i] = y6
    }
  }
}

```

```

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

```

```
## 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 no no no no no no no no no no no no no no no no no no no no
no
```

```
## Levels: no yes
```

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

```
## [1] no no no no no no no no no no no no no no no no no no no no no 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
```