Special Task

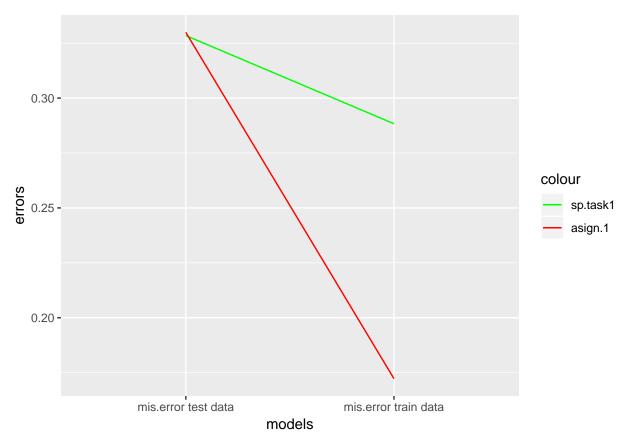
Andreas C Charitos-andch552 25 Nov 2018

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Special Task 1

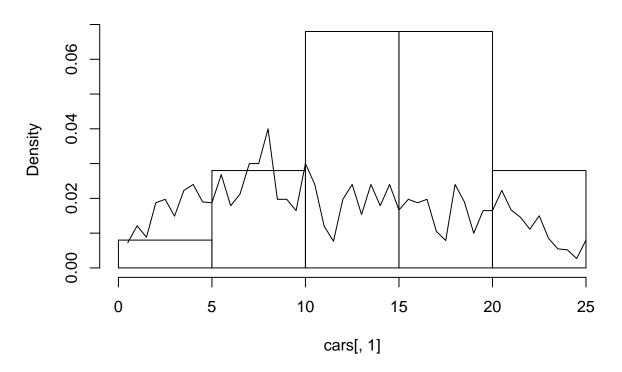
```
## sp.task1 asign.1 M
## 1 0.2883212 0.1722628 mis.error train data
## 2 0.3284672 0.3299270 mis.error test data
```



Comparing the errors for the two methods we can see that the training error from the assignment 1 is smaller compared with the one obtained from implementation of knn algorithm with the same number of neighbors. On the other side the test errors for the 2 methods are very close together.

Special Task 2

Histogram of speed



Implementing knn kernel density estimation with k=6 on the cars data set and comparing this with the a histogram of the speed column we can see that the density estimation produced looks very bumpy and doesn't provide a very good approximation on the density of the speed data.

Appendix

```
cl<-as.data.frame(cl)</pre>
  cosdisM<-cos distance(data,newdata)</pre>
  cosdisM<-as.data.frame(cosdisM)</pre>
  cosdisM$labels<-cl
  n<-ncol(cosdisM)-1</pre>
  bestM<-matrix(0,nrow=k)</pre>
  for (i in 1:n){
    ord_cosdisM<-cosdisM[order(cosdisM[,i]),]</pre>
    best_ks<-as.vector(ord_cosdisM[2:(k+1),"labels"])</pre>
    bestM<-cbind(bestM,best_ks)</pre>
  bestM<-as.data.frame(bestM)</pre>
  bestM[,1]<-NULL
  #colnames(bestM)<-rownames(newdata)</pre>
  #bestM
  \#pred_cl < -apply(bestM, 2, function(x) \{names(which(table(x) = max(table(x)))[1])\})
  pred_cl<-apply(bestM,2,function(x){ifelse(mean(x)>0.5,1,0)})
  pred_cl
}
spam<-readxl::read excel("spambase.xlsx")</pre>
#spam$Spam<-as.factor(spam$Spam)</pre>
#levels(spam$Spam)<-c("not spam", "spam")</pre>
spam<-as.data.frame(spam)</pre>
spam<-spam[sample(nrow(spam)),]</pre>
###split data
n=dim(spam)[1]
set.seed(12345)
id=sample(1:n, floor(n*0.5))
train=spam[id,]
test=spam[-id,]
train_X<-train[,!names(train)%in%c("Spam")]</pre>
train_labels<-train[,"Spam"]</pre>
test_X<-test[,!names(test)%in%c("Spam")]</pre>
test_labels<-test[,"Spam"]</pre>
#DD1<-cos_distance(train_X, train_X)</pre>
mis_error<-function(X,X1){</pre>
  n<-length(X)
  return(1-sum(diag(table(X,X1)))/n) #misclassification error function
}
```

```
K<-knearest(train_X, 30, train_X,train_labels)</pre>
mis1<-mis_error(as.vector(K),as.matrix(train_labels))</pre>
                                                            #0.2627737
K1<-knearest(train_X, 30, test_X,train_labels)</pre>
mis2<-mis_error(as.vector(K1),as.matrix(test_labels)) #0.3094891
DF<-data.frame(c(mis1,mis2),c(0.1722628,0.329927))
DF$M<-c("mis.error train data",
                 "mis.error test data")
colnames(DF)<-c("sp.task1", "asign.1", "M")</pre>
DF
library(ggplot2)
myplot<- ggplot(data = DF, aes(x =M,y=DF[,1],group=1))+</pre>
  geom_line(aes(y=DF[,1],color="asign.1"))+
  geom_line(aes(y=DF[,2],color="sp.task1"))+
  scale_color_manual(labels = c("sp.task1", "asign.1"), values = c("green", "red"))+
  labs(x="models",y="errors")
myplot
#eucl dist function given a matrix X returns
                          square matrix of distances between rows.
eucl dist<-function(X){</pre>
  n<-nrow(X)
  dist_mat_e<-matrix(0,nrow = n,ncol=n)</pre>
  for (i in 1:nrow(X)){
    for(j in 1:nrow(X)){
      dist_mat_e[i,j] < -sqrt(sum((X[i,]-X[j,])^2))
    }
  }
  dist_mat_e
#knn_density function given data, number of neghbors
                      returns kernel density estimation for each obs
knn_density<-function(data,k){</pre>
  distM<-eucl_dist(data) #as.matrix(dist(cars))</pre>
  distM<-as.data.frame(distM)</pre>
  N<-ncol(distM)
  bestM<-list()</pre>
  K<-for (i in 1:N ){</pre>
    ord_disM<-distM[order(distM[,i]),]</pre>
    best_ks<-ord_disM[(k+1),i]</pre>
    bestM[[i]]<-best_ks
    c<-unlist(bestM)</pre>
  }
  bestM<-as.data.frame(bestM)</pre>
  k/(nrow(data)*c)
```

```
dens<-knn_density(cars,6)
hist(cars[,1],freq = F,main = "Histogram of speed")
lines(seq(.5,25,0.5),dens)</pre>
```