

NYC Real Estate Sales Analysis

Overview: In this analysis, we'll explore the NYC real estate dataset to uncover trends in sale prices across different neighborhoods and property types. We'll focus on understanding the relationship between property features (e.g., neighborhood, year built) and sale price.

Install appropriate libraries & packages

```
install.packages("tidyverse")
```

```
##
```

```
## The downloaded binary packages are in
```

```
##
```

```
/var/folders/k7/3hkxc3916d94sh54b7xgkrfh0000gn/T//RtmpFEhQ1y/downloaded_packages
```

```
install.packages("knitr")
```

```
##
```

```
## The downloaded binary packages are in
```

```
##
```

```
/var/folders/k7/3hkxc3916d94sh54b7xgkrfh0000gn/T//RtmpFEhQ1y/downloaded_packages
```

```
install.packages("scales")
```

```
##
```

```
## The downloaded binary packages are in
```

```
##
```

```
/var/folders/k7/3hkxc3916d94sh54b7xgkrfh0000gn/T//RtmpFEhQ1y/downloaded_packages
```

```
install.packages("ggplot2")
```

```
##
```

```
## The downloaded binary packages are in
```

```
##
```

```
/var/folders/k7/3hkxc3916d94sh54b7xgkrfh0000gn/T//RtmpFEhQ1y/downloaded_packages
```

```
library(ggplot2)
```

```
library(tidyverse)
```

```
library(knitr)
```

```
library(readr)
```

```
library(scales)
```

```
##
## Attaching package: 'scales'

## The following object is masked from 'package:purrr':
##
##   discard

## The following object is masked from 'package:readr':
##
##   col_factor

library(dplyr)
library(lubridate)
```

Import document

```
nyc_dataset <- read_csv("~/Desktop/Data sets/nyc-rolling-sales.csv")

## New names:
## Rows: 84548 Columns: 22
## — Column specification
## _____
Delimiter: "," chr
## (10): NEIGHBORHOOD, BUILDING CLASS CATEGORY, TAX CLASS AT PRESENT, BUIL...
dbl
## (10): ...1, BOROUGH, BLOCK, LOT, ZIP CODE, RESIDENTIAL UNITS, COMMERCIA...
lgl
## (1): EASE-MENT dtm (1): SALE DATE
## i Use `spec()` to retrieve the full column specification for this data. i
## Specify the column types or set `show_col_types = FALSE` to quiet this
message.
## • `` -> `...1`
```

Clean and prepare data

```
nyc_dataset <- nyc_dataset %>%
  select(-`EASE-MENT`, -`APARTMENT NUMBER`)

colnames(nyc_dataset) <- c("id", "borough", "neighborhood",
"building_class_category",
"tax_class_present", "block", "lot",
"building_class_present",
"address", "zip_code", "residential_units",
"commercial_units",
"total_units", "land_square_feet",
"gross_square_feet",
"year_built", "tax_class_sale_time",
"building_class_sale_time",
"sale_price", "sale_date")
```

```
# Convert sale_price to numeric and clean data
nyc_dataset <- nyc_dataset %>%
  mutate(sale_price = as.numeric(gsub("[^0-9.]", "", sale_price))) %>%
  filter(!is.na(sale_price))
```

Quick overview

```
colnames(nyc_dataset) #List of column names
```

```
## [1] "id" "borough"
## [3] "neighborhood" "building_class_category"
## [5] "tax_class_present" "block"
## [7] "lot" "building_class_present"
## [9] "address" "zip_code"
## [11] "residential_units" "commercial_units"
## [13] "total_units" "land_square_feet"
## [15] "gross_square_feet" "year_built"
## [17] "tax_class_sale_time" "building_class_sale_time"
## [19] "sale_price" "sale_date"
```

```
ncol(nyc_dataset) #How many columns are in data frame?
```

```
## [1] 20
```

```
nrow(nyc_dataset) #How many rows are in data frame?
```

```
## [1] 69987
```

```
dim(nyc_dataset) #Dimensions of the data frame?
```

```
## [1] 69987 20
```

```
head(nyc_dataset) #See the first 6 rows of data frame.
```

```
## # A tibble: 6 × 20
##      id borough neighborhood building_class_category tax_class_present
##      <dbl> <dbl> <chr>          <chr>          <chr>
## 1      4      1 ALPHABET CITY 07 RENTALS - WALKUP APART... 2A
392
## 2      7      1 ALPHABET CITY 07 RENTALS - WALKUP APART... 2B
402
## 3      8      1 ALPHABET CITY 07 RENTALS - WALKUP APART... 2A
404
## 4     10      1 ALPHABET CITY 07 RENTALS - WALKUP APART... 2B
406
## 5     13      1 ALPHABET CITY 08 RENTALS - ELEVATOR APA... 2
387
## 6     15      1 ALPHABET CITY 08 RENTALS - ELEVATOR APA... 2B
400
```

```
## # i 14 more variables: lot <dbl>, building_class_present <chr>, address
<chr>,
## #   zip_code <dbl>, residential_units <dbl>, commercial_units <dbl>,
## #   total_units <dbl>, land_square_feet <chr>, gross_square_feet <chr>,
## #   year_built <dbl>, tax_class_sale_time <dbl>,
## #   building_class_sale_time <chr>, sale_price <dbl>, sale_date <dtm>
```

```
str(nyc_dataset) #See list of columns and data types (numeric, character,
etc)
```

```
## tibble [69,987 × 20] (S3: tbl_df/tbl/data.frame)
## $ id : num [1:69987] 4 7 8 10 13 15 16 17 18 19 ...
## $ borough : num [1:69987] 1 1 1 1 1 1 1 1 1 1 ...
## $ neighborhood : chr [1:69987] "ALPHABET CITY" "ALPHABET CITY"
"ALPHABET CITY" "ALPHABET CITY" ...
## $ building_class_category : chr [1:69987] "07 RENTALS - WALKUP
APARTMENTS" "07 RENTALS - WALKUP APARTMENTS" "07 RENTALS - WALKUP APARTMENTS"
"07 RENTALS - WALKUP APARTMENTS" ...
## $ tax_class_present : chr [1:69987] "2A" "2B" "2A" "2B" ...
## $ block : num [1:69987] 392 402 404 406 387 400 373 373
373 373 ...
## $ lot : num [1:69987] 6 21 55 32 153 21 40 40 40 40
...
## $ building_class_present : chr [1:69987] "C2" "C4" "C2" "C4" ...
## $ address : chr [1:69987] "153 AVENUE B" "154 EAST 7TH
STREET" "301 EAST 10TH STREET" "210 AVENUE B" ...
## $ zip_code : num [1:69987] 10009 10009 10009 10009 10009
...
## $ residential_units : num [1:69987] 5 10 6 8 24 10 0 0 0 0 ...
## $ commercial_units : num [1:69987] 0 0 0 0 0 0 0 0 0 0 ...
## $ total_units : num [1:69987] 5 10 6 8 24 10 0 0 0 0 ...
## $ land_square_feet : chr [1:69987] "1633" "2272" "2369" "1750" ...
## $ gross_square_feet : chr [1:69987] "6440" "6794" "4615" "4226" ...
## $ year_built : num [1:69987] 1900 1913 1900 1920 1920 ...
## $ tax_class_sale_time : num [1:69987] 2 2 2 2 2 2 2 2 2 2 ...
## $ building_class_sale_time: chr [1:69987] "C2" "C4" "C2" "C4" ...
## $ sale_price : num [1:69987] 6625000 3936272 8000000 3192840
16232000 ...
## $ sale_date : POSIXct[1:69987], format: "2017-07-19"
"2016-09-23" ...
```

```
summary(nyc_dataset) #Statistical summary of data. Mainly for numerics
```

```
##      id      borough      neighborhood
building_class_category
## Min.   : 4      Min.   :1.000      Length:69987      Length:69987
## 1st Qu.: 4182    1st Qu.:2.000      Class :character  Class :character
## Median : 8989    Median :3.000      Mode  :character  Mode  :character
## Mean   :10288    Mean   :2.922
## 3rd Qu.:15874    3rd Qu.:4.000
## Max.   :26738    Max.   :5.000
```

```

## tax_class_present      block      lot
building_class_present
## Length:69987      Min.   :    1   Min.   :    1.0   Length:69987
## Class :character   1st Qu.: 1348   1st Qu.:   22.0   Class :character
## Mode  :character   Median : 3378   Median :   50.0   Mode  :character
##                      Mean  : 4196   Mean  :  373.8
##                      3rd Qu.: 6186   3rd Qu.:  709.0
##                      Max.   :16319   Max.   :9106.0
##      address      zip_code      residential_units commercial_units
## Length:69987      Min.   :    0   Min.   :    0.0   Min.   :    0.0000
## Class :character   1st Qu.:10306   1st Qu.:    0.0   1st Qu.:    0.0000
## Mode  :character   Median :11209   Median :    1.0   Median :    0.0000
##                      Mean   :10741   Mean   :    1.9   Mean   :    0.1725
##                      3rd Qu.:11249   3rd Qu.:    2.0   3rd Qu.:    0.0000
##                      Max.   :11694   Max.   :1844.0   Max.   :2261.0000
##      total_units      land_square_feet      gross_square_feet      year_built
## Min.   :    0.000      Length:69987      Length:69987      Min.   :    0
## 1st Qu.:    0.000      Class :character      Class :character      1st Qu.:1920
## Median :    1.000      Mode  :character      Mode  :character      Median :1937
## Mean   :    2.092                                     Mean   :1799
## 3rd Qu.:    2.000                                     3rd Qu.:1965
## Max.   :2261.000                                     Max.   :2017
## tax_class_sale_time building_class_sale_time      sale_price
## Min.   :1.000      Length:69987      Min.   :0.000e+00
## 1st Qu.:1.000      Class :character      1st Qu.:2.250e+05
## Median :2.000      Mode  :character      Median :5.300e+05
## Mean   :1.642                                     Mean   :1.276e+06
## 3rd Qu.:2.000                                     3rd Qu.:9.500e+05
## Max.   :4.000                                     Max.   :2.210e+09
##      sale_date
## Min.   :2016-09-01 00:00:00.00
## 1st Qu.:2016-11-30 00:00:00.00
## Median :2017-02-28 00:00:00.00
## Mean   :2017-02-27 21:22:54.48
## 3rd Qu.:2017-05-31 00:00:00.00
## Max.   :2017-08-31 00:00:00.00

```

`names(nyc_dataset)`

```

## [1] "id"                "borough"
## [3] "neighborhood"      "building_class_category"
## [5] "tax_class_present" "block"
## [7] "lot"               "building_class_present"
## [9] "address"           "zip_code"
## [11] "residential_units" "commercial_units"
## [13] "total_units"       "land_square_feet"
## [15] "gross_square_feet" "year_built"
## [17] "tax_class_sale_time" "building_class_sale_time"
## [19] "sale_price"        "sale_date"

```

Quick glance summary

```
# Increase scipen to avoid scientific notation
options(scipen = 999)
```

```
# Summary statistics for sale price
summary(nyc_dataset$sale_price)
```

```
##           Min.      1st Qu.      Median      Mean      3rd Qu.      Max.
##           0.00    225000.0    530000.0   1276456.0    950000.0  2210000000.0
```

Visualizations

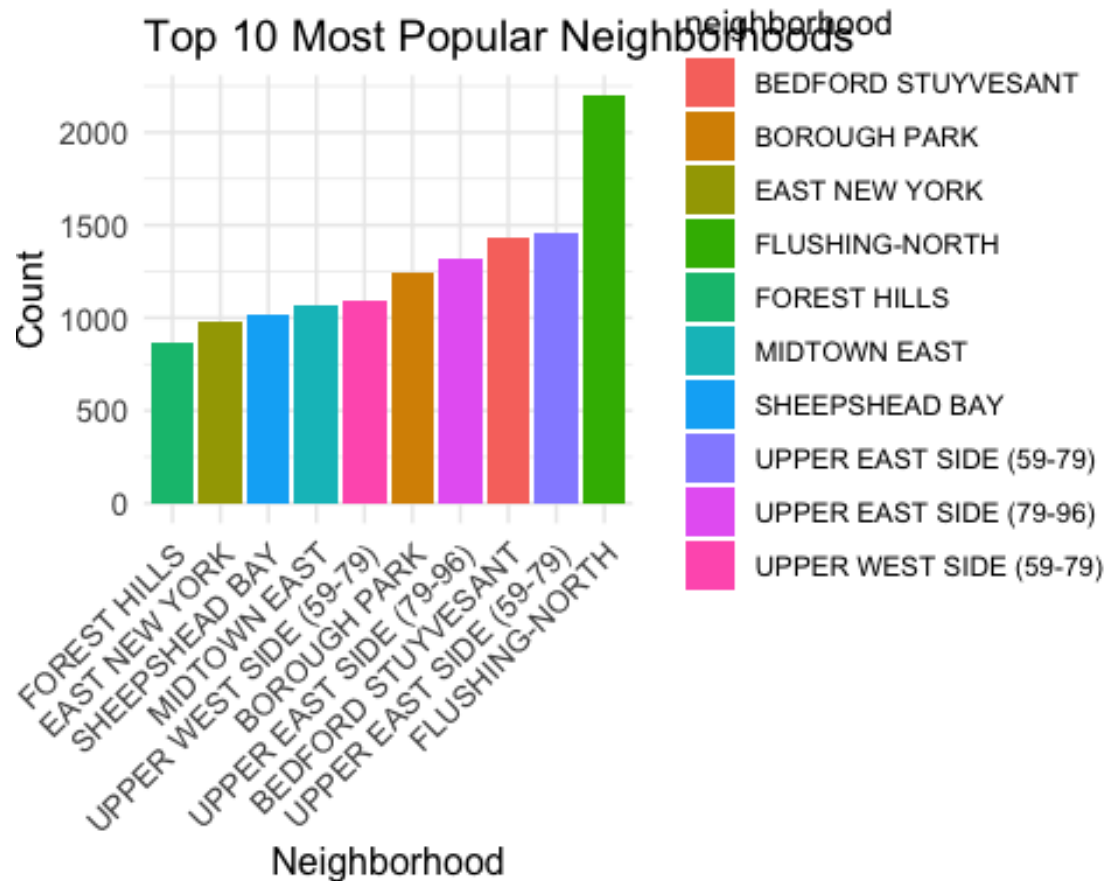
```
# Top 10 most popular neighborhoods
```

```
neighborhood_counts <- nyc_dataset %>%
  count(neighborhood)
```

```
top_neighborhoods <- neighborhood_counts %>%
  arrange(desc(n)) %>%
  top_n(10)
```

```
## Selecting by n
```

```
ggplot(top_neighborhoods, aes(x = reorder(neighborhood, n), y = n, fill =
neighborhood)) +
  geom_bar(stat = "identity") +
  labs(title = "Top 10 Most Popular Neighborhoods",
       x = "Neighborhood",
       y = "Count") +
  theme_minimal() +
  theme(
    axis.text.x = element_text(angle = 45, hjust = 1, size = 10),
    axis.title = element_text(size = 12),
    plot.title = element_text(size = 14),
    axis.text.y = element_text(size = 10),
    plot.margin = margin(5, 5, 5, 0, "pt")
  )
```

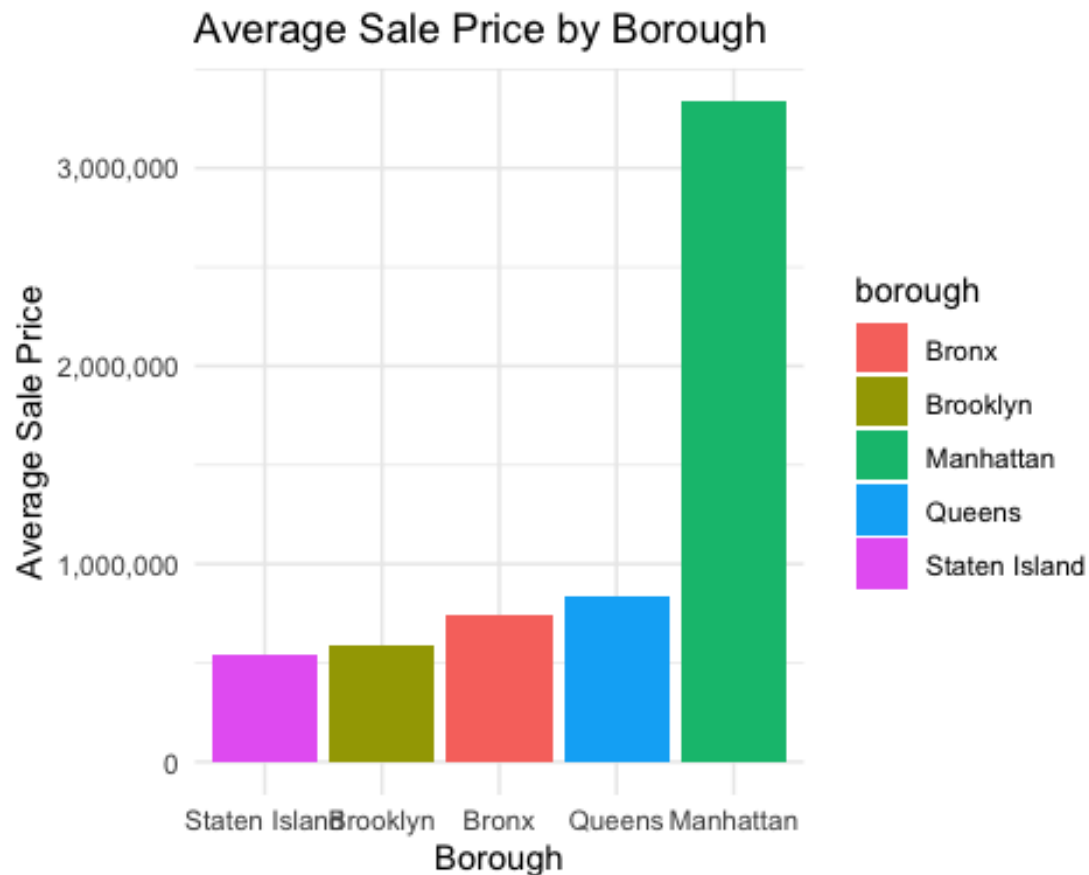


```
# Average Sale Price by Borough
nyc_dataset <- nyc_dataset %>%
  mutate(borough = recode(borough,
    `1` = "Manhattan",
    `2` = "Brooklyn",
    `3` = "Queens",
    `4` = "Bronx",
    `5` = "Staten Island"))

average_sale_price <- nyc_dataset %>%
  group_by(borough) %>%
  summarize(avg_sale_price = mean(sale_price, na.rm = TRUE))

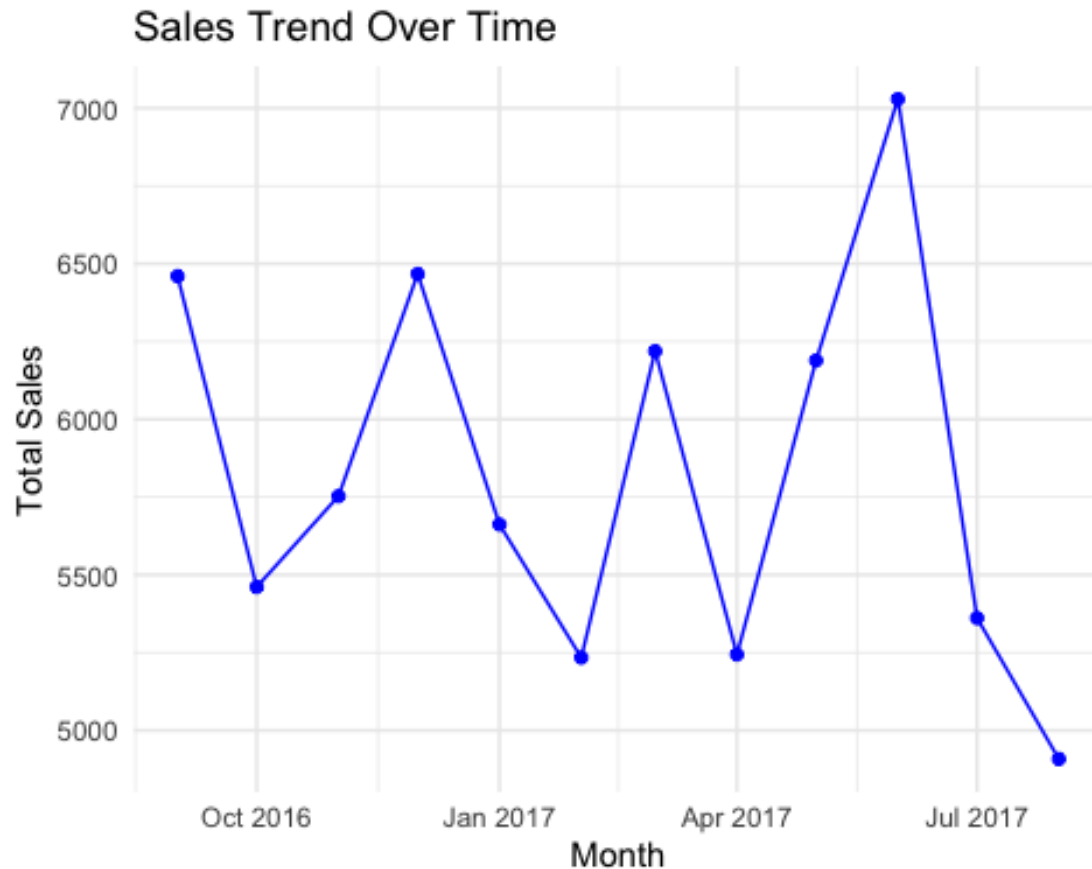
ggplot(average_sale_price, aes(x = reorder(borough, avg_sale_price), y =
  avg_sale_price, fill = borough)) +
  geom_bar(stat = "identity") +
  labs(title = "Average Sale Price by Borough",
    x = "Borough",
    y = "Average Sale Price") +
  scale_y_continuous(labels = scales::comma) +
  theme(axis.text.x = element_text(angle = 45, hjust = 1, size = 10),
    axis.title = element_text(size = 12),
    plot.title = element_text(size = 14),
```

```
plot.margin = margin(5, 5, 5, 0, "pt")) +
theme_minimal()
```



```
# Sales trend over time
monthly_sales <- nyc_dataset %>%
  mutate(month = floor_date(sale_date, "month")) %>%
  group_by(month) %>%
  summarize(total_sales = n())

ggplot(monthly_sales, aes(x = month, y = total_sales)) +
  geom_line(color = "blue") +
  geom_point(color = "blue") +
  labs(title = "Sales Trend Over Time",
       x = "Month",
       y = "Total Sales") +
  theme_minimal()
```

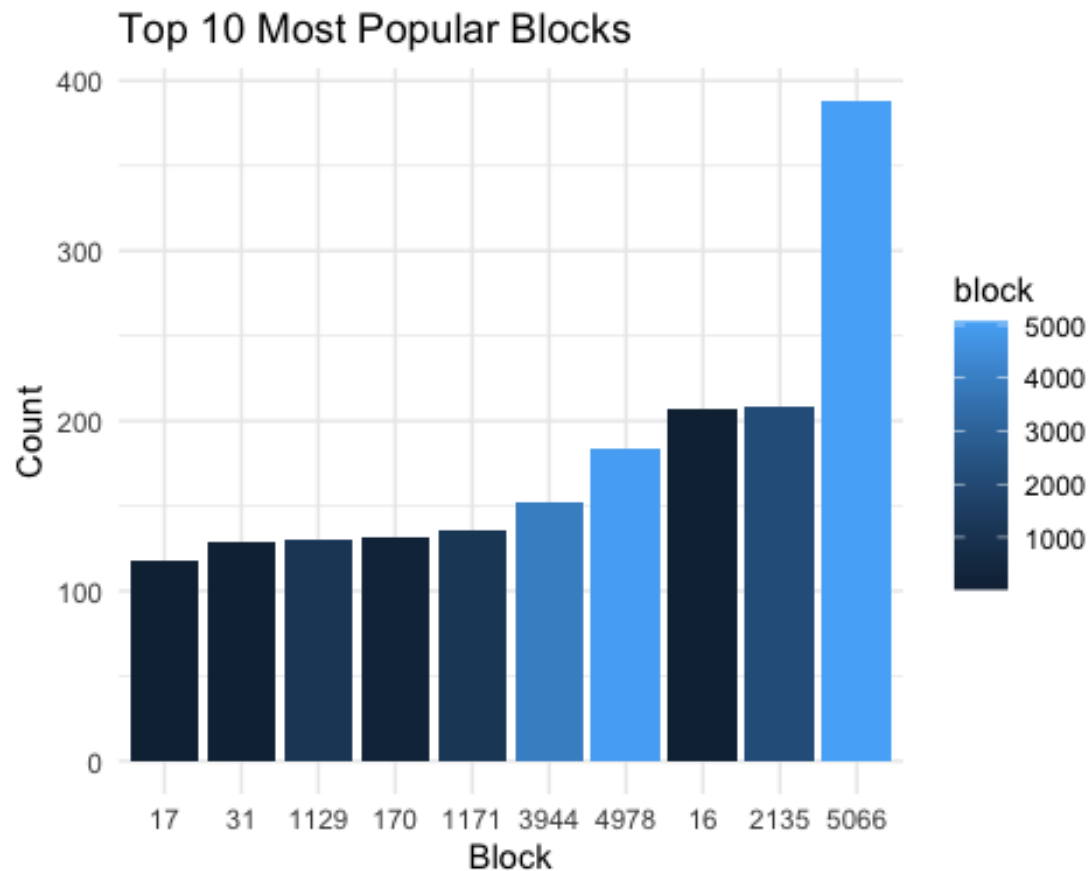



```
# Top 10 most popular blocks
block_counts <- nyc_dataset %>%
  count(block)

top_blocks <- block_counts %>%
  arrange(desc(n)) %>%
  top_n(10)

## Selecting by n

ggplot(top_blocks, aes(x = reorder(block, n), y = n, fill = block)) +
  geom_bar(stat = "identity") +
  labs(title = "Top 10 Most Popular Blocks",
       x = "Block",
       y = "Count") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1, size = 10),
        axis.title = element_text(size = 12),
        plot.title = element_text(size = 14),
        plot.margin = margin(5, 5, 5, 0, "pt")) +
  theme_minimal()
```



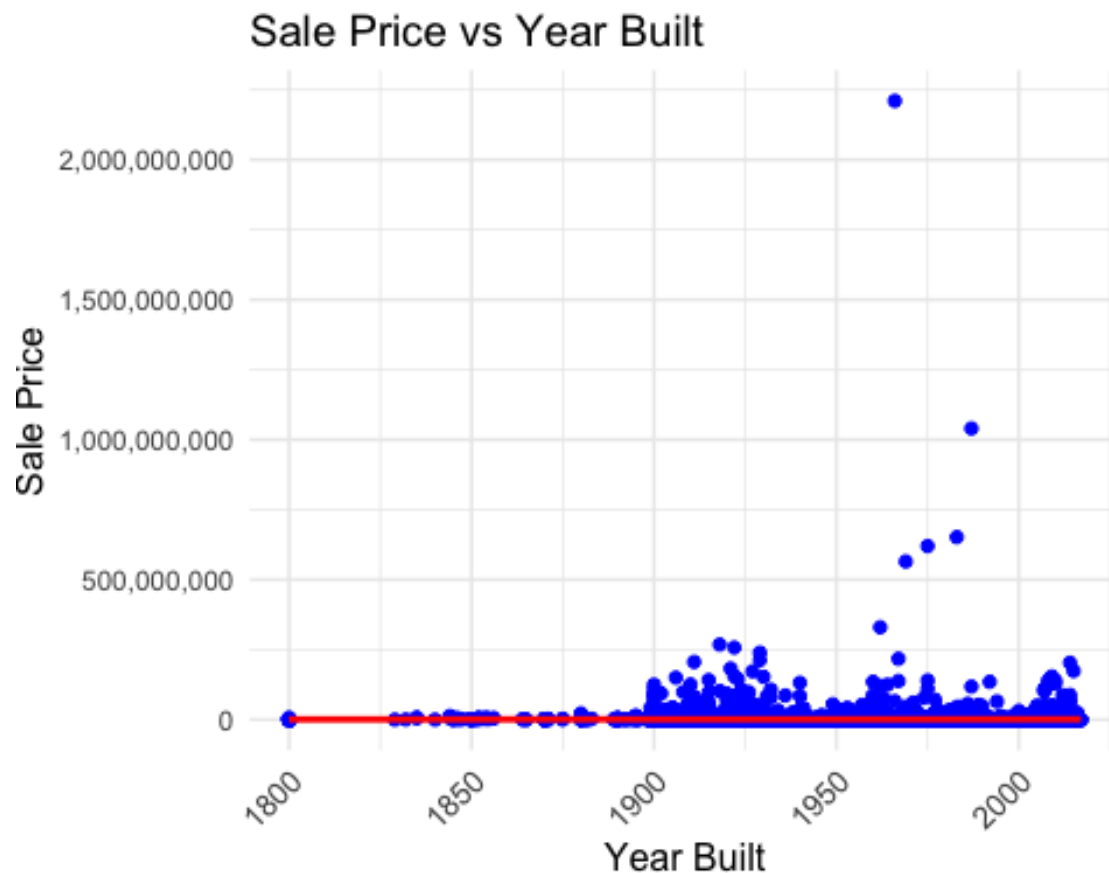
```
# Sale Price by Year Built
filtered_nyc_dataset <- nyc_dataset %>%
  filter(year_built != 0)

ggplot(filtered_nyc_dataset, aes(x = year_built, y = sale_price)) +
  geom_point(color = "blue") +
  geom_smooth(method = "lm", color = "red") +
  labs(title = "Sale Price vs Year Built",
       x = "Year Built",
       y = "Sale Price") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1, size = 10),
        axis.title = element_text(size = 12),
        plot.title = element_text(size = 14),
        plot.margin = margin(5, 5, 5, 0, "pt")) +
  scale_x_continuous(limits = c(1800, NA)) +
  scale_y_continuous(labels = scales::comma)

## `geom_smooth()` using formula = 'y ~ x'

## Warning: Removed 1 row containing non-finite outside the scale range
## (`stat_smooth()`).
```

```
## Warning: Removed 1 row containing missing values or values outside the
scale range
## (`geom_point()`).
```



```
# Sale Price vs Residential Units
ggplot(nyc_dataset, aes(x = residential_units, y = sale_price)) +
  geom_point(color = "blue") +
  geom_smooth(method = "lm", color = "red") +
  labs(title = "Sale Price vs Residential Units",
       x = "Residential Units",
       y = "Sale Price") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1, size = 10),
        axis.title = element_text(size = 12),
        plot.title = element_text(size = 14),
        plot.margin = margin(5, 5, 5, 0, "pt")) +
  scale_y_continuous(labels = scales::comma)

## `geom_smooth()` using formula = 'y ~ x'
```



Conclusion: Looking at the data, Manhattan is, on average, the most expensive and most popular borough in New York City, and the 5066 block is the most popular block within NYC. Sales trends seem to drop in the months of October, February, and April, with a significant drop in August of 2017. The year built seems to have a very slight impact on sale price compared to the number of residential units, which has a much greater impact.