

UNIVERSITY OF SOUTH FLORIDA

DEPARTMENT OF INFORMATION AND DECISION SCIENCES

Statistical Data Mining
Final Project

Real Time Webscraping through Python

Exploratory Data Analysis



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1. Introduction:

This project aims to understand and implement data retrieval from a website through Python. The process of automatically extracting meaningful data from a website, in real time, is called webscraping.

Data displayed by most websites can only be viewed using a web browser. Examples are data listings at yellow pages directories, real estate sites, social networks, industrial inventory, online shopping sites, contact databases etc. Most websites do not offer the functionality to save a copy of the data which they display to your computer. The only option then is to manually copy and paste the data displayed by the website in your browser to a local file in your computer - a very tedious job which can take many hours or sometimes days to complete.

Web Scraping is the technique of automating this process, so that instead of manually copying the data from websites, the Web Scraping software will perform the same task within a fraction of the time.

A Web Scraping software will interact with websites in the same way as your web browser. But instead of displaying the data served by the website on screen, the Web Scraping software saves the required data from the web page to a local file or database. There are multiple tools and programming languages that can achieve this. For example, WebHarvey, Python, Scraper, Outwit Hub, etc.

In this particular project, I have studied and implemented a Python based Webscraper. By studying basic python programming language through “Team lead- Introduction to Python Webscraping”, I have implemented a system in Python which can extract particular data from a Boat e-commerce website. This system extracts basic information of a boat listed on the website for sell.

Moreover, I have also done exploratory data analysis and predictive modeling on the data extracted through webscraping. For this, I have used R.

2. Python Webscrapping:

- There are two basic tasks that are used to scrape web sites:
 1. Load a web page to a string.
 2. Parse HTML from a web page to locate the interesting bits.

Python offers two excellent tools for the above tasks. I have used [BeautifulSoup](#) to do the parsing.

OBJECTIVE:

- To extract data for each boat listing on the website www.boattrader.com . Details to be extracted include Boat : Price, Class, Length, Seller Contact, etc. Furthermore, do some data visualization and statistical modeling using the data extracted.

Proposed approach:

- Using Python IDE – Pycharm 4.5 (by JetBrains), I have written a code which fetches the following details from each boat listing:

Category, Make, Hull Material , Propulsion Type , ZipCode, Length, Fuel Type, Year, Price, Class, Seller_Contact, Location

NOTE: Due to the massive amount of listings on the website, I have extracted only a sample of the entire data on the website. I have filtered for boat listings within 1000 mile radius of Tampa and extracted data for around 19064 boat listings.

- The first thing to do when writing a scraping script is to manually inspect the page(s) to scrape to determine how the data can be located.

To begin with, we are going to inspect the HTML source of the boat listing page.

- Now, after applying required filters, I got 681 pages of search results with each page containing 28 listings (19068 listings total). The html link looks as follows:

<http://www.boattrader.com/search-results/NewOrUsed-any/Type-any/Category-all/Zip-33647/Radius-200/Sort-Updated:DESC/Page-1,28?>

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- Typical Boat listing looks as follows with the boat details:

The screenshot shows a web browser displaying a boat listing on the Boat Trader website. The listing is for a 1995 Lake & Bay 20 Boca Grande boat, priced at \$23,999, located in Port Charlotte, FL. The boat details table lists the following information:

Boat Details	
Class	Power
Category	Flats Boats
Year	1995
Make	Lake & Bay
Length	20'
Propulsion Type	Other
Hull Material	Composite
Fuel Type	Gas
Location	Port Charlotte, FL
Measurements	
Propulsion	
Other Details	

Below the table, there are links for "Check This Boat's History" and "Check on Boat Loans". The boat is shown in a large image on the right, and a smaller image of the boat is shown below it. The website also features a "Request Information" form on the left side.

- Inspecting the HTML, we can typically see below structure:

```

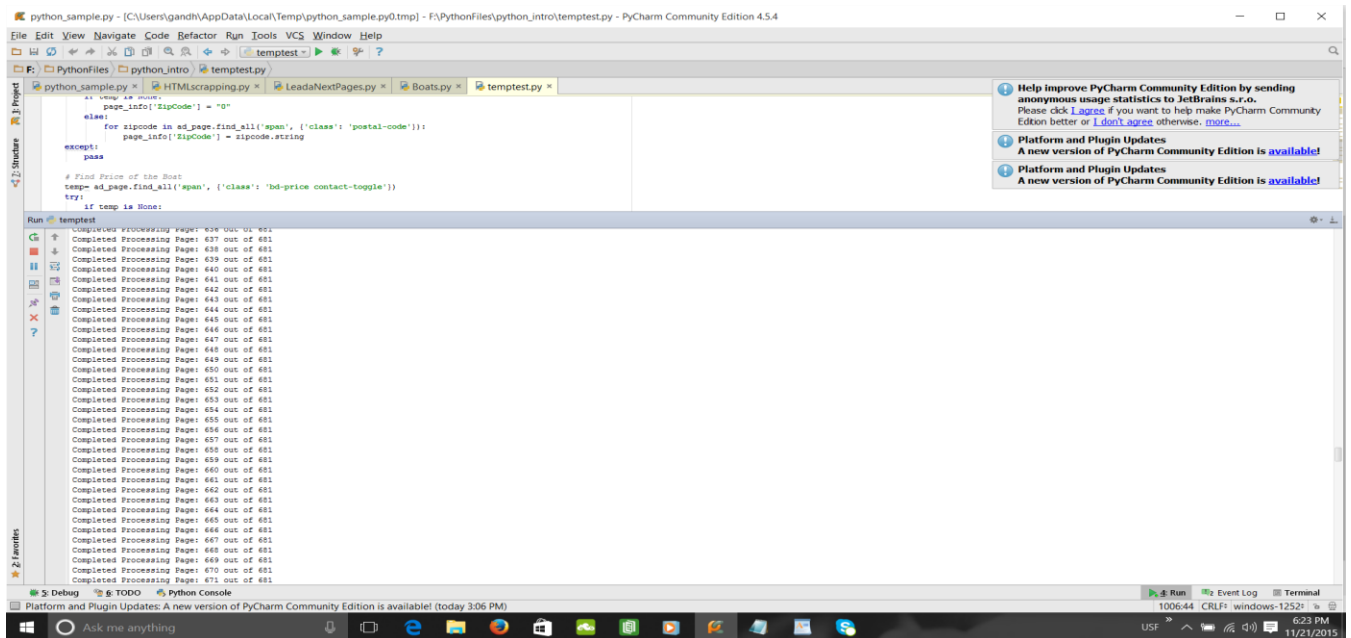
<div id="oas_Top" class="ad w728"></div>
<div class="main-content">
  <section class="boat-details">
    <div id="large-carousel" class="large-carousel"></div>
    <header class="bd-header"></header>
    <script id="listing-map" type="text/javascript">var lat = "27.0038"; var lon = "-82.2122"></script>
    <div id="directions-control"></div>
    <div class="col right"></div>
    <div class="col left">
      <section class="boat-details">
        <section id="collapsible-content-areas" class="detail-txt">
          <div class="detail-tbl">
            <div id="boat-details" class="collapse-content open">
              <div class="header"></div>
              <div class="collapsible open">
                <table>
                  <tbody>
                    <tr>
                      <th>Class</th>
                      <td>Power</td>
                    </tr>
                    <tr></tr>
                    <tr></tr>
                    <tr></tr>
                    <tr></tr>
                    <tr></tr>
                  </tbody>
                </table>
              </div>
            </div>
          </div>
        </section>
      </section>
    </div>
  </div>

```

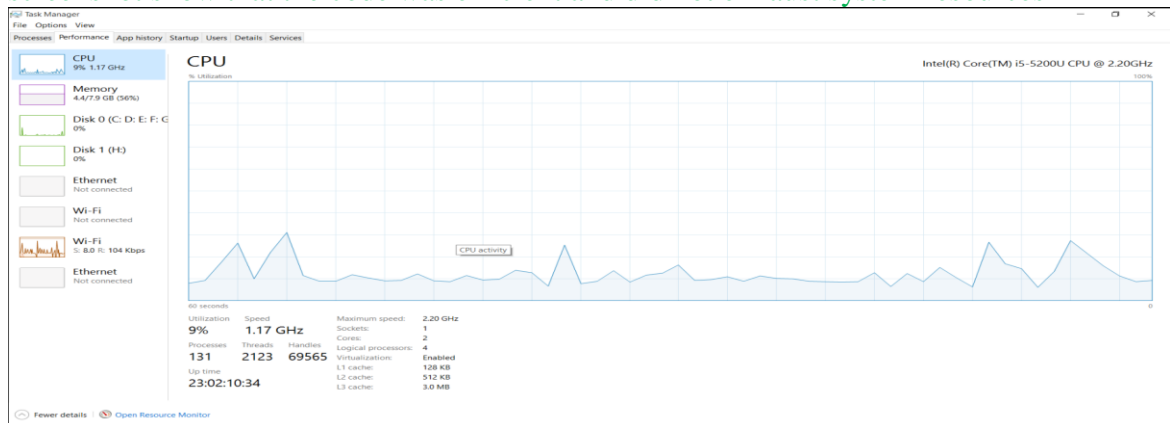
3. Webscraping Analysis and Performance Analysis:

- I used the following algorithm: (Code attached in Appendix A)

- 1) Connect to all searches link to fetch all records details
- 2) Click on Page 1 link out of n pages.
- 3) Boat listings - find all href of listings
- 4) Loop inside each listing
- 5) Find Boat Details like Class, Length, Price, etc
- 6) Go to next listing in the page and repeat step 5
- 7) At end of page listing, go to next page link
- 8) Do steps 3 to 7 for all pages
- 9) Export dataset



- Below is the performance of the system while the web-scraping was in progress. These screenshot show that the code was efficient and did not exhaust system resources



4. Data Cleaning and Data Visualization:

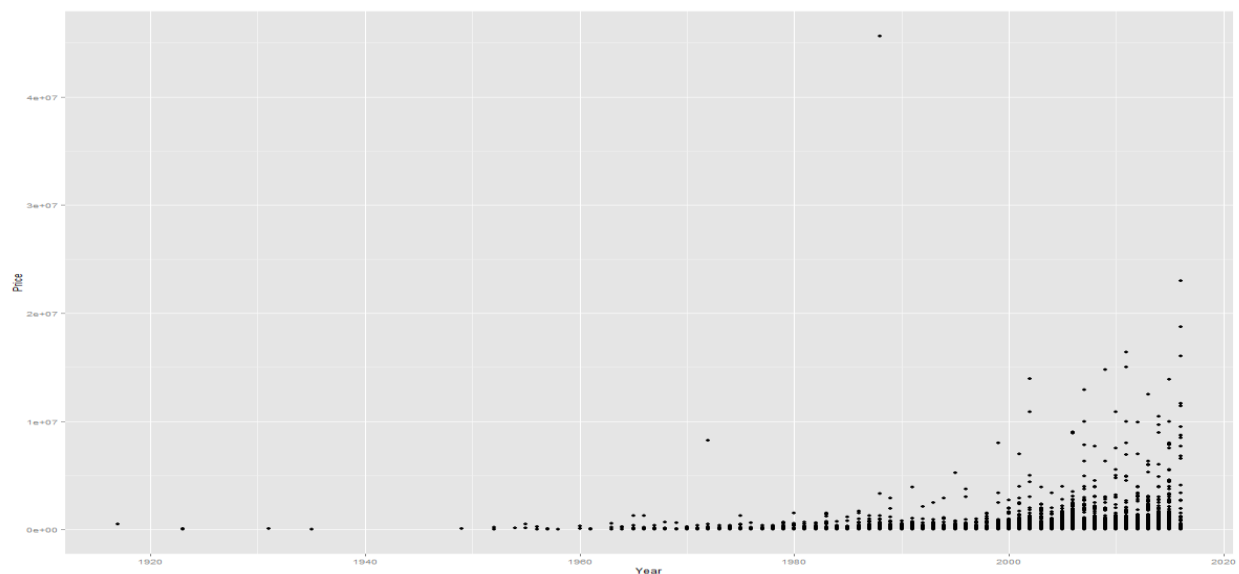
Data Cleaning:

- After the Python data extraction of 681 pages was complete, I was able to extract records of **19068** boat listings.
- However, this was raw data and required cleaning. For example, some of the listing did not mention zipcode, Seller Contact, Price. So I had to remove all such entries.
- Following stepwise data cleaning is implemented:
 - 1) Remove all entries with ZipCode = 0 (indicating blank zipcode value)
 - 2) Remove all entries with Price = Not mentioned (indicating invalid or empty price value)
 - 3) Remove all entries with Seller_contact= Blank
 - 4) Remove all entries with blank Fuel Types
 - 5) Remove all duplicates
- After doing the data cleaning, I got **8207** records. This data was then used for exploratory data analysis and statistical data modeling.
- Data cleaning is performed in R. Code is attached in Appendix B

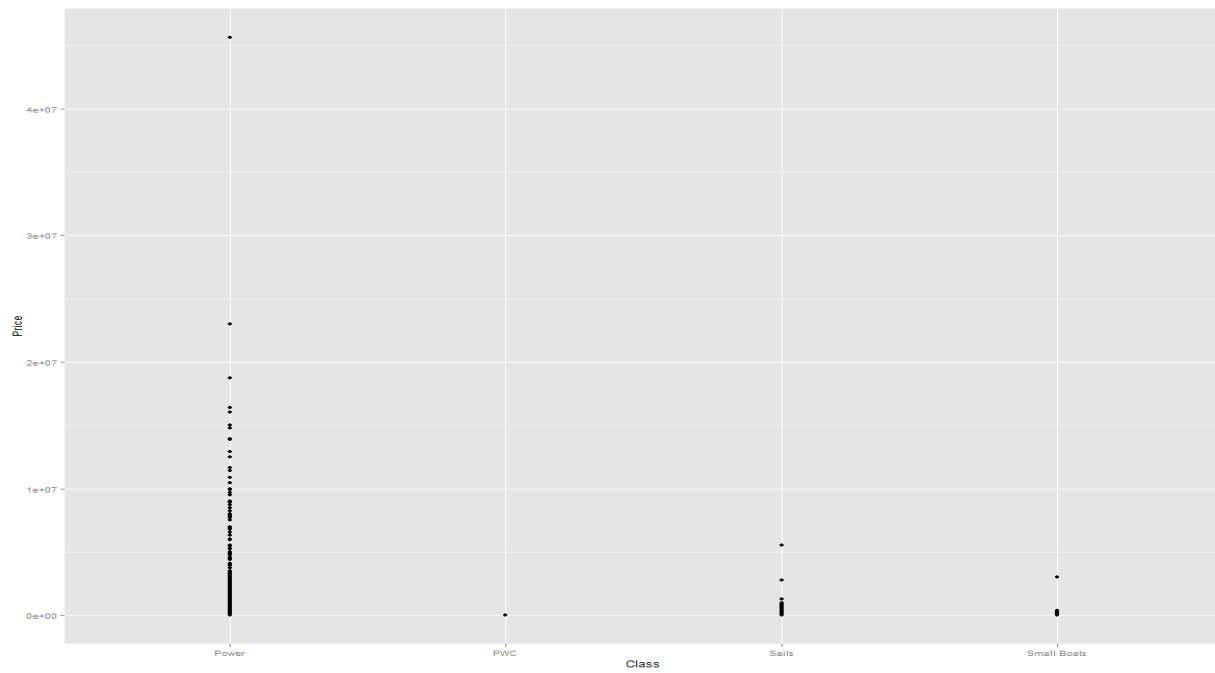
Data Visualization:

- The cleaned data is visualized in R using the **ggplot2** visualization package in R
- Below are the important visualizations done:

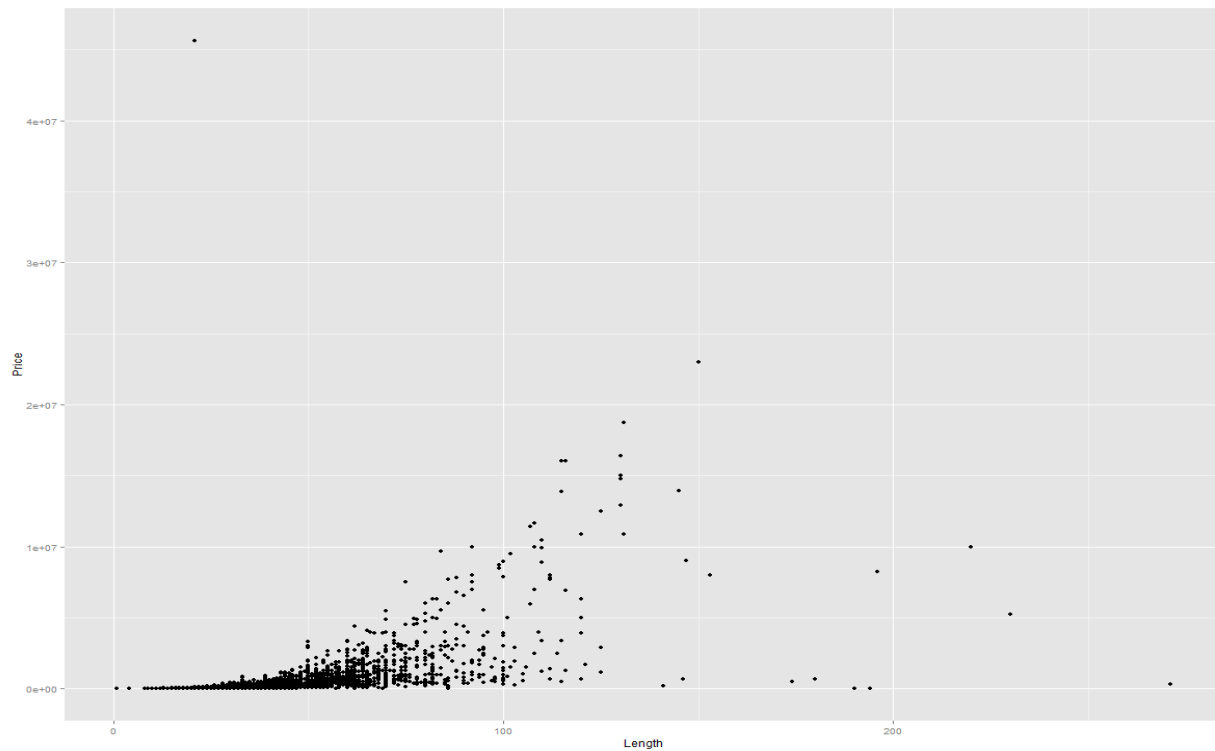
Price against Year:



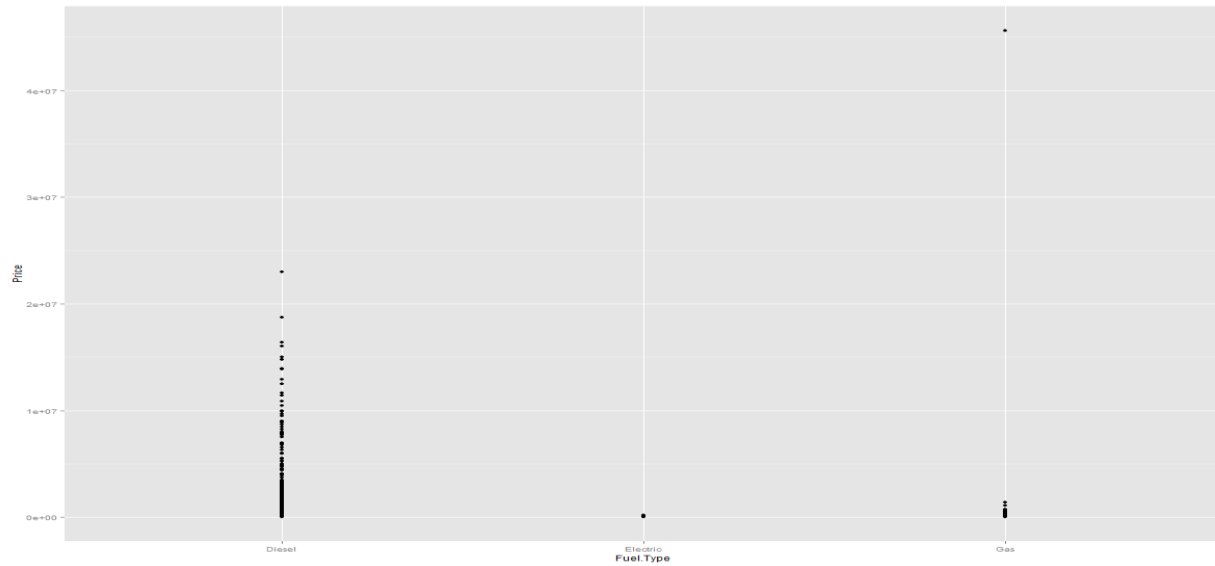
Price against Boat Class:



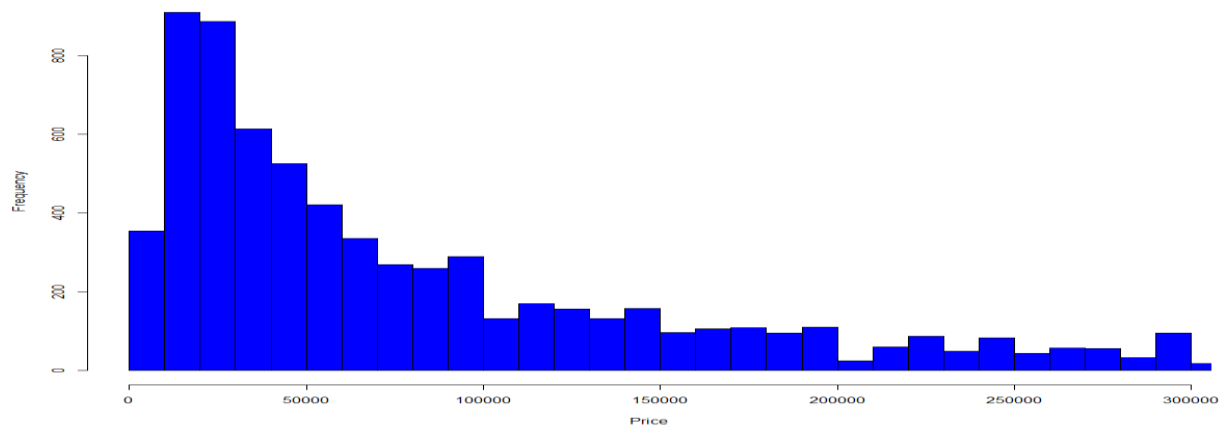
Price against Boat Length:



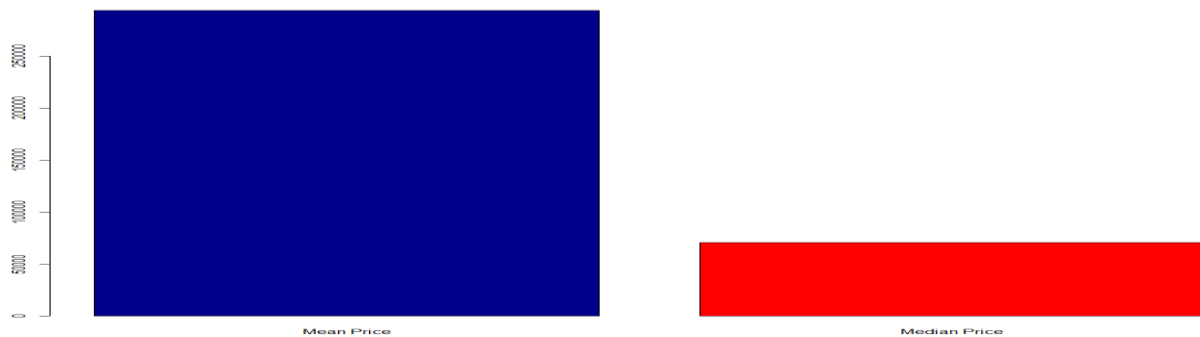
Price against Boat Fuel Type:



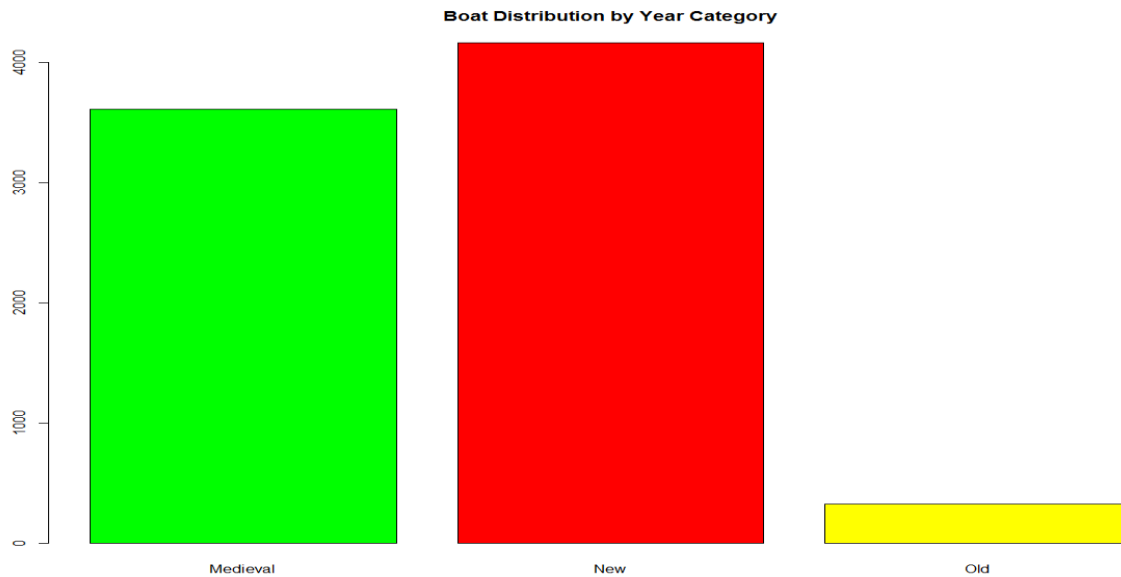
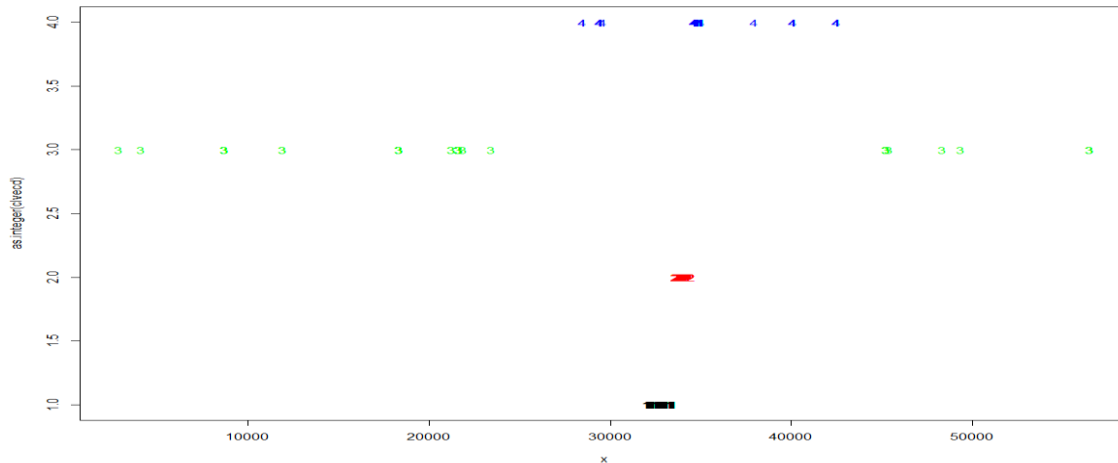
Histogram of Price



Mean Vs Median Price Distribution



Zipcode Clustering in 4 clusters:



Conclusion from Data Visualization:

- By looking at the above visualizations, we can conclude following about the data:
- 1) Price of boat increases as the Manufacture Year increases i.e Newer the boat , higher the price.
 - 2) Boats with Class as “Power” are high in Price with class=”PWC” being the least paid
 - 3) We see that boats with Length between 100-150 m have highest prices and then the prices observe decreasing trend for lengths around 200 m.
 - 4) Boats with Fuel Type= “Diesel” look to have highest price, with Electric Boats being least paid.
 - 5) The Price data is right skewed.

5. Statistical Modeling – Linear Regression:

- I have performed Linear Regression Predictive Model with:

- Price as the dependent variable and Independent variables as Boat Length, manufacture Year, and Fuel type.
- Price, Length and Year are Log transformed. As seen above, Price is right skewed. Log transformation resulted in explaining more variation in the data with adj. R2 = 86 %
- Fuel Type is a categorical variable with 3 categories: Diesel, Electric and Gas. So 2 Dummy variables are used with Diesel as baseline.

- Below is the regression summary:

```
Call:
lm(formula = log(Price) ~ log(Length) + log(Year) + as.factor(Fuel.Type))

Residuals:
    Min       1Q   Median       3Q      Max
-6.6302 -0.2930 -0.0200  0.2854  8.4269

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)   -776.30553    8.94710  -86.766 < 2e-16 ***
log(Length)     2.80901    0.02047  137.257 < 2e-16 ***
log(Year)     102.36550    1.17645   87.012 < 2e-16 ***
as.factor(Fuel.Type)Electric  -0.47631    0.16160   -2.947  0.00321 **
as.factor(Fuel.Type)Gas      -0.49117    0.01793  -27.391 < 2e-16 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.5324 on 8096 degrees of freedom
Multiple R-squared:  0.8585,    Adjusted R-squared:  0.8584
F-statistic: 1.228e+04 on 4 and 8096 DF, p-value: < 2.2e-16
```

Story around the data: Interpretation and Conclusion:

Observing the regression summary we can say that:

- 1) Price has a direct proportionality with Length and Year variables
- 2) For every percent increase in Year and Length, Price of boat increases by 102.36% and 2.80 % respectively.
- 3) Price for Electric and Gas boats drop as compared to Diesel boats. Also, Gas boats are expensive than Electric boats.
- 4) Length, Year and Fuel Type are all statistically significant variables.

Looking at the above interpretations, we see that they confer with the interpretation we had done through data visualizations.

6. Challenges and Learnings:

- Working on implementation of Boattrader.com webscraping engine, I had to learn Python from scratch. With no prior knowledge of Python, learning it was a challenging and interesting experience.
- With the help of Team Leada's "Introduction to Webscraping In Python", I was able to learn basic data structures in Python, different looping constructs and also HTML parsing in Python.
- Not only was learning python interesting, but learning to analyze a website in real time and inspecting different html tags for data extraction was a very useful boost in my programming knowledge.
- With this project, I have gained proficiency in using Python, inspecting key tags of a website for data extraction, creating excel reports in python, data visualization in R, and statistical modeling in R.
- This system implementation was a tedious. With around 19000 records to analyze, different aspects of the website had to be analyzed. The website had to be analyzed for consistency and it was important to handle exceptions.
- It took around 6 hours to fetch details of all the 19068 records.
- Important thing in this system is that it is generalized to a great extent. The user only has to enter the html link of the 1st page to be analyzed. The system then:
 - Automatically fetches total number of pages to be analyzed
 - Automatically fetches number of listings per page
 - Automatically fetches and exports to excel file, the required data of each boat listing.

7. APPENDIX:

A. PYTHON CODE:

```
##### Name : Gandhar Arvind Pathak
##### Statistical Data Mining (ISM 6137) - University of South Florida, Tampa
##### UID: U66784122
##### Final Project: Real Time Webscraping through Python and Applied Exploratory Data Analysis
##### Date: November 22, 2015

##### Objective: This code is written to fetch basic details of a Boat, from an advertisement website " http://www.boatrader.com "
##### The search limits to 1000 miles radius from the user location: precisely zipcode: 33613 (Tampa). The collected data is to be subjected to
##### exploratory data analysis for different data visualizations and data mining. This is achieved through webscraping in Python.

__author__ = 'gandh'
# Initial imports
import requests
import re # For String manipulation like removing special characters from string
import csv # To write output to CSV file

#make sure BeautifulSoup 4 is installed
from bs4 import BeautifulSoup, SoupStrainer
import bs4 # import main module first
print "Beautiful Soup Version " + bs4.__version__

# Function definations

# Opens each listing by the URL and creates a structure of the internal HTML elements
def Open_listing(each_listing):
    hml = each_listing # you'll need to define this.
    r = requests.get(hml)
    hml = r.text
    # we start with getting the soup for each page.
    list_struct = BeautifulSoup(hml, "html.parser")
    return list_struct
#Function End

# For each listing, fetches information like Price, Length, Location, Class, Seller Contact, etc
def fetch_AdInfo(ad_page):
    page_info = {} # temporary dict to hold page information-- returned as 'details' dictionary
    try:
        # Find Boat details
        for detail in ad_page.find_all('div', {'class': 'collapsible open'}):
            tables = detail.findChildren('table')
            Boat_Details_table = tables[0]
            rows = Boat_Details_table.findChildren('tr')
            for cell in rows:
                th_val = cell.findChildren('th')
                td_val = cell.findChildren('td')
                page_info[th_val[0].string] = td_val[0].string
    except:
        pass

    # find Sellers contact number
    temp = ad_page.find('div', {'class': 'contact'})
    try:
        if temp is None:
            page_info['Seller_Contact'] = "0"
        else:
            try:
                for contact in ad_page.find_all('div', {'class': 'contact'}):
                    page_info['Seller_Contact'] = re.sub("\W+", "", contact.string) # remove special characters from contact number and store in the page_info list
            except:
                pass
    except:
        pass

    # Find ZIPCODE of the Boat Location
    temp = ad_page.find('span', {'class': 'postal-code'})
    try:
        if temp is None:
            page_info['ZipCode'] = "0"
        else:
            for zipcode in ad_page.find_all('span', {'class': 'postal-code'}):
                page_info['ZipCode'] = zipcode.string
    except:
        pass

    # Find Price of the Boat
    temp = ad_page.find_all('span', {'class': 'bd-price contact-toggle'})
    try:
        if temp is None:
            page_info['Price'] = "0"
        else:
            for price in ad_page.find_all('span', {'class': 'bd-price contact-toggle'}):
                page_info['Price'] = re.sub("\W+", "", price.string) # removes '$', and spaces
    except:
        pass

    return page_info
```

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```
#Function End

# Write Ad information to excel CSV
# Make sure CSV file is created at given path

def write_to_excel(details, flag):
    inputFileName = "F:\PythonFiles\FinalProject\TestCsv.csv"
    len_of_details = details.__len__()
    for key, value in details.iteritems():
        temp = {}
        temp = value[0]
        with open(inputFileName, "ab") as f:
            w = csv.writer(f)
            if flag == 0:
                wh = csv.DictWriter(f, temp.keys())
                wh.writeheader()
                w.writerow(temp.values())
                flag = 1
            else:
                w.writerow(temp.values())
    return flag

#Function end

# Finds total number of pages and listings per page returned by the search results
def find_search_results_details(raw_html):
    r = requests.get(raw_html)
    raw_html = r.text
    lastpage_struct = BeautifulSoup(raw_html, "html.parser")
    for lastpage in lastpage_struct.find_all('a', {'class': 'last'}, href=True):
        href = lastpage['href'].encode('utf-8')
        href = href.split(",")
        listings_per_page = re.sub('\W+', "", href[1]) # removes special characters
        total_search_pages = href[0][-3:] # Extracts the last 3 characters of the string which is the total pages in the search results
        return listings_per_page, total_search_pages
#Function End

def main():
    raw_html = "http://www.boatstrader.com/search-results/NewOrUsed-any/Type-any/Category-all/Zip-33647/Radius-200/Sort-Updated:DESC/Page-1,28?" # master link to fetch data from
    page_number = 1 # maintains count of page number of the search results
    count_per_page, total_pages = find_search_results_details(raw_html)
    flag = 0 # to maintain record of headers
    while page_number <= total_pages:
        linklist = [] #Holds the individual advertisement links on each page
        details = {} # Dictionary which holds the required details of each advertisement, fetched from each page out of 'n' different advertisement pages
        raw_html = "http://www.boatstrader.com/search-results/NewOrUsed-any/Type-any/Category-all/Zip-33647/Radius-200/Sort-Updated:DESC/Page-%s,%s"%(page_number,count_per_page)# you'll
        need to define this.
        r = requests.get(raw_html)
        raw_html = r.text
        # we start with getting the soup for each page.
        bs_struct = BeautifulSoup(raw_html, "html.parser")
        # we then look for all the <li>
        for listing in bs_struct.find_all('section', {'class': 'boat-listings'}):
            for ol_tag in listing.find_all('ol', {'class': 'boat-list'}):
                for listing_link in ol_tag.find_all('li'):
                    links = listing_link.find_all('a', href=True)
                    if len(links) != 0: # make sure it found something.
                        link = links[0]
                        url = link['href'].encode('utf-8')
                        linklist.append(url) # store the URL of each listing on a particular page
        page_number += 1

    # Enter Each listing and find information

    for each_listing in linklist:
        each_listing = "http:" + each_listing
        ad_page = Open_listing(each_listing) #opens connection to each listing
        details[each_listing] = [fetch_AdInfo(ad_page)] #fetches details of the connected listing: Details like Price, Length, Location, Class, Seller Contact, etc

    flag = write_to_excel(details, flag) #write fetched information to a CSV file
    print "Completed Processing Page: %s out of %s " % (page_number - 1, total_pages)

main()
```

B. RCODE:

```
Boats <- read.csv("F:\PythonFiles\FinalProject\TestCsv.csv",stringsAsFactors = F)
attach(Boats)
Boats= Boats[,1:12]

#Data Cleaning
Zip_clean=Boats[-which(ZipCode==0 | ZipCode==" " | nchar(ZipCode)>6 | tolower(ZipCode)=="other"),]
Fuel_type=Zip_clean[-which(Zip_clean$Fuel.Type==" " | tolower(Zip_clean$Fuel.Type)=="other"),]
Price_clean=Fuel_type[-which(tolower(trimws(Fuel_type$Price))== "requestprice"),]
```

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```
Contact_clean=Price_clean[-which(nchar(Price_clean$Seller_Contact)>11),]
Contact_clean$Length=gsub("[[:punct:]]", "", Contact_clean$Length)
Final_clean=unique.data.frame(Contact_clean)
attach(Final_clean)

#Categorize Year as Old, Medieval,New
Final_clean["Year_Category"] <- NA
Final_clean$Year_Category[Final_clean$Year>=2005]= "New"
Final_clean$Year_Category[Final_clean$Year>=1980 & Final_clean$Year<2005]= "Medieval"
Final_clean$Year_Category[Final_clean$Year<1980]= "Old"
Final_clean$Price=as.numeric(Final_clean$Price)
Final_clean$Length=as.numeric(Final_clean$Length)
Final_clean$Year=as.numeric(Final_clean$Year)
attach(Final_clean)

#Write the cleaned dataset to an Excel File
library(xlsx)
write.xlsx(Final_clean, "F:/PythonFiles/FinalProject/FinalData_Cleaned.xlsx")

# Common statistics # Data Visualization
hist(Price,breaks=6000,col = "blue",xlim = c(0,mean(Price)))
meanPrice=mean(Price)
medianPrice=median(Price)
barplot(c(meanPrice,medianPrice), main="Mean Vs Median Price Distribution",names.arg=c("Mean Price","Median Price"),col=c("darkblue","red"))
barplot(table(Year_Category),main="Boat Distribution by Year Category",col = c("green","red","yellow"))
library(ggplot2)
# Price Variation along months
qplot(Year_Category,Price,data = Final_clean)
qplot(Year,Price,data = Final_clean)
qplot(Class,Price,data = Final_clean)
qplot(Fuel.Type,Price,data = Final_clean)
qplot(Length,Price,data = Final_clean)
qplot(Hull.Material,Price,data = Final_clean)

#Regression Model

Scraplm=lm(log(Price)~log(Length)+log(Year)+as.factor(Fuel.Type))
summary(Scraplm);
library(mgcv)
mod=gam(Price ~ s(Year) + s(Length))
summary(mod)

#clustering
names(Final_clean)
mod2 <- kmeans(dist(Final_clean[,5]),4) #For Zipcode
clus <- as.numeric(mod2$cluster)
by(Final_clean[,5],as.factor(clus),summary)

library(fpc)
plotcluster(Final_clean$ZipCode,mod2$cluster)
```

8. Literature Survey / References:

- 1) Team leada course “ Introduction to Webscraping in Python:
- 2) Webscraping 101 - <http://www.gregreda.com/2013/03/03/web-scraping-101-with-python/>
- 3) Easy Webscraping with Python - <http://blog.miguelgrinberg.com/post/easy-web-scraping-with-python>