Greater London Housing Prices

Nigel Brown

2020/10/13

${\bf Contents}$

1	Intr	roduction	2
2	Dat	a Wrangling	2
	2.1	Features added to the data	2
3	Exploritory Data Analysis		
	3.1	How does the avg price increase year on year ?	7
	3.2	Is there a trend of which month most sales occur on ? \dots	7
	3.3	Which are the 10 most expensive districts based on the number of properties sold over 3 million GBP?	8
	3.4	How are properties sold for greater than 3 million GBP clustered ?	9
	3.5	Which are the 10 least expensive districts based on the number of properties under 100K GBP?	9
	3.6	How are properties under 100k GBP clustered ?	10
	3.7	Are the sales evenly spread across Greater London ?	10
	3.8	Which type of property sold the most?	11
	3.9	Which year had the most sales	12
	3.10	How often do properties change ownership ?	13
	3.11	How are house prices distributed?	14
4	Met	${ m thods/Analysis}$	15
5	Res	sults	15
6	Cor	nclusion	15
7	Disc	Disclaimers	
8	Apr	Appendix	

1 Introduction

The goal of this project is to predict house prices in the Greater London Area based on the type of property and the area of London that the property resides in.

The dataset for the project is made from data sourced from HM Land Registry and FreeMapTools. This data has been pre-processed into a dataset titled lhd.rds. The data chosen is a 10 year period 2009-2019. The reason 2020 has been left out of the data is that is an incomplete year at the time of analysis. The data contains 4 different types of property detached; flats/maisonettes; semi-detached and terraced. It doesn't contain the number of rooms or the total area of the property, therefore a 2 bedroom flat is categorized the same as a 4 bedroom flat and a prediction can only be made on the property type of a given district.

The price paid data set from the HM Land Registry is joined together with spatial data to add latitude and longitude for the properties based on their postcodes. This will allow for mapping of the data.

2 Data Wrangling

As a first step data, that is in logical form (true/false) is converted into integers 1 or 0. Next all columns containing character data are factorized.

2.1 Features added to the data

Next a new feature is created based on the outward code contained in a postcode i.e the sub-district and another feature containing the number of times a property was observed to be sold in the 10 year time period is added.

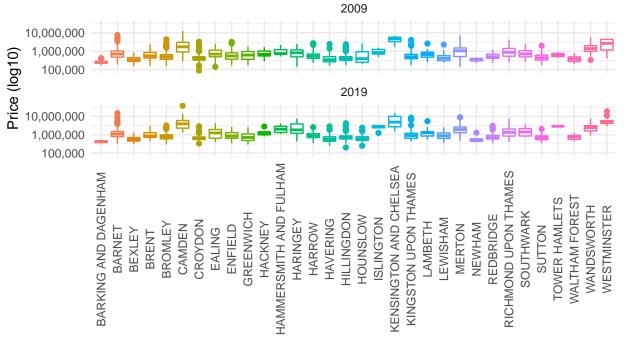
3 Exploritory Data Analysis

Now that the dataset is cleaned, an exploration of the data is performed. The data consists of 477382 rows of data, each row observes a single sale. The are 12 features in the dataset, with price being the outcome feature

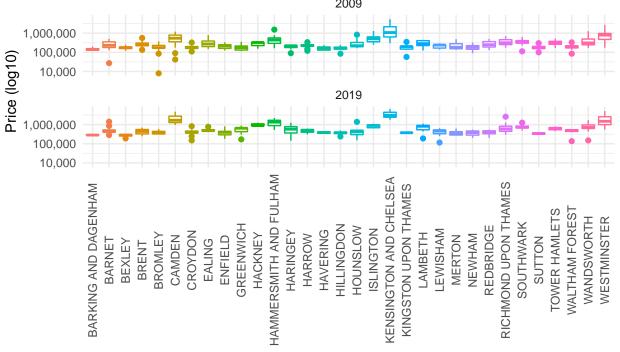
First the price movement for each property type is plotted

Movement of detached house prices per district 2009 – 2019

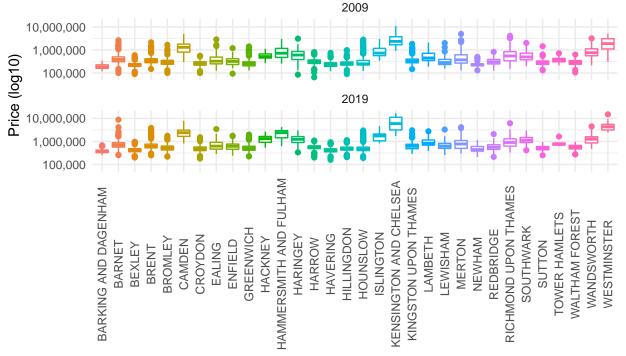
The spread of prices in Hounslow and Westminster has narrowed, while in Kensington and Chelsea it has widened



Movement of flat/masonette prices per district 2009 – 2019 Camden & Kensigton and Chelsea have seen the largest price growth

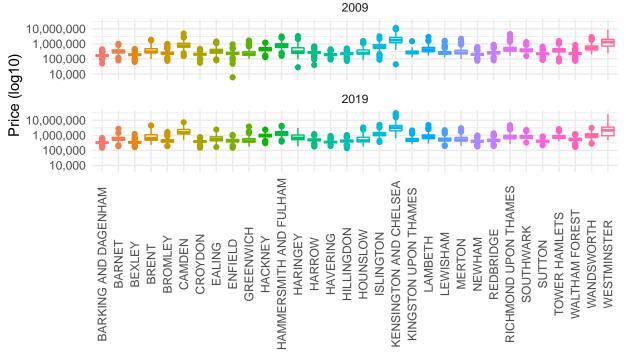


Movement of semi-detached house prices per district 2009 – 2019 Kensington and Chelsea has the biggest jump in prices



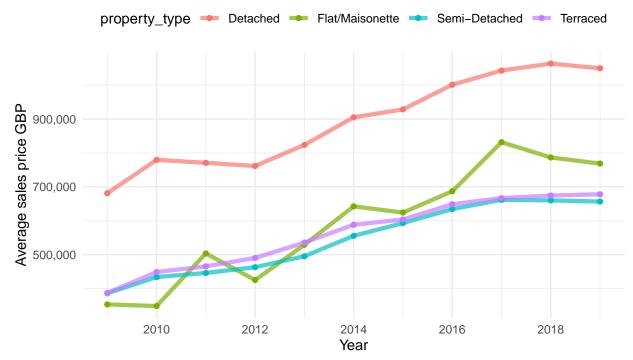
Movement of terraced house prices per district 2009 – 2019

The price increase has been similar across all districts



3.1 How does the avg price increase year on year?

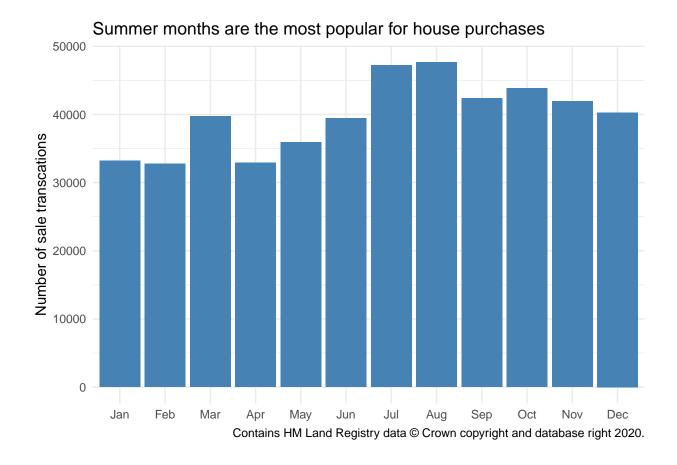
The increase average housing cost by type in Greater London A distinct slowdown in the increase since 2017



Contains HM Land Registry data © Crown copyright and database right 2020.

3.2 Is there a trend of which month most sales occur on?

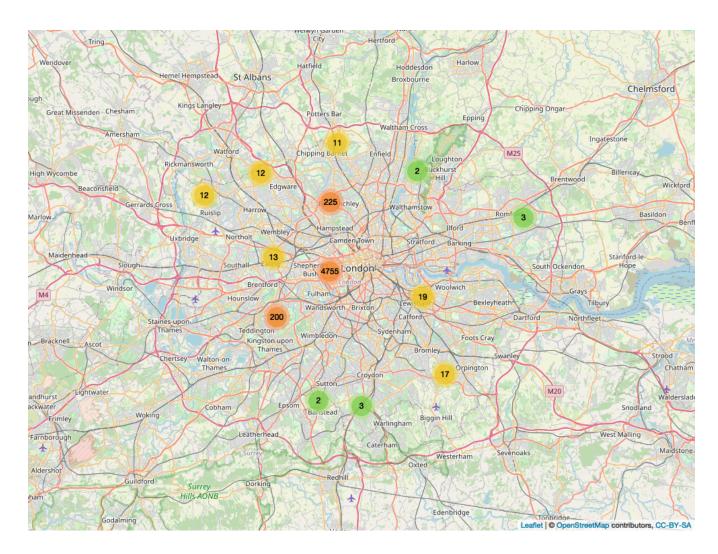
`summarise()` ungrouping output (override with `.groups` argument)



3.3 Which are the 10 most expensive districts based on the number of properties sold over 3 million GBP?

district	$properties_greater_than_3_million_GBP$
KENSINGTON AND CHELSEA	1927
WESTMINSTER	958
CAMDEN	648
MERTON	287
RICHMOND UPON THAMES	268
WANDSWORTH	225
BARNET	217
HAMMERSMITH AND FULHAM	211
HARINGEY	102
ISLINGTON	96

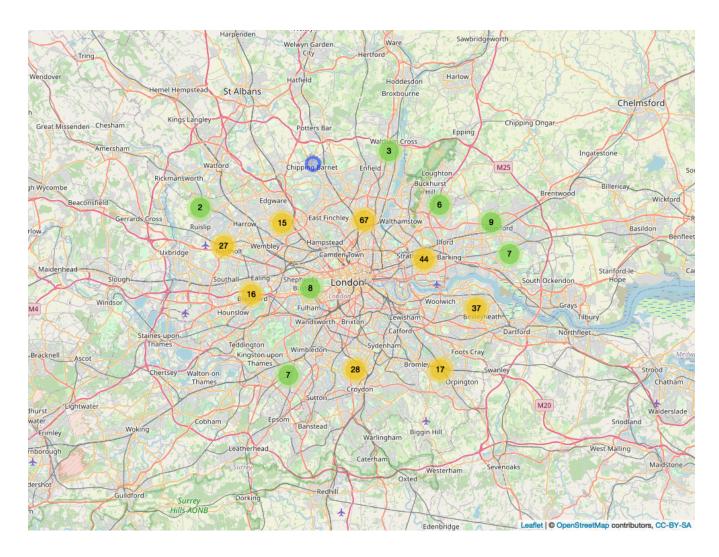
3.4 How are properties sold for greater than 3 million GBP clustered?



3.5 Which are the 10 least expensive districts based on the number of properties under 100K GBP?

district	properties_less_than_100K_GBP
BEXLEY	31
WALTHAM FOREST	22
NEWHAM	21
ENFIELD	20
CROYDON	19
BROMLEY	17
HARROW	16
EALING	13
HARINGEY	12
BARNET	11

3.6 How are properties under 100k GBP clustered?



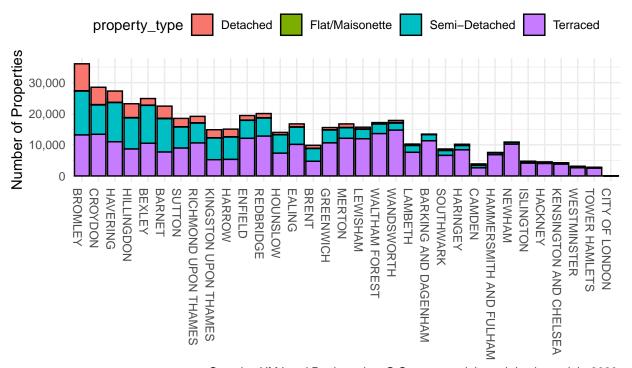
As can be seen from the above analysis some districts contain both top end and bottom end properties, such as Barnet and Haringey.

3.7 Are the sales evenly spread across Greater London?

`summarise()` regrouping output by 'district' (override with `.groups` argument)

Number of observed sales by district

Terraced housing dominates the Greater London housing market



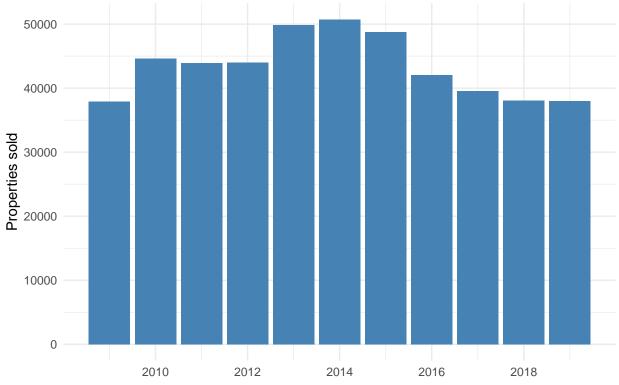
Contains HM Land Registry data © Crown copyright and database right 2020.

3.8 Which type of property sold the most?

property_type	$number_of_properties$
Terraced	275212
Semi-Detached	149397
Detached	49792
Flat/Maisonette	2981

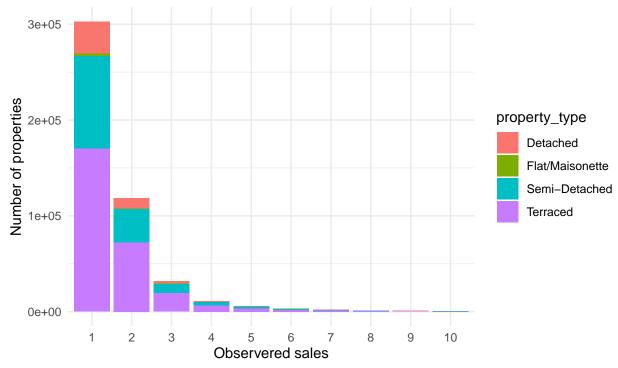
3.9 Which year had the most sales

Since 2014 the volume of house sales has been on the decline.

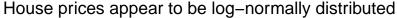


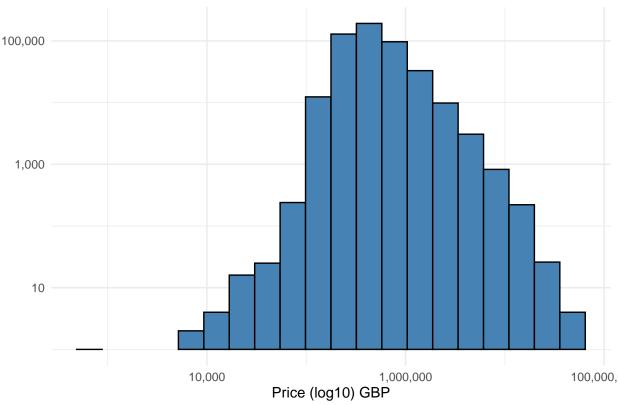
3.10 How often do properties change ownership?

5% of properties sold more than 3 times in the 10 year period. Properties sold more than 10 times excluded from plot.



3.11 How are house prices distributed?





glimpse(lhd)

```
## Rows: 477,382
## Columns: 12
## $ price
                      <int> 283000, 182000, 242000, 420000, 470000, 316250, 2...
                      <fct> BR6 7DQ, TN16 3RA, SM4 4PT, NW2 1QA, BR2 7DD, BR5...
## $ postcode
                     <fct> T, T, T, S, S, S, S, S, T, T, T, T, T, T, S, T, S...
## $ property_type
## $ new build
                     ## $ district
                      <fct> BROMLEY, BROMLEY, MERTON, BARNET, BROMLEY, BROMLE...
## $ address
                      <fct> 6 CLIFTON CLOSE, 18 ST MARYS GREEN, 25 KINGSBRIDG...
## $ transaction_year <int> 2009, 2009, 2009, 2009, 2009, 2009, 2009, 2009, 2009, 2...
## $ transaction_month <int> 12, 9, 4, 9, 10, 6, 6, 2, 2, 2, 10, 9, 3, 7, 10, ...
## $ latitude
                      <dbl> 51.36033, 51.30605, 51.38303, 51.56470, 51.37647,...
## $ longitude
                      <dbl> 0.06765773, 0.02851643, -0.21894415, -0.21283547,...
## $ outward_code
                     <fct> BR6, TN1, SM4, NW2, BR2, BR5, SM2, N21, TW1, RM8,...
## $ num_of_sales
                      <int> 2, 1, 1, 1, 1, 1, 1, 3, 8, 6, 1, 1, 2, 2, 2, 2...
```

nlevels(lhd\$outward_code)

[1] 190

nlevels(lhd\$district)

[1] 33

nlevels(lhd\$postcode)

[1] 79137

nlevels(lhd\$address)

[1] 377515

nlevels(lhd\$property_type)

[1] 4

- 4 Methods/Analysis
- 5 Results
- 6 Conclusion

7 Disclaimers

Contains HM Land Registry data \odot Crown copyright and database right 2020. This data is licensed under the Open Government Licence v3.0

Free Map Tools

Contains Ordnance Survey data © Crown copyright and database right 2020.

Contains Royal Mail data © Royal Mail copyright and database right 2020.

Source: Office for National Statistics licensed under the Open Government Licence v3.0

8 Appendix