Bangkok Housing Price Prediction

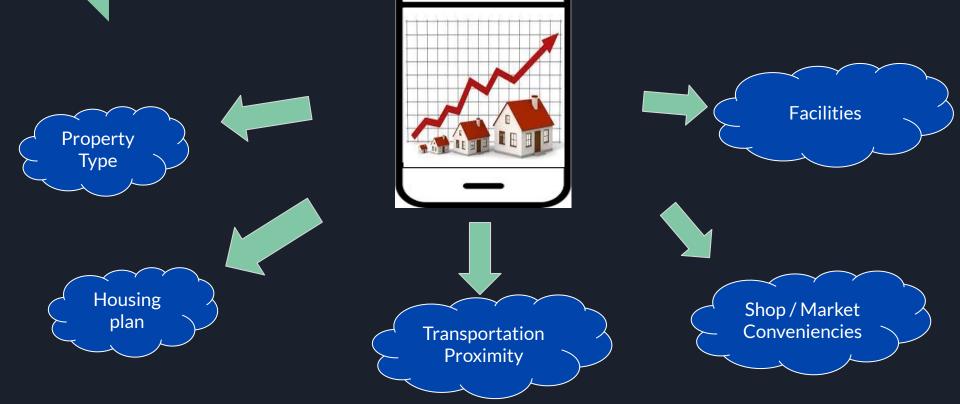
What is the price of my property?



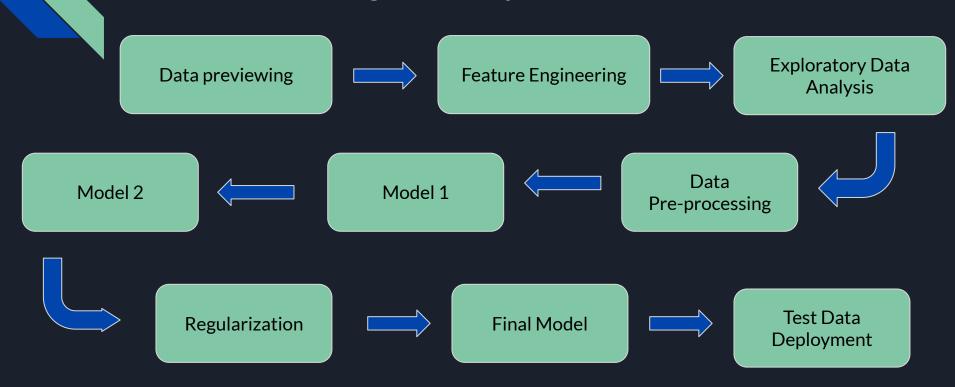




Problem Statement



Model Building Journey



Data Previewing

- Multiple missing values
- 8 object data columns

#	Column	Non-Null Count	Dtype
0	id	14271 non-null	int64
1	province	14271 non-null	object
2	district	14271 non-null	object
3	subdistrict	14260 non-null	object
4	address	14271 non-null	object
5	property_type	14271 non-null	object
6	total_units	10509 non-null	float64
7	bedrooms	14228 non-null	float64
8	baths	14236 non-null	float64
9	floor_area	14271 non-null	int64
10	floor_level	8093 non-null	float64
11	land_area	4917 non-null	float64
12	latitude	14271 non-null	float64
13	longitude	14271 non-null	float64
14	nearby_stations	14271 non-null	int64
15	nearby_station_distance	7228 non-null	object
16	nearby_bus_stops	6009 non-null	float64
17	nearby_supermarkets	13885 non-null	float64
18	nearby_shops	14271 non-null	int64
19	year_built	14271 non-null	int64
20	month_built	8397 non-null	object
21	facilities	14271 non-null	object
22	price	14271 non-null	int64
dtyp	es: float64(9), int64(6),	object(8)	

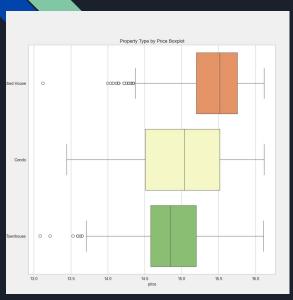
id	0
province	0
district	0
subdistrict	11
address	0
property_type	0
total_units	3762
bedrooms	43
baths	35
floor_area	0
floor_level	6178
land_area	9354
latitude	0
longitude	0
nearby_stations	0
nearby_station_distance	7043
nearby_bus_stops	8262
nearby_supermarkets	386
nearby_shops	0
year_built	0
month_built	5874
facilities	0
price	0
dtype: int64	

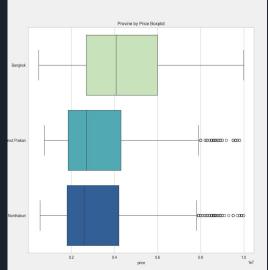
Feature Engineering

- Target Encoding: District, Subdistrict, Province,
 Property_type columns
- Numeric Encoding Month_built column
- Count_facilities column: Count each facility
- Closest_station_name and Closest_station_distance

nearby_station_distance	closest_station_name	closest_station_distance
[[E7 Ekkamai BTS, 270], [E6 Thong Lo BTS, 800]]	E7 Ekkamai BTS	270
[[E7 Ekkamai BTS, 270], [E6 Thong Lo BTS, 800]]	E7 Ekkamai BTS	270
[[E9 On Nut BTS, 110]]	E9 On Nut BTS	110

Exploratory Data Analysis





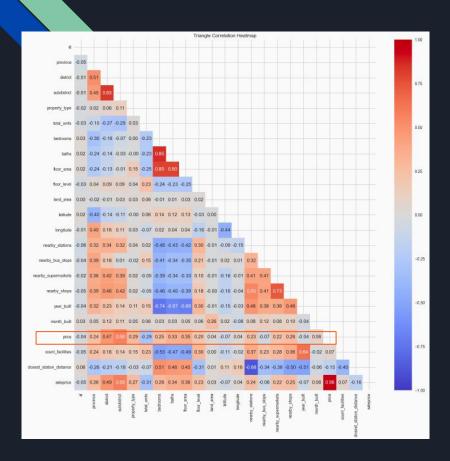
Rak Wan newi newi newi newi newi nemi nemi nemi nemi nemi nemi nemi nem	0	00000	0 000			
hiein lang Buri Khae Buri Khae N Na Sur					0000	

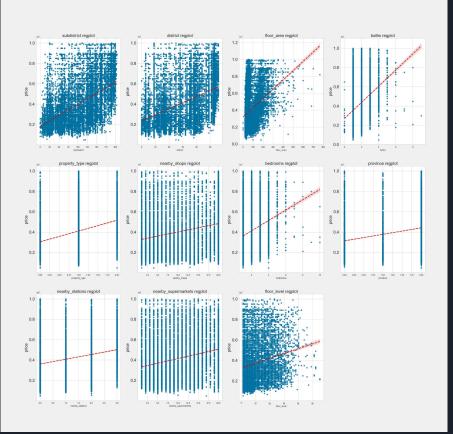
saleprice
15.458860
15.129510
14.982282
14.941575

	saleprice
bedrooms	
6.0	15.630790
5.0	15.552675
7.0	15.539077
9.0	15.404475
8.0	15.366126
4.0	15.358923
10.0	15.222124
2.0	15.208586
3.0	15.195864
1.0	14.913156



Correlation Analysis





Feature Selection

```
price
                             1.000000
saleprice
                             0.962380
subdistrict
                             0.564869
district
                             0.473325
floor area
                             0.351357
baths
                             0.334650
                             0.285423
property type
nearby shops
                             0.257855
bedrooms
                             0.254158
                             0.239900
province
nearby_stations
                            0.232143
nearby supermarkets
                             0.224702
floor_level
                             0.198645
month built
                             0.079473
count_facilities
                             0.069456
longitude
                             0.037417
land area
                             0.036735
year_built
                           -0.042557
id
                           -0.044732
latitude
                           -0.065560
nearby bus stops
                           -0.066557
closest_station_distance
                           -0.152466
total units
                           -0.285896
```

- Selected features are set for X dataframe
- Price for Y dataframe

Pre-processing

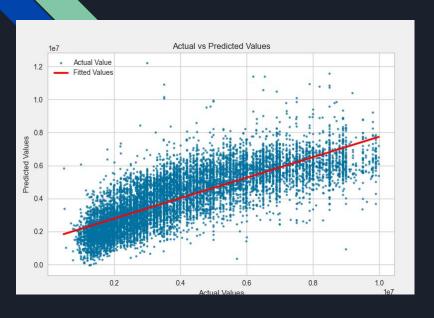
KNN-Imputer

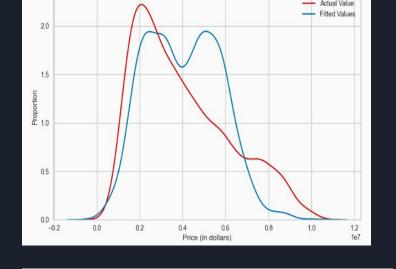
```
[{'K': 1, 'RMSE': 1377283.3377121552},
{'K': 2, 'RMSE': 1373359.9953259628},
{'K': 3. 'RMSE': 1373497.8253319028},
{'K': 4, 'RMSE': 1371375.2827217935},
{'K': 5, 'RMSE': 1372416.397510357},
{'K': 6, 'RMSE': 1373266.8728558398},
{'K': 7, 'RMSE': 1372352.5412769017},
{'K': 8, 'RMSE': 1372658.577194074},
{'K': 9, 'RMSE': 1372868.397009458},
{'K': 10, 'RMSE': 1372597.0730115585},
{'K': 11, 'RMSE': 1372455.7726511175},
{'K': 12, 'RMSE': 1372345.5270683996},
{'K': 13, 'RMSE': 1372567.3684354573},
{'K': 14, 'RMSE': 1372808.0569123065},
{'K': 15, 'RMSE': 1372833.6433913363},
{'K': 16, 'RMSE': 1372629.328407314},
{'K': 17, 'RMSE': 1373212.4456665257},
{'K': 18, 'RMSE': 1373402.4419220332},
{'K': 19, 'RMSE': 1373374.8489120326}]
```

- Perform KNN Imputer from K:1-19, search the K with the lowest RMSE
- Step1: Transforming X_train with KNN Imputer
- Step2: Fit Train KNN with Linear Regression model and evaluate its RMSE

Standard-Scaler

Model1: Linear Regression





Actual vs Fitted Values

R-Square Train: 0.6193559381542455

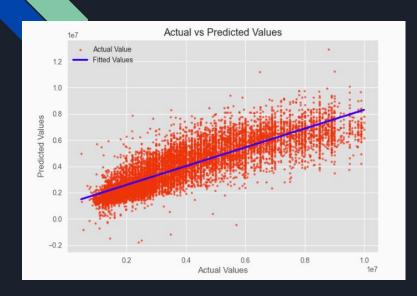
R-Square Validation: 0.6092927658135293

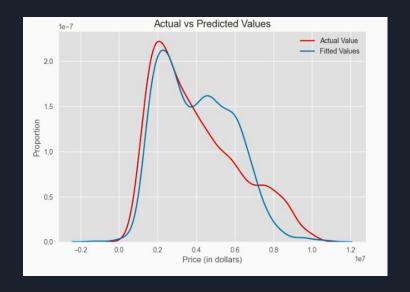
RMSE of Train: 1346780.1298077581

RMSE of Validation : 1371375.2827217935

- The model is underfit
Performance on train and test data
are resemblance

Model2: Polynomial-Featured Linear Regression





R-Square Train: 0.7155729816477747

R-Square Validation: 0.7002739334580338

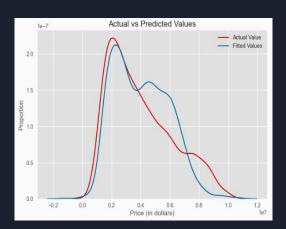
RMSE of Train: 1164186.5969208174

RMSE of Validation : 1201137.8386152948

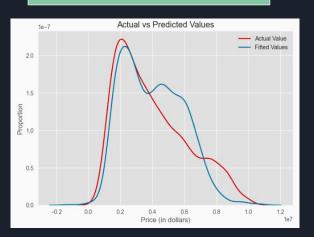
- The model is underfit
- Performance on train and test data are resemblance
 - Performance improved

Polynomial Regression Tuning

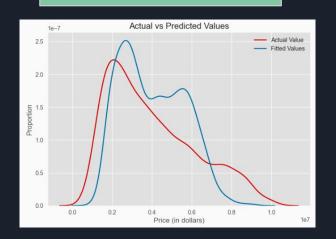
Ridge Regularization



Lasso Regularization



Elastic Net Regularization



R-Square Train: 0.7155349372270392 R-Square Validation: 0.7005436883743024

RMSE of Train : 1164264.4540175162

RMSE of Validation : 1200597.201998452

R-Square Train: 0.7155729299162299

R-Square Validation: 0.7002846888592379

RMSE of Train: 1164186.702791857

RMSE of Validation : 1201116.2875446773

R-Square Train: 0.6575774224070181

R-Square Validation: 0.6448924379011504

RMSE of Train: 1164186.702791857

RMSE of Validation : 1201116.2875446773