The Battle of Neighborhoods

The UK House Price Index: Price cost Analysis

Coursera_Capstone

Content Flow:

- Problem Background and Objective
- Dataset Sources
- Data Cleaning
- Methodology : Data Analysis
- Results
- Data Visualization
- Discussion of Result on Prime Cases
- Conclusion

Problem Background & Objective

- UK current Real Estate Situation
- Remarkable Accommodation
- Major choice Factors
- Stack holder Identifications
- Curiosity to learn the future prediction
- Best Place among all according to requirement
- Profit margin Consideration
- ► How DATA-SCIENCE help them ?

Data Sources:

- Three types of data required:
- 1) Past Data of trending: 'http://landregistry.data.gov.uk/'
- Reliability: Govt. of UK, since 1995 to August 2018
- ▶ 2) Geological Google Map Data: 'Google Cloud Platform/ Google Maps'
- 3) Four squre location data: JSON formatted file. 'https://foursquare.com/'

Data Cleaning:

- Identify the Nan Values.
- Remove all such Rows with out data due to big dataset compare to missing values)
- Give column name as per common understanding with out the use of the 'jargon' naming
- Drop the unnecessary data column after identifying the required columns.
- Check random dataset for the desire pre processing result.

Methodology: (Steps)

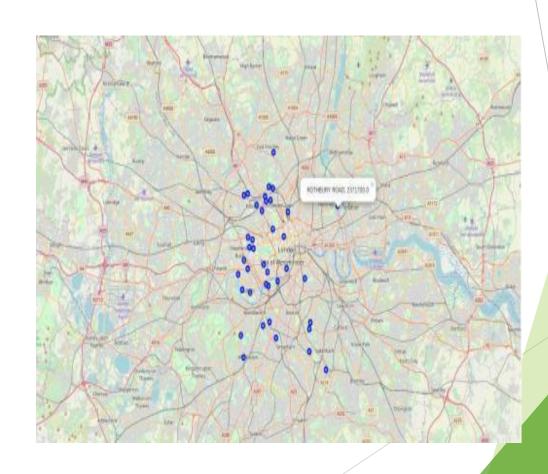
- ▶ 1) First in the script import all the require libraries such as pandas, numpy, matplotlib and folium. We need to import the JSON file of the particular region and requests statement for query analysis.
- 2. First, pass the necessary data from the price paid dataset which includes the transactions received at HM Land Registry in the period from the first to the last day of August 2018. Load that data with the local directory saving path method, which save in jupetor notebook local folder.
- > 3. Then we need to give understandable label to the columns and it is gone by the list method of python, in perfect order of the dataset. But we only require two columns for the initial analysis.
- ▶ 4. The data is cleansed and any data of sales agreements which predates 2016 is dropped from the dataset.
- ▶ 5. The data is further condensed by selecting it only for the city of London which is area of choice in this project. And our main focus is for the certain limit of price range.

Methodology:(Cont....)

- 6. Unique "Street names" in the city of London where recent transactions for sale of property were done are filtered from the dataset.
- > 7. Location coordinates (latitude, longitude) of these street names are fetched by making API calls to Google Maps. A separate one-time Python script was developed to fetch this data and store it in a CSV file. By merging two datasets we can easily filtered out the data for the require street name.
- ▶ 8. The average price of property on each of these streets is determined by taking a mean on recent transactions of sale of property on respective streets.
- 9. Based upon the budget of the client, the current average prices are compared and all recommendations for the locations are made by plotting them on map of London. The locations popups are labelled with the respective street names and their average property price.
- ▶ 10. The recommended locations are further fed into Foursquare API calls to determine various venues in proximity to them. All reported venues are then tabulated and presented to the user.
- ▶ 11. Important facilities like Hospitals, Grocery stores, Elementary schools, High Schools are searched in vicinity of each location and then reported in a tabular form to the user.

Results: Analysis via Visualization What this data say?

Street	Latitude	Longitude	Avg_Price
DULWICH WOOD AVENUE	51.425586	-0.082416	2.297000e+06
SOUTH HILL PARK	51.557134	-0.164343	2.466667e+06
TEIGNMOUTH ROAD	51.550139	-0.214496	2.295000e+06
BURNSALL STREET	51.489042	-0.166883	2.286500e+06
FORDWYCH ROAD	51.551511	-0.206736	2.290000e+06
PORTEN ROAD	51.498603	-0.214120	2.200000e+06
ALBERT BRIDGE ROAD	51.477861	-0.164743	2.383333e+06
EDITH VILLAS	51.491665	-0.206556	2.402500e+06
WESTBOURNE GROVE	51.514797	-0.197071	2.300000e+06
LADBROKE ROAD	51.508776	-0.203410	2.261250e+06



Data Analysis:

Venue Categor	Venue Longitude	Venue Latitude	Venue	Street Longitude	Street Latitude	Street	
Ba	-0.083425	51.424580	Beer Rebellion	-0.082416	51.425586	DULWICH WOOD AVENUE	0
Indian Restaura	-0.086488	51.427795	The Indian Dining Club	-0.082416	51.425586	DULWICH WOOD AVENUE	1
Pu	-0.086168	51.427880	The Paxton	-0.082416	51.425586	DULWICH WOOD AVENUE	2
Train Statio	-0.083959	51.424530	Gipsy Hill Railway Station (GIP)	-0.082416	51.425586	DULWICH WOOD AVENUE	3
Breakfast Spi	-0.083836	51.424425	Brown & Green @ The Station	-0.082416	51.425586	DULWICH WOOD AVENUE	4
Italian Restaura	-0.086131	51.427591	Manuel's Restaurant and Bar	-0.082416	51.425586	DULWICH WOOD AVENUE	5
Bookstor	-0.166277	51.555513	Daunt Books Hampstead	-0.164343	51.557134	SOUTH HILL PARK	6
Lak	-0.165973	51.559300	Hampstead Heath Ponds	-0.164343	51.557134	SOUTH HILL PARK	7
Pa	-0.164921	51.559622	Hampstead Heath	-0.164343	51.557134	SOUTH HILL PARK	8
Indian Restaura	-0.166312	51.555476	Paradise	-0.164343	51.557134	SOUTH HILL PARK	9
Museu	-0.167975	51.555745	Keats House	-0.164343	51.557134	SOUTH HILL PARK	10
Bake	-0.165586	51.554494	karma bread	-0.164343	51.557134	SOUTH HILL PARK	11
Pu	-0.165697	51.554733	The Garden Gate	-0.164343	51.557134	SOUTH HILL PARK	12
Thai Restaura	-0.164737	51.554115	The Little Thai	-0.164343	51.557134	SOUTH HILL PARK	13
Scenic Looko	-0.159639	51.559661	Parliament Hill	-0.164343	51.557134	SOUTH HILL PARK	14
Ca	-0.164975	51.554174	Silverberry Deli & Kitchen	-0.164343	51.557134	SOUTH HILL PARK	15
Pu	-0.168806	51.556968	The Freemasons Arms	-0.164343	51.557134	SOUTH HILL PARK	16
Greek Restaura	-0.165561	51.554423	Zara	-0.164343	51.557134	SOUTH HILL PARK	17
Gastropu	-0.161576	51.553420	The Stag	-0.164343	51.557134	SOUTH HILL PARK	18
Grocery Sto	-0.165119	51.553956	M&S Simply Food	-0.164343	51.557134	SOUTH HILL PARK	19
Italian Restaura	-0.166230	51.555340	Mimmo la Bufala	-0.164343	51.557134	SOUTH HILL PARK	20

	Justinanis	r acimy manie	racinty category	Distance	raumy Lamuus	raciny congnuou
0	DULWICH WOOD AVENUE	Tesco	Grocery Store	1311	51.436213	-0.090561
1	DULWICH WOOD AVENUE	Tesco	Supermarket	3911	51.397043	-0.049559
2	DULWICH WOOD AVENUE	M&S Simply Food	Grocery Store	4513	51.389183	-0.111039
3	DULWICH WOOD AVENUE	Tesco	Grocery Store	3677	51.445271	-0.124966
4	DULWICH WOOD AVENUE	Tesco	Grocery Store	5315	51.438880	-0.155979
5	DULWICH WOOD AVENUE	Sainsbury's Local	Grocery Store	3749	51.446901	-0.124255
6	DULWICH WOOD AVENUE	telferscot primary school	Elementary School	4250	51.442807	-0.137087
7	DULWICH WOOD AVENUE	M&S Streatham Hill Foodhall	Grocery Store	3419	51.439372	-0.126443
8	DULWICH WOOD AVENUE	The Co-operative Food	Grocery Store	2306	51.440017	-0.106273
9	DULWICH WOOD AVENUE	Tesco	Supermarket	3216	51.398649	-0.099187
10	DULWICH WOOD AVENUE	Golden Jubilee Wing	Hospital	4836	51.468522	-0.093121
11	DULWICH WOOD AVENUE	Tesco Express	Grocery Store	3634	51.401783	-0.118253
12	DULWICH WOOD AVENUE	The Co-operative Food	Grocery Store	5529	51.471577	-0.112531
13	DULWICH WOOD AVENUE	Tesco Express	Grocery Store	4642	51.465227	-0.103222
14	DULWICH WOOD AVENUE	M&S Foodhall	Grocery Store	3386	51.426259	-0.131190
15	DULWICH WOOD AVENUE	The Co-operative Food	Supermarket	4366	51.432608	-0.020514
16	DULWICH WOOD AVENUE	Aldi	Supermarket	5169	51.448675	-0.017781
17	DULWICH WOOD AVENUE	The Co-operative Food	Supermarket	4934	51.381969	-0.069717
18	DULWICH WOOD AVENUE	Sainsbury's Local	Grocery Store	3155	51.440406	-0.121180
19	DULWICH WOOD AVENUE	Tesco	Grocery Store	6212	51.470177	-0.028570
20	DULWICH WOOD AVENUE	Balgowan Primary School	Elementary School	3811	51.404480	-0.039182
21	DULWICH WOOD AVENUE	Tesco	Grocery Store	5228	51.468581	-0.112767
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Discussion:

- What they choose from the table data ?
- ▶ Q.1:with Elder Family member...

▶ Q.2: Having children....

▶ Q.3: College student...

Conclusion:

▶ Deciding Factors :

▶Price

▶ Requirements

Thank you, for your participation...

►Questions?