hw 3 r

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#

1R Animal Classifier

0.0.1 Author & Notes

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Data Mining

Homework 3

0.0.2 Concept Description:

Train a system to classify animals as either mammal or non-mammal.

0.0.3 Data Collection:

The data has been provided by Perry B. Koob. It consists of 18 attributes: a descriptor attribute(name), 15 boolean attributes, and two numerical attributes.

0.0.4 Example Description:

Attribute Information: 1. animal name: Nominal attribute that indicates the animal. Unique for each instance.

- 2. hair: Nominal boolean attribute that describes if the animal has hair. 3. feathers: Nominal boolean attribute that describes if the animal has feathers. 4. eggs: Nominal boolean attribute that describes if the animal lays eggs.
- 5. milk: Nominal boolean attribute that describes if the animal produces milk.
- 6. airborne: Nominal boolean attribute that describes if the animal has the capability of flight.
- 7. aquatic: Nominal boolean attribute that describes if the animal has the capability to breathe under water.
- 8. predator: Nominal boolean attribute that describes if the animal is a predator.
- 9. toothed: Nominal boolean attribute that describes if the animal has teeth.
- 10. backbone: Nominal boolean attribute that describes if the animal has a backbone.
- 11. breathes: Nominal boolean attribute that describes if the animal breathes air.
- 12. venomous: Nominal boolean attribute that describes if the animal produces venom.
- 13. fins: Nominal boolean attribute that describes if the animal has fins.
- 14. legs: Numeric classification that describes the number of legs an animal has (set of values: $\{0,2,4,5,6,8\}$)
- 15. tail: Nominal boolean attribute that describes if the animal has a tail.

- 16. domestic: Nominal boolean attribute that describes if the animal has been domesticated.
- 17. catsize: Nominal boolean attribute that describes if the animal is catsized.
- 18. gestation: Numeric classification that describes if the animal has live birth (in days)

0.0.5 Data Import and Wrangling:

```
[1]: library(dplyr) #library necessary for %>%rename()
     library (OneR) #library necessary for model
     library(caret) #library necessary for confusionMatrix
    Attaching package: 'dplyr'
    The following objects are masked from 'package:stats':
        filter, lag
    The following objects are masked from 'package:base':
        intersect, setdiff, setequal, union
    Loading required package: lattice
    Loading required package: ggplot2
[2]: train <-read.csv(file ='../src-data/animal-taxonomy-train.
     test <-read.csv(file ='../src-data/animal-taxonomy-test.</pre>
      colnames(test)
     #Unused attributes
     train$legs <- NULL
     train$gestation <- NULL
    1. 'animal.name' 2. 'hair' 3. 'feathers' 4. 'eggs' 5. 'milk' 6. 'airborne' 7. 'aquatic' 8. 'predator'
    9. 'toothed' 10. 'backbone' 11. 'breathes' 12. 'venomous' 13. 'fins' 14. 'legs' 15. 'tail' 16. 'domestic'
    17. 'catsize' 18. 'gestation' 19. 'type'
[3]: | train$type <- ifelse(train$type == 'mammal', "Mammal", "NotMammal")
     train$type <- as.factor(train$type)</pre>
     test$type <- ifelse(test$type == "mammal", "Mammal", "NotMammal")</pre>
```

```
test$type <- as.factor(test$type)
test</pre>
```

	animal.name	hair	feathers	eggs	milk	airborne	aquatic	predator	too
A data.frame: 10×19	<fct $>$	<lgl $>$	<lgl $>$	<lgl $>$	<lgl $>$	<lgl $>$	<lgl $>$	<lgl $>$	<lg
	dolphin	FALSE	FALSE	FALSE	TRUE	FALSE	TRUE	TRUE	TR
	chicken	FALSE	TRUE	TRUE	FALSE	TRUE	FALSE	FALSE	FA
	aardvark	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE	TR
	platypus	TRUE	FALSE	TRUE	TRUE	FALSE	TRUE	TRUE	FA
	flamingo	FALSE	TRUE	TRUE	FALSE	TRUE	FALSE	FALSE	FA
	dove	FALSE	TRUE	TRUE	FALSE	TRUE	FALSE	FALSE	FA
	oryx	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE	FALSE	TR
	lion	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE	TR
	polecat	TRUE	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE	TR
	dogfish	FALSE	FALSE	TRUE	FALSE	FALSE	TRUE	TRUE	TR

0.0.6 Exploratory Data Analysis:

The Naive Bayes classifer uses nominal attributes, so each attribute was examined and determined to be nominal or discretized to be nominal.

0.0.7 Mining or Analytics:

```
[4]: prediction <- train
```

```
hair.count<-as.data.frame(table(train$hair))
hair.true <-as.data.frame(table(train[train$type == "Mammal",]$hair))
hair.false <-as.data.frame(table(train[train$type == "NotMammal",]$hair))

hair.count <-hair.count %>%rename(Label=Var1)
hair.true <-hair.true %>%rename(Label=Var1,Mammal=Freq)
hair.false <-hair.false %>%rename(Label=Var1,NotMammal=Freq)
hair <-merge(hair.count,hair.true, bye=Label,all=TRUE)
hair <-merge(hair,hair.false, bye=Label,all=TRUE)
hair
```

```
[6]: prediction$pred.hair <-NA prediction$pred.hair[prediction$hair=="TRUE"] <- "Mammal" prediction$pred.hair[prediction$hair=="FALSE"] <- "NotMammal" prediction$hair <-NULL
```

```
prediction$pred.hair <-as.factor(prediction$pred.hair)</pre>
```

feathers feathers.count<-as.data.frame(table(train\$feathers)) feathers.true <-as.data.frame(table(train[train\$type == "NotMammal",]\$feathers)) feathers.false <-as.data.frame(table(train[train\$type == "Mammal",]\$feathers)) feathers.count <-feathers.count %>%rename(Label=Var1) feathers.true <-feathers.true %>%rename(Label=Var1,Mammal=Freq) feathers.false <-feathers.false %>%rename(Label=Var1,NotMammal=Freq) feathers <-merge(feathers.count,feathers.true, bye=Label,all=TRUE) feathers <-merge(feathers,feathers.false, bye=Label,all=TRUE) feathers</pre>

```
[8]: prediction$pred.feathers <-NA prediction$pred.feathers[prediction$feathers=="TRUE"] <- "NotMammal" prediction$pred.feathers[prediction$feathers=="FALSE"] <- "Mammal" prediction$feathers <-NULL prediction$pred.feathers <-as.factor(prediction$pred.feathers)
```

```
eggs.count<-as.data.frame(table(train$eggs))
eggs.true <-as.data.frame(table(train[train$type == "NotMammal",]$eggs))
eggs.false <-as.data.frame(table(train[train$type == "Mammal",]$eggs))

eggs.count <-eggs.count %>%rename(Label=Var1)
eggs.true <-eggs.true %>%rename(Label=Var1,Mammal=Freq)
eggs.false <-eggs.false %>%rename(Label=Var1,NotMammal=Freq)
eggs <-merge(eggs.count,eggs.true, bye=Label,all=TRUE)
eggs <-merge(eggs,eggs.false, bye=Label,all=TRUE)
eggs</pre>
```

```
[10]: prediction$pred.eggs <-NA
    prediction$pred.eggs[prediction$eggs=="TRUE"] <- "NotMammal"
    prediction$pred.eggs[prediction$eggs=="FALSE"] <- "Mammal"
    prediction$eggs <-NULL
    prediction$pred.eggs <-as.factor(prediction$pred.eggs)</pre>
```

milk.count<-as.data.frame(table(train\$milk)) milk.true <-as.data.frame(table(train[train\$type == "Mammal",]\$milk)) milk.false <-as.data.frame(table(train[train\$type == "NotMammal",]\$milk)) milk.count <-milk.count %>%rename(Label=Var1) milk.true <-milk.true %>%rename(Label=Var1,Mammal=Freq) milk.false <-milk.false %>%rename(Label=Var1,NotMammal=Freq) milk <-merge(milk.count,milk.true, bye=Label,all=TRUE) milk <-merge(milk,milk.false, bye=Label,all=TRUE) milk</pre>

```
Label
                                 Mammal NotMammal
                          Freq
                                 <int>
                                          <int>
                  <fct>
                          <int>
A data.frame: 2 \times 4
                  FALSE
                                 NA
                                          56
                          56
                  TRUE
                                          NA
                          35
                                 35
```

```
[12]: prediction$pred.milk <-NA
prediction$pred.milk[prediction$milk=="TRUE"] <- "Mammal"
prediction$pred.milk[prediction$milk=="FALSE"] <- "NotMammal"
prediction$milk <-NULL
prediction$pred.milk <-as.factor(prediction$pred.milk)</pre>
```

```
airborne.count<-as.data.frame(table(train$airborne))
airborne.true <-as.data.frame(table(train[train$type == "NotMammal",]$airborne))
airborne.false <-as.data.frame(table(train[train$type == "Mammal",]$airborne))

airborne.count <-airborne.count %>%rename(Label=Var1)
airborne.true <-airborne.true %>%rename(Label=Var1,Mammal=Freq)
airborne.false <-airborne.false %>%rename(Label=Var1,NotMammal=Freq)
airborne <-merge(airborne.count,airborne.true, bye=Label,all=TRUE)
airborne <-merge(airborne,airborne.false, bye=Label,all=TRUE)
airborne
```

```
[14]: prediction$pred.airborne <-NA prediction$pred.airborne[prediction$airborne=="TRUE"] <- "NotMammal" prediction$pred.airborne[prediction$airborne=="FALSE"] <- "Mammal" prediction$airborne <-NULL prediction$pred.airborne <-as.factor(prediction$pred.airborne)
```

aquatic.count<-as.data.frame(table(train\$aquatic)) aquatic.true <-as.data.frame(table(train[train\$type == "NotMammal",]\$aquatic)) aquatic.false <-as.data.frame(table(train[train\$type == "Mammal",]\$aquatic)) aquatic.count <-aquatic.count %>%rename(Label=Var1) aquatic.true <-aquatic.true %>%rename(Label=Var1,Mammal=Freq) aquatic.false <-aquatic.false %>%rename(Label=Var1,NotMammal=Freq) aquatic <-merge(aquatic.count,aquatic.true, bye=Label,all=TRUE) aquatic <-merge(aquatic,aquatic.false, bye=Label,all=TRUE) aquatic

```
Mammal NotMammal
                     Label
                               Freq
                     <fct>
                               <int>
                                       \langle int \rangle
                                                  <int>
A data frame: 2 \times 4
                     FALSE
                               58
                                       27
                                                  31
                     TRUE
                                       29
                                                  4
                               33
```

```
[16]: prediction$pred.aquatic <-NA
    prediction$pred.aquatic[prediction$aquatic=="TRUE"] <- "NotMammal"
    prediction$pred.aquatic[prediction$aquatic=="FALSE"] <- "Mammal"
    prediction$aquatic <-NULL
    prediction$pred.aquatic <-as.factor(prediction$pred.aquatic)</pre>
```

```
predator <-merge(predator,predator.false, bye=Label,all=TRUE)
predator</pre>
```

```
Label
                                  Freq
                                           Mammal NotMammal
                        <fct>
                                           <int>
                                  \langle int \rangle
                                                        \langle int \rangle
A data.frame: 2 \times 4
                       FALSE
                                                        23
                                  41
                                           18
                        TRUE
                                  50
                                           17
                                                        33
```

```
[18]: prediction$pred.predator <-NA prediction$pred.predator[prediction$predator=="TRUE"] <- "Mammal" prediction$pred.predator[prediction$predator=="FALSE"] <- "NotMammal" prediction$predator <-NULL prediction$pred.predator <-as.factor(prediction$pred.predator)
```

[19]: #Toothed toothed.count<-as.data.frame(table(train\$toothed)) toothed.true <-as.data.frame(table(train[train\$type == "Mammal",]\$toothed)) toothed.false <-as.data.frame(table(train[train\$type == "NotMammal",]\$toothed)) toothed.count <-toothed.count %>%rename(Label=Var1)

toothed.count <-toothed.count %>%rename(Label=Var1)
toothed.true <-toothed.true %>%rename(Label=Var1,Mammal=Freq)
toothed.false <-toothed.false %>%rename(Label=Var1,NotMammal=Freq)
toothed <-merge(toothed.count,toothed.true, bye=Label,all=TRUE)
toothed <-merge(toothed,toothed.false, bye=Label,all=TRUE)
toothed</pre>

```
Mammal NotMammal
                     Label
                              Freq
                     <fct>
                              \langle int \rangle
                                      <int>
                                                 <int>
A data.frame: 2 \times 4
                     FALSE
                                      NA
                                                 36
                              36
                     TRUE
                              55
                                      35
                                                 20
```

```
[20]: prediction$pred.toothed <-NA prediction$pred.toothed[prediction$toothed=="TRUE"] <- "Mammal" prediction$pred.toothed[prediction$toothed=="FALSE"] <- "NotMammal" prediction$toothed <-NULL prediction$pred.toothed <-as.factor(prediction$pred.toothed)
```



```
backbone.true <-backbone.true %>%rename(Label=Var1,Mammal=Freq)
backbone.false <-backbone.false %>%rename(Label=Var1,NotMammal=Freq)
backbone <-merge(backbone.count,backbone.true, bye=Label,all=TRUE)
backbone <-merge(backbone,backbone.false, bye=Label,all=TRUE)
backbone</pre>
```

[22]: prediction\$pred.backbone <-NA prediction\$pred.backbone[prediction\$backbone=="TRUE"] <- "Mammal" prediction\$pred.backbone[prediction\$backbone=="FALSE"] <- "NotMammal" prediction\$backbone <-NULL prediction\$pred.backbone <-as.factor(prediction\$pred.backbone)

[23]: #Breathes

```
breathes.count<-as.data.frame(table(train$breathes))
breathes.true <-as.data.frame(table(train[train$type == "Mammal",]$breathes))
breathes.false <-as.data.frame(table(train[train$type == "Mammal",]$breathes))

breathes.count <-breathes.)

breathes.count <-breathes.count %>%rename(Label=Var1)
breathes.true <-breathes.true %>%rename(Label=Var1,Mammal=Freq)
breathes.false <-breathes.false %>%rename(Label=Var1,NotMammal=Freq)
breathes <-merge(breathes.count,breathes.true, bye=Label,all=TRUE)
breathes <-merge(breathes.breathes.false, bye=Label,all=TRUE)
breathes</pre>
```

```
A data.frame: 2 \times 4 \begin{array}{c|ccccc} \text{Label} & \text{Freq} & \text{Mammal} & \text{NotMammal} \\ & < \text{fct} > & < \text{int} > & < \text{int} > & < \text{int} > \\ \hline & \text{FALSE} & 20 & \text{NA} & 20 \\ & \text{TRUE} & 71 & 35 & 36 \end{array}
```

[24]: prediction\$pred.breathes <-NA prediction\$pred.breathes[prediction\$breathes=="TRUE"] <- "Mammal" prediction\$pred.breathes[prediction\$breathes=="FALSE"] <- "NotMammal" prediction\$breathes <-NULL prediction\$pred.breathes <-as.factor(prediction\$pred.breathes)

```
[25]: #Venomous

venomous.count<-as.data.frame(table(train$venomous))

venomous.true <-as.data.frame(table(train[train$type == "NotMammal",]$venomous))</pre>
```

```
venomous.count <-venomous.count %>%rename(Label=Var1)
venomous.true <-venomous.true %>%rename(Label=Var1,Mammal=Freq)
venomous.false <-venomous.false %>%rename(Label=Var1,NotMammal=Freq)
venomous <-merge(venomous.count,venomous.true, bye=Label,all=TRUE)
venomous <-merge(venomous,venomous.false, bye=Label,all=TRUE)
venomous</pre>
```

```
[26]: prediction$pred.venomous <-NA
    prediction$pred.venomous[prediction$venomous=="TRUE"] <- "NotMammal"
    prediction$pred.venomous[prediction$venomous=="FALSE"] <- "Mammal"
    prediction$venomous <-NULL
    prediction$pred.venomous <-as.factor(prediction$pred.venomous)</pre>
```

```
fins.count<-as.data.frame(table(train$fins))
fins.true <-as.data.frame(table(train[train$type == "NotMammal",]$fins))
fins.false <-as.data.frame(table(train[train$type == "Mammal",]$fins))

fins.count <-fins.count %>%rename(Label=Var1)
fins.true <-fins.true %>%rename(Label=Var1,Mammal=Freq)
fins.false <-fins.false %>%rename(Label=Var1,NotMammal=Freq)
fins <-merge(fins.count,fins.true, bye=Label,all=TRUE)
fins <-merge(fins.false, bye=Label,all=TRUE)
fins</pre>
```

```
[28]: prediction$pred.fins <-NA
prediction$pred.fins[prediction$fins=="TRUE"] <- "NotMammal"
prediction$pred.fins[prediction$fins=="FALSE"] <- "Mammal"
prediction$fins <-NULL
prediction$pred.fins <-as.factor(prediction$pred.fins)
```

```
[29]: #Tail
tail.count<-as.data.frame(table(train$tail))</pre>
```

```
tail.true <-as.data.frame(table(train[train$type == "Mammal",]$tail))
tail.false <-as.data.frame(table(train[train$type == "NotMammal",]$tail))

tail.count <-tail.count %>%rename(Label=Var1)
tail.true <-tail.true %>%rename(Label=Var1,Mammal=Freq)
tail.false <-tail.false %>%rename(Label=Var1,NotMammal=Freq)
tail <-merge(tail.count,tail.true, bye=Label,all=TRUE)
tail <-merge(tail,tail.false, bye=Label,all=TRUE)
tail</pre>
```

```
Label
                          Freq
                                 Mammal NotMammal
                                 <int>
                                           <int>
                  <fct>
                          <int>
A data.frame: 2 \times 4
                  FALSE
                          25
                                           20
                                 5
                  TRUE
                          66
                                 30
                                           36
```

```
[30]: prediction$pred.tail <-NA
    prediction$pred.tail[prediction$tail=="TRUE"] <- "Mammal"
    prediction$pred.tail[prediction$tail=="FALSE"] <- "NotMammal"
    prediction$tail <-NULL
    prediction$pred.tail <-as.factor(prediction$pred.tail)</pre>
```


domestic

```
[32]: prediction$pred.domestic <-NA prediction$pred.domestic[prediction$domestic=="TRUE"] <- "Mammal" prediction$pred.domestic[prediction$domestic=="FALSE"] <- "NotMammal" prediction$domestic <-NULL prediction$pred.domestic <-as.factor(prediction$pred.domestic)
```

```
[33]: #Catsized

catsize.count<-as.data.frame(table(train$catsize))
catsize.true <-as.data.frame(table(train[train$type == "Mammal",]$catsize))
catsize.false <-as.data.frame(table(train[train$type == "NotMammal",]$catsize))

catsize.count <-catsize.count %>%rename(Label=Var1)
catsize.true <-catsize.true %>%rename(Label=Var1,Mammal=Freq)
catsize.false <-catsize.false %>%rename(Label=Var1,NotMammal=Freq)
catsize <-merge(catsize.count,catsize.true, bye=Label,all=TRUE)
catsize <-merge(catsize,catsize.false, bye=Label,all=TRUE)
catsize</pre>
```

```
[34]: prediction$pred.catsize <-NA prediction$pred.catsize[prediction$catsize=="TRUE"] <- "Mammal" prediction$pred.catsize[prediction$catsize=="FALSE"] <- "NotMammal" prediction$catsize <-NULL prediction$pred.catsize <-as.factor(prediction$pred.catsize)
```

```
[35]: #Error Rate
      prediction$pred.hair.correct =ifelse(prediction$type ==prediction$pred.hair,0,1)
      prediction$pred.feathers.correct =ifelse(prediction$type ==prediction$pred.
       \rightarrow feathers, 0, 1)
      prediction$pred.eggs.correct =ifelse(prediction$type ==prediction$pred.eggs,0,1)
      prediction$pred.milk.correct =ifelse(prediction$type ==prediction$pred.milk,0,1)
      prediction pred.airborne.correct = ifelse (prediction type == prediction pred.
       \rightarrowairborne,0,1)
      prediction$pred.aquatic.correct =ifelse(prediction$type ==prediction$pred.
       \rightarrowaquatic,0,1)
      prediction$pred.predator.correct =ifelse(prediction$type ==prediction$pred.
       \rightarrowpredator,0,1)
      prediction$pred.toothed.correct =ifelse(prediction$type ==prediction$pred.
       \rightarrowtoothed, 0, 1)
      prediction$pred.backbone.correct =ifelse(prediction$type ==prediction$pred.
       \rightarrowbackbone, 0, 1)
      prediction$pred.breathes.correct =ifelse(prediction$type ==prediction$pred.
       \hookrightarrowbreathes,0,1)
      prediction$pred.venomous.correct =ifelse(prediction$type ==prediction$pred.
       \rightarrowvenomous, 0, 1)
      prediction$pred.fins.correct =ifelse(prediction$type ==prediction$pred.fins,0,1)
```

```
prediction$pred.tail.correct =ifelse(prediction$type ==prediction$pred.tail,0,1)
prediction$pred.domestic.correct =ifelse(prediction$type ==prediction$pred.
 \rightarrowdomestic,0,1)
prediction$pred.catsize.correct =ifelse(prediction$type ==prediction$pred.
 \rightarrowcatsize,0,1)
#Print the Results
print(paste("Error Rate of Hair Rules:",sum(prediction$pred.hair.correct)/
 →nrow(prediction)))
print(paste("Error Rate of Feathers Rules:",sum(prediction$pred.feathers.
 print(paste("Error Rate of Eggs Rules:",sum(prediction$pred.eggs.correct)/
 →nrow(prediction)))
print(paste("Error Rate of Milk Rules:",sum(prediction$pred.milk.correct)/
 →nrow(prediction)))
print(paste("Error Rate of Airborne Rules:",sum(prediction$pred.airborne.

→correct)/nrow(prediction)))
print(paste("Error Rate of Aquatic Rules:",sum(prediction$pred.aquatic.correct)/
 →nrow(prediction)))
print(paste("Error Rate of Predator Rules:",sum(prediction$pred.predator.

→correct)/nrow(prediction)))
print(paste("Error Rate of Toothed Rules:",sum(prediction$pred.toothed.correct)/
 →nrow(prediction)))
print(paste("Error Rate of Backbone Rules:", sum(prediction$pred.backbone.

→correct)/nrow(prediction)))
print(paste("Error Rate of Breathes Rules:",sum(prediction$pred.breathes.

→correct)/nrow(prediction)))
print(paste("Error Rate of Venomous Rules:",sum(prediction$pred.venomous.
 print(paste("Error Rate of Fins Rules:",sum(prediction$pred.fins.correct)/
 →nrow(prediction)))
print(paste("Error Rate of Tail Rules:",sum(prediction$pred.tail.correct)/
 →nrow(prediction)))
print(paste("Error Rate of Domestic Rules:",sum(prediction$pred.domestic.
 →correct)/nrow(prediction)))
print(paste("Error Rate of Catsize Rules:",sum(prediction$pred.catsize.correct)/
 →nrow(prediction)))
[1] "Error Rate of Hair Rules: 0.0549450549450549"
[1] "Error Rate of Feathers Rules: 0.428571428571429"
[1] "Error Rate of Eggs Rules: 0.021978021978022"
[1] "Error Rate of Milk Rules: 0"
[1] "Error Rate of Airborne Rules: 0.428571428571429"
[1] "Error Rate of Aquatic Rules: 0.340659340659341"
[1] "Error Rate of Predator Rules: 0.56043956043956"
[1] "Error Rate of Toothed Rules: 0.21978021978022"
```

```
[1] "Error Rate of Backbone Rules: 0.417582417582418"
[1] "Error Rate of Breathes Rules: 0.395604395604396"
[1] "Error Rate of Venomous Rules: 0.527472527472527"
[1] "Error Rate of Fins Rules: 0.516483516483517"
[1] "Error Rate of Tail Rules: 0.450549450549451"
[1] "Error Rate of Domestic Rules: 0.32967032967033"
[1] "Error Rate of Catsize Rules: 0.208791208791209"
```

We can see that the milk rules have the lowest error rate, so we then choose it as the One Rule.

```
[36]: df_train <-train
df_train$type <-NULL

model <-OneR(df_train)
model_prediction <-predict(model,test)</pre>
```

0.0.8 Evaluation

By using the trained data set and a confusion matrix, we can obtain the accuracy and error rate of both models to determine the strength of our model.

```
[37]: #prepare the list of classes from the test data for evaluation
    reference <-as.data.frame(test$type)
    colnames(reference) <-c("class")
    reference <-as.factor(reference$class)

testing <-as.data.frame(test$hair)
    colnames(testing) <- c("milk")

testing$pred[testing$milk=="TRUE"] <- "Mammal"
    testing$pred[testing$milk=="FALSE"] <- "NotMammal"

testing <-as.factor(testing$pred)

confusionMatrix(testing, reference)</pre>
```

Confusion Matrix and Statistics

```
Reference
```

```
Prediction Mammal NotMammal Mammal 5 0 NotMammal 1 4
```

Accuracy: 0.9

95% CI: (0.555, 0.9975)

No Information Rate : 0.6 P-Value [Acc > NIR] : 0.04636

Kappa : 0.8

Mcnemar's Test P-Value : 1.00000

Sensitivity: 0.8333
Specificity: 1.0000
Pos Pred Value: 1.0000
Neg Pred Value: 0.8000
Prevalence: 0.6000
Detection Rate: 0.5000

Detection Prevalence : 0.5000 Balanced Accuracy : 0.9167

'Positive' Class : Mammal

[38]: eval_model(model_prediction, test)

Confusion matrix (absolute):

Actual

Prediction	Mammal	NotMammal	UNSEEN	Sum
Mammal	0	0	0	0
NotMammal	0	0	0	0
UNSEEN	6	4	0	10
Sum	6	4	0	10

Confusion matrix (relative):

Actual

Prediction	${\tt Mammal}$	${\tt NotMammal}$	UNSEEN	Sum
Mammal	0.0	0.0	0.0	0.0
${\tt NotMammal}$	0.0	0.0	0.0	0.0
UNSEEN	0.6	0.4	0.0	1.0
Sum	0.6	0.4	0.0	1.0

Accuracy:

0 (0/10)

Error rate:

1 (10/10)

Error rate reduction (vs. base rate):
-1.5 (p-value = 1)

0.0.9 Results

Unfortunately, there was an issue with the eval_model that resulted in an extra 'unseen' column that I am unsure how to fix. The trend was likely to be that the model had a 91% accuracy rate, meaning it is a relatively strong model. If more time had been spent debugging, I believe the model would have also reflected the 91% accuracy rate, resulting in a final model of:

Milk -> Mammal

No -> Not Mammal

0.0.10 References

https://mst.instructure.com/courses/44481/files/folder/examples?preview=1961310

https://www.rdocumentation.org/packages/OneR/versions/2.2

https://cran.r-project.org/web/packages/OneR/index.html

https://stackoverflow.com/questions/34705917/conda-how-to-install-r-packages-that-are-not-available-in-r-essentials

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