

Project 2

Bangkok House Price Prediction



Problem Statement

"..AP Thai CMO is thinking of launching new project in 2024, aiming to found accommodation in Bangkok and suburban areas with data-driven pricing strategy. What he needs are

- (1) To see which factors are significant to real estates price?
- (2) An acceptably accurate price prediction model.

So, he requests his Data Science team to deliver the result within 2 weeks.."

Dataset Overview

• 14,271 as number of observation

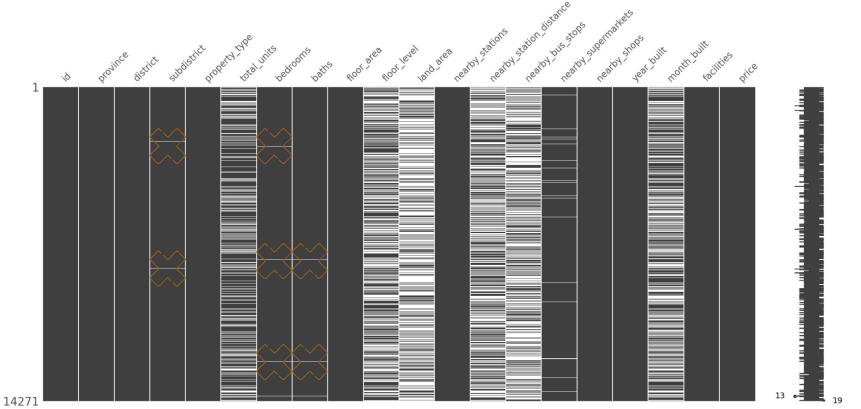
23 features from which 10 features having null value

Having 3 property types: Condo-65%, Townhouse-20% and Detached House-15%

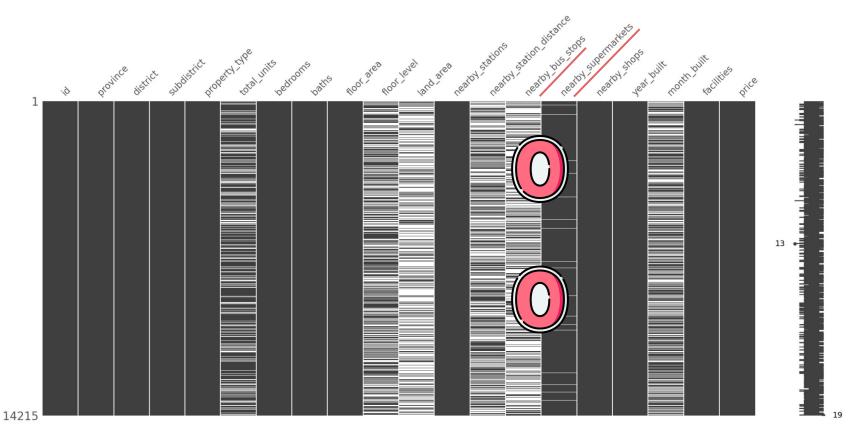
• 'id', 'address', 'longitude' and 'latitude' features will be excluded from this analysis



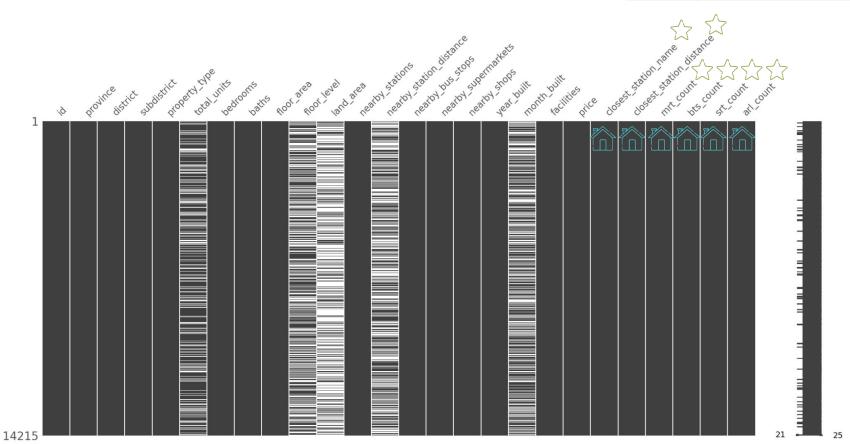
Observations contain missing value in columns 'subdistrict', 'bedrooms' and 'baths' were dropped



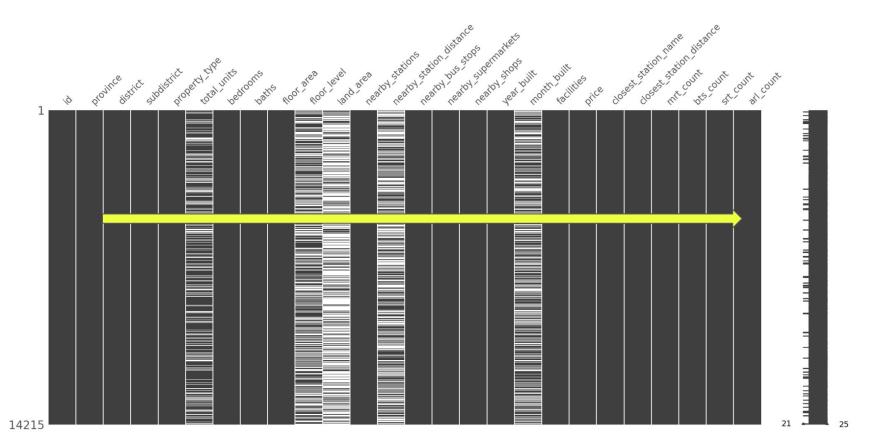
Replace Missing Value with zero oin 'nearby_bus_stops' and 'nearby_supermarkets' since the minimum values in those columns are above zero. Thus, the missing values are assumed to be zero.



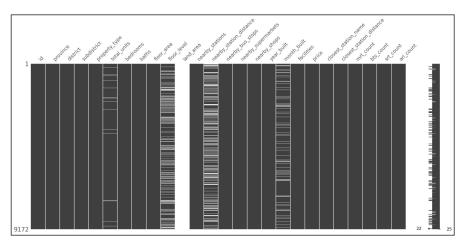
6 numeric features were added to DataFrame which were extracted from 'nearby_station_distance'

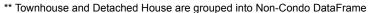


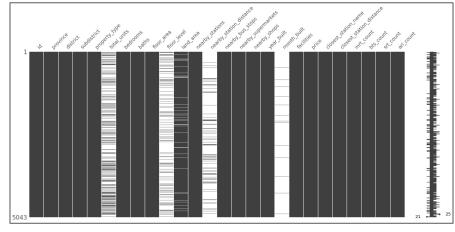
The remaining missing value seems to have a pattern in horizontal trend



Divide DataFrame into Condo-DataFrame and Non-Condo DataFrame for ease of missing value dealing





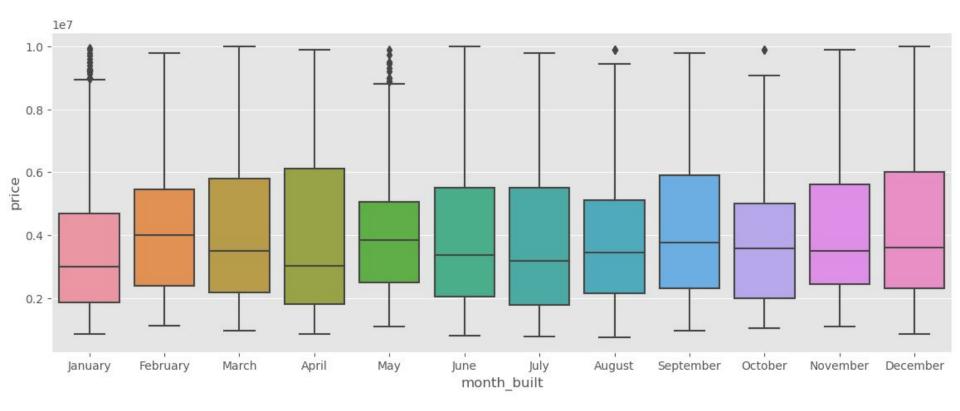


- **Condo DataFrame** → Impute by subdistrict grouped median
 - → Drop some columns
 - → Dummify string columns
 - → Standardize the features scale

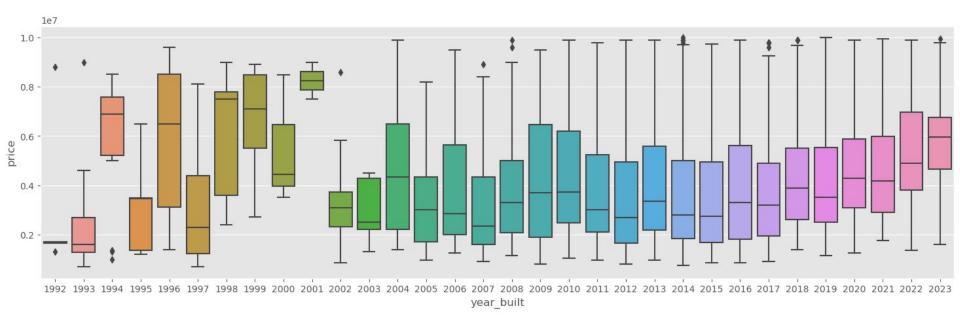
Non-Condo DataFrame \rightarrow Drop some columns

- → Dummify string columns
- → Scale Standardize

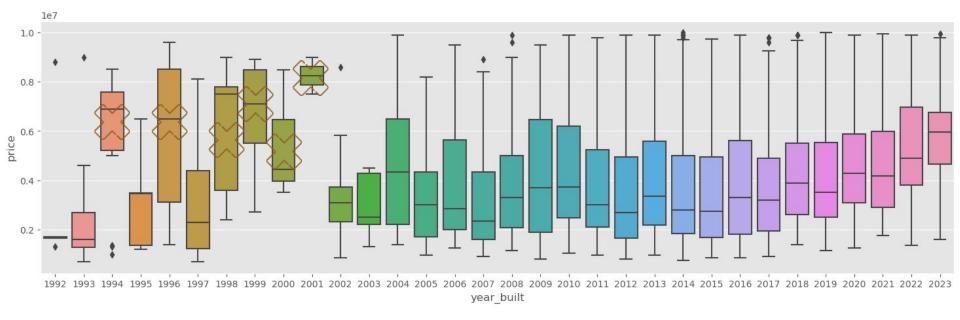
condo_df: 'month_built' column was dropped as the distributions of 'price' in each month are not different.



condo_df: 'price' from 'year_built' 2002 seems to get higher gradually but before 'year_built' 2002 having high fluctuation, so some observations were dropped in order the get the distribution in the same trend.



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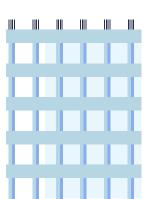


Modeling

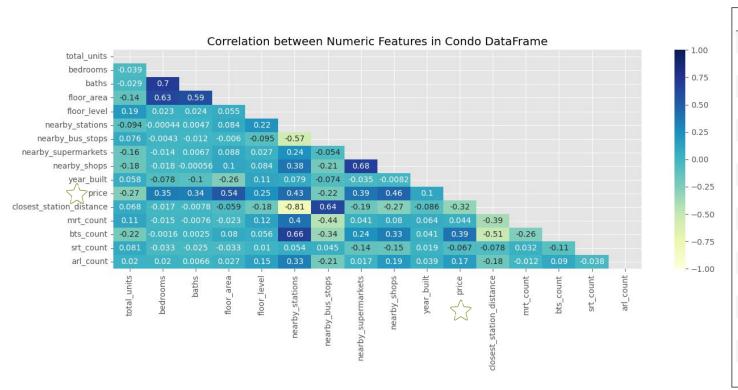
 Use LassoCV Model to predict the price for Condo and Non-condo property types

 To eliminate Multicollinearity between features when a large number of columns are added to the model

To prevent overfitting incident as the number of features is high



condo_df: 'floor_area' has strongest <u>positive</u> correlation with 'price' 'closet_station_distance' has strongest <u>negative</u> correlation with 'price'



	9
	price
price	1.000000
floor_area	0.538680
nearby_shops	0.459276
nearby_stations	0.429618
bts_count	0.390942
nearby_supermarkets	0.390459
bedrooms	0.348176
baths	0.335953
floor_level	0.247946
arl_count	0.174884
year_built	0.104080
mrt_count	0.043958
srt_count	-0.066988
nearby_bus_stops	-0.216240
total_units	-0.267610
closest_station_distance	-0.342243

condo_df : LassoCV Model (1)

17 features used

- R^2 Score Training : 0.67913
- RMSE Training : 1,236,913.73 THB

- R² Score Test : 0.59483
- RMSE Test : 1,394,385.29 THB

	Coefficient(LassoCV)	
floor_area	1.054034e+06	
year_built	4.779500e+05	
floor_level	3.514049e+05	
nearby_shops	3.417091e+05	
bts_count	2.785592e+05	
nearby_supermarkets	2.506112e+05	
nearby_stations	1.061332e+05	
baths	8.798380e+04	
arl_count	6.744622e+04	
mrt_count	3.147875e+04	
bedrooms	3.101113e+04	
srt_count	-0.000000e+00	
nearby_bus_stops	-4.197816e+04	
closest_station_distance	-7.588427e+04	
total_units	-2.594722e+05	
province_Nonthaburi	-3.014386e+05	
province_Samut Prakan	-3.224617e+05	

condo_df : LassoCV Model (2)

- 188 features used, 'district' and 'closest station name' are considered
- Drop 'province'

- *R*² Score Training : 0.82149
- RMSE Training: 922,563.70 THB

- R² Score Test: 0.75651
- RMSE Test : 1,080,946 THB

Top 5 positive factors to 'price'

	Coefficient(LassoCV)
floor_area	902392.427581
closest_station_name_NO STATION	749624.678145
district_Watthana	569997.869994
year_built	463330.139981
district_Khlong Toei	395114.679163

Bottom 5 negative factors to 'price'

	Coefficient(LassoCV)
closest_station_name_PP15 Bang Son MRT	-97097.602070
closest_station_name_PP03 Sam Yaek Bang Yai MRT	-112795.468275
closest_station_name_PP13 Yaek Tiwanon MRT	-122762.164325
closest_station_name_PP10 Bang Krasor MRT	-176240.598474
closest_station_distance	-954337.285492

condo_df : LassoCV Model (3)

250 features used, 'facilities' is considered

• *R*² Score - Training : 0.83660

RMSE - Training : 882,666.25 THB

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Top 5 positive factors to 'price'

2	Coefficient(LassoCV)
floor_area	933445.702345
closest_station_name_NO STATION	645681.134843
district_Watthana	524577.149013
year_built	401975.753450
district_Khlong Toei	342879.909802

Bottom 5 negative factors to 'price'

• R² Score - Test : 0.76722

• RMSE - Test : 1,056,890.35 THB

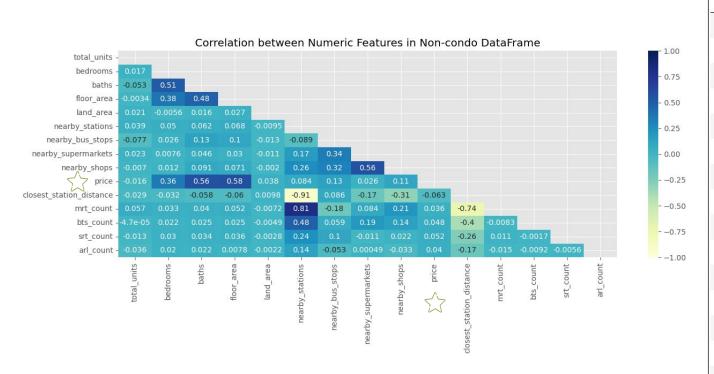


	Coefficient(LassoCV)
total_units	-107237.369214
closest_station_name_PP03 Sam Yaek Bang Yai MRT	-135356.272369
closest_station_name_PP13 Yaek Tiwanon MRT	-147282.383687
closest_station_name_PP10 Bang Krasor MRT	-161931.682796
closest_station_distance	-876662.493607

condo_df : Model 3 is best performing model

	LassoCV Model 1	LassoCV Model 2	LassoCV Model 3
Number of Features	17	188	250
R^2 - Score : Training Set	0.67913	0.82149	0.83660
RMSE : Training Set (THB)	1,236,913.73	922,563.70	882,666.25
R^2 - Score : Test Set	0.59483	0.75651	0.76722
RMSE : Test Set (THB)	1,394,385.29	1,080,946	1,056,890.35

non_condo_df: 'floor_area' has strongest positive correlation with 'price'



	price	
price	1.000000	
floor_area	0.579923	
baths	0.562876	
bedrooms	0.356604	
nearby_bus_stops	0.134385	
nearby_shops	0.108877	
nearby_stations	0.084206	
srt_count	0.052115	
bts_count	0.047740	
arl_count	0.039918	
land_area	0.037896	
mrt_count	0.035675	
nearby_supermarkets	0.025911	
total_units	-0.016065	
closest_station_distance	-0.063402	

non_condo_df : LassoCV Model(4)

17 features used

- *R*² Score Training : 0.58887
- RMSE Training : 1,368,795.91 THB

- R² Score Test : 0.54628
- RMSE Test : 1,425,545.88 THB

	Coefficient(LassoCV)
baths	769282.390392
floor_area	488845.551418
nearby_stations	120082.061558
nearby_shops	117020.931938
nearby_bus_stops	87550.149099
land_area	42364.944936
srt_count	26418.669983
arl_count	18342.710354
bts_count	-0.000000
total_units	-0.000000
closest_station_distance	0.000000
bedrooms	-18590.317664
mrt_count	-23186.592932
nearby_supermarkets	-34131.981625
province_Samut Prakan	-165775.103390
province_Nonthaburi	-187804.210543
property_type_Townhouse	-832889.936857

non_condo_df : LassoCV Model(5)

- 143 features used, 'district', 'closest_station_name' and 'property_type' are considered
- Drop 'province'

- *R*² Score Training : 0.66545
- RMSE Training : 1,234,739.77 THB

- R² Score Test : 0.61654
- RMSE Test : 1,310,524.22 THB

Top 5 positive factors to 'price'

	Coefficient(LassoCV)
floor_area	902392.427581
closest_station_name_NO STATION	749624.678145
district_Watthana	569997.869994
year_built	463330.139981
district_Khlong Toei	395114.679163

Bottom 5 negative factors to 'price'

	Coefficient(LassoCV)
closest_station_name_PP15 Bang Son MRT	-97097.602070
closest_station_name_PP03 Sam Yaek Bang Yai MRT	-112795.468275
closest_station_name_PP13 Yaek Tiwanon MRT	-122762.164325
closest_station_name_PP10 Bang Krasor MRT	-176240.598474
closest_station_distance	-954337.285492

non_condo_df : Model 6 is best performing model

	LassoCV Model 4	LassoCV Model 5	LassoCV Model 6
Number of Features	17	143	206
R ² - Score : Training Set	0.58887	0.66545	0.69549
RMSE : Training Set (THB)	1,368,795.91	1,234,739.77	1,1178,004.21
R ² - Score : Test Set	0.54628	0.61654	0.64561
RMSE : Test Set (THB)	1,425,545.88	1,310,524.22	1,259,870.66

Recommendation and Summary



For Condo, the lassoCV model 3 is the best model with highest R2-Score and the overall average predicted price error is around 1,056,890 THB. The total area of the condo is the most positive factor when it increases by 1 meter square, the predicted price will increase around +933,445 THB while the distance from the closest railway station is the most negative factor when it increases by 1 meter, the predicted price will decrease around -876,662 THB

For Non-condo, Detached house and Townhouse, the lassoCV model 6 is the best model with highest R2-Score and the overall average predicted price error is around 1,259,870 THB. The number of baths is the most positive factor when it increases by 1 unit, the predicted price will increase around +636,543 THB while the 'townhouse' property type is the most negative factor which give the sales price cheaper than that of detached house around -802,908 THB in the same environment.

