

# **MELBOURNE CITY OPEN DATA PLAYGROUND**

# **CLUE Business establishment's location and industry classification**

# **Exploratory Data Analysis**

Date	Author/Contributor	Change	
17-Nov-2021	Steven Tuften	Initial Draft	

#### **ATTRIBUTIONS**

Jupyter Notebook derivative of data exploration notebook and d2i\_tools.py created by Albert Hon in T2 2021.

#### Package/Library Imports

```
In [1]: import os import to simport time from urllib.request import urlopen import json from datetime import datetime import numby as np import pandas as pd from sodapy import Socrata import geopandas import plotly.express as px from shapely.geometry import Polygon, Point from d2i_tools import * import warnings warnings.simplefilter("ignore")
```

## Constants

```
In [2]: dataset_id = 'vesm-c7r2' # Melbourne CLUE Business establishment's location and industry classification
geoJSON_Id = 'aia8-ryiq' # Melbourne CLUE Block polygons in GeoJSON format

apptoken = os.environ.get("SODAPY_APPTOKEN") # Anonymous app token
domain = "data.melbourne.vic.gov.au"
client = Socrata(domain, apptoken) # Open Dataset connection

WARNING:root:Requests made without an app_token will be subject to strict throttling limits.
```

## [01] Retrieve dataset Metadata

```
print('Selected metadata for the dataset of interest')
          metadata_df[metadata_df.id.isin([dataset_id])].T
          Selected metadata for the dataset of interest
Out[3]:
                                                            Business establishment and industry classifica...
                                                       id
                                                                                            vesm-c7r2
                                                                                           [bs7n-5veh]
                                                parent fxf
                                                             Data collected as part of the City of Melbourn...
                                               description
                                              data_upd_at
                                                                              2021-11-02T22:13:48.000Z
                                              pv_last_wk
                                                                                                   10
                                                                                                   67
                                              pv_last_mth
                                                                                                 2340
                                                 pv_total
                                                                                                  753
                                          download_count
                                               categories
                                                                  [economy, demographics, transportation]
                                         domain_category
                                             domain_tags [business, census of land use and employment, ...
                                                              [{'key': 'Quality_Known-Issues', 'value': 'Non...
                                         domain_metadata
                                   Quality What's-included
                                                                           Full dataset has been included
                                 Quality_Update-frequency
                                                                                       Every two years
                                   Quality_Reliability-level
                                                                                     Reliable and timely
                                     How-to-use_Linked-to
                                                                                                 NaN
            Data-management_Source-data-update-frequency
                                                                                       Every two years
                                    Quality_Known-Issues
                                                                                                 None
                           How-to-use Further-information
                                                                      http://www.melbourne.vic.gov.au/clue
                            Quality_Data-quality-statement A team of up to 6 surveyors conducts a field s...
```

## [02] Display first few rows

Below are the first 3 rows of this dataset:

In [3]: metadata\_df = loadClientDatasetsMetadata(client)

```
In [4]: dataresource = client.get_all(dataset_id)
    dataset = pd.DataFrame.from_dict(dataresource)
    print(f'The shape of dataset is {dataset.shape}.')
    print('Below are the first 3 rows of this dataset:')
    dataset.head(3).T

The shape of dataset is (20036, 11).
```

Out[4]:

	0	1	2
clue_small_area	Melbourne (CBD)	Melbourne (CBD)	Melbourne (CBD)
location	{'latitude': '-37.82121122', 'needs_recoding':	{'latitude': '-37.82121122', 'needs_recoding':	{'latitude': '-37.82121122', 'needs_recoding':
y_coordinate	-37.82121122	-37.82121122	-37.82121122
census_year	2020	2020	2020
anzsic4_code	0	9511	0
x_coordinate	144.9568736	144.9568736	144.9568736
block_id	1	1	1
anzsic4_description	Vacant Space	Hairdressing and Beauty Services	Vacant Space
property_id	108843	108843	108843
bps_base_id	108843	108843	108843
trading_name	62 Rebecca Walk MELBOURNE VIC 3000	14 Rebecca Walk MELBOURNE VIC 3000	86 Rebecca Walk MELBOURNE VIC 3000

# [03] Data Pre-processing

Cast Data types before analysis

```
In [5]: dataset[['census_year', 'anzsic4_code','block_id']] = dataset[['census_year', 'anzsic4_code','block_id']].astype(int)
dataset[['x_coordinate', 'y_coordinate']] = dataset[['x_coordinate', 'y_coordinate']].astype(float)
          dataset = dataset.convert_dtypes() # convert remaining to string
          dataset.dtypes
Out[5]: clue_small_area
                                     string
          location
                                     object
          y_coordinate
                                    float64
          census_year
                                      Int32
          anzsic4_code
                                      Int32
          x_coordinate
                                    float64
          block_id
                                      Int32
          anzsic4_description
                                     string
          property_id
                                     string
          bps_base_id
                                     string
          trading_name
                                     string
          dtype: object
```

#### Are there any missing values?

In [7]: dataset[dataset['x\_coordinate'].isnull()]

Out[7]:

	clue_small_area	location	y_coordinate	census_year	anzsic4_code	x_coordinate	block_id	anzsic4_description	property_id	bps_base_id	trading_name
92	Melbourne (CBD)	NaN	NaN	2020	4512	NaN	5	Takeaway Food Services	101345	101345	Kiosk 12, Campbell Arcade MELBOURNE VIC 3000
93	Melbourne (CBD)	NaN	NaN	2020	0	NaN	5	Vacant Space	101345	101345	Shop 3-4, Campbell Arcade MELBOURNE VIC 3000
94	Melbourne (CBD)	NaN	NaN	2020	0	NaN	5	Vacant Space	101345	101345	Shop 8A, Campbell Arcade MELBOURNE VIC 3000
95	Melbourne (CBD)	NaN	NaN	2020	0	NaN	5	Vacant Space	101345	101345	Shop 5, Campbell Arcade MELBOURNE VIC 3000
96	Melbourne (CBD)	NaN	NaN	2020	0	NaN	5	Vacant Space	101345	101345	Shop 9, Campbell Arcade MELBOURNE VIC 3000
97	Melbourne (CBD)	NaN	NaN	2020	0	NaN	5	Vacant Space	101345	101345	Shop 8, Campbell Arcade MELBOURNE VIC 3000
98	Melbourne (CBD)	NaN	NaN	2020	0	NaN	5	Vacant Space	101345	101345	Shop 11, Campbell Arcade MELBOURNE VIC 3000
99	Melbourne (CBD)	NaN	NaN	2020	5910	NaN	5	Internet Service Providers and Web Search Portals	101345	101345	Shop 6-7, Campbell Arcade MELBOURNE VIC 3000
100	Melbourne (CBD)	NaN	NaN	2020	4244	NaN	5	Newspaper and Book Retailing	101345	101345	Shop 10, Campbell Arcade MELBOURNE VIC 3000
101	Melbourne (CBD)	NaN	NaN	2020	4242	NaN	5	Entertainment Media Retailing	101345	101345	Shop 1, Campbell Arcade MELBOURNE VIC 3000
102	Melbourne (CBD)	NaN	NaN	2020	0	NaN	5	Vacant Space	101345	101345	Kiosk 13, Campbell Arcade MELBOURNE VIC 3000
4157	Melbourne (CBD)	NaN	NaN	2020	2630	NaN	34	Electricity Distribution	110942	110942	Substation Opposite 77 Queen Street, MELBOURNE
6292	Melbourne (CBD)	NaN	NaN	2020	7530	NaN	45	Local Government Administration	111492	111492	Melbourne Visitor Booth Bourke Street MELBOURN
12254	Carlton	NaN	NaN	2020	2630	NaN	259	Electricity Distribution	111450	111450	Substation Opposite 1 Barry Street, CARLTON VI
12660	Carlton	NaN	NaN	2020	2620	NaN	273	Electricity Transmission	111010	111010	Substation 5, MacPherson Street CARLTON NORTH $\dots$
13766	North Melbourne	NaN	NaN	2020	0	NaN	380	Vacant Space	103467	103467	254 Errol Street NORTH MELBOURNE VIC 3051
13783	West Melbourne (Residential)	NaN	NaN	2020	0	NaN	403	Vacant Space	109755	109755	487-491 Victoria Street WEST MELBOURNE VIC 3003
14274	West Melbourne (Industrial)	NaN	NaN	2020	2630	NaN	502	Electricity Distribution	111448	111448	Substation Lloyd Street WEST MELBOURNE VIC 3003
14447	Kensington	NaN	NaN	2020	8922	NaN	509	Nature Reserves and Conservation Parks Operation	565481	565481	Bellair Street Reserve Bellair Street KENSINGT
14647	East Melbourne	NaN	NaN	2020	2630	NaN	604	Electricity Distribution	545681	545681	Substation 36 Opposite 96 Simpson Street, EAST
14790	East Melbourne	NaN	NaN	2020	7530	NaN	613	Local Government Administration	111447	111447	Building Opposite 172 Powlett Street, EAST MEL
15544	Southbank	NaN	NaN	2020	2630	NaN	739	Electricity Distribution	567772	567772	Substation 99A Sturt Street SOUTHBANK VIC 3006
17623	Parkville	NaN	NaN	2020	2630	NaN	901	Electricity Distribution	545572	545572	Substation 146 Opposite 29-53 levers Street, P
17916	Parkville	NaN	NaN	2020	5809	NaN	931	Other Telecommunications Services	107423	107423	Antenna Off Brens Drive, PARKVILLE VIC 3052
17950	Parkville	NaN	NaN	2020	5809	NaN	931	Other Telecommunications Services	525698	525698	Antenna Off Brens Drive, PARKVILLE VIC 3052
19067	Docklands	NaN	NaN	2020	9499	NaN	1108	Other Repair and Maintenance n.e.c.	678365	678365	125 Harbour Esplanade DOCKLANDS VIC 3008
19736	North Melbourne	NaN	NaN	2020	2812	NaN	2385	Sewerage and Drainage Services	618476	618476	Pumping Station No.2 330 Macaulay Road KENSING
19737	North Melbourne	NaN	NaN	2020	2812	NaN	2385	Sewerage and Drainage Services	618527	618527	Pumping Station No.5 Sutton Street NORTH MELBO
19995	Kensington	NaN	NaN	2020	2812	NaN	2540	Sewerage and Drainage Services	618478	618478	Pumping Station No.1 Smith Street KENSINGTON V

## Drop rows with no latitude or longitude?

We will not be using the latitude and longitude at property level so we can leave these two rows in the dataset.

In [8]: ## If we wanted to drop these rows we would use the following two commands.

#dataset = dataset.dropna(axis=0)
#print(dataset.isnull().sum())

# [04] Analyse data in Aggregate

Count of Business Establishments by CLUE small area

```
In [9]: groupbyfields = ['clue_small_area']
aggregatebyfields = {'anzsic4_code': ["count"]}

maxByBlock = pd.DataFrame(dataset.groupby(groupbyfields, as_index=False).agg(aggregatebyfields))
maxByBlock.head(10)

Out[9]:
```

	clue_small_area	anzsic4_code
		count
0	Carlton	1346
1	Docklands	1648
2	East Melbourne	645
3	Kensington	566
4	Melbourne (CBD)	11056
5	Melbourne (Remainder)	369
6	North Melbourne	1308
7	Parkville	449
8	Port Melbourne	674
9	South Yarra	63

# Count of Business Establishments by ANZSIC4 Code

```
In [10]: groupbyfields = ['anzsic4_description']
aggregatebyfields = {'anzsic4_code': ["count"]}

maxByBlock = pd.DataFrame(dataset.groupby(groupbyfields, as_index=False).agg(aggregatebyfields))
maxByBlock.head(10)
```

Out[10]:

	anzsic4_description	alizaic4_code	
		count	
0	Accommodation	342	
1	Accounting Services	249	
2	Adult, Community and Other Education n.e.c.	68	
3	Advertising Services	66	
4	Aged Care Residential Services	13	
5	Air Conditioning and Heating Services	4	
6	Air and Space Transport	12	
7	Aircraft Manufacturing and Repair Services	1	
8	Airport Operations and Other Air Transport Sup	4	
9	Alternative Health Services	23	

Count of Business Establishments by ANZSIC4 Code & CLUE small area

Out[11]:

```
In [11]: groupbyfields = ['clue_small_area','anzsic4_description']
    aggregatebyfields = {'anzsic4_code': ["count"]}

maxByBlock = pd.DataFrame(dataset.groupby(groupbyfields, as_index=False).agg(aggregatebyfields))
    maxByBlock.head(40)
```

clue\_small\_area anzsic4\_description anzsic4\_code 0 86 Carlton Accommodation Carlton Accounting Services Adult, Community and Other Education n.e.c. Carlton 3 Carlton Advertising Services Carlton Aged Care Residential Services 5 Carlton Alternative Health Services Amusement Parks and Centres Operation Carlton Carlton Antique and Used Goods Retailing Architectural Services 9 Carlton Arts Education 10 Carlton Automotive Electrical Services 11 Carlton Bakery Product Manufacturing (Non-factory based) 12 Carlton 13 Carlton Book Publishing 14 Carlton Brothel Keeping and Prostitution Services 15 Building and Other Industrial Cleaning Services Carlton Business and Professional Association Services 16 Carlton 17 Carlton Cafes and Restaurants 170 18 Carlton Car Retailing 19 Catering Services 20 Carlton Child Care Services 21 Carlton Chiropractic and Osteopathic Services 22 Carlton Clothing Manufacturing 23 Clothing Retailing Carlton 24 Clothing and Footwear Repair Clothing and Footwear Wholesaling 25 Carlton 26 Clubs (Hospitality) Carlton 27 Carlton Commission-Based Wholesaling 28 Communications Equipment Manufacturing Carlton 29 Computer System Design and Related Services 30 Carlton Computer and Computer Peripheral Retailing 31 Carlton Concreting Services 32 Convenience Store 10 Carlton 33 Creative Artists, Musicians, Writers and Perfo... Carlton 34 Credit Union Operation 35 Carlton Defence 36 Carlton Dental Services 37 Carlton Educational Support Services 38 Carlton Electrical Services 39 Electrical, Electronic and Gas Appliance Retai...

Count of Business Establishments by Block Id

```
In [12]: groupbyfields = ['block_id']
aggregatebyfields = {'anzsic4_code': ["count"]}

maxByBlock = pd.DataFrame(dataset.groupby(groupbyfields, as_index=False).agg(aggregatebyfields))
maxByBlock.head(10)

Out[12]:
Plack id aggregate of the control of
```

	block_id	anzsic4_code
		count
0	1	41
1	2	4
2	4	47
3	5	11
4	6	44
5	11	93
6	12	99
7	13	96
8	14	284
9	15	364

## Count of Business Establishments by CLUE small area, Block Id and ANZSIC4 Code

```
In [13]: groupbyfields = ['clue_small_area','block_id','anzsic4_description']
aggregatebyfields = {'anzsic4_code': ["count"]}

maxByBlock = pd.DataFrame(dataset.groupby(groupbyfields, as_index=False).agg(aggregatebyfields))
maxByBlock.head(10)
```

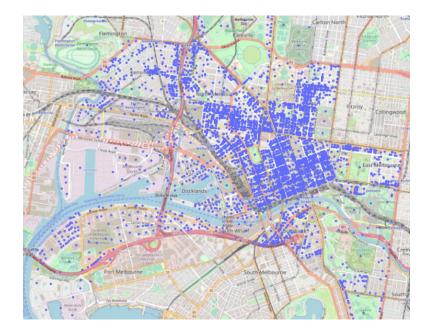
## Out[13]:

	clue_small_area block_id		anzsic4_description	anzsic4_code
				count
0	Carlton	201	Nature Reserves and Conservation Parks Operation	1
1	Carlton	201	Other Goods and Equipment Rental and Hiring n	1
2	Carlton	202	Accommodation	1
3	Carlton	203	Accommodation	3
4	Carlton	203	Cafes and Restaurants	2
5	Carlton	203	Child Care Services	2
6	Carlton	203	Creative Artists, Musicians, Writers and Perfo	1
7	Carlton	203	Vacant Space	1
8	Carlton	204	Accommodation	2
9	Carlton	204	Cafes and Restaurants	1

# Plot Business Establishments by Location on map

```
In [14]: groupbyfields = ['clue_small_area','block_id','y_coordinate','x_coordinate']
          aggregatebyfields = {'anzsic4_code': ["count"]}
          businessesByLocn = pd.DataFrame(dataset.groupby(groupbyfields, as_index=False).agg(aggregatebyfields))
          businessesByLocn.head(10)
Out[14]:
              clue_small_area block_id y_coordinate x_coordinate anzsic4_code
                                201
                                       -37.794677
                                                   144.965947
                     Carlton
                     Carlton
                                 201
                                       -37.794367
                                                   144.966228
                     Carlton
                                      -37.794573
                                                   144.965299
                                       -37.796707
                                                   144.965534
                                       -37.796680
                                                   144.964900
                     Carlton
                                203
                                      -37.796307
                                                   144 965281
                                                   144.965304
                     Carlton
                                203
                                       -37.796147
                     Carlton
                                 203
                                       -37.796069
                                                   144.965138
                                                   144.965758
                                       -37.795946
                                                   144.965213
In [15]: fig = px.scatter_mapbox(businessesByLocn, lat="y_coordinate", lon="x_coordinate",
                                    hover_name="clue_small_area",
hover_data=["clue_small_area", "block_id"],
                                    title='Business Establishments by Location for 2020',
                                    zoom=12.5,
                                    center = {"lat": -37.813, "lon": 144.945},
                                    width=950, height=800)
         widtn=950, height=800)
fig.update_layout(mapbox_style="open-street-map")
fig.show()
```

## Business Establishments by Location for 2020



# **Plot Dwelling Density by Block**

```
In [16]: groupbyfields = ['block_id', 'clue_small_area', 'anzsic4_description']
    aggregatebyfields = ('anzsic4_code': ["count"]}

businessesByBlock = pd.DataFrame(datasepuby(groupbyfields, as_index=False).agg(aggregatebyfields))
businessesByBlock.columns = businessesByBlock.columns.map(''.join) # flatten column header
businessesByBlock.rename(columns={'clue_small_area': 'clue_area'}, inplace=True) #rename to match GeoJSON extract
businessesByBlock.rename(columns=('anzsic4_codecount': 'business_count'}, inplace=True) #rename to match GeoJSON extract
businessesByBlock.head(10)
Out[16]:
```

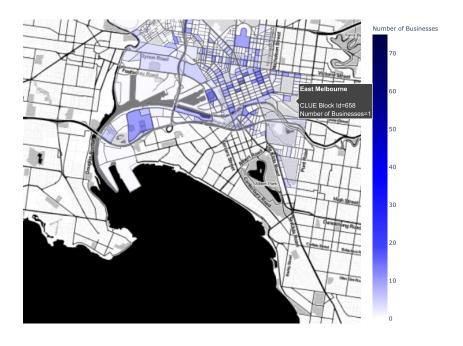
	block_id	clue_area	anzsic4_description	business_count
0	1	Melbourne (CBD)	Air and Space Transport	1
1	1	Melbourne (CBD)	Architectural Services	1
2	1	Melbourne (CBD)	Cafes and Restaurants	2
3	1	Melbourne (CBD)	Computer System Design and Related Services	1
4	1	Melbourne (CBD)	Convenience Store	1
5	1	Melbourne (CBD)	Credit Reporting and Debt Collection Services	1
6	1	Melbourne (CBD)	Electricity Distribution	1
7	1	Melbourne (CBD)	General Insurance	1
8	1	Melbourne (CBD)	Hairdressing and Beauty Services	1
9	1	Melbourne (CBD)	Liquor Retailing	1

#### Get Block Polygon data in GeoJSON format

Load the CLUE Blocks in GeoJSON format and verify the location keys.

Illustrate Business Establishment Density using a Chloropleth Map using Block regions defined by the GeoJSON data

#### Business Establishments by CLUE Block Id for 2020



In [ ]: