

DATA INTENSIVE COMPUTING

Data Economy: A Real Case Study



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Problem 3A: EDA on Brooklyn Rolling Sales data

Step 1: Cleaning the data and Performing EDA

```
21 # Performing cleaning on the data
22 data_brooklyn$SALE.PRICE.N <- as.numeric(gsub("[^[:digit:]]", "", data_brooklyn$SALE.PRICE))
23 count(is.na(data_brooklyn$SALE.PRICE.N))
24 names(data_brooklyn) <- tolower(names(data_brooklyn))
25
26 data_brooklyn$gross.sqft <- as.numeric(gsub("[^[:digit:]]", "", data_brooklyn$gross.square.feet))
27 data_brooklyn$land.sqft <- as.numeric(gsub("[^[:digit:]]", "", data_brooklyn$land.square.feet))
28 data_brooklyn$sale.date <- as.Date(data_brooklyn$sale.date)
29 data_brooklyn$year.built <- as.numeric(as.character(data_brooklyn$year.built))
30 head(data_brooklyn)
31
40 data_brooklyn.sale <- data_brooklyn[data_brooklyn$sale.price.n!=0,]
41 head(data_brooklyn)
42
61 # Removing outliers
62 data_brooklyn.homes$outliers <- (log(data_brooklyn.homes$sale.price.n) <=5) + 0
63 data_brooklyn.homes <- data_brooklyn.homes[which(data_brooklyn.homes$outliers==0),]
64 plot(log(data_brooklyn.homes$gross.sqft), log(data_brooklyn.homes$sale.price.n))
65
```

After cleaning The data looks something like this

```
> head(data_brooklyn.sale)
  borough      neighborhood      building.class.category tax.class.at.present block lot
1      3                15 CONDOS - 2-10 UNIT RESIDENTIAL                814 1103
2      3                15 CONDOS - 2-10 UNIT RESIDENTIAL                814 1105
3      3                15 CONDOS - 2-10 UNIT RESIDENTIAL               1967 1401
4      3                15 CONDOS - 2-10 UNIT RESIDENTIAL               1967 1402
5      3                15 CONDOS - 2-10 UNIT RESIDENTIAL               1967 1403
6      3                15 CONDOS - 2-10 UNIT RESIDENTIAL               1967 1404
  ease.ment building.class.at.present      address apart.ment.number zip.code
1      NA                342 53RD STREET                11220
2      NA                342 53RD STREET                11220
3      NA                290 GREENE AVE                11238
4      NA                290 GREENE AVE                11238
5      NA                290 GREENE AVE                11238
6      NA                290 GREENE AVE                11238
  residential.units commercial.units total.units land.square.feet gross.square.feet year.built
1                0                0            0                0                0            0
2                0                0            0                0                0            0
3                0                0            0                0                0            0
4                0                0            0                0                0            0
5                0                0            0                0                0            0
6                0                0            0                0                0            0
  tax.class.at.time.of.sale building.class.at.time.of.sale sale.price sale.date sale.price.n gross.sqft land.sqft
1                2                R1    $403,572 2013-07-09    403572            0            0
2                2                R1    $218,010 2013-07-12    218010            0            0
3                2                R1    $952,311 2013-04-25    952311            0            0
4                2                R1    $842,692 2013-04-25    842692            0            0
5                2                R1    $815,288 2013-04-25    815288            0            0
6                2                R1    $815,288 2013-04-25    815288            0            0
```

	Neighborhood	Total_Sales
1	BEDFORD STUYVESANT	754228259
2	PARK SLOPE	733389041
3	WILLIAMSBURG-NORTH	577846277
4	BROOKLYN HEIGHTS	540126620
5	CROWN HEIGHTS	454188002
6	WILLIAMSBURG-SOUTH	440947016

We also performed aggregation on Sales Period so as to get intuition on the Total Sales for a particular month

```

85 # Analysing Sales Data by aggregating over months
86 Sale_Dates <- data_brooklyn.sale$sale.date
87 Sale_Price <- data_brooklyn.sale$sale.price.n
88 Sale_Period <- as.yearmon(Sale_Dates, "%b-%y")
89 Sale_Period_Frame <- data.frame(Sale_Period, Sale_Price)
90 Cum_Sale_Period_Frame <- aggregate(Sale_Price ~ Sale_Period,
91                                   Sale_Period_Frame, function(x) sum(as.numeric(x)))
92 colnames(Cum_Sale_Period_Frame) <- c("Sale_Period", "Total_Sales")
93 Cum_Sale_Period_Frame
94

```

Cumulative Total Sales over the year

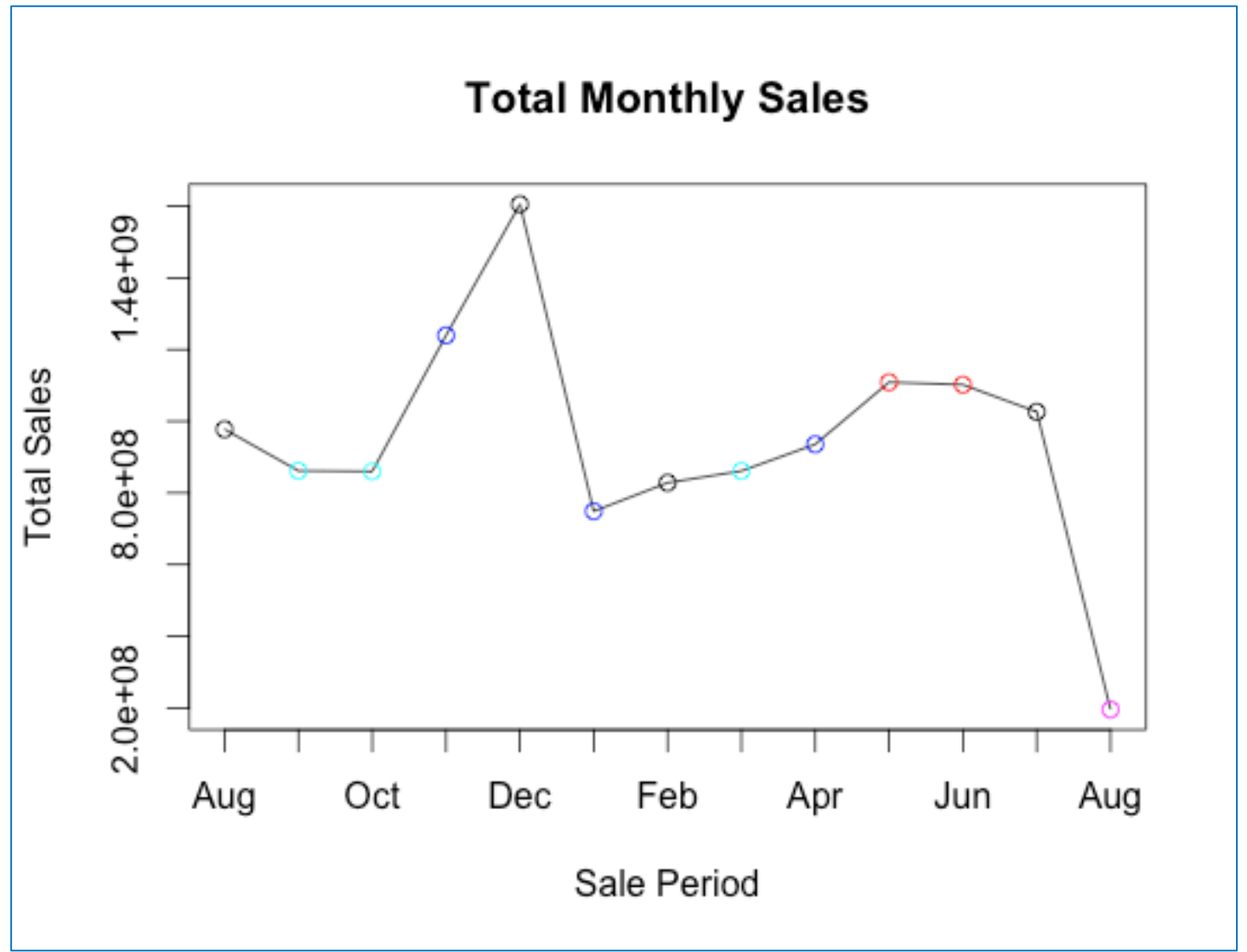
```

> Cum_Sale_Period_Frame
  Sale_Period Total_Sales
1    Aug 2012  977471505
2    Sep 2012  861661453
3    Oct 2012  859888461
4    Nov 2012 1239527524
5    Dec 2012 1605319345
6    Jan 2013  748783668
7    Feb 2013  828278777
8    Mar 2013  861116653
9    Apr 2013  935646324
10   May 2013 1108659450
11   Jun 2013 1101898050
12   Jul 2013 1026217841
13   Aug 2013 195467166

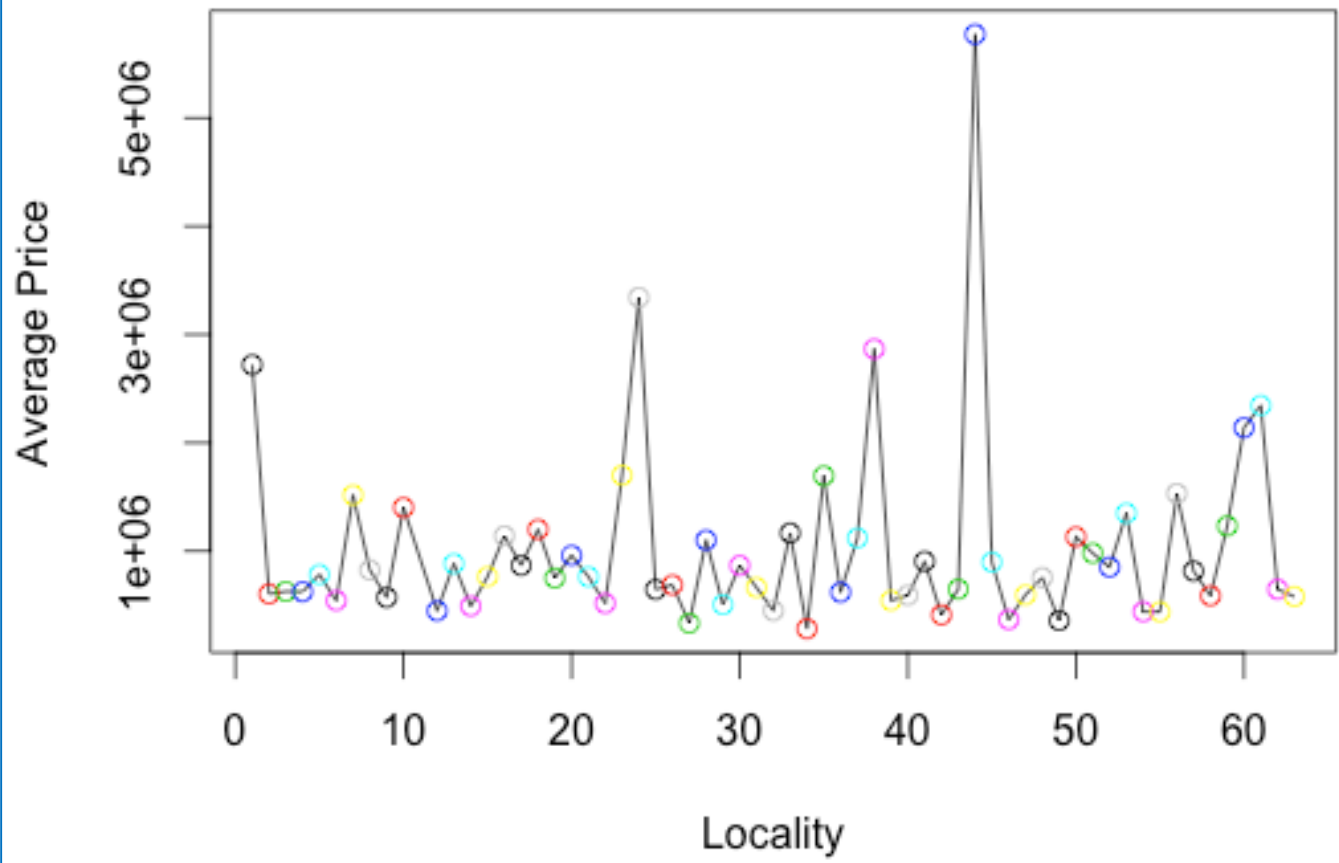
```

Step 2: Generating graphs

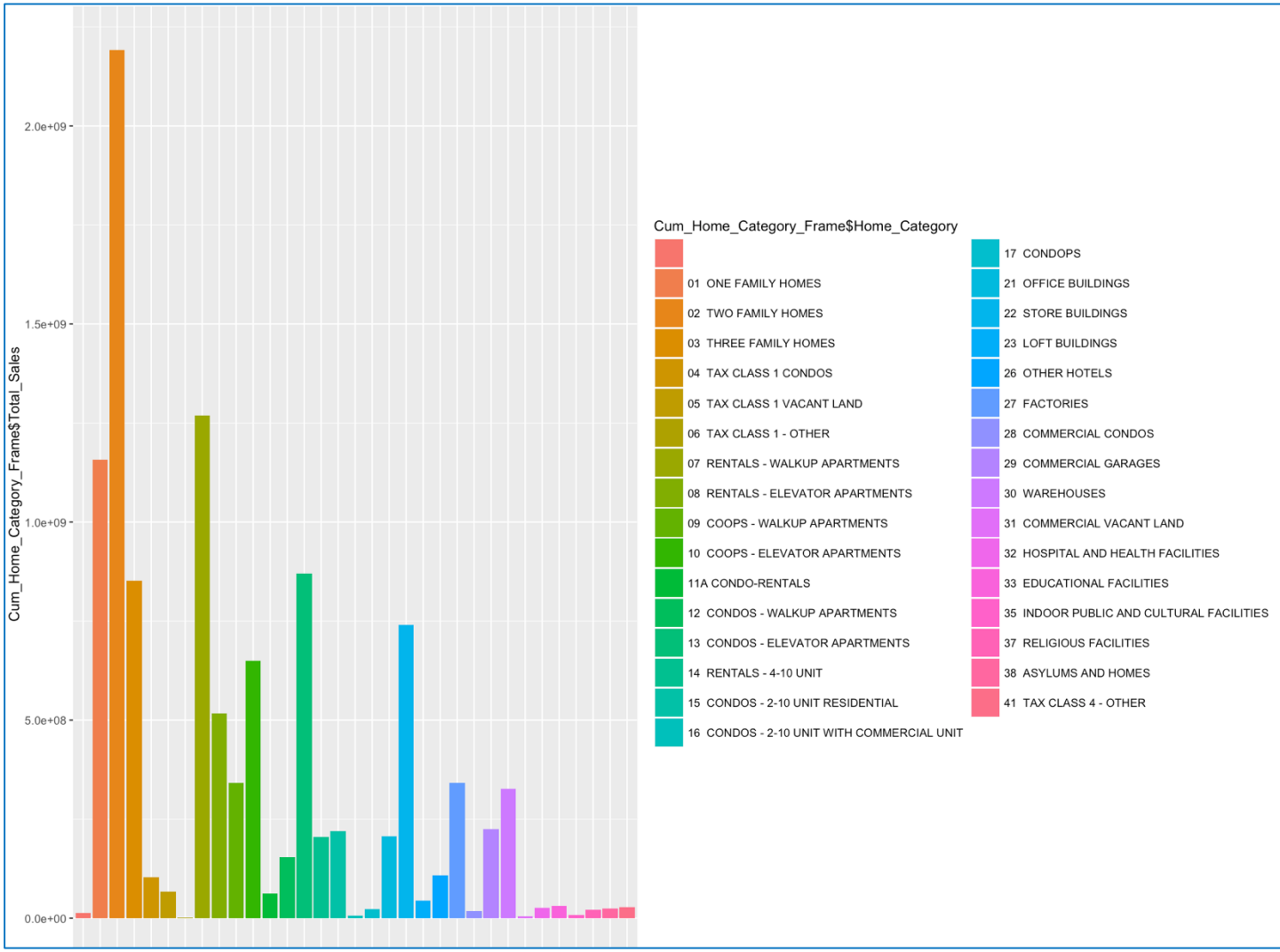
The next step is to gather meaning information on the sales for the Brooklyn by generating different graphs and plots



Neighborhood vs Average Price



Graph for Cumulative Total Sales over the year vs Home Category



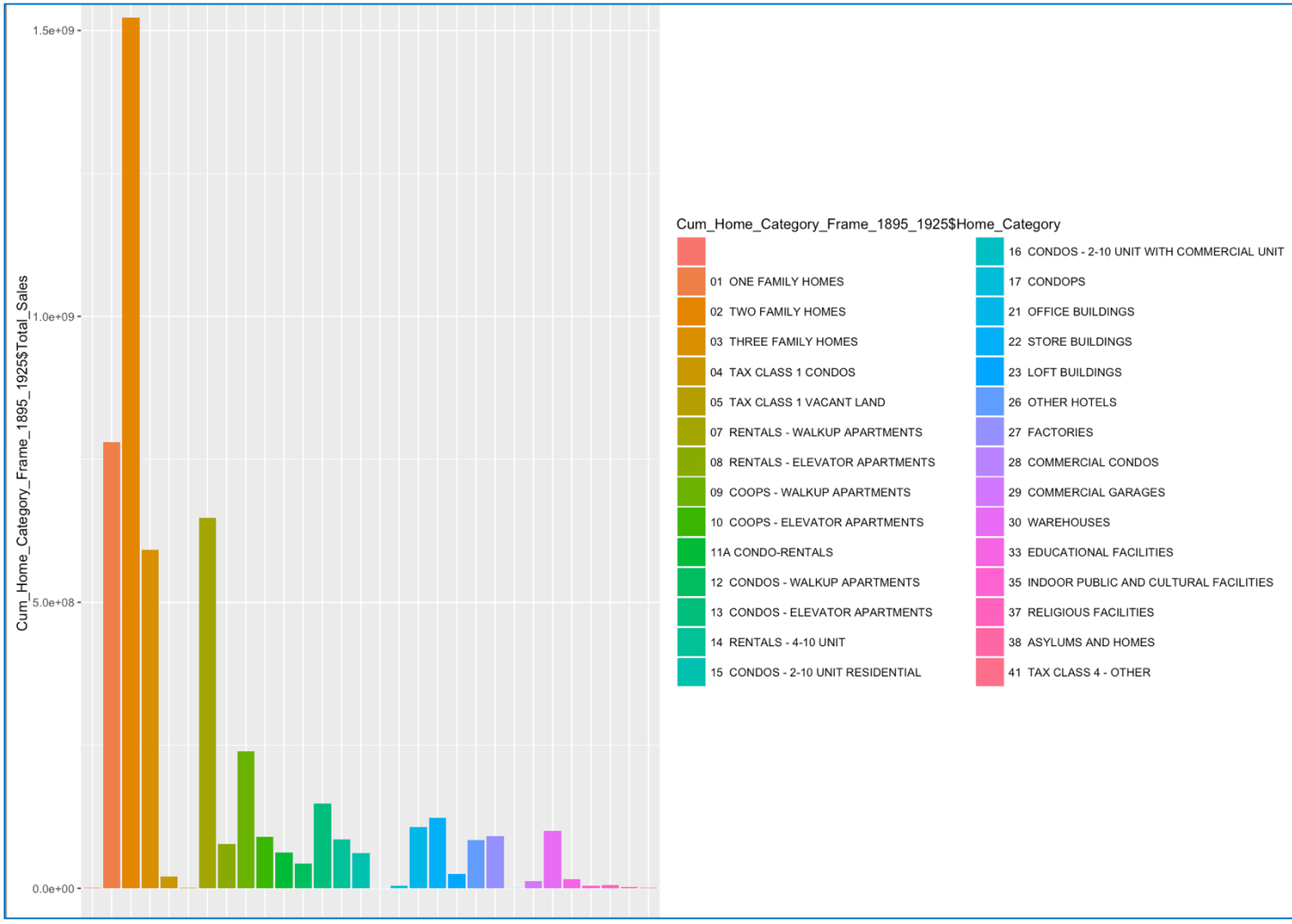
Step 3: Analysis in details

Now we take a step ahead and split the Home built dates into groups of 30 years each and analyze how the sales of Brooklyn are affecting over the years.

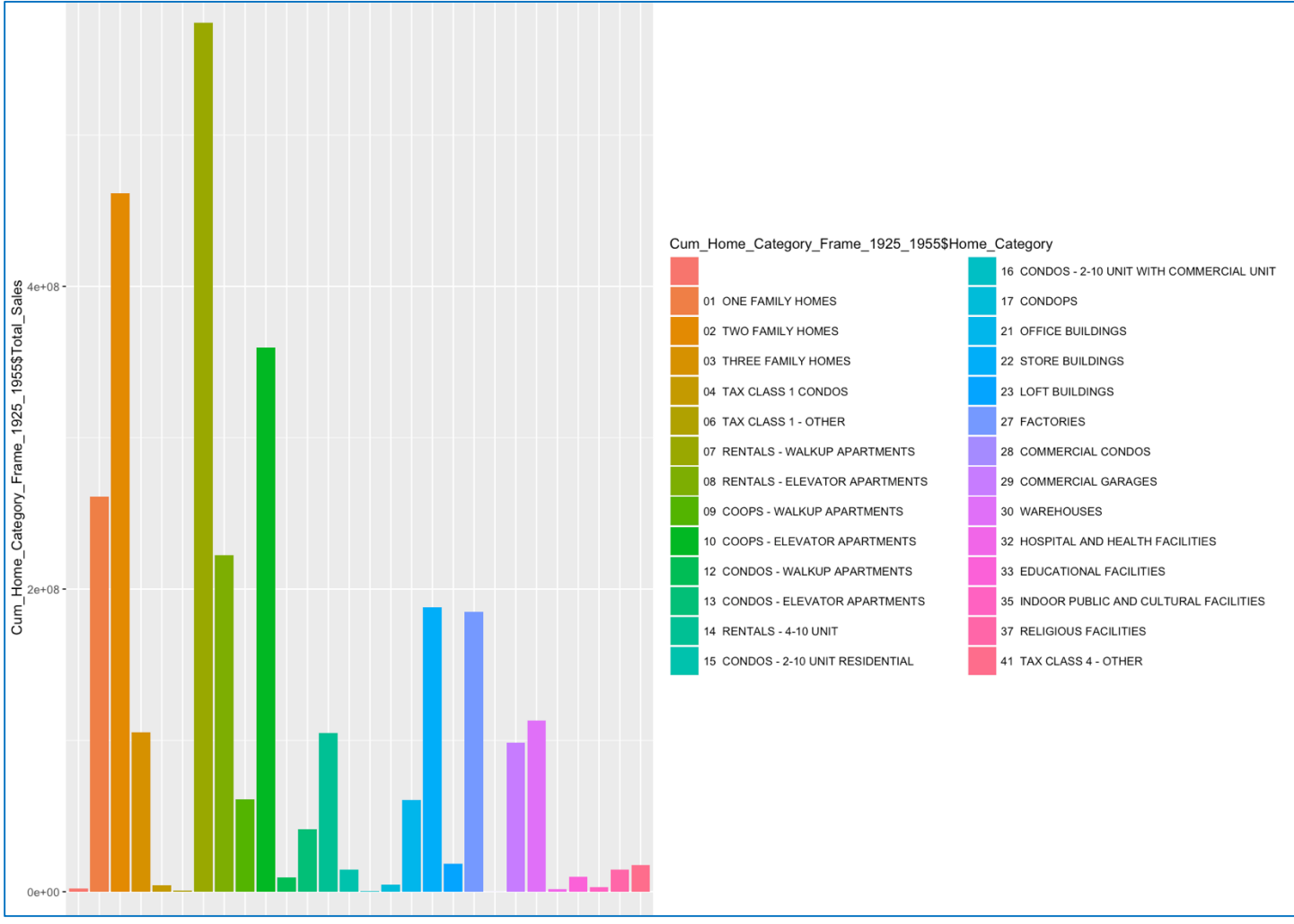
This will also give us intuition that which home categories got popular during these different time frames

Graphs for Cumulative Total Sales over the year vs Home Category

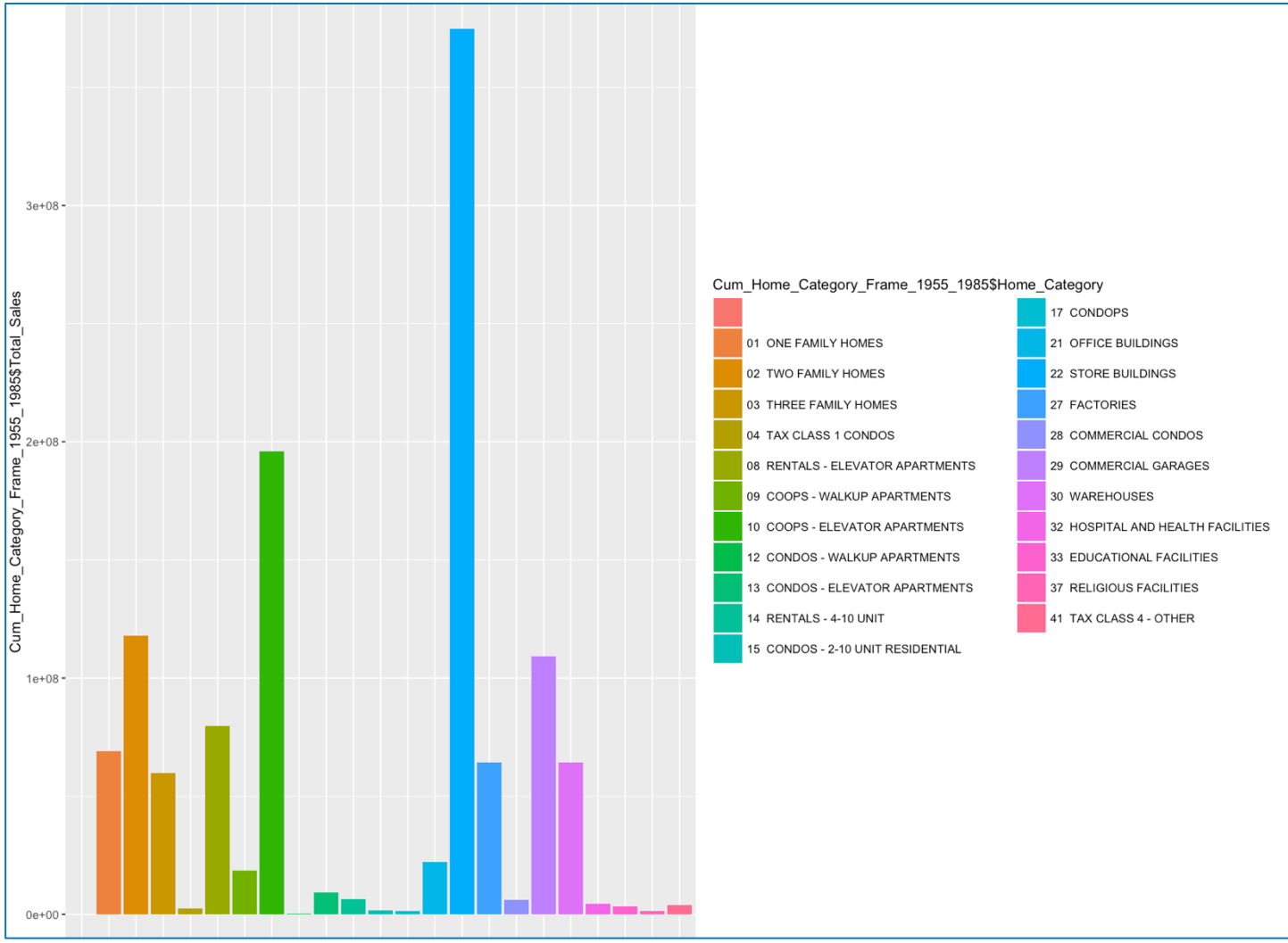
Period 1895 - 1925



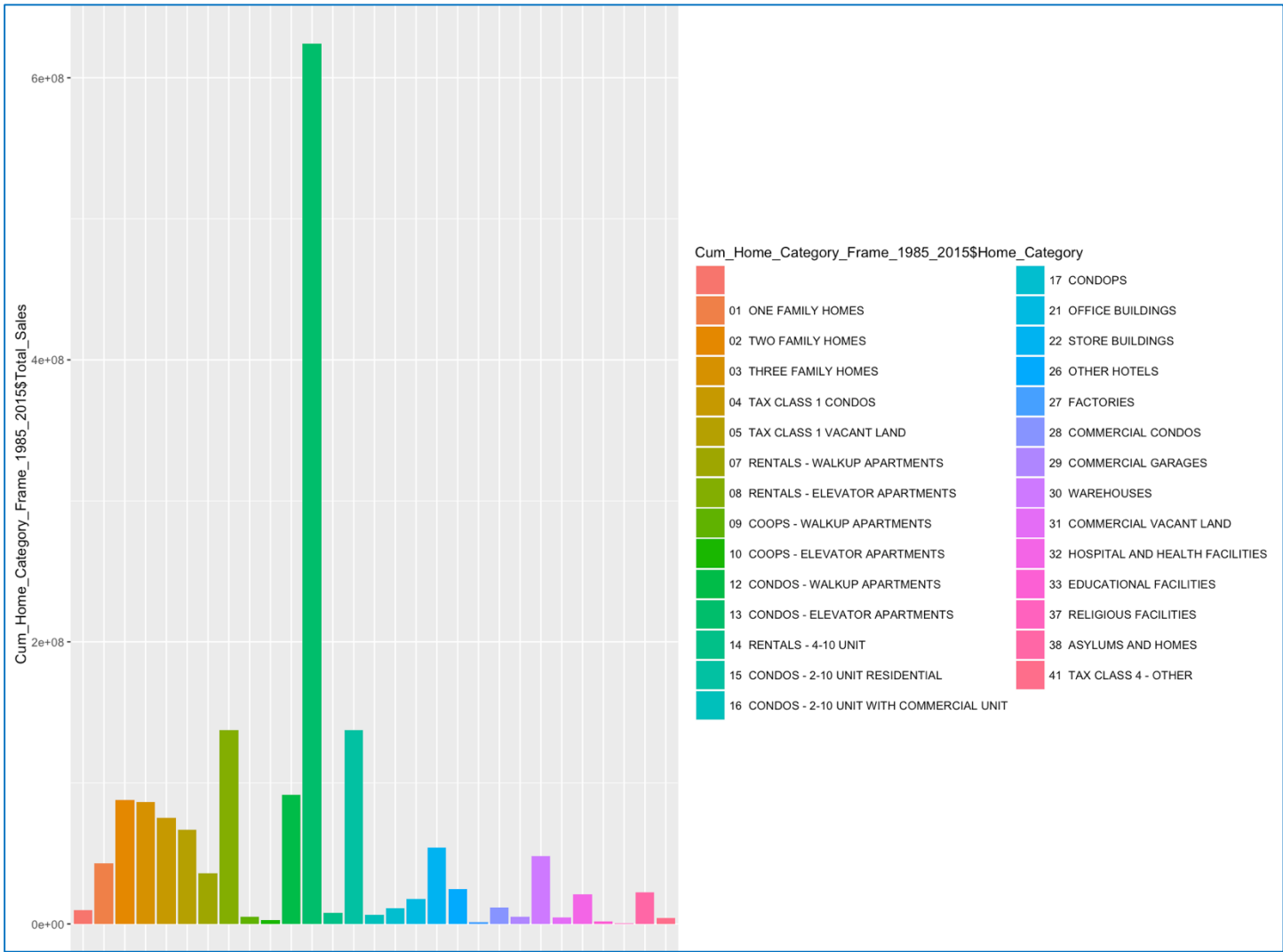
Period 1925-1955



Period 1955-1985



Period 1985-2015



Problem 3B: EDA on data for Manhattan, Queens, Bronx and State Island

In this problem we extend our analysis on data for other locations. We

Step 1:

In the first step we will do cleaning on data. For example, the data cleaning process for Manhattan is described below. We will follow the same process for other locations as well

```
4 # collecting and cleaning data for bronx
5 data_bronx<- read.xls("rollingsales_bronx.xls",pattern="BOROUGH")
6 names(data_bronx) <- tolower(names(data_bronx))
7 data_bronx$sale.price.n <- as.numeric(gsub("[^:digit:]", "", data_bronx$sale.price))
8 data_bronx$sale.date <- as.Date(data_bronx$sale.date)
9 data_bronx$year.built <- as.numeric(as.character(data_bronx$year.built))
10 data_bronx <- data_bronx[data_bronx$sale.price.n!=0,]
11 data_bronx <- data_bronx[data_bronx$year.built !=0, ]
12 data_bronx_frame <- data.frame(data_bronx$neighborhood, data_bronx$building.class.category,
13                               data_bronx$year.built, data_bronx$sale.price.n, data_bronx$sale.date)
14 data_bronx_frame$city_name <- "bronx"
15 colnames(data_bronx_frame) <- c("Neighborhood", "Home_Category", "Year_Built",
16                               "Sale_Price", "Sale_Date", "City_Name")
17 head(data_bronx_frame)
18
```

Step 2:

Now to analyze data for all the locations and compare them, we need to perform some aggregation so that its easy for us to do some analysis.

Since its sales data, so we decided to prepare cumulative sales report aggregated over the months for each location

```
86 # Total Monthly Sales for manhattan
87 manhattan_sale_dates <- data_manhattan_frame$Sale_Date
88 manhattan_sale_price <- data_manhattan_frame$Sale_Price
89 manhattan_sale_period <- as.yearmon(manhattan_sale_dates, "%b-%y")
90 manhattan_sale_period_frame <- data.frame(manhattan_sale_period, manhattan_sale_price)
91 Cum_manhattan_sale_period_frame <- aggregate(manhattan_sale_price ~ manhattan_sale_period,
92                                              manhattan_sale_period_frame, function(x) sum(as.numeric(x)))
93 colnames(Cum_manhattan_sale_period_frame) <- c("Sale_Period", "Total_Sales")
94 Cum_manhattan_sale_period_frame
95
```

Cumulative Sales Report of Manhattan

```
> Cum_manhattan_sale_period_frame
Sale_Period Total_Sales
1      Aug 2012  3156456343
2      Sep 2012  2431564752
3      Oct 2012  3501959004
4      Nov 2012  3055128566
5      Dec 2012  9767822979
6      Jan 2013  1970663705
7      Feb 2013  1699322318
8      Mar 2013  5194577564
9      Apr 2013  2612873487
10     May 2013  3225134016
11     Jun 2013  4800550442
12     Jul 2013  2948817514
13     Aug 2013  1045684307
```

Cumulative Sales Report of Bronx

```
> Cum_bronx_sale_period_frame
Sale_Period Total_Sales
1      Aug 2012  288923568
2      Sep 2012  155982875
3      Oct 2012  212528548
4      Nov 2012  190602548
5      Dec 2012  569294931
6      Jan 2013  102444352
7      Feb 2013  156180170
8      Mar 2013  166035985
9      Apr 2013  160981961
10     May 2013  192790621
11     Jun 2013  274493826
12     Jul 2013  266120898
13     Aug 2013  14165132
```

Cumulative Sales Report of Queens

```
> Cum_queens_sale_period_frame
Sale_Period Total_Sales
1      Aug 2012   585897994
2      Sep 2012   619968661
3      Oct 2012   524397483
4      Nov 2012   704928058
5      Dec 2012  1261615579
6      Jan 2013   523754224
7      Feb 2013   492639885
8      Mar 2013   471830280
9      Apr 2013   697097835
10     May 2013   698096417
11     Jun 2013   797794174
12     Jul 2013   742304012
13     Aug 2013   95004881
```

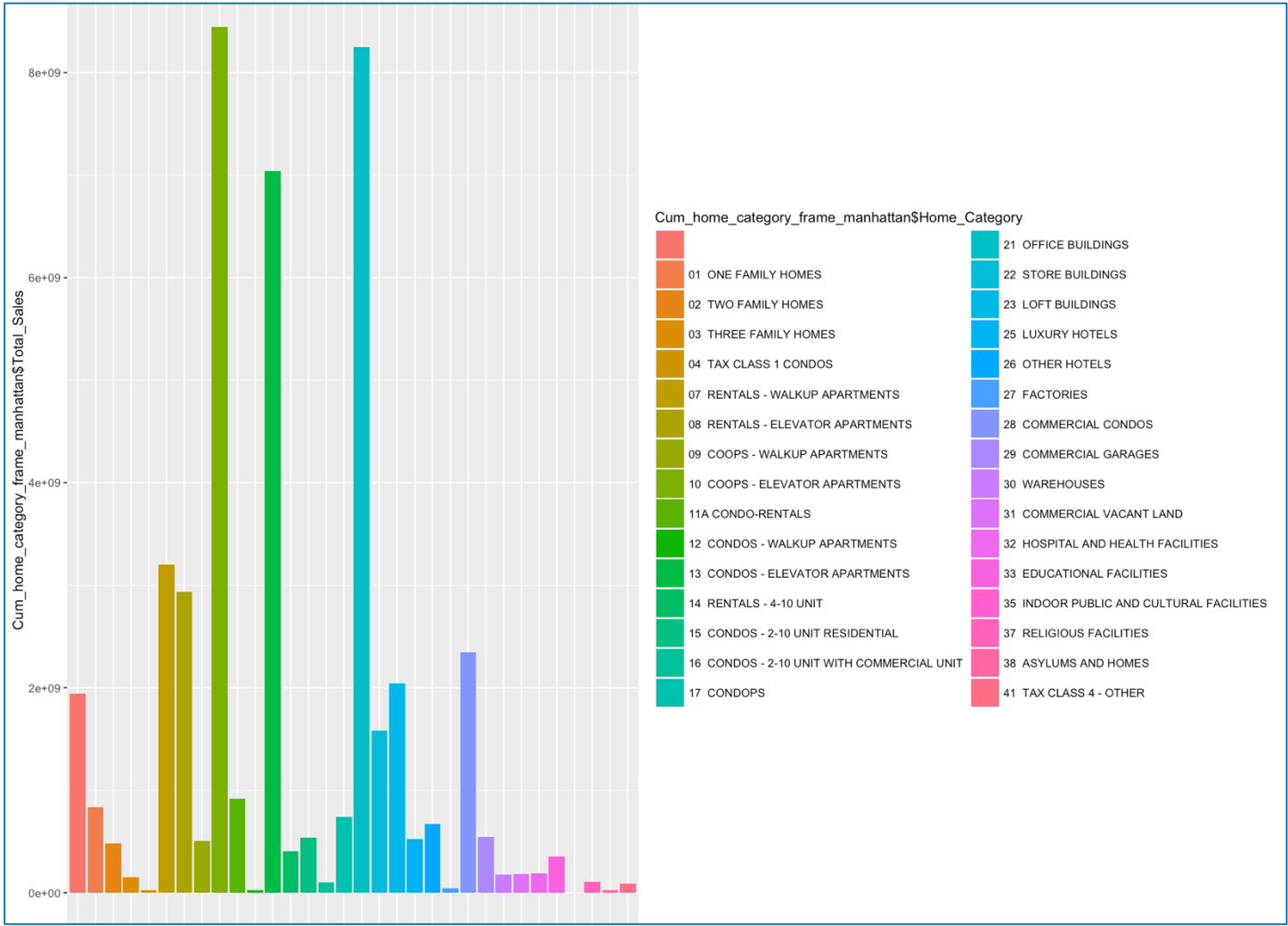
Cumulative Sales Report of Staten Island

```
> Cum_statenisland_sale_period_frame
Sale_Period Total_Sales
1      Aug 2012   154489576
2      Sep 2012   125426555
3      Oct 2012   130826584
4      Nov 2012   107127162
5      Dec 2012   153455706
6      Jan 2013   119104497
7      Feb 2013   119654868
8      Mar 2013   115236623
9      Apr 2013   133466945
10     May 2013   166663287
11     Jun 2013   176750027
12     Jul 2013   114421100
13     Aug 2013   2002500
```

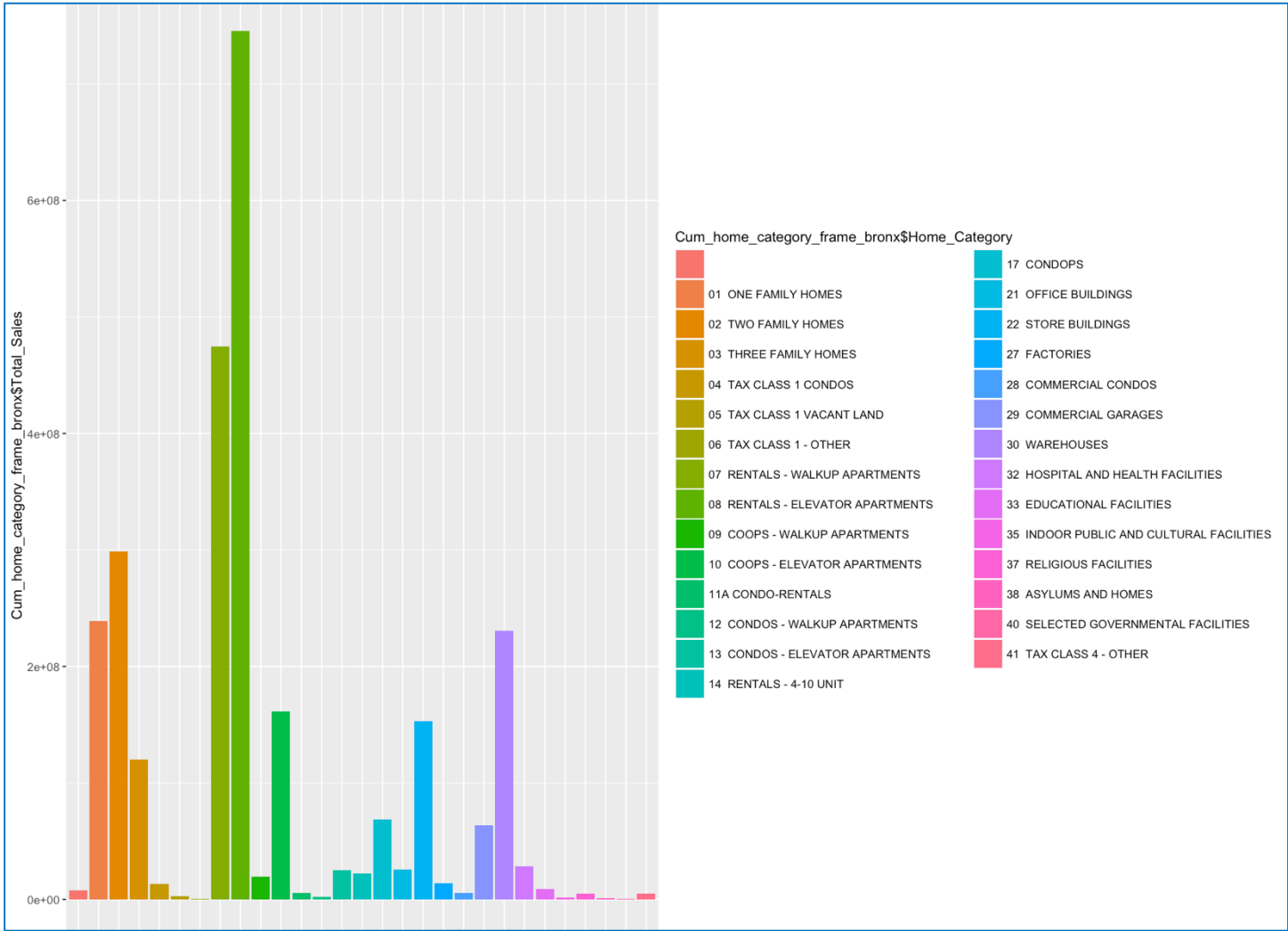
Step 3: Generating Graphs

Now we will plot graphs for the generated sales and analyze the cost of living

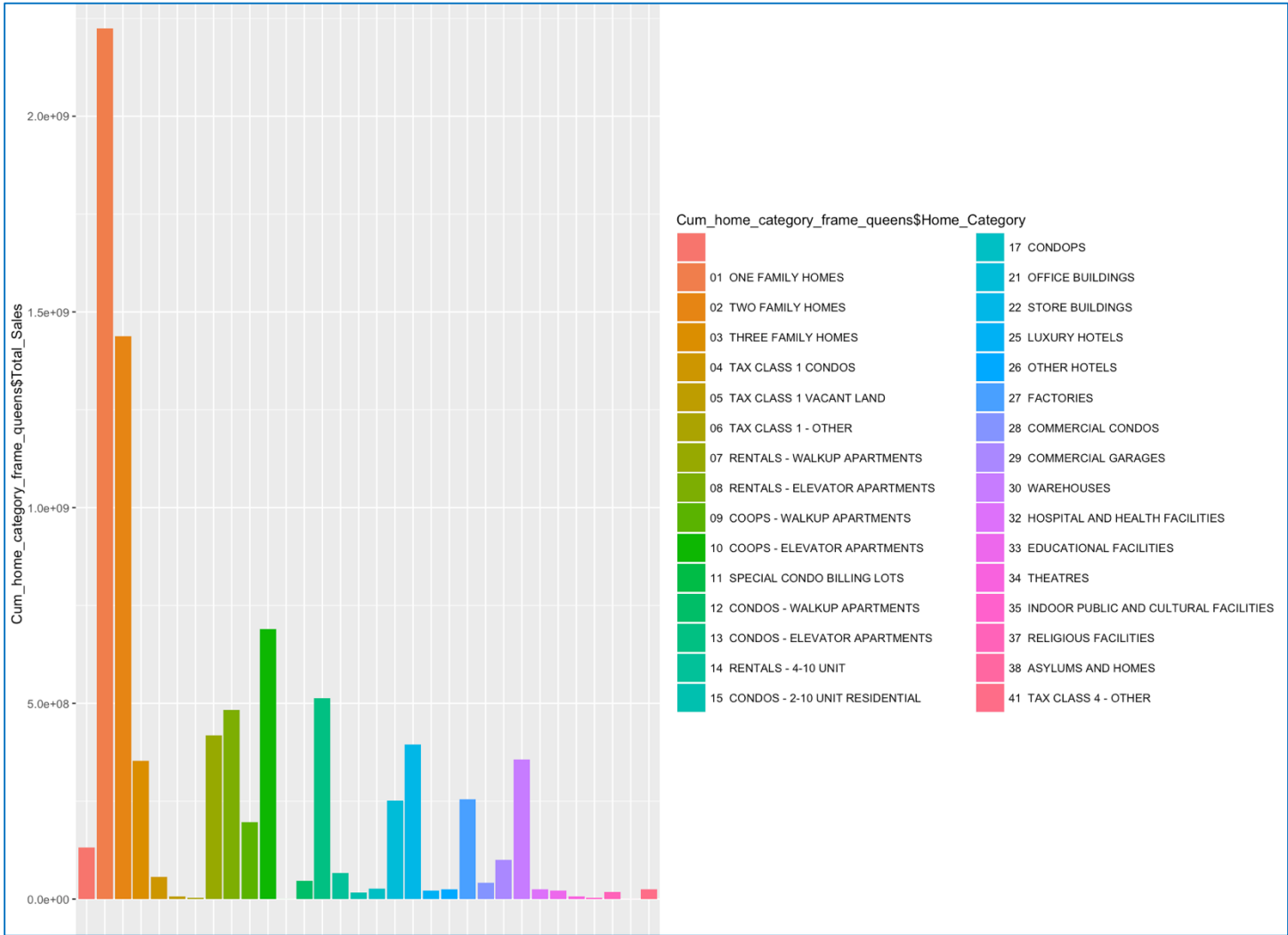
Graph on Cumulative Sales Report of Manhattan



Graph on Cumulative Sales Report of Bronx



Graph on Cumulative Sales Report of Queens



Graph on Cumulative Sales Report of Staten Island

