Assignment 1 - Documentation INFO 7390 Advanced Data Science and Architecture



Working with Edgar datasets: Wrangling, Pre-processing and Exploratory Data Analysis

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OVERVIEW:

Programming Language and Libraries

We have used Python programming language and following libraries to solve this assignment.

Libraries used:

- urllib allows access to websites via program
- **BeautifulSoup(bs4)** web scraping
- **OS** operating system dependant functionality
- sys access to variables used by the interpreter
- **zipfile** used for zipping files
- csv write to .csv format
- **logging** track events
- **boto** python package that provides interface to Amazon Web Services
- **boto.s3** creates connection to interact with the server
- **shutil** to remove files and directory
- Glob to read files efficiently
- Pandas put data from csv files into data frames
- **itertools** creates iterators for efficient looping
- **argparse** to get command line inputs and arguments
- **numpy** for computing on data
- **statistics** to obtain functions like mode
- matplotlib—2D plotting library for visualization

Tools used:Jupyter Notebook, Amazon S3 bucket, Docker, Github, Sublime Command line.

Logging:

We have used the logging module in python which will log the operations with the timestamp, levelname, file name, line number and message that we have customized.

There are 5 levels in logging module:

Debug: Detailed information, typically of the interest only when diagnosing problems

Info: Confirmation that things are working as expected

Warning: An indication that something unexpected happenedor indicative of some problem in the near future

Error: Due to more problem, the software has not been able to perform some function

Critical: A serious error, indicating that the program itself may be unable to continue running

We have used 3 levels- Info, error and warning in our logs.

Below two handlers are used to log the output.

File handler – This willlog all the level messagesinto the log file.

Stream handler – This will log the level messages into console.

Logging code snippet:

```
#create logger
logger = logging.getLogger()
logger.setLevel(logging.DEBUG) #setting level to debug
#creates file handler which logs all the level messages
fh = logging.FileHandler('Log_file2.log' , mode = 'w')
fh.setLevel(logging.DEBUG) #setting level to handler to debug
formatter = logging.Formatter('[%(asctime)s] %(levelname)8s --- %(message)s' +
                              (%(filename)s:%(lineno)s)' ,datefmt = '%Y-%m-%d %H:%M:%S')
if (logger.hasHandlers()):
   logger.handlers.clear()
#if handler_console is none :
handler_console = logging.StreamHandler(stream = sys.stdout)
#then add it back
handler console.setFormatter(formatter)
handler console.setLevel(logging.INFO)
logger.addHandler(handler_console)
fh.setFormatter(formatter)
logger.addHandler(fh)
```

```
#create logger
logger = logging.getLogger()
logger.setLevel(logging.DEBUG) #setting Level to debug
#creates file handler which logs all the level messages
fh = logging.FileHandler('Log_file1.log' , mode = 'w')
fh.setLevel(logging.DEBUG) #setting level to handler to debug
formatter = logging.Formatter('[%(asctime)s] %(levelname)8s --- %(message)s' +
                             '(%(filename)s:%(lineno)s)' ,datefmt = '%Y-%m-%d %H:%M:%S')
if (logger.hasHandlers()):
    logger.handlers.clear()
#if handler_console is none :
handler_console = logging.StreamHandler(stream = sys.stdout)
#then add it back
handler_console.setFormatter(formatter)
handler_console.setLevel(logging.INFO)
logger.addHandler(handler_console)
fh.setFormatter(formatter)
logger.addHandler(fh)
```

ABSTRACT

PROBLEM 1:

In this part of the question, we have automated the generation of URL based on the CIK and Accession number. Once the URL is formed, all the 10-q links will be parsed and tables from the forms will be scraped and stored as individual tables which we have later zipped and pushed it to S3 bucket. The whole process is automated using Docker as the last step of this problem.

PROBLEM 2:

This part deals with the extraction of the log files present on the link given below:

https://www.sec.gov/dera/data/edgar-log-file-data-set.html

and then after downloading the zipped files inside a directory, these files are unzipped to access the CSVs and the data inside them. A series of pre-processing and data wrangling is applied on the 1st day(next one if the previous is not available) of each month of a particular year. The data is converted to a data frame and missing data is handled first, then the summary metrics in both statistical and graphical form is calculated. The analysis ends with calculating the anomalies in the data, if any.

APPROACH

Problem 1 Data wrangling Edgar data from text files:

Part 1: Parse files

Methodology:

- 1. Using the cik and accession number of a company, the url which contains the list of form is generated.
- 2. From that url, the url with 10-q forms are accessed.
- 3. From the 10-q form all the tables are searched.
 - a) Searched for statistical data which has '\$' and '%' in it.
- 4. Once the tables are separated from the other content, the data in each table is extracted.
 - a) The data is present inside the tag which is present inside tag.
- b) To go inside the tag we are first searching for all tag in which is present in .
 - c) So once tag is searched, the data is extracted in the form of text.
- d) There is a chance for the data not to be clean, so we are removing all the unwanted content.
- 5. After extracting all the data from the font , we are storing that in list form .
- 6. This is then converted to csv format where the rows are formed by the data from the tag. All these csv files are stored in a separate folder which is generated automatically.
- 7. This folder which contains the csv files is then stored in a zip file.

Process overview:

Step 1:

Generating URL:

- 1. The standard URL part for all the EDGAR data irrespective of their cik and accession number is stored in a variable called 'url'.
- 2. cik is generally a 10-digit number with '0's' preceding it. To generate the proper url we need to strip out the unwanted digits. So using .lstrip() function all the 0's are stripped.
- 3. To create the proper url, we need to omit all the '-' in accession number which is done using replace() function. This replaces ever '-' with "".
- 4. The newly created URL is stored in a variable called 'new_url'.
- 5. If the 'new_url' exist the log will output "Url request is successful" or else it will log "HTTP url not found" and exit the loop.
- 6. If the 'new_url' works, then the url is been read and that is parsed. This is stored in another variable called "new_url_2".

This is the output with proper cik and accession number.

Step 2:

Finding the 10-q form:

- 1. From the "new_url_2" which holds the parsed html page, all the <a> tags are found and that is stored in a variable called "containers".
- 2. In that containers, all the "href" labels are searched and the one with "10-q.htm" will be stored in "final_url".
- 3. The "final_url" is then parsed and stored in "scrape".

The final_url is printed for the given cik and accession number.

Step 3:

Finding Tables from the form and cleaning it:

The 10-q form has not only tables, but also other types of headers and footers. So in order to extract all the tables first we need to separate out the tables from the unwanted content.

1. All the tables in a html page will be inside the "table" tag which will always be present inside the "div" tag. So after inspecting the page, using select() function 'div table' tag is being found. select() function helps in finding two tags simultaneously. All the content under 'div table' is stored inside 'scrape_page' variable.

Output of scrape_page is printed below.

```
w Roman, serif" style="font-size:10.0pt;">New York</font></u></b>
<font face="Times New Roman,serif" style="font-size:10.0p</p>
t;"> </font>
<b><u><font face="Times Ne</pre>
w Roman, serif" style="font-size:10.0pt;">13-0871985</font></u></b>
(/tr>
n,serif" style="font-size:10.0pt;">(State of incorporation)</font>
```

2. Once that is filtered out, the tables inside this should be filtered out. For that, the only thing that differentiates the tables from other content is '\$' and '%'. So that is found by iterating through tag. This value is stored in the variable called "rawlist_tables". Now, the "rawlist_tables" holds the contents that is present inside the tables in that page.

Step 4:

Extracting data from the table and Cleansing it:

```
or tables 2 in rawlist tables:
   final list = []
   for rows 1 in tables 2.find all('tr'):
       row=[]
       for columns 1 in rows 1.findAll('td'):
           para = columns 1.find all('p')
           if len(para)>0:
               for para 1 in para:
                   new_para = para_1.get_text().replace("\n"," ")
                   new_para_1 = new para.replace("\xa0","")
                   row.append(new para 1)
           else :
               new_col=columns_l.get_text().replace("\n"," ")
               new col 2=new col.replace("\xa0","")
               row.append(new col 2)
       final list.append(row)
```

1. Once the tables are found, the next step is to scrape the data from them. For this, again tag searched for the content of tables.

```
cfont color="black" face="Times New Roman,serif"
  per share amounts)
cfont color="black" face="Times New Roman,serif"
  style="font-size:8.0pt;"> c/font>
  cfont color="black" face="Times New Roman,serif"
  style=
```

- 2. Sometimes , tag will not have any content . To omit all those cases the len(styles) condition is used.
- 3. The content inside the tag will hold other unwanted characters such as "\n", "\xao" which has to be replaced. So function replace() is used to do that.
- 4. Once the table is ready, all the table data is appended to final_list.

The data that is appended to the final_list

Step 5:

Saving it as .csv files:

```
with open(os.path.join('FINAL_TABLES' , 'DATA' + str(rawlist_tables.index(tables_2)) + '.csv'), 'w') as csv_file:
    csv_writer = csv.writer(csv_file)
    csv_writer.writerows(final_list)
    logger.info("Table has been extracted in csv format")
```

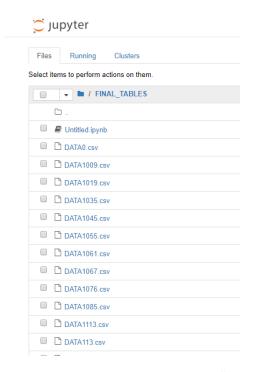
1. Directory 'FINAL_TABLES' have been formed using "os.mkdir()" function.

```
if not os.path.exists('FINAL_TABLES'):
    try:
        os.makedirs('FINAL_TABLES')
    except OSError as exc: # Guard against race condition
        logger.error("OSError" + str(exc))
```

- 2. The tables are iterated by indexing "tables_2".
- 3. All the files are written as csv files using "csv.writer()" function.

```
[2018-02-14 22:50:13]
                         INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
[2018-02-14 22:50:13]
                         INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
[2018-02-14 22:50:13]
                       INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
[2018-02-14 22:50:13] INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
[2018-02-14 22:50:13] INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
[2018-02-14 22:50:13] INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
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[2018-02-14 22:50:14] INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
                       INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
[2018-02-14 22:50:14]
                        INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
[2018-02-14 22:50:14]
[2018-02-14 22:50:14]
                         INFO --- Table has been extracted in csv format(<ipython-input-4-54ef8afc0651>:150)
```

The log output



All the csv files are created and stored in FINAL_TABLES folder.

Desired output:

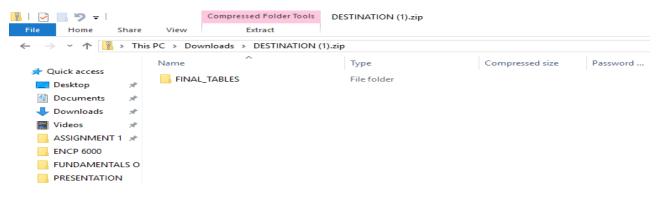
			ThreeMonth	s Ended	Septemb	Nine Month	ns Ended	Septen	nber 30,	
(Dollarsinn	millionsexce	ot pers	2013			2012			2013	2012
Revenue:										
Services	\$		14,225		\$	14,626		\$	42,811	\$ 44,279
Sales			8,987			9,642			27,735	29,424
Financing			509			479			1,506	1,500
Total reve	enue		23,720			24,747			72,052	75,203
Cost:										
Services			9,098			9,515			27,950	29,285
Sales			2,975			3,242			9,108	10,003
Financing			268			258			805	784
Total cost	:		12,341			13,016			37,863	40,072

Step 6:

Zipping all the files:

```
# zipping all the csv files
zip_file = zipfile.ZipFile('DESTINATION.zip' , 'w',zipfile.ZIP_DEFLATED)
for tables_2 in rawlist_tables:
    zip_file.write(os.path.join('FINAL_TABLES' , 'DATA' + str(rawlist_tables.index(tables_2)) + '.csv'))
    logger.info("All files are zipped")
    zip_file.close()
```

1. zipfile() function is used here. Using that "DESTINATION.zip" file is created and all the files in the FINAL_TABLES are zipped in here.



Part 2: Dockerize this pipeline

Docker Steps to build and run the code:

To build and run the image locally

- 1. create a docker file using below line
 - vim mydockerfile
- 2. Build the image

docker build -f mydockerfile -t assignpart1

3. Run the image to create container

docker run -e CIK=cik -e Accession_number=accno -e

Access_key=acckey -e Secret_key=seckey -e Input_location=iloc -tiassignpart

4. Tag the image

docker tag <image_id>dhanisha/assignpart1:latest

- 5. Push the docker image to docker hub docker push dhanisha/assignpart1
- 6. To pull the docker image from docker hub docker pull dhanisha/assignpart1
- 7. Run the image pulled from docker hub using following commands docker run -e CIk=cik -e Accession_number=accno -e

Access_key=acckey -e Secret_key=seckey -e Input_location=iloc -ti dhanisha/assignpart1

Giving Inputs:

- a) Here we are using arg parser to pass arguments that the interpreter needs.
- b) It takes cik, accession number, accesskey, secret key and then the location(to create bucket) in the same order.
- c) After it takes the input, it takes the length of the parameter.

Using Amazon S3 bucket:

Amazon Simple Storage Service (Amazon S3) is a web service that provides highly scalable cloud storage. Amazon S3 provides easy to use object storage, with a simple web service interface to store and get any amount of data from anywhere on the web.

Use library Boto s3 to access Amazon Simple Storage Service (Amazon S3) using AWS Python SDK.

Follow below steps to upload files to amazon S3

- 1. Create S3 client passing access key, secret key as a parameter
- 2. Create a bucket giving it a name and passing location as a parameter
- 3. Upload the file to bucket specifying bucket name and file name.

```
🖯 '''BucketAlreadyOwnedByYou errors will only be returned outside of the US Standard region.
Inside the US Standard region (i.e. when you don't specify a location constraint), attempting to recreate a bucket you a
logger.info("Uploading files to amazon")
     buck_name="ads-partl-assignment"
     S3_client = boto3.client('s3', Input_location, aws_access_key_id= Access_key, aws_secret_access_key= Secret_key)
     if Input_location == 'us-east-1':
            S3_client.create_bucket(
                Bucket=buck name
     else:
      S3_client.create_bucket(
              Bucket=buck_name,
                CreateBucketConfiguration={'LocationConstraint': Input location},
     logging.info("connection successful")
     S3 client.upload file("Extracted tables.zip", buck name, "Extracted tables.zip")
     S3 client.upload file("Log filel.log", buck name, "Log filel.log")
    logging.info("Files uploaded successfully")
Rexcept Exception as e:
     logging.error("Error uploading files to Amazon s3" + str(e))
```

Handling exceptions:

- Validation of CIK and Accession number:
 If the CIK and Accession number given is incorrect, it logs Invalid CIK and accession number and exits the code. Moreover, it throws HTTP Error as the url won't be formed.
 - 2. Validating Amazon access key and Secret key:
 When proper credentials are given, it connects to the HTTP and forms a s3 bucket in aws or else it throws a HTTP error and exits the program.

Problem 2Missing Data Analysis:

Methodology:

- **1.** The user is asked to input a year, awsaccessKey ,awssecretKey and the location.
 - On the basis of the year provided, an url will be generated, which will help the user to access the log files from EDGAR site of the entered year.
- 2. All the scenarios are taken into consideration to check if all the four inputs are valid or not. If even one of them is wrong, an error is thrown.

- **3.** Apart from the validations on the inputs, during the file extractions, if the log file is not available for any month or for any day, then the next available files are tried, if none of the data is available then the exception is thrown.
- **4.** The data extracted is cleaned via missing data calculation and then summary metrics is calculated for the same followed by visualizations.
- **5.** Finally, the results are stored in one directory and uploaded to AWS 3.

Process overview:

Step 1:

Handle Missing values:

- a. Reading the csv file in to a panda data frame and Finding out the null values in each column of the dataframe using :
 - #data.isnull().sum()
- b. Working on different columns with null values:
 - <u>b.1</u> Browser: Using the "MAXIMUM LIKELIHOOD" method and replacing the NaN values with the maximum used browser for that day of the month of the given year using the below commands:

```
#d =df.apply(pd.value_counts)
#data['browser'].replace(np.nan,d.index[0], inplace = True)
```

<u>b.</u>2Size (NUMERICAL DATA): The NaN values of the size column are handled via "IMPUTATION" method. The values of file size are grouped on the basis of extention and then the NaN values are replaced by the mean of the size. The same is performed on 4 different set of extentions (.xml,.htm.,.txt,.).

```
#s=data[['extention','size']].groupby(data['extention'].str.contains('txt'))
['size'].mean().reset_index(name='mean').sort_values(['mean'],ascendin
g=False)
```

```
#data.loc[(data['size'].isnull()) & (data['extention'].str.contains('txt'))]
#data.loc[(data['size'].isnull()) & (data['extention'].str.contains('txt'))] = s
#data.loc[(data['size'].isnull()) & (data['extention'].str.contains('txt'))]
```

<u>b.3</u> For the CATEGORICAL DATA columns (cik,accession,ip,date,time): Since, these are the basis of all the predictions and conclusions, therefore dropping all the rows which have NaN values in these coloumns using the below commands.

```
#data.dropna(subset=['cik'],inplace=True)
#data.dropna(subset=['accession'],inplace=True)
#data.dropna(subset=['ip'],inplace=True)
#data.dropna(subset=['date'],inplace=True)
#data.dropna(subset=['time'],inplace=True)
```

<u>b.</u>4For the categorical columns (code,zone,extention,idx,find): These columns are used for calculating the summary metrics, therefore dropping the NaN value rows

will make a problem in the analysis, therefore performing imputation and filling the values with the max value of the corresponding column.

```
#data['code'].fillna(data['code'].max(),inplace=True)
#data['zone'].fillna(data['zone'].max(),inplace=True)
#data['extention'].fillna(data['extention'].max(),inplace=True)
#data['idx'].fillna(data['idx'].max(),inplace=True)
#data['find'].fillna(data['find'].max(),inplace=True)

b.5For all the remaining columns:

#data['norefer'].fillna(1,inplace=True)
#data['noagent'].fillna(1,inplace=True)
#data['crawler'].fillna(0,inplace=True)
```

As soon as the different functions are performed, they are getting logged in the logger file, indicating every steps for missing data handling.

After all the steps are built successfully, the value of the NaN values throughout the data is calculated again and the expected output is:



------Missing data handling is achieved------

Step 2:

Summary Metrics:

a. The cleaned data is now worked upon for various summary metrics both in statistical and graphical form. The first command used is the **describe()** command to get a brief view of the data statistics.

b. STATICTICAL METHOD:

<u>b.</u>1To calculate which class of IP addresses(Class A, Class B, Class C) has done the maximum EDGAR fillings using mode:

The following analysis helps us in finding out the group of companies which file the maximum of Q10 forms, or who are searched the most , which will give us a

Prediction parameter that which companies are likely to follow the trend, this is done on the basis of IP classes. Class A companies are the big companies likes Google and Yahoo, and reduces with Class B and Class C.

```
ClassAlist = []
ClassBlist = []
Others = []
for i in range(0,len(df.index)):
   a = df.iloc[i]
   octects = a.str.split('.',expand = True)
    o = octects.iloc[0,0]
    if 0<int(o)<127:</pre>
        ClassAlist.append(int(o))
    elif 128<int(o)<182:</pre>
       ClassBlist.append(int(o))
        Others.append(int(o))
if len(ClassAlist) > (len(ClassBlist) | len(Others)):
    logger.info("The Companies who have filled maximum are the ones having ClassA ips ")
if len(ClassBlist) > (len(ClassAlist) | len(Others)):
   logger.info("The Companies who have filled maximum are the ones having ClassB ips ")
if len(Others) > (len(ClassBlist) | len(ClassAlist)):
    logger.info("The Companies who have filled maximum are the ones having ClassC ips ")
```

b.2Mean and Median sizes for each Browser:

The following summary metric helps in understanding what browsers are the most used browsers, or where the trend is moving towards like which browsers are about to be closed.

```
logger.info("Mean and Median sizes for each Browser")
brow_df = data.groupby('browser').agg({'size':['mean', 'median'],'crawler': len})
brow_df.columns = ['_'.join(col) for col in brow_df.columns]
data.reset_index(drop=True)
print(brow_df)
```

<u>b.3</u>Distinct count of ip per month i.e. per log file and status code on the basis ofip:

```
#Compute distinct count of ip per month i.e. per log file
  ipcount_df = df['ip'].nunique()
  logger.info("Compute distinct count of ip per month i.e. per log file")
  print(ipcount_df)

#Computing the count of status code on the basis of ip
  StCo_count=data[['code','ip']].groupby(['code'])['ip'].count().reset_index(name='count')
  logger.info("Computing the count of status code on the basis of ip")
  print(StCo_count)
  data.reset_index(drop=True)
```

<u>b.4</u>To find out the 15 top searched CIKs:

The following summary metric is yet another prediction parameter, about the future scope of the top 15 maximum searched companies on the basis of their CIKs

```
#To find out the 15 top searched CIKs
    cik_df = pd.DataFrame(data, columns = ['cik'])
    d = cik_df.apply(pd.value_counts)
    logger.info("Top 15 most searched CIKs with the count")
    d.head(15)
    data.reset_index(drop=True)
```

b.5Everything on per day basis

Some of the values are measured on per day basis:

```
#1. Average of Size
Avg_size=data[['date','size']].groupby(['date'])['size'].mean().reset_index(name='mean')
logger.info("Average of file size is computed")
print(Avg_size)
#2. Number Of Requests
Req_day=data[['date','ip']].groupby(['date'])['ip'].count().reset_index(name='count')
logger.info("Number of request per day is computed")
print(Req_day)
```

-----Summary Metrics(Statistical) is Calculated Successfully------

c. GRAPHICAL METHOD:

The same set of metrics is visualized using graphs:

```
#graph of no of status codes by browser
   try:
       logger.info("graphical analysis started")
       Num of codes=data[['browser','code']].groupby(['browser'])['code'].count().reset index(name = 'count code').sort values
(['count_code'],ascending=False)
       data.reset_index(drop = True)
       print(Num of codes)
       x= np.array(range(len(Num of codes)))
       y= Num_of_codes['count_code']
       xticks1 = Num_of_codes['browser']
       plt.xticks(x,xticks1)
       plt.bar(x,y)
       plt.title('Count of status code for all the browsers')
       plt.ylabel('Count of codes')
       plt.xlabel('Browsers')
       plt.savefig('Graphical_images/countsperbrowser'+ str(i) +'.png',dpi=100)
       #plt.savefig(os.path.join('Graphical_images',str(val),'countsperbrowser.png'),dpi=100)
       #plt.show()
       plt.clf()
       logger.info("graphical analysis end")
   except Exception as e:
       logger.error(str(e))
       logging.error("Error plotting the graph ")
```

```
#graph for max cik(10) by IP used
   try:
       logger.info("graphical analysis started")
       Num_of_CIKs=data[['cik','ip']].groupby(['cik'])['ip'].count().reset_index(name='count').sort_values(['count'],ascending=False).head
(10)
       data.reset_index(drop=True)
       print(Num_of_CIKs)
       x = np.array(range(len(Num_of_CIKs)))
       y = Num_of_CIKs['count']
       xticks2 = Num_of_CIKs['cik']
       plt.xticks(x, xticks2)
       plt.bar(x,y)
       plt.title('Top 10 CIKs by IPs')
       plt.ylabel('Count of IPs')
       plt.xlabel('CIK-')
       #plt.savefig(os.path.join('Graphical_images',str(val),'CIKsbyIPcount.png'),dpi=100)
       plt.savefig('Graphical_images/CIKsbyIPcount'+ str(i) +'.png',dpi=100)
       #plt.show()
       plt.clf()
       logger.info("graphical analysis end")
   except Exception as e:
       logger.error(str(e))
       logging.error("Error plotting the graph ")
        TORRING GILLOL / FILLOL PROCEETING THE READIL 1
#Graph of Mean of file size on the basis of code status
    try:
        Mean_size=data[['code','size']].groupby(['code'])['size'].mean().reset_index(name='mean').sort_values(['mean'],ascending=False)
        data.reset_index(drop=True)
        print(Mean_size)
        x = np.array(range(len(Mean_size)))
        y = Mean_size['mean']
        xticks3 = Mean_size['code']
        plt.xticks(x, xticks3)
        plt.bar(x,y)
        plt.title('filesize by codes')
        plt.ylabel('mean size')
        plt.xlabel('Code')
        {\it \#plt.savefig(os.path.join('Graphical\_images', str(val), 'MeanSizeByCode.png'), dpi=100)}
        plt.savefig('Graphical_images/MeanSizeByCode'+str(i) +'.png',dpi=100)
        #plt.show()
        plt.clf()
    except Exception as e:
        logger.error(str(e))
        logging.error("Error plotting the graph ")
```

The output for the graphical summary metrics are given below for the particular year: 2018

```
#Graph for average file size by extension
       Avg size=data[['extention','size']].groupby(['extention'])['size'].mean().reset index(name='mean').sort values
(['mean'],ascending=False).head(20)
       data.reset index(drop=True)
       print(Avg size)
       x = np.array(range(len(Avg_size)))
       y = Avg size['mean']
       xticks4 = Avg_size['extention']
       plt.xticks(x, xticks4)
       plt.bar(x,y)
       plt.title('Avg File size by extention')
       plt.ylabel('MeanFileSize')
       plt.xlabel('Extention')
       #plt.savefig(os.path.join('Graphical images',str(val),'filesizebyextention.png'),dpi=100)
       plt.savefig('Graphical images/filesizebyextention'+str(i) +'.png',dpi=100)
       #plt.show()
       plt.clf()
    except Exception as e:
       logger.error(str(e))
       logging.error("Error plotting the graph ")
```

Step 3:

Checking Anomalies:

Graphical approach is used for calculating the Anomalies.

Anomaly detection is performed on the size column. Since the column used is a single data frame, boxplot is used. However, a better visualization is possible through Scatter, and bar plot. Only if size is visualized with respect to some other variable.

```
#Anomalies in FileSize
      try:
            logger.info("Anomalies analysis started")
            data.boxplot(column='size',vert=True,sym='',whis=10,showfliers=False)
            plt.xticks(rotation=70)
            plt.title('Anomalies displayed on the file size')
            plt.vlabel('size')
            #plt.savefig(os.path.join('Graphical_images',str(val),'Anomalies.png'),dpi=100
            plt.savefig('Graphical images/'+'Anomalies'+ str(i) +'.png',dpi=100)
            #plt.show()
            logger.info("Anomalies analysis ended")
      except Exception as e:
            logger.error(str(e))
            logging.error("Error plotting the graph ")
      i = i+1
#uploading files to amazon s3
""BucketAlreadyOwnedByYou errors will only be returned outside of the US Standard region.
Inside the US Standard region (i.e. when you don't specify a location constraint), attempting to recreate a bucket you already own will
succeed.''
try:
   buck name="ads-part2-assign"
   S3_client = boto3.client('s3', Location, aws_access_key_id= accessKey, aws_secret_access_key=secretKey)
   if Location == 'us-east-1':
          S3_client.create_bucket(
             Bucket=buck name,
   else:
      S3 client.create bucket(
             Bucket=buck name,
             CreateBucketConfiguration={'LocationConstraint': Location},
   logging.info("Connection is successful")
   S3_client.upload_file("Graphical_images", buck_name, "Graphical_images")
S3_client.upload_file("'Log_file2.log'", buck_name, "'Log_file2.log'")
   logging.info("Files uploaded successfully")
except Exception as e:
   logging.error("Error uploading files to Amazon s3" + str(e))
```

EVERYTHING FOR YEAR 2008:

ZIPPED and UNZIPPED Directories:

Two directories are created inside the docker container for storing all the zipped files we fetched using the url. And then all the zipped CSVs are then extracted inside the directory unzipped file.

```
[2018-02-16 10:17:04] INFO --- Zipped file directory created!! (ADSAssign1Part2Final.py:87)
[2018-02-16 10:17:04] INFO --- UnZipped file directory created!! (ADSAssign1Part2Final.py:95)
```

RETRIVING THE ZIPPED files from Internet

```
[2018-02-16 10:17:19]INFO --- Retrieving zipped log file (ADSAssign1Part2Final.py:108)[2018-02-16 10:17:25]INFO --- Retrieving zipped log file (ADSAssign1Part2Final.py:108)[2018-02-16 10:17:37]INFO --- Retrieving zipped log file (ADSAssign1Part2Final.py:108)[2018-02-16 10:17:52]INFO --- Retrieving zipped log file (ADSAssign1Part2Final.py:108)[2018-02-16 10:17:59]INFO --- Retrieving zipped log file (ADSAssign1Part2Final.py:108)
```

EXTRACTING the CSVs month by month for the first day available:

```
[2018-02-16 10:19:06] INFO --- log20080101.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080201.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080301.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080301.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080401.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080501.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080601.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080601.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080601.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080601.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080601.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080601.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125) INFO --- log20080601.csv successfully extracted to folder: unzippedfiles.
```

GRAPHICAL_IMAGES directory is created to store all the png files at one place :

```
[2018-02-16 10:19:22] INFO --- log20081201.csv successfully extracted to folder: unzippedfiles. (ADSAssign1Part2Final.py:125)
[2018-02-16 10:19:22] INFO --- Graphical_images directory created!! (ADSAssign1Part2Final.py:144)
```

MISSING DATA ANALYSIS

a. Started with the calculation of which columns have missing value and how many are missing.

b. All the handling functions explained above are performed on this file of 2018 [2018-02-16 10:19:06] INFO --- log20080101.csv

```
2018-02-16 10:19:24]
                         INFO --- Finding the NaN values in each coloumn (ADSAssign1Part2Final.py:150)
                         INFO --- Working on the Browser Coloumn (ADSAssign1Part2Final.py:151)
[2018-02-16 10:19:24]
                         INFO --- Grouping the values of all the browsers and storing in a dataframe (ADSAssign1Part2Final.py:154)
[2018-02-16 10:19:24]
                         INFO --- Counting the frequency of each browser type used in descending order (ADSAssign1Part2Final.py:157)
[2018-02-16 10:19:24]
                         INFO --- Selecting the brower at the index θ(as it will be the maximum used browser) (ADSAssign1Part2Final.py:160)
[2018-02-16 10:19:24]
                         INFO --- confirming that no NaN values are present on the browser coloumn (ADSAssign1Part2Final.py:163)
[2018-02-16 10:19:24]
                         INFO --- Working on the Size Coloumn (ADSAssign1Part2Final.py:164)
[2018-02-16 10:19:24]
                         INFO --- Replacing the file size for ext : txt, by the mean of all the file size corresponding to txt (ADSAssign1Part2Final.py:165)
[2018-02-16 10:19:24]
                         INFO --- Replacing the file size with NaN values for ext : htm, by the mean of all the file size corresponding to htm (ADSAssign1Part2Final.
2018-02-16 10:19:27]
[2018-02-16 10:19:28]
                         INFO --- Replacing the file size with NaN values for ext : xml, by the mean of all the file size corresponding to xml (ADSAssign1Part2Final.
[2018-02-16 10:19:29]
                         INFO --- To check how many NaN values are remaining (ADSAssign1Part2Final.py:183)
                         INFO --- Replacing the file size for rest of the files with the mean of file size of txt extension, as it is the max used (ADSAssign1Part2F:
[2018-02-16 10:19:29]
l.py:184)
```

After performing this, repeating a again, we can see that all the missing values are being taken care of :

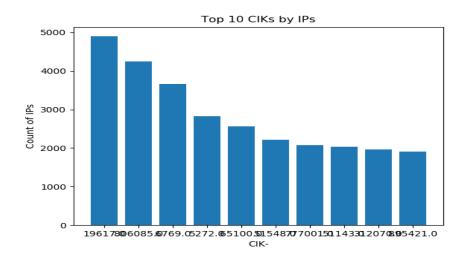
```
[2018-02-16 10:19:29]
                          INFO --- Working on all other coloumns (ADSAssign1Part2Final.py:187)
                          INFO --- If cik,Accession,ip,date are empty fields drop the records (ADSAssign1Part2Final.py:188)
[2018-02-16 10:19:29]
                          INFO --- Calculating the max categorical value in other coloumns( code, zone, extention, idx, find) and filling the NaNs (ADSAssign1Part2Final.py
[2018-02-16 10:19:29]
:194)
[2018-02-16 10:19:29]
                          INFO --- Filling empty values with Categorical Values for coloumns (norefer, noagent, nocrawler) (ADSAssign1Part2Final.py:201)
            0
date
time
            0
zone
cik
accession
extention
code
            0
size
            0
idx
            0
norefer
noagent
ind
crawler
prowser
dtype: int64
                       INFO --- Missing data is handled successfully (ADSAssign1Part2Final.py:206)
```

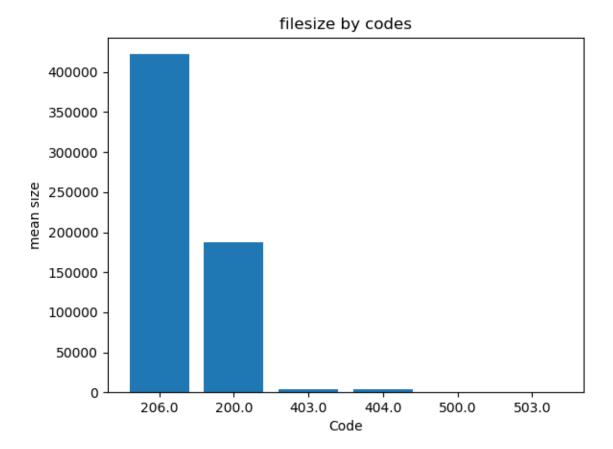
SUMMARY METRICS for the same

```
dtype: int64
                         INFO --- Missing data is handled successfully (ADSAssign1Part2Final.py:206)
[2018-02-16 10:19:29]
                         INFO --- Calculating Summary metrics of clean data (ADSAssign1Part2Final.py:209)
[2018-02-16 10:19:29]
                         INFO --- To calculate which class of IP addresses(Class A, Class B , Class C) has done the maximum EDGAR fillings using mode (ADSAssign1Part2F
[2018-02-16 10:19:30]
inal.py:212)
[2018-02-16 10:23:02]
                         INFO --- The Companies who have filled maximum are the ones having ClassA ips (ADSAssign1Part2Final.py:228)
                         INFO --- Mean and Median sizes for each Browser (ADSAssign1Part2Final.py:235)
[2018-02-16 10:23:02]
           size_mean size_median crawler_len
prowser
        1.155820e+05
                           9346.0
fox
                                          23.0
iem
        3.717001e+06
                        3717001.0
                                          1.0
        5.509095e+05
                         550909.5
                                          2.0
lin
       4.606733e+05
                         15420.5
                                        256.0
        3.048609e+05
                          8109.0
                                        1291.0
nac
mie
        3.722820e+05
                         15433.0
                                        277.0
       4.817409e+03
                         3559.0
                                         22.0
rim
       1.322456e+06
                         312067.5
                                         22.0
        1.650593e+05
                          9207.5
                                     159174.0
[2018-02-16 10:23:02]
                         INFO --- Top 15 most searched CIKs with the count (ADSAssign1Part2Final.py:244)
[2018-02-16 10:23:02]
                         INFO --- Compute distinct count of ip per month i.e. per log file (ADSAssign1Part2Final.py:250)
3895
[2018-02-16 10:23:02]
                         INFO --- Computing the count of status code on the basis of ip (ADSAssign1Part2Final.py:255)
   code count
  200.0 160021
  206.0
            956
2018-02-16 10:23:02]
                         INFO --- Average of file size is computed (ADSAssign1Part2Final.py:262)
        date
                       mean
  2008-01-01 167162.000143
                         INFO --- Number of request per day is computed (ADSAssign1Part2Final.py:266)
[2018-02-16 10:23:02]
        date count
  2008-01-01 161068
[2018-02-16 10:23:02]
                         INFO --- Mean of file size on the basis of code status (ADSAssign1Part2Final.py:270)
   code
                 mean
  200.0 167240.206804
  206.0 169654.670502
  404.0 3451.000000
                        INFO --- Summary metrics computed succesfully!! (ADSAssign1Part2Final.py:273)
2018-02-16 10:23:02]
```

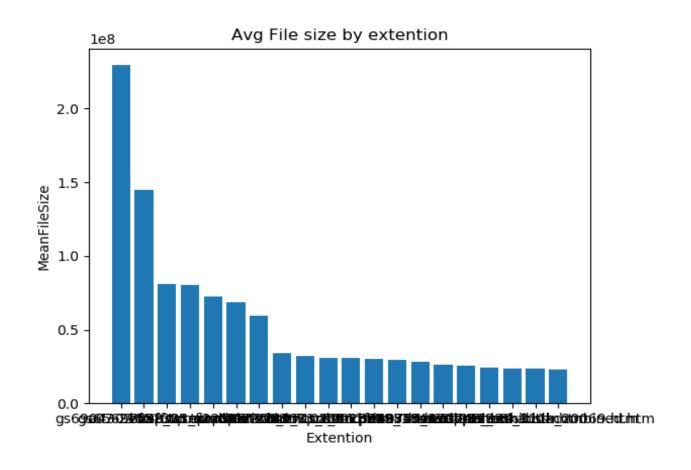
GRAPHICAL REPRESENTATIONS

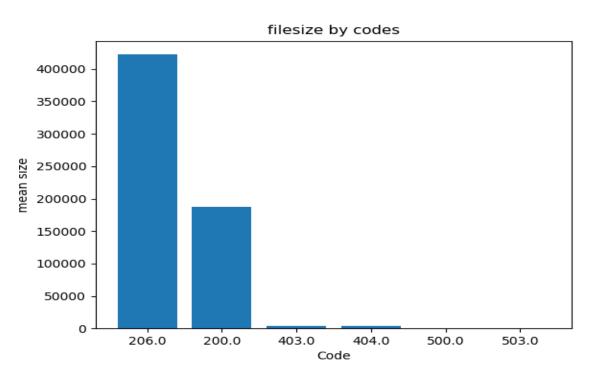
```
INFO --- graphical analysis started (ADSAssign1Part2Final.py:280)
[2018-02-16 10:53:27]
            count_code
   browser
10
       win
                 753160
       mac
                   6378
       lin
                   5492
       mie
                   1699
       fox
                    233
       rim
                     47
       opr
       iem
                     10
                      6
       iph
       saf
                      2
       oth
[2018-02-16 10:53:27]
                            INFO --- graphical analysis end (ADSAssign1Part2Final.py:296)
[2018-02-16 10:53:27]
                            INFO --- graphical analysis started (ADSAssign1Part2Final.py:303)
                count
382
       19617.0
                  4896
5844
      806085.0
                  4242
141
        6769.0
                  3666
        5272.0
104
                  2823
1277
       65100.0
                  2565
1050
       51548.0
                  2209
      777001.0
4878
                  2082
1040
       51143.0
                  2027
2551 312070.0
                  1959
9771 895421.0
                  1908
                            INFO --- graphical analysis end (ADSAssign1Part2Final.py:319)
[2018-02-16 10:53:27]
   code
                    mean
   206.0
          422346.549425
          187669.064187
  200.0
  403.0
            3632.000000
   404.0
             3451.000000
  500.0
             525.000000
   503.0
              323.000000
                        extention
35023
                gs6904302-fwp.txt 2.294876e+08
35020
                gs6569724-fwp.txt 1.445442e+08
2526
              a07-21558_25npx.htm 8.108014e+07
           v051383_fwp.htm 8.008723e+07
boafuturecomplete.htm 7.264694e+07
d22815.htm 6.887880e+07
50413
15161
20683
```

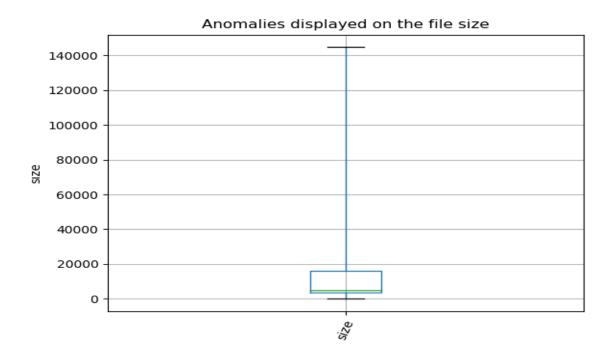




```
code
                    mean
   206.0
          422346.549425
          187669.064187
   200.0
   403.0
            3632.000000
   404.0
            3451.000000
   500.0
             525.000000
   503.0
             323.000000
                        extention
               gs6904302-fwp.txt
35023
                                   2.294876e+08
35020
               gs6569724-fwp.txt
                                   1.445442e+08
2526
             a07-21558_25npx.htm
                                   8.108014e+07
50413
                 v051383_fwp.htm
                                   8.008723e+07
15161
           boafuturecomplete.htm
                                   7.264694e+07
                       d22815.htm
20683
                                   6.887880e+07
                       d21731.htm
20602
                                   5.974119e+07
5458
              a08-7260_2ncsr.htm
                                    3.403204e+07
             a08-2249_1posam.htm
4354
                                   3.236706e+07
5417
                a08-7101 1nq.htm
                                   3.103957e+07
           brd031_0001331971.txt
15331
                                   3.091128e+07
16493
                 c23411fe497.htm
                                   3.022326e+07
             a06-25229_18497.htm
1289
                                   2.947193e+07
                 p74855nvcsr.htm
                                   2.794462e+07
44116
20684
              d22845_wrapper.htm
                                   2.612327e+07
25298
                       e61297.htm
                                   2.568189e+07
                                   2.402662e+07
20950
               d421743_ex4-1.htm
              a07-17329_311k.htm
2166
                                   2.348830e+07
       exh991to8kabslb_20069.htm
                                   2.348815e+07
45139
       psa_exhibits-combined.htm
                                   2.292322e+07
```







RESULTS and PREDICTIONS BASED ON PREVIOUS ANALYSIS

1. To calculate which class of IP addresses(Class A, Class B, Class C) has done the maximum EDGAR fillings using mode:

For the following the maximun IP address are the Class Alps.

2. Mean and Median sizes for each Browser on the basis of crawler_len:

_	size_mean	size_median	crawler_len
browser			
fox	1.155820e+05	9346.0	23.0
iem	3.717001e+06	3717001.0	1.0
iph	5.509095e+05	550909.5	2.0
lin	4.606733e+05	15420.5	256.0
mac	3.048609e+05	8109.0	1291.0
mie	3.722820e+05	15433.0	277.0
opr	4.817409e+03	3559.0	22.0
rim	1.322456e+06	312067.5	22.0
win	1.650593e+05	9207.5	159174.0

3. Compute distinct count of ip per month i.e. per log file

Value: 3895

4. Computing the count of status code on the basis ofip

```
code count
0 200.0 160021
1 206.0 956
2 404.0 91
```

5. Mean of file size on the basis of code status

```
code mean
0 200.0 167240.206804
1 206.0 169654.670502
2 404.0 3451.000000
```

6. As per the graph (Boxplot), no significant anomalies are found.

Docker Steps to build and run the code:

To build and run locally: Commands for Problem 2:

- 1. Create a docker file vim dockerfile
- 2. Build the image locally docker build -f dockerfile -t assignpart2
- 3. Run the image
 docker run -e year=yr -e accessKey=acckey -e secretKey=seckey -e
 location=loc -ti assignpart2
- 4. Tag the image docker tag <imageid>dhanisha/assignpart2:latest
- 5. Push the image to docker hub docker push dhanisha/assignpart2
- 6. To pull the image from hub docker pull dhanisha/assignpart2
- 7. To run the image that is pulled from the hub docker run -e year=yr -e accessKey=acckey -e secretKey=seckey -e locaiton=loc -tidhanisha/assignpart2

Uploading files to Amazon S3 bucket:

```
#uploading files to amazon s3
U'BucketAlreadyOwnedByYou errors will only be returned outside of the US Standard region.

Inside the US Standard region (i.e. when you don't specify a location constraint), attempting to recreate a bound of the US Standard region.
∃try:
      buck name="ads-part2-assignment"
      S3 client = boto3.client('s3', Location, aws access key id= accessKey, aws secret access key=secretKey)
      if Location == 'us-east-1':
              S3_client.create_bucket(
                   Bucket=buck_name,
      else:
         S3_client.create_bucket(
                  Bucket=buck name,
                   CreateBucketConfiguration={'LocationConstraint': Location},
      logging.info("Connection is successful")
      S3_client.upload_file("Graphical_images", buck_name, "Graphical_images")
      S3_client.upload_file("'Log_file2.log'", buck_name, "'Log_file2.log'")
      logging.info("Files uploaded successfully")
pexcept Exception as e:
      logging.error("Error uploading files to Amazon s3" + str(e))
```

CONCLUSION

We were able to extract the data from the 10-q form for the given cik and accession number and perform data science feature such a data wrangling, data pre-processing, Exploratory Data Analysis, anomaly detection via statistical and graphical methods on log file data for the year provided. We were also able to handle exceptions.

REFERENCES

- https://stackoverflow.com/questions/33388555/unable-to-install-boto3
- https://stackoverflow.com/questions/4770635/s3-error-the-difference-between-the-request-time-and-the-current-time-is-too-la
- https://www.sec.gov/dera/data/edgar-log-file-data-set.html
- https://old.datahub.io/dataset/edgar
- https://docs.python.org/3/library/urllib.parse.html
- https://docs.python.org/3/
- https://boto3.readthedocs.io/en/latest/
- https://docs.docker.com/engine/reference/commandline/build/
- https://docs.docker.com/get-started/
- https://docs.docker.com/engine/reference/commandline/run/
- https://docs.aws.amazon.com/AmazonS3/latest/gsg/CreatingABucket.html
- https://docs.aws.amazon.com/aws-sdk-php/v3/guide/examples/s3-examples.html