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#####IS 470 Data
Exploration------
# Part 1: Pokemon Data
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# This dataset contains information on 800 Pokemon from six generations of
Pokemon.
# VARIABLE DESCRIPTIONS:
#number: The entry number of the Pokemon
#name: The English name of the Pokemon
#type1: The Primary Type of the Pokemon
#type2: The Secondary Type of the Pokemon
#hp: The Base HP of the Pokemon
#attack: The Base Attack of the Pokemon
#defense: The Base Defense of the Pokemon
#sp.atk: The Base Special Attack of the Pokemon
#sp.def: The Base Special Defense of the Pokemon
#speed: The Base Speed of the Pokemon
#generation: The numbered generation which the Pokemon was first introduced
#legendary: Denotes if the Pokemon is legendary.
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### 1. Import and clean data
# Import data from csv
pokemon <- read.csv(file = "pokemon.csv", stringsAsFactors = FALSE)</pre>
#this is done to import the data in R.
#FALSE imlies that at this stage all the variables will be stored as vectors.
#However, later categorical variables should be stored as factors.
# str() shows the structure of data
str(pokemon)
# examine the number of rows and cols
nrow(pokemon) #shows total number of rows
ncol(pokemon)
# Show the head and tail rows of a data frame
head(pokemon) # head will show first 6 lines of data frame.
pokemon[1:6,] #on the left side, specify rows (in this case 1 through 6),
#on the right, after comma, we specify the number of columns.
head (pokemon, n=1) #n=1 will only give the first row of dataset.
tail(pokemon) #tail will give the last 6 rows
# summary() shows the mean and the five-number statistics indicating the
spread
#of each column's values
summary (pokemon) #will show summary statistics for all the variables. However,
#if categorical variables are stored as vectors, then it will show meaningless
#statistics of categorical variables.
# Remove unique identifiers (pokemon number and Name) from further analysis.
pokemon \langle -pokemon[,c(-1,-2)] \rangle #before transforming categorical variables into
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#numerical you first need to remove rows that do not provide useful info. In

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#this case pokemon name and number do not provide useful info.
# Change categorical variables to factors
str(pokemon)
pokemon$Type.1 <- factor(pokemon$Type.1) #we assign Type.1 factor into original
#variable (vector) Type.1. We transform it this way.
#then we transform the rest of the categorical variables.
pokemon$Type.2 <- factor(pokemon$Type.2)</pre>
pokemon$Generation <- factor(pokemon$Generation)</pre>
pokemon$Legendary <- factor(pokemon$Legendary)</pre>
str(pokemon)
summary (pokemon)
summary(pokemon$Type.1) #this will give the summary of frequencies of all
#categories in the Type.1 variable.
# set missing values in Type.2 as none
levels(pokemon$Type.2) #this will show all of the unique categories (levels)
#in Type.2 pokemon
levels(pokemon$Type.2)[1]#this will show the first level, in this case its
#missing.
levels(pokemon$Type.2)[1] <- "none"#it is good practice to assign none to the
#missing values.
str(pokemon)
summary(pokemon)
### 2. understanding a single variable: numeric variables
# Show summary of one or more columns
summary (pokemon$Attack) #this summary function, applied on numerical variable
#"Attack", shows summary statistics of this variable.
summary(pokemon[,c("Attack", "Defense")]) #summary of 2 numberic variables. We
#need to use combine functio since there are more than 1 variables.
Alternative
#would be writing c(4,5)
summary(pokemon$Sp..Atk + pokemon$Sp..Def) #Summary info of a NEW variable
which
#is a sum of special attack and special defence.
which (pokemon\$Sp..Atk + pokemon\$Sp..Def == 340) #will show pokemon that has a
#of special attack and defence = to 340
# obtain the mean, median, and range of a numeric variable
mean(pokemon$Attack)
median(pokemon$Attack)
range(pokemon$Attack)
# use quantile to calculate the five-number summary for Attack
quantile(pokemon$Attack)
# IOR
IQR(pokemon$Attack) #difference between 1st and 3d quartile
# boxplot of numeric variables
boxplot(pokemon$Attack, main="Boxplot of Attack in the pokemon data set",
ylab="Attack")
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boxplot(pokemon$Defense, main="Boxplot of Defense in the pokemon data set",
ylab="Defense")
boxplot(pokemon[which(pokemon$Generation==1),4], main="Boxplot of Attack of
the 1st generation pokemon", ylab="Attack")
boxplot(pokemon[which(pokemon$Generation==1),5], main="Boxplot of Defense of
the 1st generation pokemon", ylab="Defense")
# histograms of a numeric variable
hist(pokemon$Attack, main = "Histogram of Attack in the pokemon data set",
xlab = "Attack")
hist(pokemon$Defense, main = "Histogram of Defense in the pokemon data set",
xlab = "Defense")
hist(pokemon$HP, main = "Histogram of HP in the pokemon data set", xlab =
"HP")
# variance and standard deviation of a numeric varaible
var(pokemon$Attack)
sd(pokemon$Attack)
### 3. Exploring categorical variables
# Summary of categorical variable
summary(pokemon$Type.1) #gives count
nlevels(pokemon$Type.1) #will tell how many unique categories
# Plot categorical variable
plot(pokemon$Type.1, main = "Plot of Type.1 in the pokemon data set", xlab =
"Type.1")
table(pokemon$Type.1) #will show table with frequency
sort(table(pokemon$Type.1))
# Run prop.table
Type table = table(pokemon$Type.1) #THis creates a table for Type 1, you have
#to create it before you can find proportions
prop.table(Type table) #this will show proportions
### 4. Understand relationships of multiple variables
# scatter plot: two numeric variables
plot(pokemon$Attack, pokemon$Defense) #scatter plot of 2 numberic variables
# Generate correlation coefficients of two numeric variables in a 2x2 matrix
\# cor(X,Y) lies between -1 and 1. zero means no correlation. 1 or -1 indicates
full correlation
# positive value means positive correlation and negative values mean negative
relationships
cor(pokemon[,c("Attack", "Defense")])
cor(pokemon[,c(4,5)])
# Generate the correlation matrix of all numeric variables
cor(pokemon[,3:8])
# Generate 2D scatter plots
pairs(pokemon[,3:8]) #scatterplots for columns 3 to 8
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## Examine relationships between numeric variables and factors
# boxplot groups values of a numeric variable based on the values of a factor
boxplot(Attack~Type.1, data = pokemon)
boxplot(Attack~Type.1, data = pokemon[which(pokemon$Legendary=='True'),])
boxplot(Attack~Type.1, data = pokemon[which(pokemon$Legendary=='False'),])
boxplot(HP~Legendary, data = pokemon)
boxplot(HP~Type.1, data = pokemon)
# Part 2: CarAuction Data
###
# This dataset contains information of cars purchased at the Auction.
# VARIABLE DESCRIPTIONS:
#Auction: Auction provider at which the vehicle was purchased
#Color: Vehicle Color
#IsBadBuy: Identifies if the kicked vehicle was an avoidable purchase
#MMRCurrentAuctionAveragePrice: Acquisition price for this vehicle in average
condition as of current day
#Size: The size category of the vehicle (Compact, SUV, etc.)
#TopThreeAmericanName: Identifies if the manufacturer is one of the top three
American manufacturers
#VehBCost: Acquisition cost paid for the vehicle at time of purchase
#VehicleAge: The Years elapsed since the manufacturer's year
#VehOdo: The vehicles odometer reading
#WarrantyCost: Warranty price (term=36month and millage=36K)
#WheelType: The vehicle wheel type description (Alloy, Covers)
### 1. Import and clean data
# Import data from csv
carAuction <- read.csv(file = "carAuction.csv", stringsAsFactors = FALSE)</pre>
### I spend 3 hrs on the first part. I wrote a lot of comments. But every now
#and then this software DISCONNECTS. first 5-6 times my comments were
#still there. But the last time it happened they were ALL gone.
# str() shows the structure of data
str(carAuction)
# summary() shows the mean and the five-number statistics indicating the
spread of each column's values
summary(carAuction)
# Change all categorical variables to factors
#first, check which variables are categorical
carAuction$Auction = factor(carAuction$Auction)
carAuction$Color = factor(carAuction$Color)
carAuction$IsBadBuy = factor(carAuction$IsBadBuy)
carAuction$Size = factor(carAuction$Size)
carAuction$TopThreeAmericanName = factor(carAuction$TopThreeAmericanName)
carAuction$WheelType = factor(carAuction$WheelType)
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str(carAuction)

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summary(carAuction)
### 2. understanding a single variable: numerical variables
# Show summary of VehOdo
summary(carAuction$VehOdo)
# obtain the mean, median, and range of WarrantyCost
mean(carAuction$WarrantyCost)
median(carAuction$WarrantyCost)
range(carAuction$WarrantyCost)
# use quantile to calculate the five-number summary for WarrantyCost
quantile(carAuction$WarrantyCost)
# display the IQR of WarrantyCost
IQR(carAuction$WarrantyCost)
# boxplot of numeric variables: VehBCost and VehicleAge
boxplot(carAuction$VehBCost, main="Boxplot of vehicle cost", ylab="Cost")
boxplot(carAuction$VehicleAge, main="Boxplot of vehicle age", ylab="Age")
boxplot(VehBCost~VehicleAge, data = carAuction)
# histograms of VehOdo
hist(carAuction$VehOdo, main = "Histogram of VehOdo", xlab = "VehOdo")
### 3. Exploring categorical variables
# Show the number of cars in different WheelType
nlevels(carAuction$WheelType)
# Show the proportion of cars in different WheelType
Type table = table(carAuction$WheelType)
prop.table(Type table)
### 4. Understand relationships of multiple variables
# scatter plot: VehBCost and MMRCurrentAuctionAveragePrice
plot(carAuction$VehBCost, carAuction$MMRCurrentAuctionAveragePrice)
# Generate correlation coefficients of VehBCost and
MMRCurrentAuctionAveragePrice
cor(carAuction[,c("VehBCost", "MMRCurrentAuctionAveragePrice")])
# Generate the correlation matrix of all numeric variables (you can specify
the numeric variable names or their column numbers in a c() function)
cor(carAuction[,c('MMRCurrentAuctionAveragePrice','VehBCost','VehicleAge','VehOdo',
'WarrantyCost')])
## Examine relationships between numeric variables and factors
# boxplot VehBCost based on IsBadBuy
boxplot(VehBCost~IsBadBuy, data = carAuction)
# Question: list one thing you learned from the carAuction data exploration.
#I learned to save my work in R as I go. All of my comments for the first
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#section were gone after software disconencted me. I will need to doing again to print it out and study.

#aslo: I learned that VehBCOst and MMRAverageprice are positively
#corelated.
#also: I learned to write function to get summary statistics.
# Spent about 2hours on first section, since i was pausing and writing
comments,
#but they are all gone :(