

Case Study

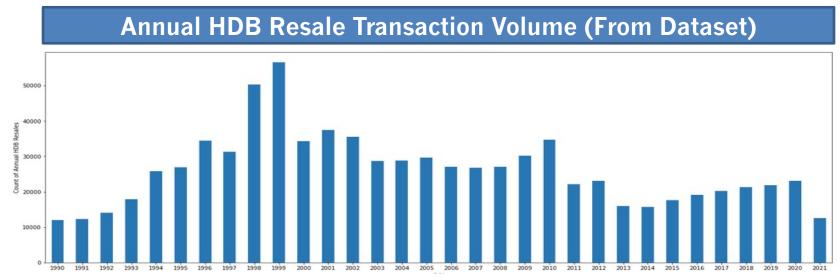


HDB Resale Price Estimator

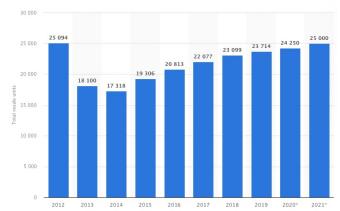
BACKGROUND INFORMATION - RESALE VOLUME



> HDB dataset given shows at least 10k HDB resale transactions per annum, and external sources show in recent years this number goes past 20k.



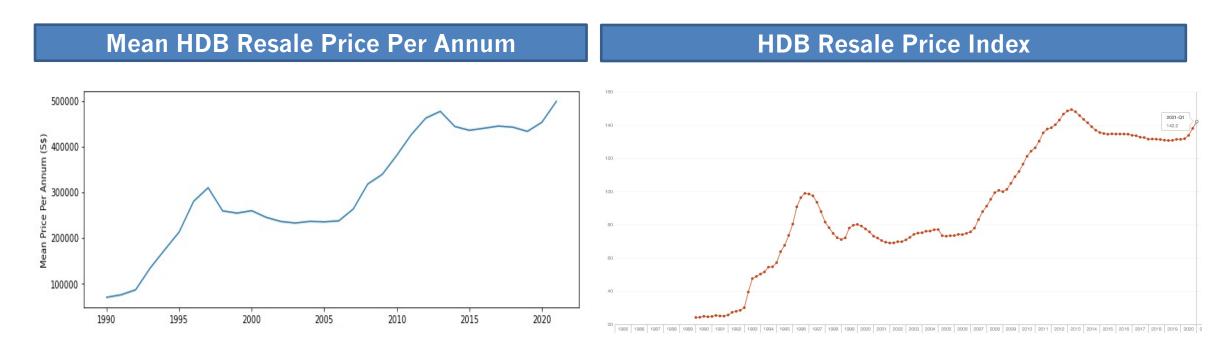
HDB Resale Estimates



BACKGROUND INFORMATION – RESALE PRICE



- > HDB resale prices have remained buoyant since its conception
- > One of the reasons for the resilience of the public housing market is it is a basic necessity and the most affordable type of housing in Singapore.
- > Demand for public housing is unlikely to evaporate even during economic downturns.



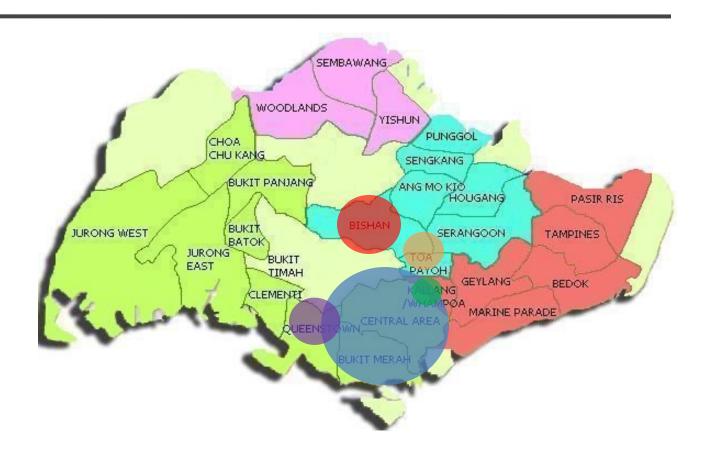
Source: HDB Dataset, Data.gov.sg

BACKGROUND INFO – MOST EXPENSIVE NEIGHBOURHOODS



TOWNS WITH THE MOST SALES OVER \$\$1M

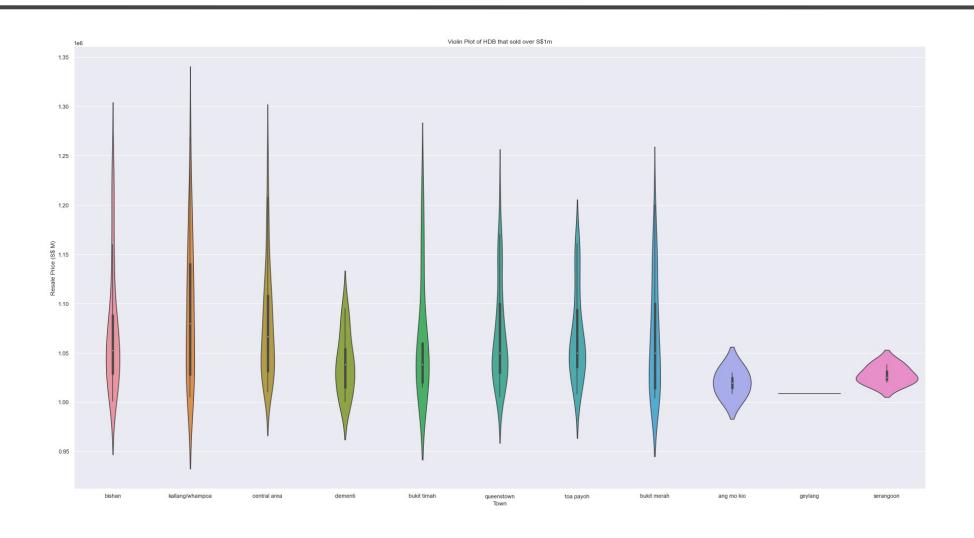
Town Name	Count
Central Area (CBD)	112
Bishan	68
Queenstown	49
Toa Payoh	38
Kallang/Whampoah	29



Source: HDB Dataset Notes: Graph not to scale

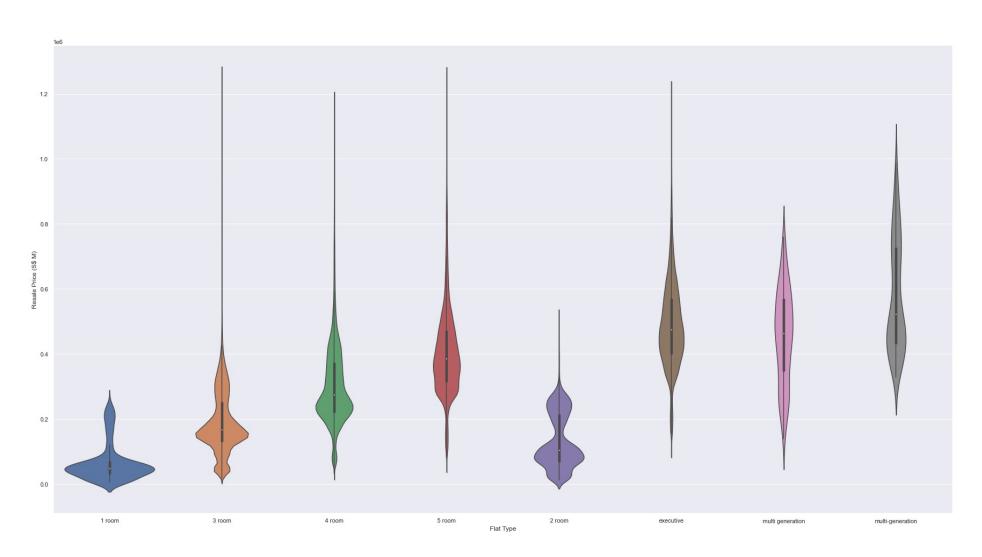
BACKGROUND INFO – MOST EXPENSIVE NEIGHBOURHOODS Institute of Data





BACKGROUND INFO – MOST EXPENSIVE FLAT TYPE





BACKGROUND INFORMATION – GOVT REGULATION



The ownership and transaction of HDB flats are regulated. Only Singaporeans and permanent residents can own HDB flats. Each resident household can own only one HDB flat, which will be their primary place of residence. Each flat owner must live in his flat for a minimum of five years before he can sell the flat.

➤ There are other rules and regulations to ensure that HDB flats are treated as a home rather than a trading asset, which contribute to its price stability.



BUSINESS PROBLEM



At least 80% of Singaporean resident population live in HDB's, of which 90% own their own homes (HDB.gov.sg). So it is very likely that most Singaporeans either are current homeowners or are on the market to purchase a HDB.

But due to government regulation, the public housing market is not as liquid as the private housing market, so most consumers have little knowledge on the market.

Homeowners and buyers often rely on the expertise of real estate agents that specialize in the HDB sector and have the potential to "go in blind" without fully understanding the worth of the asset in question.

We propose a data driven approach for homeowners and potential buyers to better understand the market and use this tool as a point of negotiation to value this asset class.

THE DATASET



- ➤ Given 5 datasets with information on HDB resale transaction information from 1990 to 2020 were merged
- ➤ Around 25k duplicate entries were removed
- > Final dataset is 834,508 entries with 11 feature columns

	month	town	flat_type	block	street_name	storey_range	floor_area_sqm	flat_mode	l lease_commence_date	resale_price
0	1990-01	ANG MO KIO	1 ROOM	309	ANG MO KIO AVE 1	10 TO 12	31.0	IMPROVED	1977	7 9000
1	1990-01	ANG MO KIO	1 ROOM	309	ANG MO KIO AVE 1	04 TO 06	31.0	IMPROVED	1977	7 6000
2	1990-01	ANG MO KIO	1 ROOM	309	ANG MO KIO AVE 1	10 TO 12	31.0	IMPROVED	1977	7 8000
3	1990-01	ANG MO KIO	1 ROOM	309	ANG MO KIO AVE 1	07 TO 09	31.0	IMPROVED	1977	6000
4	1990-01	ANG MO KIO	3 ROOM	216	ANG MO KIO AVE 1	04 TO 06	73.0	NEW GENERATION	1976	47200
	month	town	flat_type	block	street_name	storey_range	floor_area_sqm	flat_model i	ease_commence_date	esale_price
0	2000-01	ANG MO KIO	3 ROOM	170	ANG MO KIO AVE 4	07 TO 09	69.0	Improved	1986	147000.0
1	2000-01	ANG MO KIO	3 ROOM	174	ANG MO KIO AVE 4	04 TO 06	61.0	Improved	1986	144000.0
2	2000-01	ANG MO KIO	3 ROOM	216	ANG MO KIO AVE 1	07 TO 09	73.0	New Generation	1976	159000.0
3	2000-01	ANG MO KIO	3 ROOM	215	ANG MO KIO AVE 1	07 TO 09	73.0	New Generation	1976	167000.0
4	2000-01	ANG MO KIO	3 ROOM	218	ANG MO KIO AVE 1	07 TO 09	67.0	New Generation	1976	163000.0
	month	town	flat_type	block	street_name	storey_range	floor_area_sqm	flat_model	lease_commence_date	resale_price
0	2012-03	ANG MO KIO	2 ROOM	172	ANG MO KIO AVE 4	06 TO 10	45.0	Improved	1986	250000.0
1	2012-03	ANG MO KIO	2 ROOM	510	ANG MO KIO AVE 8	01 TO 06	44.0	Improved	1980	265000.0
2	2012-03	ANG MO KIO	з поом	610	ANG MO KIO AVE 4	06 TO 10	68.0	New Generation	1980	315000.0
3	2012-03	ANG MO KIO	3 ROOM	474	ANG MO KIO AVE 10	01 TO 06	67.0	New Generation	1984	320000.0
4	2012-03	ANG MO KIO	3 ROOM	604	ANG MO KIO AVE 5	06 TO 10	67.0	New Generation	1980	321000.0
	month	town	flat_type	block	street_name	storey_range	floor_area_sqm	flat_model	lease_commence_date	resale_price
0	2015-01	ANG MO KIO	3 ROOM	174	ANG MO KIO AVE 4	07 TO 09	60.0	Improved	1986	255000.0
1	2015-01	ANG MO KIO	з воом	541	ANG MO KIO AVE 10	01 TO 03	68.0	New Generation	1981	275000.0
2	2015-01	ANG MO KIO	з поом	163	ANG MO KIO AVE 4	01 TO 03	69.0	New Generation	1980	285000.0
3	2015-01	ANG MO KIO	з воом	446	ANG MO KIO AVE 10	01 TO 03	68.0	New Generation	1979	290000.0
4	2015-01	ANG MO KIO	3 ROOM	557	ANG MO KIO AVE 10	07 TO 09	68.0	New Generation	1980	290000.0
	month	town	flat_type	block	street_name	storey_range	floor_area_sqm	flat_model	lease_commence_date	resale_price
0	2017-01	ANG MO KIO	2 ROOM	406	ANG MO KIO AVE 10	10 TO 12	44.0	Improved	1979	232000.0
1	2017-01	ANG MO KIO	3 ROOM	108	ANG MO KIO AVE 4	01 TO 03	67.0	New Generation	1978	250000.0
2	2017-01	ANG MO KIO	3 ROOM	602	ANG MO KIO AVE 5	01 TO 03	67.0	New Generation	1980	262000.0
-										
3	2017-01	ANG MO KIO	3 ROOM	465	ANG MO KIO AVE 10	04 TO 06	68.0	New Generation	1980	265000.0



Parties in the market should use the next few slides to have a better understanding of their asset in order to purchase/sell a property at least near the median value of previous transactions

A regression model was also built for homeowners/buyers that proposes a purchase/sale price





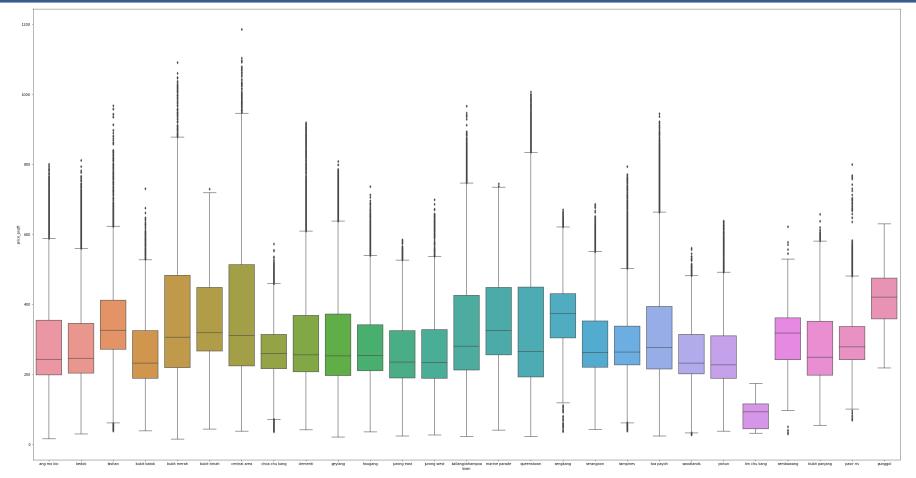
Source: HDB Dataset

Notes: The figures are subject to rounding errors



- > Pungoll has the highest median price per square foot
- ➤ Lim Chu Kang has the lowest median price per square foot

Boxplot of Town to Price PSF

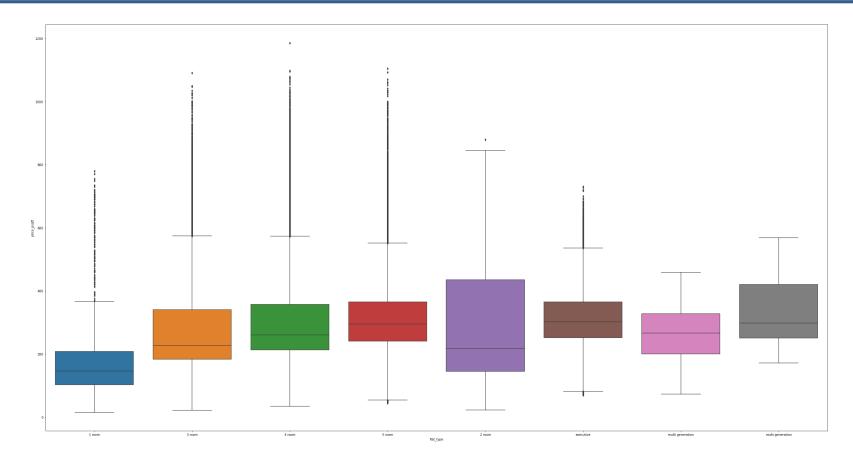


Source: HDB Dataset



- ➤ Multi-generation flats have the highest median price per square foot
- > 1-room flats have the lowest median price per square foot

Boxplot of Flat Type to Price PSF

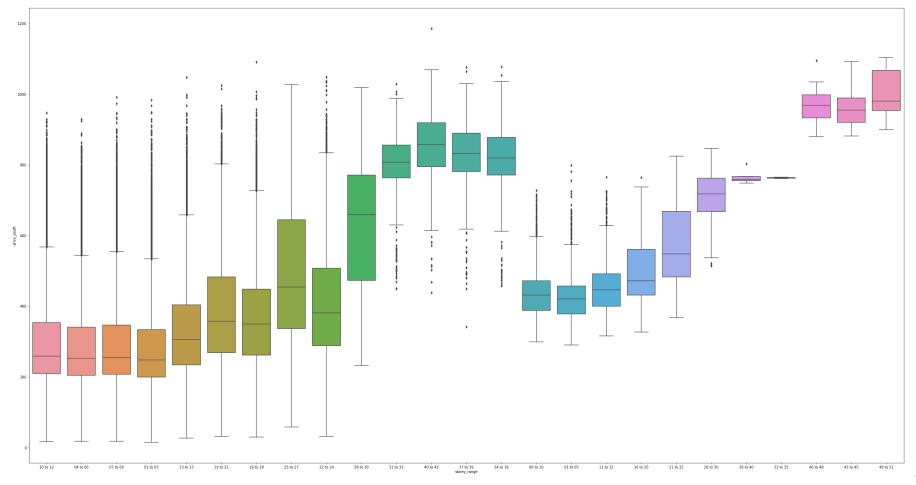


Source: HDB Dataset



➤ Higher floors command the highest prices psf

Boxplot of Storey Range to Price PSF



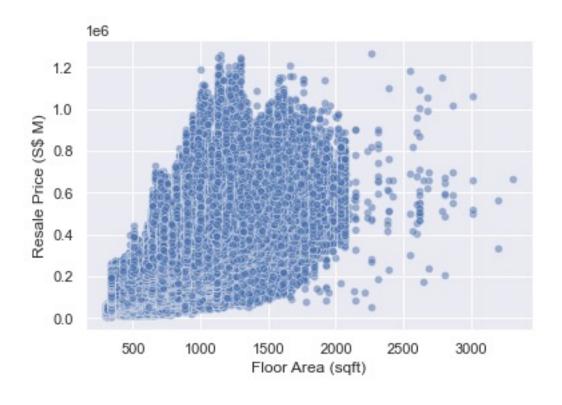
Source: HDB Dataset

MODEL

LINEAR RELATIONSHIP IDENTIFIED







FEATURE SELECTION (SIMPLE LR)



> Floor area by sqft picked as predictor for simple LR

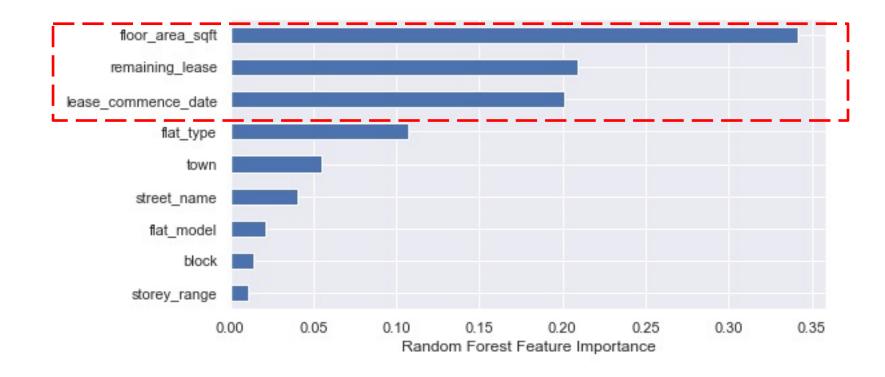
Triangle Correlation Heatmap



MULTI-FEATURE SELECTION



> Top 3 features by RandomForest picked.



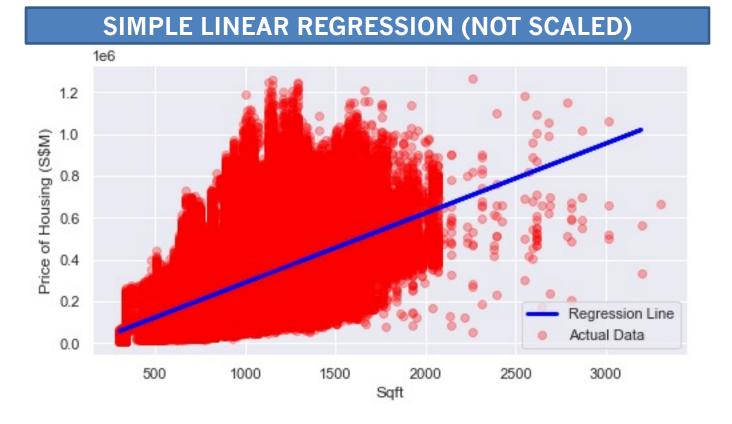
Source: HDB Dataset

Notes: RandomForest was used as it can handle both continuous and categorical data at ease

MODEL SELECTION



Model Type	Features	R2	RMSE (S\$)
Simple LR (Not Scaled)	1	0.38	119,710
Simple LR (Scaled)	1	0.38	119,513
Multi LR (Scaled)	9	0.66	88,307
RFR	9	0.97	24,182
RRR	3	0.89	49,880



PREDICTION TESTING



Predictor Features	Characteristic	town flat_type block street_name storey_range flat_model lease_commence_date remaining_lease floor_area_sqft price_psqft resale_price date
Town	Tampines	2020-09-01 18 5 944 196 0 7 20 66 1657.6406 446.417637 740000.0
Flat Type	Executive	
Block	942	<pre>In [132]: 1 RFR.predict(x_pred) #- predicted price is SGD 850k. 14% higher than actual value of S\$740k [Parallel(n_jobs=8)]: Using backend ThreadingBackend with 8 concurrent workers. [Parallel(n_jobs=8)]: Done 2 tasks elapsed: 0.0s</pre>
Street Name	Tampines Ave 5	[Parallel(n_jobs=8)]: Done 9 tasks elapsed: 0.0s [Parallel(n_jobs=8)]: Done 16 tasks elapsed: 0.0s [Parallel(n_jobs=8)]: Done 25 tasks elapsed: 0.0s
Storey Range	10 to 12	[Parallel(n_jobs=8)]: Done 34 tasks elapsed: 0.0s [Parallel(n_jobs=8)]: Done 45 tasks elapsed: 0.0s [Parallel(n_jobs=8)]: Done 56 tasks elapsed: 0.0s [Parallel(n_jobs=8)]: Done 69 tasks elapsed: 0.0s
Flat Model	Maisonette	[Parallel(n_jobs=8)]: Done 82 tasks elapsed: 0.0s [Parallel(n_jobs=8)]: Done 96 out of 100 elapsed: 0.0s remaining: 0.0s [Parallel(n_jobs=8)]: Done 100 out of 100 elapsed: 0.0s finished
Lease Commence Date	1992	Out[132]: array([850383.76])
		In [134]: 1 850/740
Remaining Lease	66	Out[134]: 1.1486486486486487
Floor Area Saft	1.657	

IMPROVEMENTS

IMPROVEMENTS FOR FUTURE



- ➤ More work needs to be done on feature engineering. Price psf imputed from target variable too simplistic
- > Could include geospatial data as a new feature and identify proximity to CBD, public transport, schools as these could possible increase asset valuations

Thank you