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# DATA230 Decision Trees and CART (in-class demo code)
# Don't forget to install the following R packages first!
#install.packages("naniar")
#install.packages("rpart")
#install.packages("rpart.plot")
#install.packages("ggformula")
library(naniar) #naniar is a package to make it easier to summarise and handle missing values
data("iris")
#iris <- subset(iris,Species!="setosa") #remove "setosa" species</pre>
any na(iris) #Dtree algorithms cannot take N/A value, we need to check the N/A value first.
n <- nrow(iris)</pre>
set.seed(1117) #specify seeds
#new <- iris[sample(n),]</pre>
t_idx <- sample(seq_len(n), size = round(0.7 * n)) #random sample 70% data as training data,
the rest 30% is the test data
traindata <- iris[t idx,]</pre>
testdata <- iris[ - t_idx,]</pre>
library(rpart) #"rpart" package is used to implement CART, the split default to Gini, it can
be replace to Information
library(rpart.plot)
tree <- rpart(Species ~ ., data = traindata,</pre>
                method = "class") #change to anova for numerical
#pruning process, find the best stopping point.
printcp(tree) #find the best/optimal stopping point (CP, complexity paramater)
tree.pruned <- prune(tree,cp = tree$cptable[which.min(tree$cptable[,"xerror"]),"CP"])</pre>
rpart.plot(tree.pruned,digits=2)
future <- predict(tree.pruned, testdata, type="class")</pre>
future <- as.data.frame(future)</pre>
final <- cbind(future, testdata)</pre>
confusion <- table(final$Species,final$future, dnn = c("truth", "predicted"))</pre>
accuracy <- sum(diag(confusion)) / sum(confusion)</pre>
accuracy
#plot the iris dataset on "Sepal Length + Sepal Width" and "Petal Length + Petal Width" plane
library(ggformula)
scatterplot1=gf point(Sepal.Length ~ Sepal.Width, data = iris, color = ~ Species) %>%
  gf labs(title = "Figure 1: Scatterplot of Iris Data")
scatterplot2=gf_point(Petal.Length ~ Petal.Width, data = iris, color = ~ Species) %>%
  gf labs(title = "Figure 1: Scatterplot of Iris Data")
scatterplot1
scatterplot2
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