Assignment 1: Analysis of Bank Filings Data — Data Wrangling and Exploratory data analysis using Tableau Case Study

Under guidance of:

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Part 1: Analyzing Banks with asset values greater

than \$10 Billion

Process workflow:

Step 1: Web Scraping using Python 3

Step 2: Cleaning the data

Step 3: Creating stacked and unstacked CSV file using Pandas

Step 4: Pre-processing the data for Analysis

Step 5: Creating Dashboards in Tableau

Step 1: Web Scraping using Python 3

Modules Used :-

BeautifulSoup

To parse HTML document

Requests

To create a request and a response to the website

Panda

To create stacked and unstacked CSV file

```
BeautifulSoup
   requests.get("https://www.ffiec.gov/nicpubweb/nicweb/MCSGreaterThan100.aspx")
 parking the page using BeautifulSoup module
      BeautifulSoup(html_doc, 'html.parser')
 reading option tag to know all the quarters
option_tags = soup.findAll('option')
      to save data of all the files
           option tags:
   # preparing the AJAX call to load data "quarter" wise
```

Step 2 : Cleaning the data

- Extracting the RSSD ID from the bank name
- Made RSSD ID unique column so that no new row gets created if Bank Changes their name in any quarter
- Sanitizing the bank name
- Extracting column headers
- Creating a new column for quarters
- Removing the Rank column

Step 3: Creating stacked and unstacked CSV file using Pandas

- Creating a 2 dimensional structure for the CSV file.
- Pivoting the data frame to create unstacked version of the file



File: Python code for Step 1-3

Output CSV: Stacked Version

	Α	В	С	D	Е
1	RSSD ID	Institution Name	Location	TotalAssets	Date
2	1039502	JPMORGAN CHASE & CO.	NEW YORK, NY	\$2,466,096,000	20160630
3	1073757	BANK OF AMERICA CORPORATION	CHARLOTTE, NC	\$2,189,811,000	20160630
4	1120754	WELLS FARGO & COMPANY	SAN FRANCISCO, CA	\$1,889,235,000	20160630
5	1951350	CITIGROUP INC.	NEW YORK, NY	\$1,818,771,000	20160630
6	2380443	GOLDMAN SACHS GROUP, INC., THE	NEW YORK, NY	\$896,870,000	20160630
7	2162966	MORGAN STANLEY	NEW YORK, NY	\$828,873,000	20160630
8	1119794	U.S. BANCORP	MINNEAPOLIS, MN	\$438,463,000	20160630
9	3587146	BANK OF NEW YORK MELLON CORPORATION, T	NEW YORK, NY	\$372,351,000	20160630
10	1069778	PNC FINANCIAL SERVICES GROUP, INC., THE	PITTSBURGH, PA	\$361,528,406	20160630
11	2277860	CAPITAL ONE FINANCIAL CORPORATION	MCLEAN, VA	\$339,247,718	20160630
12	3232316	HSBC NORTH AMERICA HOLDINGS INC.	NEW YORK, NY	\$295,534,689	20160630
13	3606542	TD GROUP US HOLDINGS LLC	WILMINGTON, DE	\$276,317,370	20160630
14	1607170	TEACHERS INSURANCE & ANNUITY ASSOCIATION	NEW YORK, NY	\$276,045,408	20160630
15	1111435	STATE STREET CORPORATION	BOSTON, MA	\$255,396,733	20160630
16	1074156	BB&T CORPORATION	WINSTON SALEM, NC	\$221,858,615	20160630
17	1131787	SUNTRUST BANKS, INC.	ATLANTA, GA	\$199,276,480	20160630
18	1026632	CHARLES SCHWAB CORPORATION, THE	SAN FRANCISCO, CA	\$198,052,000	20160630
19	1275216	AMERICAN EXPRESS COMPANY	NEW YORK, NY	\$159,632,000	20160630
20	1562859	ALLY FINANCIAL INC.	DETROIT, MI	\$157,931,000	20160630
21	3226762	RBC USA HOLDCO CORPORATION	NEW YORK, NY	\$151,710,605	20160630
22	1132449	CITIZENS FINANCIAL GROUP, INC.	PROVIDENCE, RI	\$145,568,297	20160630
23	1447376	UNITED SERVICES AUTOMOBILE ASSOCIATION	SAN ANTONIO, TX	\$144,819,412	20160630
24	3840207	STATE FARM MUTUAL AUTOMOBILE INSURANCE	BLOOMINGTON, IL	\$143,801,553	20160630
25	1070345	FIFTH THIRD BANCORP	CINCINNATI, OH	\$143,625,325	20160630
26	1245415	BMO FINANCIAL CORP.	WILMINGTON, DE	\$132,007,952	20160630
27	3981856	SANTANDER HOLDINGS USA, INC.	BOSTON, MA	\$126,502,203	20160630
28	3242838	REGIONS FINANCIAL CORPORATION	BIRMINGHAM, AL	\$126,378,482	20160630
29	1037003	M&T BANK CORPORATION	BUFFALO, NY	\$123,820,584	20160630
30	1199611	NORTHERN TRUST CORPORATION	CHICAGO, IL	\$121,509,559	20160630

Output CSV: Unstacked Version

	Α	В	С	D	Е	F	G	Н	1	J	K
1	RSSD ID	Institution Name	Location	20121231	20130331	20130630	20130930	20131231	20140331	20140630	20140
2	1020180	BREMER FINANCIAL CORPORATION	SAINT PAUL, MN								
3	1020902	FIRST NATIONAL OF NEBRASKA, INC.	OMAHA, NE	\$16,409,360	\$15,890,934	\$15,466,494	\$15,977,890	\$16,271,487	\$16,573,510	\$16,754,901	\$17,082,
4	1025309	BANK OF HAWAII CORPORATION	HONOLULU, HI	\$13,789,923	\$13,563,474	\$13,787,068	\$13,916,694	\$14,127,598	\$14,333,411	\$14,884,625	\$14,577,
5	1025608	BANCWEST CORPORATION	HONOLULU, HI	\$79,869,488	\$78,851,727	\$80,069,917	\$81,729,444	\$83,527,474	\$84,945,155	\$86,617,152	\$86,894,
6	1025608	FIRST HAWAIIAN, INC.	HONOLULU, HI								
7	1026632	CHARLES SCHWAB CORPORATION, THE	SAN FRANCISCO, CA	\$133,637,000	\$133,324,000	\$135,907,000	\$140,211,000	\$143,642,000	\$144,066,000	\$143,401,000	\$147,445,
8	1027004	ZIONS BANCORPORATION	SALT LAKE CITY, UT	\$55,511,918	\$54,110,564	\$54,904,540	\$55,188,312	\$56,031,127	\$56,080,844	\$55,111,275	\$55,458,
9	1027518	CITY NATIONAL CORPORATION	LOS ANGELES, CA	\$28,618,492	\$27,433,754	\$27,379,501	\$29,059,404	\$29,717,951	\$29,851,542	\$30,819,092	\$32,015,
10	1031449	SVB FINANCIAL GROUP	SANTA CLARA, CA	\$22,766,602	\$22,802,242	\$22,165,054	\$23,757,075	\$26,417,306	\$29,724,778	\$33,322,533	\$36,044,
11	1032473	DEUTSCHE BANK TRUST CORPORATION	NEW YORK, NY	\$74,148,000	\$75,523,000	\$71,992,000	\$66,067,000	\$66,926,000	\$72,603,000	\$69,406,000	\$60,715,
12	1036967	CIT GROUP INC.	LIVINGSTON, NJ	\$44,012,251	\$44,563,888	\$44,631,022	\$46,223,981	\$47,138,960	\$48,578,081	\$44,152,666	\$46,480,
13	1037003	M&T BANK CORPORATION	BUFFALO, NY	\$82,985,468	\$82,794,833	\$83,229,005	\$84,427,485	\$85,162,391	\$88,530,360	\$90,835,002	\$97,230,
14	1039502	JPMORGAN CHASE & CO.	NEW YORK, NY	\$2,359,141,000	\$2,389,349,000	\$2,439,494,000	\$2,463,309,000	\$2,415,689,000	\$2,476,650,000	\$2,519,995,000	\$2,526,655,
15	1048773	VALLEY NATIONAL BANCORP	WAYNE, NJ	\$16,012,646	\$16,028,703	\$15,977,202	\$15,976,943	\$16,156,541	\$16,344,464	\$16,335,967	\$16,726,
16	1049341	COMMERCE BANCSHARES, INC.	KANSAS CITY, MO	\$22,176,815	\$22,240,695	\$21,921,322	\$22,462,282	\$23,081,892	\$22,820,527	\$23,016,162	\$22,710,
17	1049828	UMB FINANCIAL CORPORATION	KANSAS CITY, MO	\$14,927,196	\$15,705,470	\$15,253,217	\$16,184,233	\$16,911,852	\$15,945,830	\$15,562,690	\$16,284,
18	1060627	FIRSTBANK HOLDING COMPANY	LAKEWOOD, CO	\$12,874,974	\$13,078,898	\$13,022,264	\$13,169,177	\$13,384,848	\$14,009,813	\$13,941,733	\$14,052,
19	1068025	KEYCORP	CLEVELAND, OH	\$89,425,613	\$89,441,210	\$90,858,779	\$91,016,413	\$92,991,716	\$90,928,218	\$91,934,784	\$89,884,
20	1068191	HUNTINGTON BANCSHARES INCORPOR	COLUMBUS, OH	\$56,153,185	\$56,054,966	\$56,113,687	\$56,648,251	\$59,476,344	\$61,145,753	\$63,797,113	\$64,330,
21	1069778	PNC FINANCIAL SERVICES GROUP, INC.	PITTSBURGH, PA	\$305,285,879	\$300,945,933	\$304,547,667	\$308,912,603	\$320,596,232	\$323,586,973	\$327,251,474	\$334,602,
22	1070345	FIFTH THIRD BANCORP	CINCINNATI, OH	\$121,500,604	\$120,789,641	\$122,642,935	\$124,939,419	\$129,685,180	\$129,654,487	\$132,562,382	\$134,187,
23	1070804	FIRSTMERIT CORPORATION	AKRON, OH	\$14,916,022	\$15,273,159	\$23,534,261	\$24,140,030	\$23,912,451	\$24,500,602	\$24,566,414	\$24,610,
24	1073757	BANK OF AMERICA CORPORATION	CHARLOTTE, NC	\$2,212,004,452	\$2,176,625,000	\$2,125,686,000	\$2,128,706,000	\$2,104,995,000	\$2,152,533,000	\$2,172,001,000	\$2,126,138,
25	1074156	BB&T CORPORATION	WINSTON SALEM, NC	\$183,872,371	\$180,836,757	\$182,735,153	\$181,050,008	\$183,009,992	\$184,651,158	\$188,012,330	\$187,021,
26	1075612	FIRST CITIZENS BANCSHARES, INC.	RALEIGH, NC	\$21,283,651	\$21,351,012	\$21,308,822	\$21,511,352	\$21,199,091	\$22,154,997	\$22,062,840	\$21,942,
27	1076217	UNITED BANKSHARES, INC.	CHARLESTON, WV						\$11,886,320	\$12,051,710	\$12,085,
20	1070530	DDV/A COMBACC DANICCHAREC INIC	HOHETON TV			¢60 674 076	סדת דפת חדי	C71 OCE 176	¢74 0E7 117	C7E 7A7 100	Ć70 107 i

Step 4: Pre-processing the data for Analysis

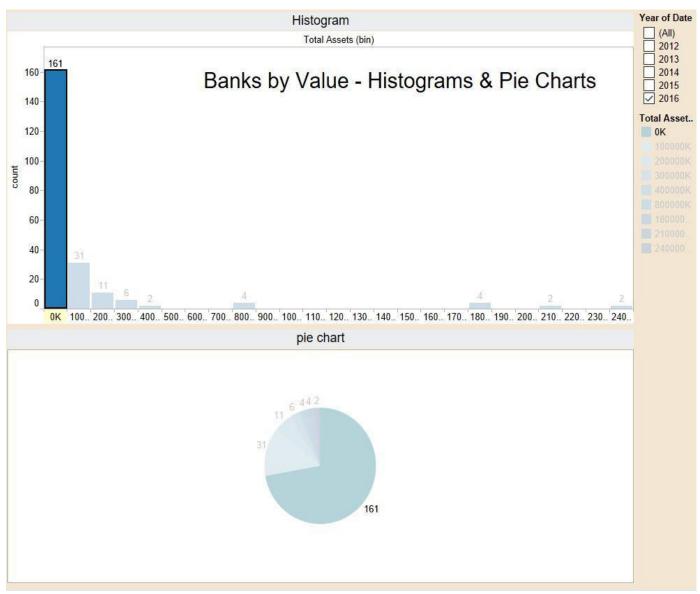
 Splitting the Location into city and state for analysis purpose on Tableau

Rssd Id # nic_ma	Institution Name Abc nic_main_data	Location Abc nic_main_data	City -⊕	State ==	Total Assets # nic_main_d	Date ⊟ nic_ma	Total Assets (bin)	count =#
1039502	JPMORGAN CHAS	NEW YORK, NY	NEW YORK	NY	2,466,096,000.00	6/30/2016	2,400,000,000	1
1073757	BANK OF AMERIC	CHARLOTTE, NC	CHARLOTTE	NC	2,189,811,000.00	6/30/2016	2,100,000,000	1
1120754	WELLS FARGO &	SAN FRANCISCO,	SAN FRANCISCO	CA	1,889,235,000.00	6/30/2016	1,800,000,000	1
1951350	CITIGROUP INC.	NEW YORK, NY	NEW YORK	NY	1,818,771,000.00	6/30/2016	1,800,000,000	1
2380443	GOLDMAN SACH	NEW YORK, NY	NEW YORK	NY	896,870,000.00	6/30/2016	800,000,000	1
2162966	MORGAN STANLEY	NEW YORK, NY	NEW YORK	NY	828,873,000.00	6/30/2016	800,000,000	1
1119794	U.S. BANCORP	MINNEAPOLIS, MN	MINNEAPOLIS	MN	438,463,000.00	6/30/2016	400,000,000	1
3587146	BANK OF NEW YO	NEW YORK, NY	NEW YORK	NY	372,351,000.00	6/30/2016	300,000,000	1
1069778	PNC FINANCIAL S	PITTSBURGH, PA	PITTSBURGH	PA	361,528,406.00	6/30/2016	300,000,000	1
2277860	CAPITAL ONE FIN	MCLEAN, VA	MCLEAN	VA	339,247,718.00	6/30/2016	300,000,000	1
3232316	HSBC NORTH AM	NEW YORK, NY	NEW YORK	NY	295,534,689.00	6/30/2016	200,000,000	1
3606542	TD GROUP US HO	WILMINGTON, DE	WILMINGTON	DE	276,317,370.00	6/30/2016	200,000,000	1

Step 5: Creating Dashboards in Tableau

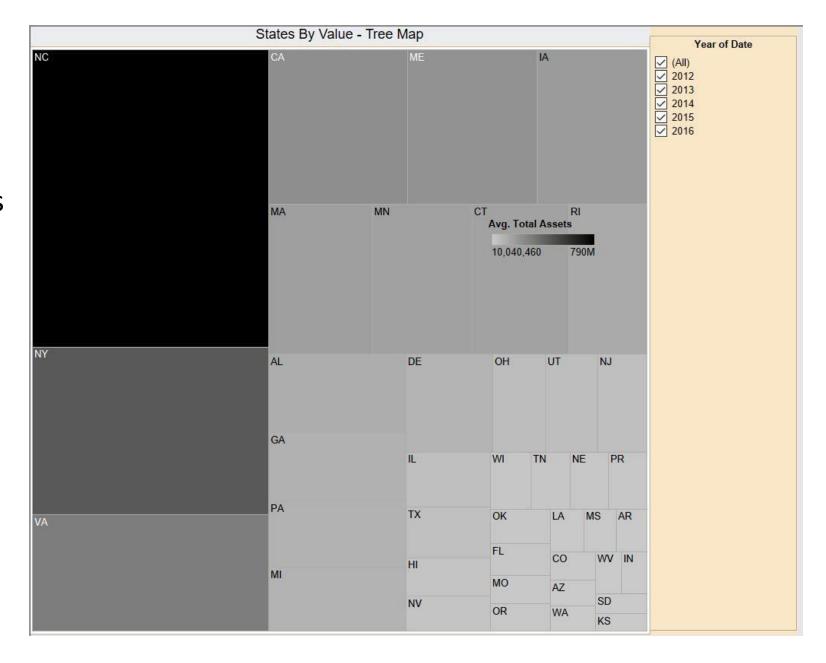
Banks by Value:

- Histograms shows the count of companies in the range on the x axis.
- Pie chart shows the count of companies in the range. Gives us insights about the percentage of the total value.



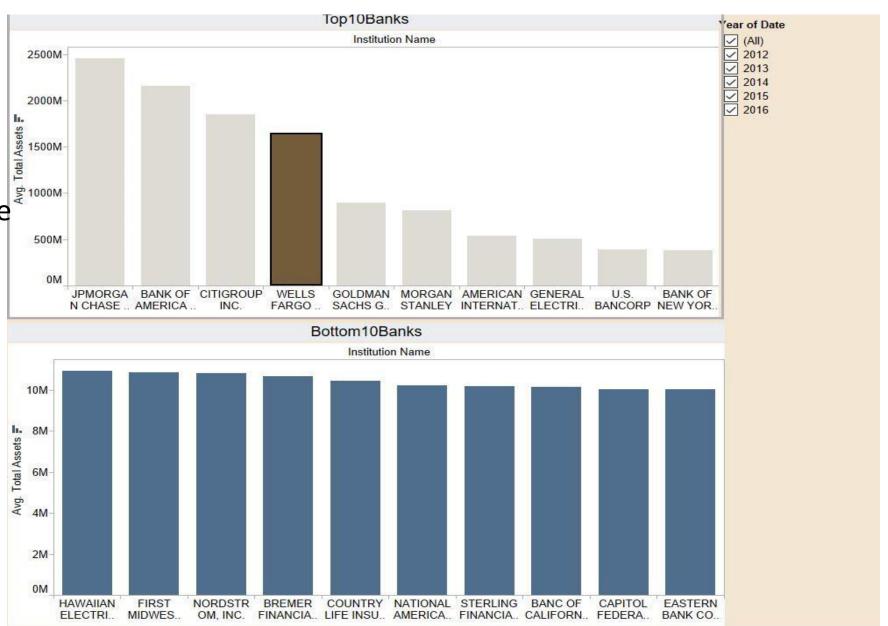
States by Value:

- Tree map here shows the assets of companies grouped by the state of where they are located.
- The values are filtered according to the Year. If multiple years are selected it takes the average of the values



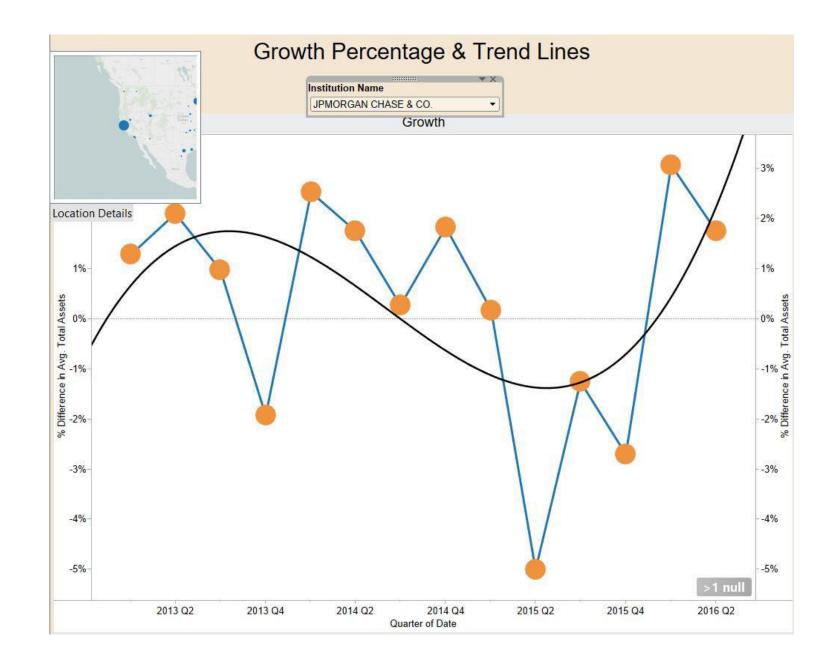
Top and Bottom 10 banks:

- The bar chart shows the top 10 and bottom 10 banks by value of their assets
- The values are filtered according to the Year. If multiple years are selected it takes the average of the values



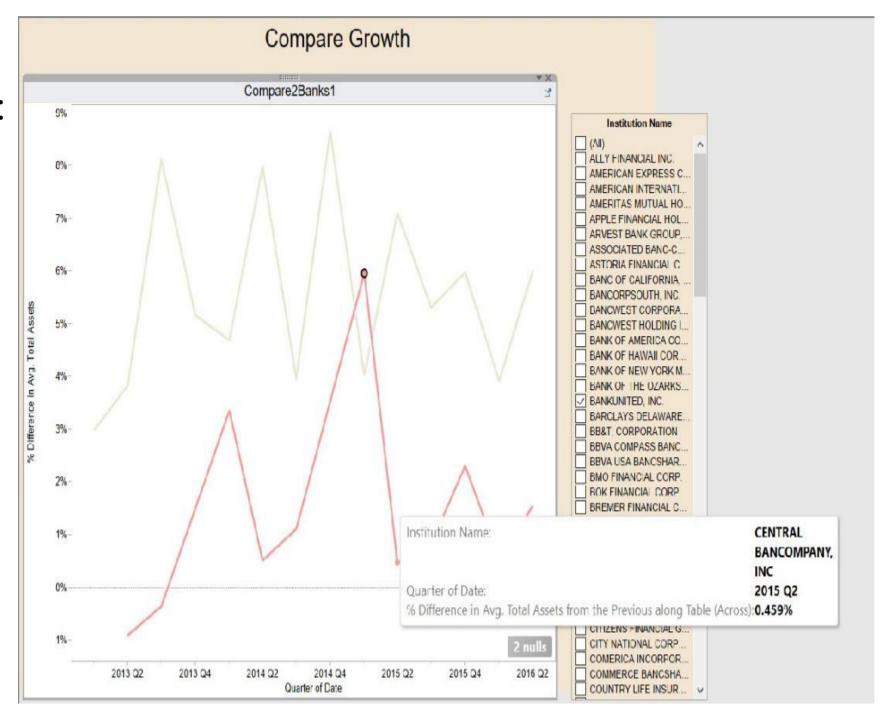
Growth Percentage:

- Shows the percentage growth of assets & trend lines with respect to successive quarters
- The figure shows the trend line for JP Morgan, which shows their assets are expected to increase in the next quarter.



Growth comparison:

- The graph compares the asset growth of companies selected for every quarter.
- The figure shows that % asset growth of Central bankcompany Inc. is generally lower than Bankunited Inc



Total Asset value:

 The figure shows total asset values grouped by the states for every quarter.

Total Asset values by State and Quarter Pivot Quarter of ... State Q1 121,529,351 ^ 121,535,448 Q2 Q3 120,376,869 Q4 121,283,033 AR Q1 14,511,482 Q2 14,647,082 Year of Date ✓ (All) ✓ 2012 ✓ 2013 Q3 14,852,699 Q4 14,545,194 Q1 AZ 13,249,991 **2014** 13,407,486 Q2 ✓ 2015 Q3 12,122,197 ✓ 2016 Q4 12,437,794 247,663,193 CA Q1 Q2 232,556,947 Q3 Q4 CO 14,563,989 14,592,854 Q2 14,106,401 Q3 14,088,943 Q4 CT 164,188,052 Q1 Q2 152,306,640 Q3 183,318,179 Avg. Total Assets Q4 159,341,482 DE Q1 94,400,183 10,040,460 794,707,861 Q2 94,824,919 Q3 98.747.308 91,879,149 Q4 FL Q1 21,462,311 22,206,296 Q2 Q3 21,205,504 20,831,379 Q4 GA Q1 105,751,372 Q2 106,679,523 Q3 104.509.145

Part 2: Data Scraping for BHCPR Average Reports

Process workflow:

Step 1: Scrape WebPage to get all the reqd. PDF files (Peer 1)

Step 2 : Scraping the PDF file

Step 3: Cleaning the data

Step 4: Creating and saving the data into CSV file

Step 5: Repeating Step 2-4 for all Peer 1 files

Note: Used Python2.7

Step 1: Scrape WebPage to get all the PDFs

Modules used:

Beautiful Soup

To parse HTML document

Requests

To create a request and a response to the website

Slate

To scrape PDFs

Urllib

Fetch file from HTTPResponse

```
BeautifulSoup
requests.get["https://www.fflec.gov/nicpubseb/content/BHCPREFT/BHCPR_Posr_htm"]
      the page using Beautifulloup medule
   BeautifulSoup(html_doc, "html.parser")
     soup.find_all("div",{"class":"contentfull"})
       iii mydivs[ii].findAll("table") :
       - rows[1].findAll("td")
      HILDORIUS III MORREIT
        # married links late w limit
        list_of_links.append("https://www.ffiec.gov/micpulweb/content/BMCPRPT/" = link.a.geti'href')
linked file | List_of_Links:

    urtlib.urlopen(linked_file)
```

Step 2 : Scraping the PDF file

- Used "slate 0.5.2" package for scraping all PDFs.
- "slate" is a wrapper package for PDFMiner
- Fetched all PDF file links using "urllib" library

Step 3 : Cleaning the Data

- Identified the rows & columns by recognizing the pattern of formatting in the PDF file
- Created a list of lists containing the data of PDF file
- Removed all the white spaces
- Removed all redundant lines
- Aligned the headers & data in respective positions

Step 4: Creating and saving the data into CSV

 Created File name as reqd. by extracting it from the online PDF file itself i.e. from

https://www.ffiec.gov/nicpubweb/content/BHCPRRPT/REPORTS/BHCPR PEER/June2016/PeerGroup 1 June2016.pdf

```
to "Peer1_2016_June.csv"
```

• Using "csv" module, made a CSV file for each quarter

Output CSV

Less: Provision for Loan and Lease Losses	0.21	0.32	0.33	0.49	1.1
Plus: Realized G/L on HTM Sec	0	0	0	0	0
Plus: Realized G/L on AFS Sec	0.03	0.03	0.04	0.05	0.06
Plus: Other Tax Equiv Adjustments	0	0	0	0	0
Equals: Pretax Net Oper Inc (TE)	1.44	1.23	1.28	1.12	0.77
Less: Applicable Income Taxes (TE)	0.48	0.42	0.42	0.39	0.29
Less: Minority Interest	0	0	0.01	0.01	0
Equals: Net Operating Income	0.95	0.81	0.88	0.72	0.51
Plus: Net Extraordinary Items	0	0	0	0	0
Equals: Net Income	0.95	0.81	0.87	0.72	0.51
Memo: Net Income (Last Four Qtrs)	0.94	0.75	0.86	0.72	0.52
MARGIN ANALYSIS:					
Avg Earning Assets / Avg Assets	90.18	90.01	89.87	90.3	89.93
Avg Int-Bearing Funds / Avg Assets	68.17	69.01	68.65	70.86	72.85
Int Income (TE) / Avg Earning Assets	3.83	4.11	4.07	4.32	4.54
Int Expense / Avg Earning Assets	0.62	0.75	0.74	0.88	1.07
Net Int Inc (TE) / Avg Earn Assets	3.15	3.31	3.28	3.37	3.39
YIELD OR COST:					
Total Loans and Leases (TE)	4.74	5.09	5.06	5.34	5.44
Interest-Bearing Bank Balances	0.31	0.32	0.31	0.34	0.35
Fed Funds Sold & Reverse Repos	0.43	0.36	0.4	0.38	0.45
Trading Assets	1.29	1.4	1.25	1.43	1.37
Total Earning Assets	3.77	4.06	4.03	4.26	4.46

Step 5: Repeating Step 2-4 for all Peer 1 files

- Traverse through the list of links of PDF files fetched in Step 1
- For every PDF file, scrape, clean & save a new CSV



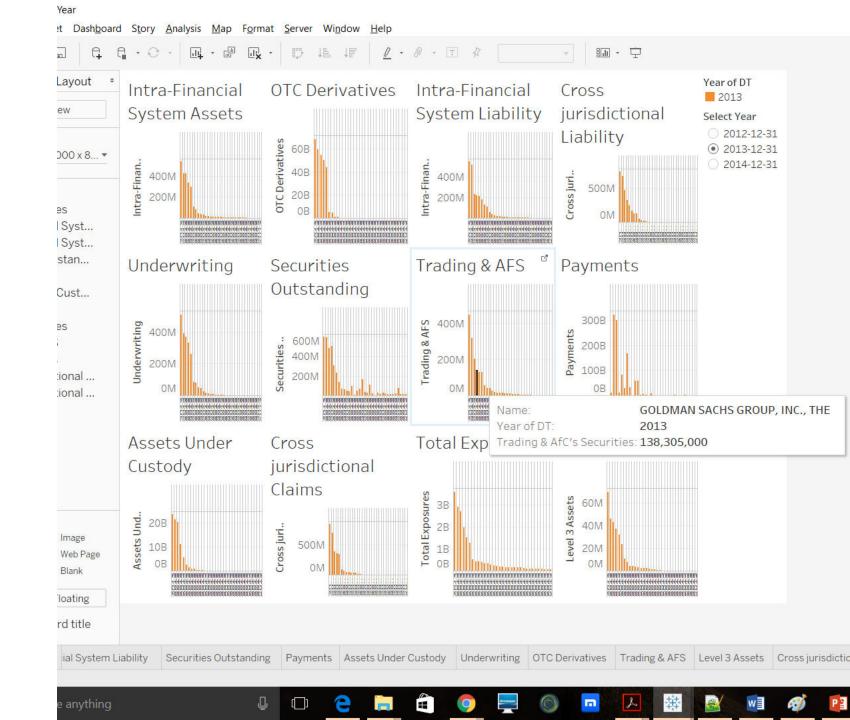
File: Python code for Part 2

Part 3:

Analysis of Banking Organization Systemic Risk Reports for the years 2012, 2013 and 2014

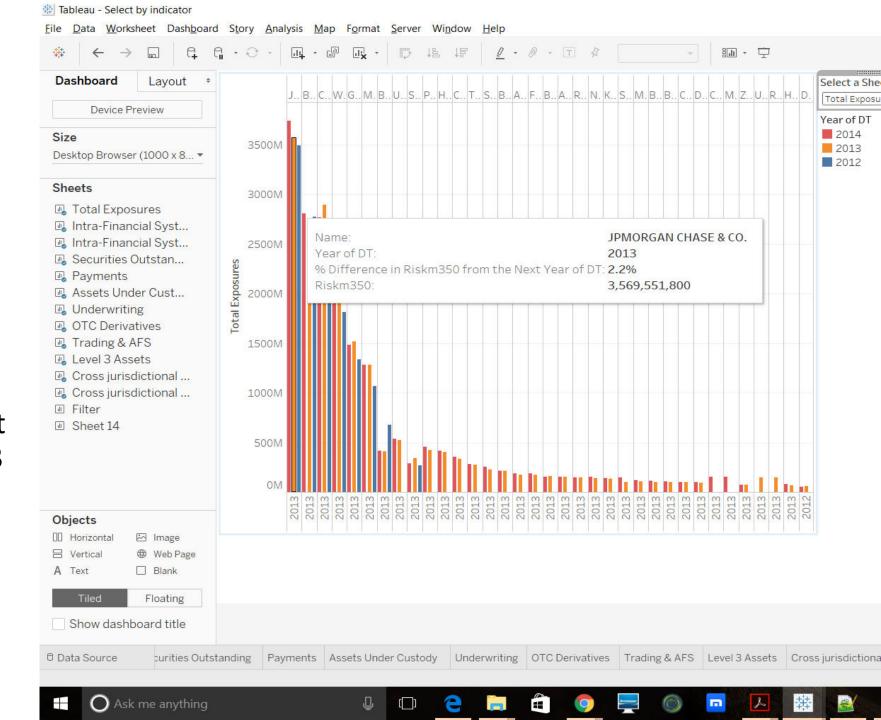
Graphs Dashboard

- 12 bar charts for Measures for each Bank.
- Filtering based on Year.



Time Series Dashboard

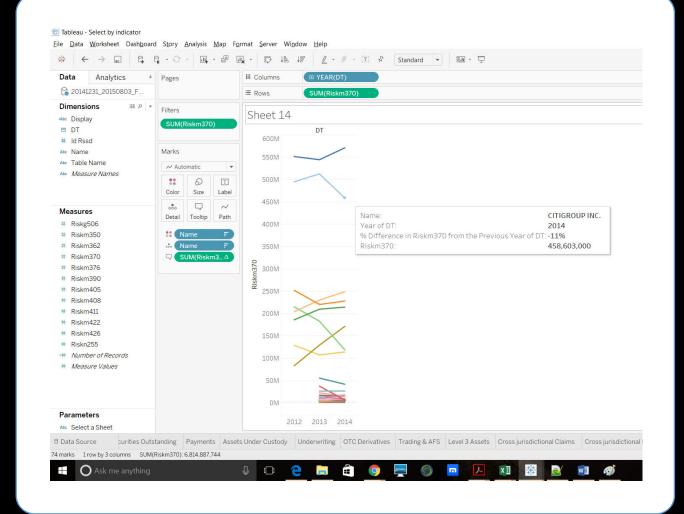
- Shows Year to Year comparison in form of Percentage Change values.
- Sheet Selector as
 Dashboard filter, to select
 Measure value over the 3 years.

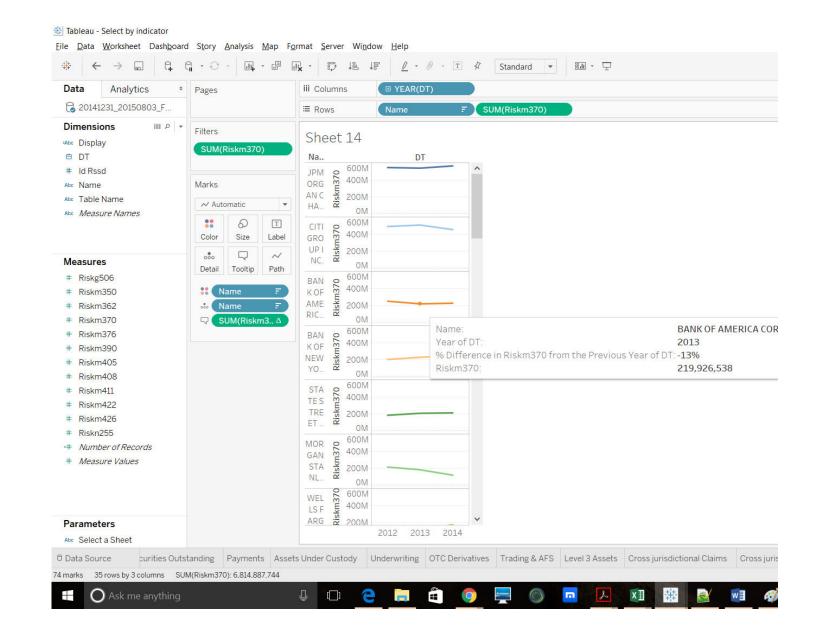


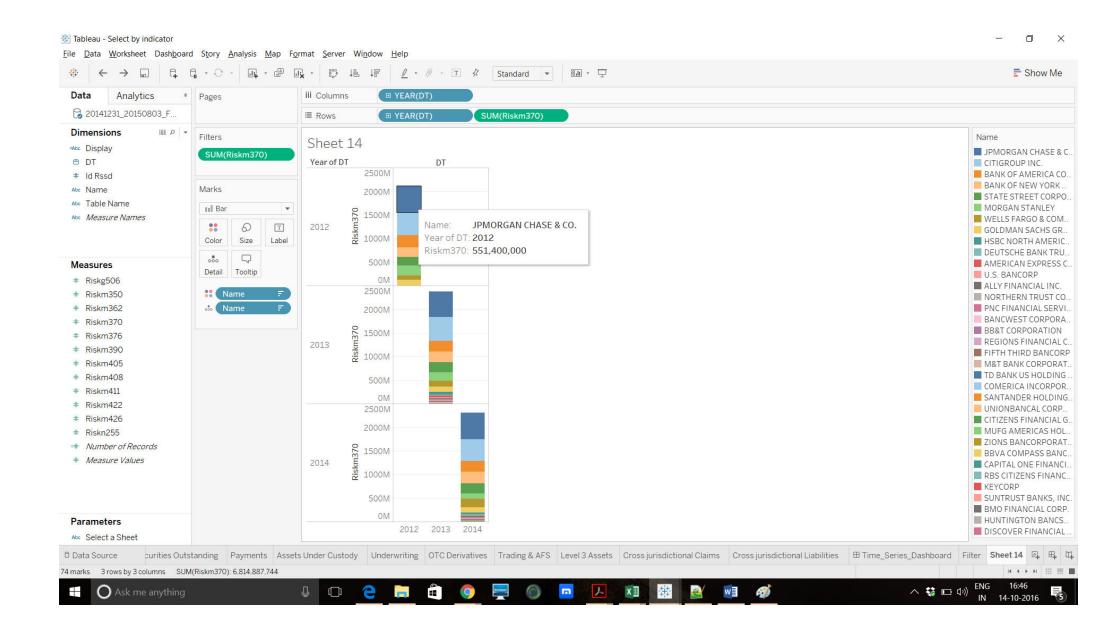
Alternative Visualization techniques

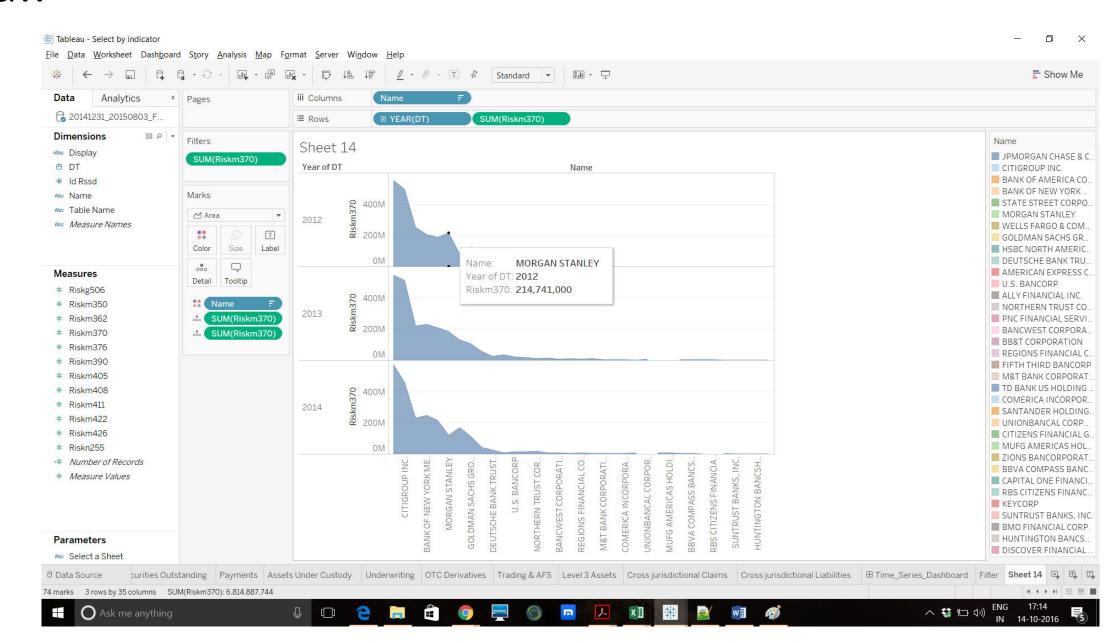
Line Graph

- Progressive visual of a measure for each bank over the years.
- Tooltip Indicates percentage change.
- Indicates the rise & fall of a measure for each bank.



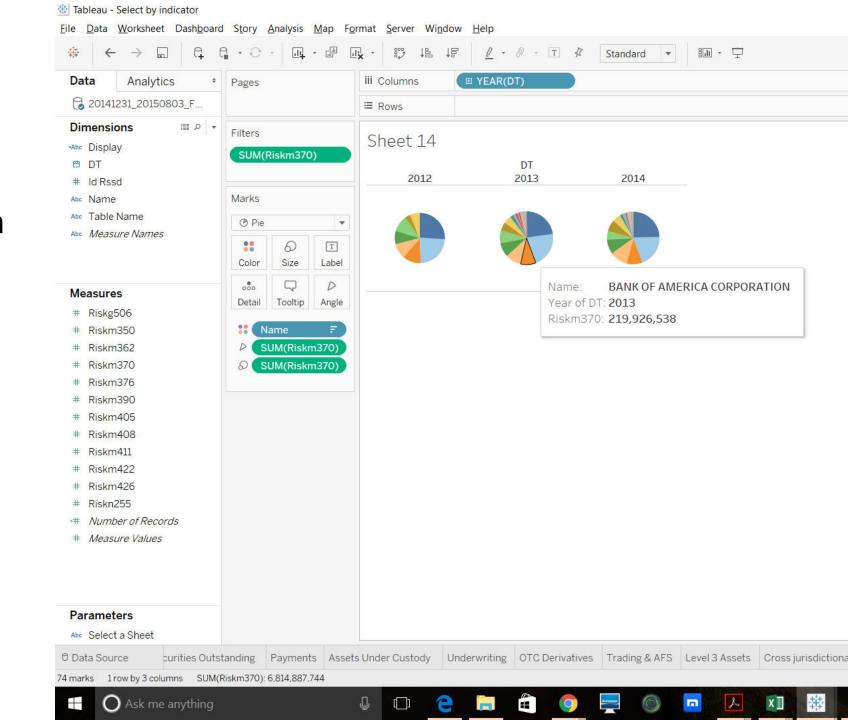






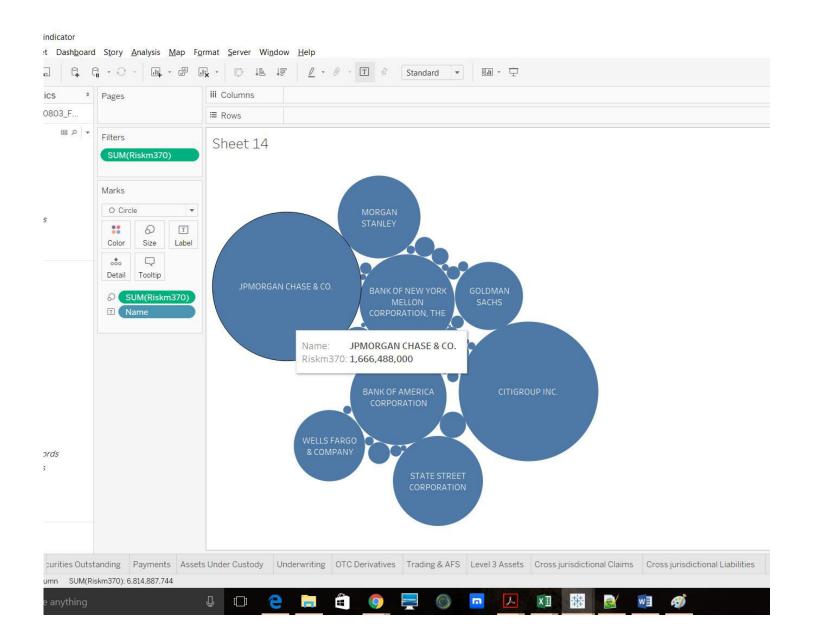
Pie Chart

- Year Wise Distribution of individual measure for all the banks.
- Chart shows data for all the banks & its measure distribution for each year.



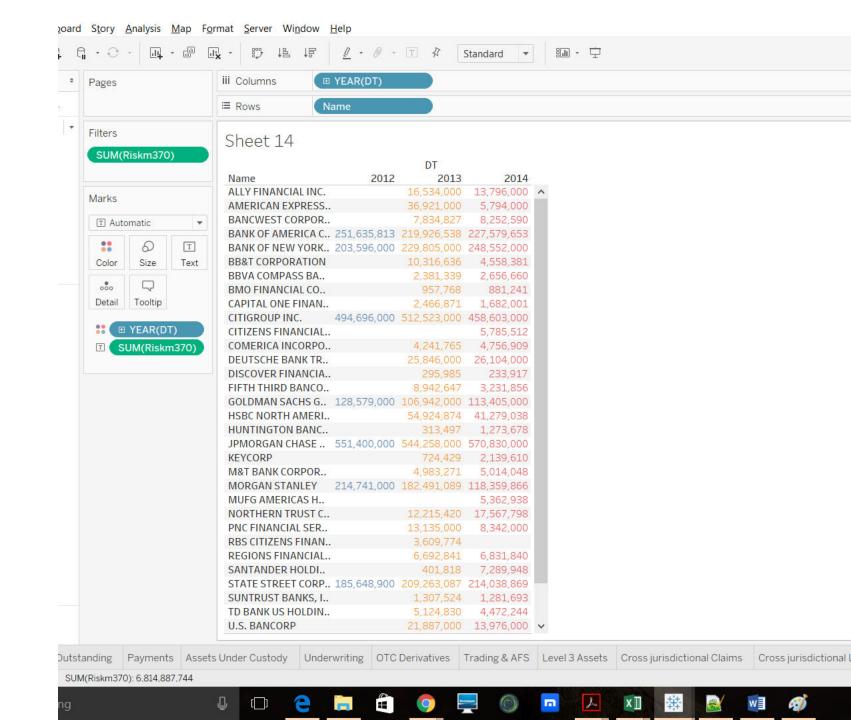
Packed Bubbles

- Total Bank Distribution of a measure value computed for all years.
- Combined Distribution for all banks for all the years for an individual measure.



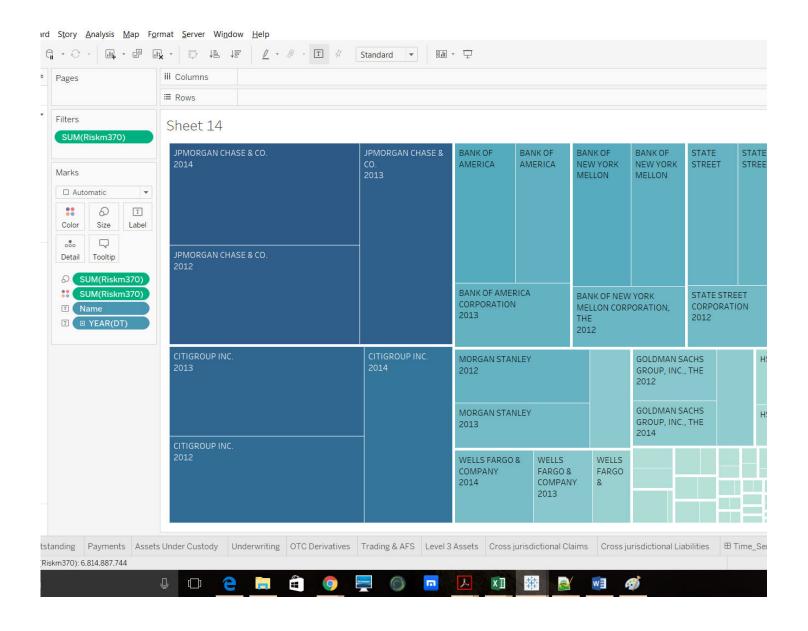
Text Tables/Highlight Tables

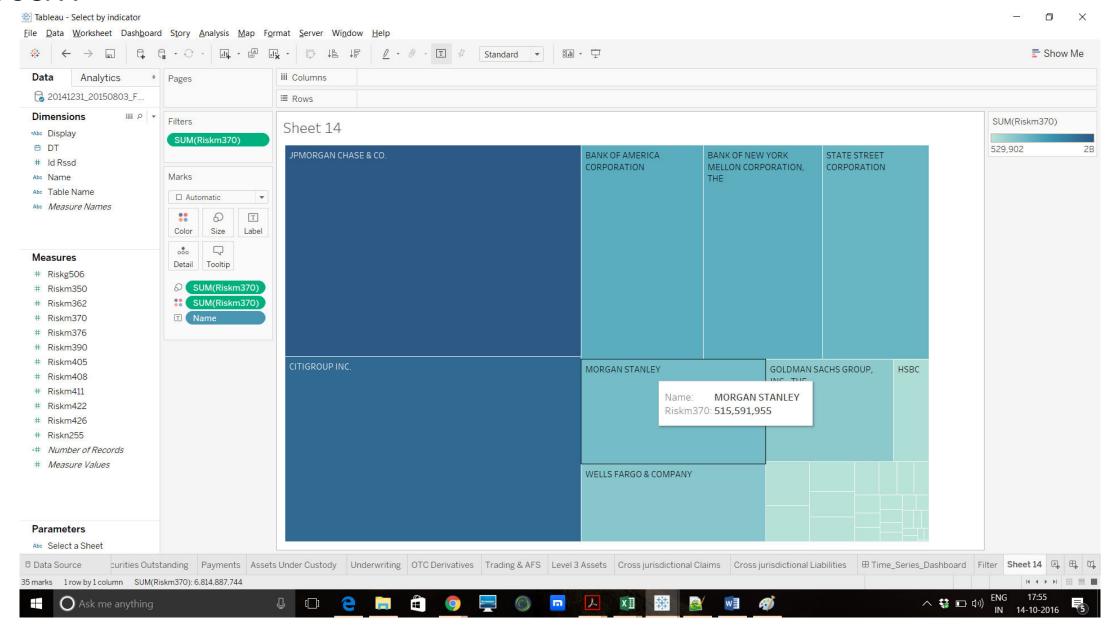
 Tabular data for bank assets over 3 years.



Tree Maps

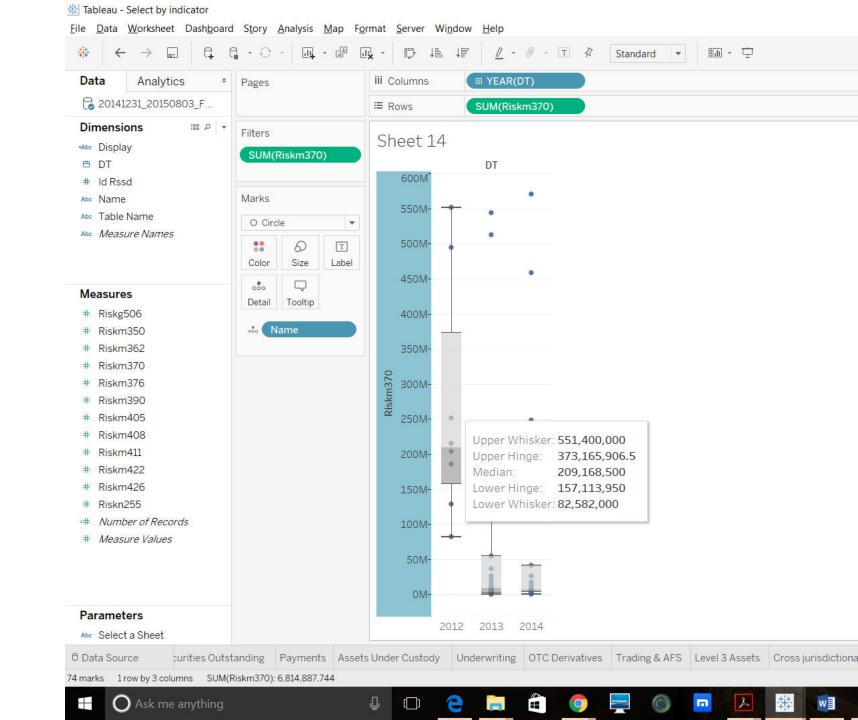
- Year wise measure distribution for banks.
- Coloring the rectangles by measure value





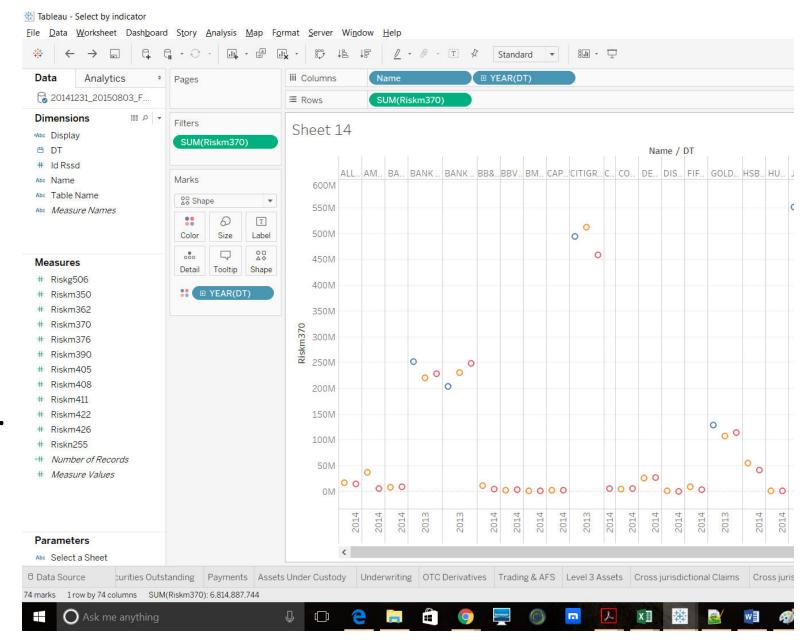
Box Plots

- Shows Median of the measure value along with 1st & 3rd Quartiles.
- Shows the skewness of data.
- Shows details about the outliers



Side-by-Side Circle Views

- Technique to accentuate data on scatter plots.
- Tells about relative concentration of data.



Thank you!