

HDB Resale Price

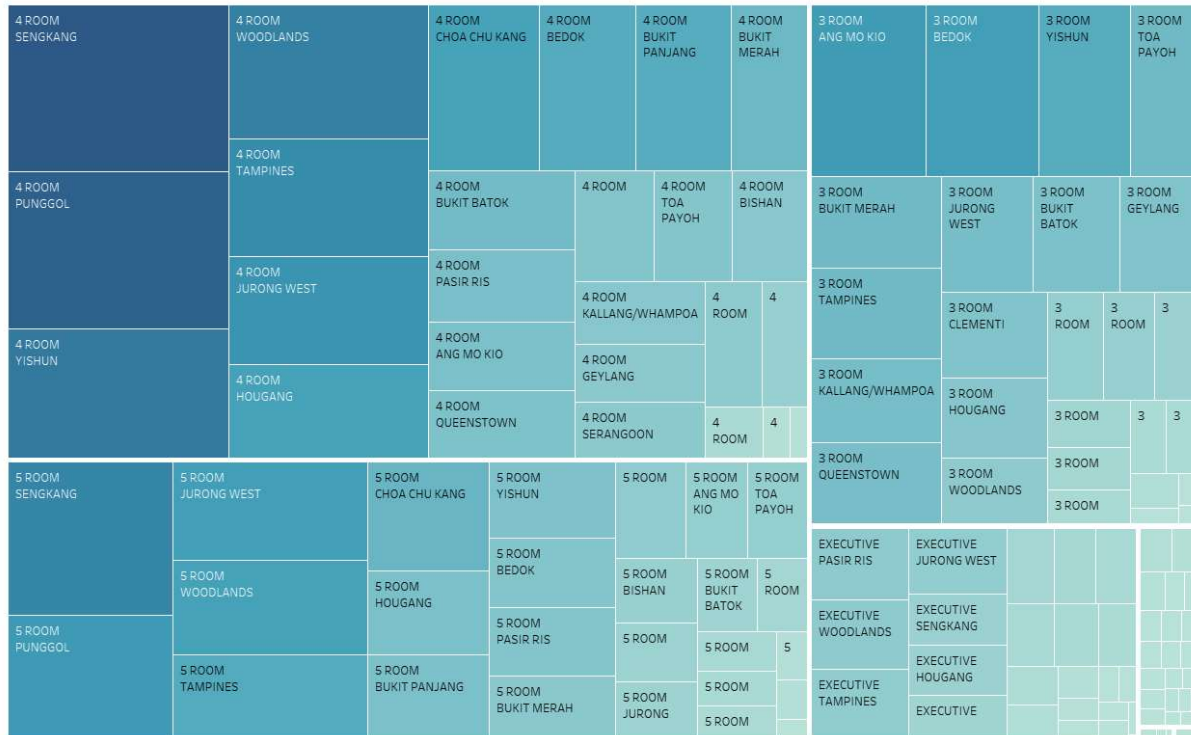
Machine Learning Linear Regression

Problem Statement

- With HDB resale price on the rise, there is a need to better predict sale price in order to budget for future purchase.
- Price prediction is based on several factors such as location, flat type and lease remaining.
- Dataset is obtained from singstat.gov.sg and data covers transaction from Jan 2017 to Jan 2022.

Data Observations

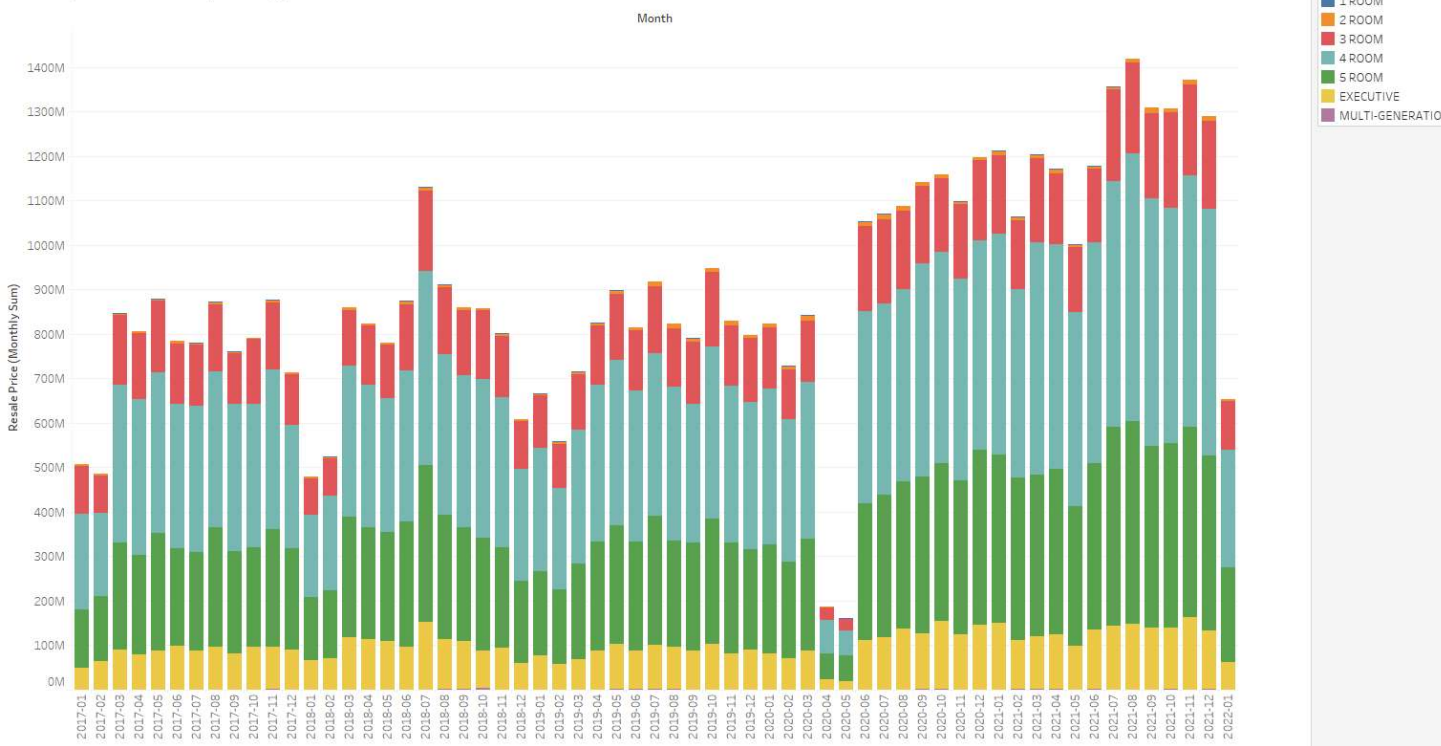
Flat Type and Location



- Majority of resale flat type are 4-room and 5 room in Sengkang, Punggol, Yishun, Woodlands, Tampines.

Data Observations

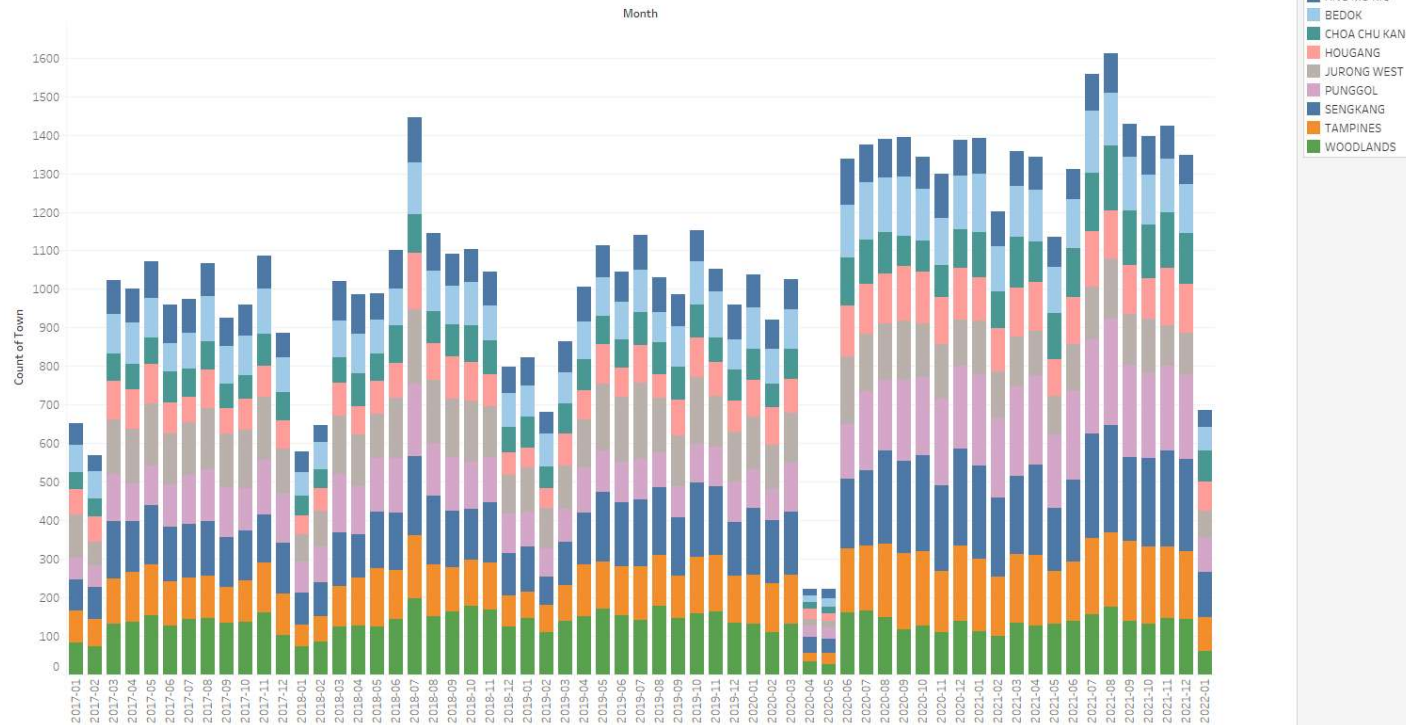
Monthly Transaction by Flat Type



- 4 room and 5 room make up the bulk of sale.
- Sale were low in April/May 2020 due to circuit breaker
- Trend of rising prices over the years

Data Observations

Monthly Transaction by Town



- Number of transaction in Singapore's 10 most populous town.
- Sale were low in April/May 2020 due to circuit breaker
- Trend of rising prices over the years
- Sengkang and Punggol had the most number of sale

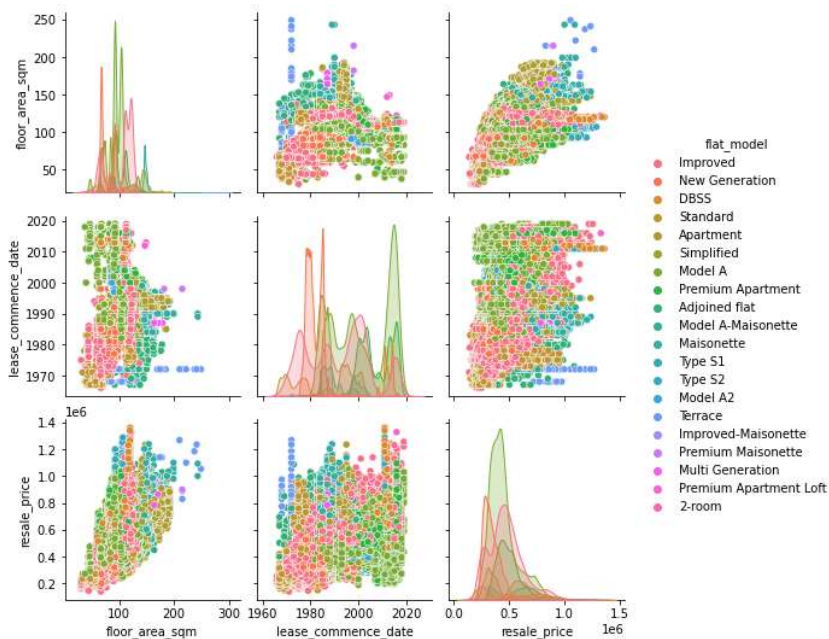
EDA

```
Shape of Dataframe: (117932, 11)
Checking if columns have null values.
month            0
town             0
flat_type        0
block            0
street_name      0
storey_range     0
floor_area_sqm   0
flat_model       0
lease_commence_date 0
remaining_lease  0
resale_price     0
dtype: int64
Index(['month', 'town', 'flat_type', 'block', 'street_name', 'storey_range',
      'floor_area_sqm', 'flat_model', 'lease_commence_date',
      'remaining_lease', 'resale_price'],
      dtype='object')
Unique values in each column
month            61
town             26
flat_type         7
block           2566
street_name      556
storey_range     17
floor_area_sqm   169
flat_model       20
lease_commence_date 54
remaining_lease   647
resale_price     3000
dtype: int64
      floor_area_sqm  lease_commence_date  resale_price
count  117932.000000      117932.000000  1.179320e+05
mean     97.843225        1995.091909  4.604087e+05
std     24.124616         13.441321  1.592940e+05
min     31.000000        1966.000000  1.400000e+05
25%     82.000000        1985.000000  3.450000e+05
50%     94.000000        1996.000000  4.300000e+05
75%    113.000000        2005.000000  5.400000e+05
max    249.000000        2019.000000  1.360000e+06
```

```
[5 rows x 11 columns]
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 117932 entries, 0 to 117931
Data columns (total 11 columns):
#   Column              Non-Null Count  Dtype
---  ---
0   month               117932 non-null  object
1   town               117932 non-null  object
2   flat_type          117932 non-null  object
3   block              117932 non-null  object
4   street_name        117932 non-null  object
5   storey_range       117932 non-null  object
6   floor_area_sqm     117932 non-null  float64
7   flat_model         117932 non-null  object
8   lease_commence_date 117932 non-null  int64
9   remaining_lease    117932 non-null  object
10  resale_price        117932 non-null  float64
dtypes: float64(2), int64(1), object(8)
```

- Dataframe has 117,932 entries with 11 columns
- No null values found.
- Datatype in Dataframe contains object, float and integer.

EDA



- Strong correlation of 0.621 between floor_area_sqm and resale_price. This suggests that a higher floor area will result in a higher resale price which is generally true.
- Lease commence date also has a correlation with resale price. Generally, a later lease commence date will fetch a higher resale price as there are more years remaining in the flat.
- Floor area and lease commence date do not seem to have a correlation given a score of 0.151.

```
dtype: float64
      floor_area_sqm  lease_commence_date  resale_price
floor_area_sqm      1.000000          0.150695      0.621973
lease_commence_date  0.150695          1.000000      0.348542
resale_price         0.621973          0.348542      1.000000
```


Results and Suggestions

```
Results on Test Data
RMSE: 44656.32
R2 Score: 0.92109
```

	predicted_price	resale_price	Difference_%
0	477662.535119	440000.0	0.078848
1	803011.461195	850000.0	-0.058515
2	526788.491474	475000.0	0.098310
3	255119.548716	232000.0	0.090622
4	538409.386456	570000.0	-0.058674
...
41272	549108.378666	533000.0	0.029336
41273	408411.422826	290000.0	0.289932
41274	458967.356733	430000.0	0.063114
41275	418604.337808	408000.0	0.025333
41276	184190.634209	238000.0	-0.292140

```
[41277 rows x 3 columns]
Max % difference: 0.411
Min % difference: -67.421
```

- LinearRegression model was used with accuracy of 92%
- Predicted price differs from -67% to 0.4%. (See attached excel file for full comparison)
- Random Forest model may be tested to see if it brings about greater accuracy.
- As it stands, current model with 92% accuracy is sufficient.
- Past sales data from before 2017 may be used to train model for greater accuracy.