

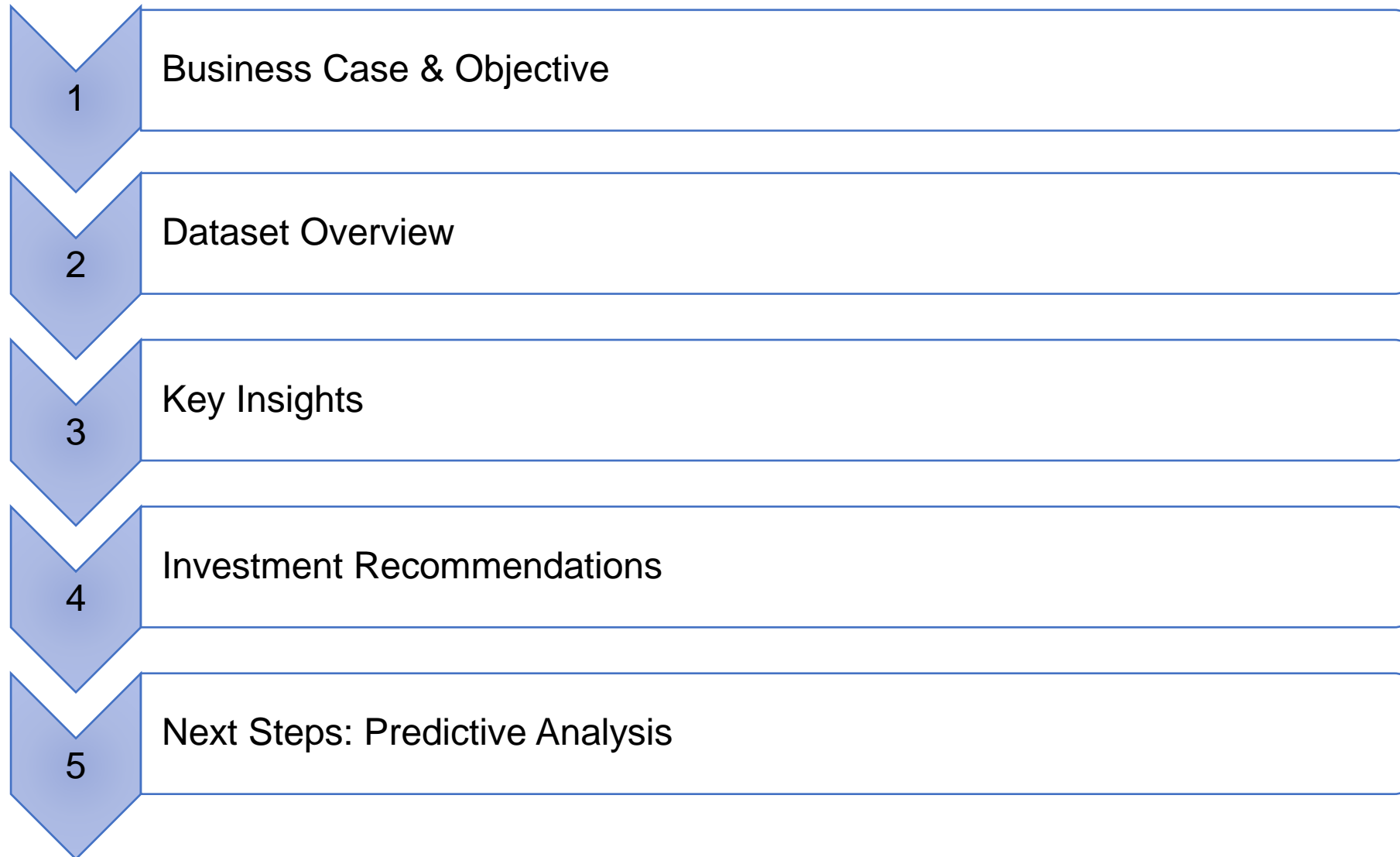
“Descriptive and Explorative Analysis for Smart Real Estate”

Course: Data Science for Business

Study Program: Business Consulting Masters

WiSe 24-25

Agenda



1. Business Case & Objective (1/2)

Business Case



1. BUSINESS CASE & OBJECTIVE

2. DATASET OVERVIEW

3. KEY INSIGHTS

4. INVESTMENT IDEAS

5. NEXT STEPS: PREDICTIVE ANALYSIS

1. Business Case & Objective (2/2)

Objectives



- 1. Analyze and assess the UK housing dataset.**
- 2. Identify key trends and patterns in house prices and sales.**
- 3. Propose actionable investment strategies based on descriptive insights.**
- 4. Plan deeper predictive analysis for future insights (Phase 2).**

2. Dataset Overview (1/4)

Dataset Source: <https://www.kaggle.com/hm-land-registry/uk-housing-prices-paid>

```
In [22]: df.head()
```

```
Out[22]:
```

```
Transaction unique identifier  Price  Date of Transfer \
0 {81B82214-7FBC-4129-9F6B-4956B4A663AD}  25000  1995-08-18 00:00
1 {8046EC72-1466-42D6-A753-4956BF7CD8A2}  42500  1995-08-09 00:00
2 {278D581A-5BF3-4FCE-AF62-4956D87691E6}  45000  1995-06-30 00:00
3 {1D861C06-A416-4865-973C-4956DB12CD12}  43150  1995-11-24 00:00
4 {DD8645FD-A815-43A6-A7BA-4956E58F1874}  18899  1995-06-23 00:00
```

```
Property Type Old/New Duration  Town/City  District \
0          T      N          F    OLDHAM    OLDHAM
1          S      N          F    GRAYS    THURROCK
2          T      N          F HIGHBRIDGE  SEDGEMOOR
3          T      N          F   BEDFORD  NORTH BEDFORDSHIRE
4          S      N          F  WAKEFIELD    LEEDS
```

```
County PPDCategory Type Record Status - monthly file only
0 GREATER MANCHESTER          A          A
1 THURROCK          A          A
2 SOMERSET          A          A
3 BEDFORDSHIRE          A          A
4 WEST YORKSHIRE          A          A
```

```
In [23]: df.columns
```

```
Out[23]:
```

```
Index(['Transaction unique identifier', 'Price', 'Date of Transfer',  
      'Property Type', 'Old/New', 'Duration', 'Town/City', 'District',  
      'County', 'PPDCategory Type', 'Record Status - monthly file only'],  
      dtype='object')
```

```
In [27]: df.shape
```

```
Out[27]: (22489348, 11)
```

Data Overview

Columns and records available

2. Dataset Overview (2/4)

```
In [25]: df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 22489348 entries, 0 to 22489347
Data columns (total 11 columns):
#   Column                                Dtype
---  -
0   Transaction unique identifier         object
1   Price                                int64
2   Date of Transfer                     object
3   Property Type                        object
4   Old/New                             object
5   Duration                             object
6   Town/City                           object
7   District                            object
8   County                              object
9   PPDCategory Type                    object
10  Record Status - monthly file only    object
dtypes: int64(1), object(10)
memory usage: 1.8+ GB
```

```
In [26]: df.isnull().sum()
Out[26]:
Transaction unique identifier    0
Price                            0
Date of Transfer                 0
Property Type                    0
Old/New                          0
Duration                         0
Town/City                        0
District                         0
County                           0
PPDCategory Type                 0
Record Status - monthly file only 0
dtype: int64
```

Datatype of Columns

No. of null values

2. Dataset Overview (3/4)

- *Extracting `Year` and `Month` from `Date of Transfer`*
- *Converting datatype of `Date of Transfer` to datetime*

```
...: pandarallel.initialize()
...: df['Date of Transfer'] = df['Date of Transfer'].parallel_apply(lambda x : x.split(' ')[0])
...: df['Year'] = df['Date of Transfer'].str[:4]
...: df['Date of Transfer'].iloc[1].split(" ")[0]
...: df['Date of Transfer'] = df['Date of Transfer'].parallel_apply(lambda x:
__import__('datetime').datetime.strptime(x, '%Y-%m-%d'))
...: df['Year'] = df['Date of Transfer'].dt.year
...: df['Month'] = df['Date of Transfer'].dt.month
INFO: Pandarallel will run on 8 workers.
INFO: Pandarallel will use standard multiprocessing data transfer (pipe) to transfer data between the
main process and workers.
```

WARNING: You are on Windows. If you detect any issue with pandarallel, be sure you checked out the
Troubleshooting page:
<https://nalepae.github.io/pandarallel/troubleshooting/>

```
In [4]: df.head()
```

```
Out[4]:
```

	Price	Date of Transfer	Property Type	Old/New	Town/City \
0	25000	1995-08-18	T	N	OLDHAM
1	42500	1995-08-09	S	N	GRAYS
2	45000	1995-06-30	T	N	HIGHBRIDGE
3	43150	1995-11-24	T	N	BEDFORD
4	18899	1995-06-23	S	N	WAKEFIELD

	District	County	Year	Month
0	OLDHAM	GREATER MANCHESTER	1995	8
1	THURROCK	THURROCK	1995	8
2	SEDGEMOOR	SOMERSET	1995	6
3	NORTH BEDFORDSHIRE	BEDFORDSHIRE	1995	11
4	LEEDS	WEST YORKSHIRE	1995	6

Drop unwanted columns

```
In [39]: df.drop(columns = 'Transaction unique identifier', axis = 1, inplace = True)
...: df.drop(columns = 'Duration', axis = 1, inplace = True)
...: df.drop(columns = 'PPDCategory Type', axis = 1, inplace = True)
...: df.drop(columns = 'Record Status - monthly file only', axis = 1, inplace = True)
...: df.shape
Out[39]: (22489348, 7)
```


3. Key Insights (1/9)

Exploratory Data Analysis for Investments



In what should we invest ?

When should we invest ?

Where should we invest ?

1. BUSINESS CASE & OBJECTIVE

2. DATASET OVERVIEW

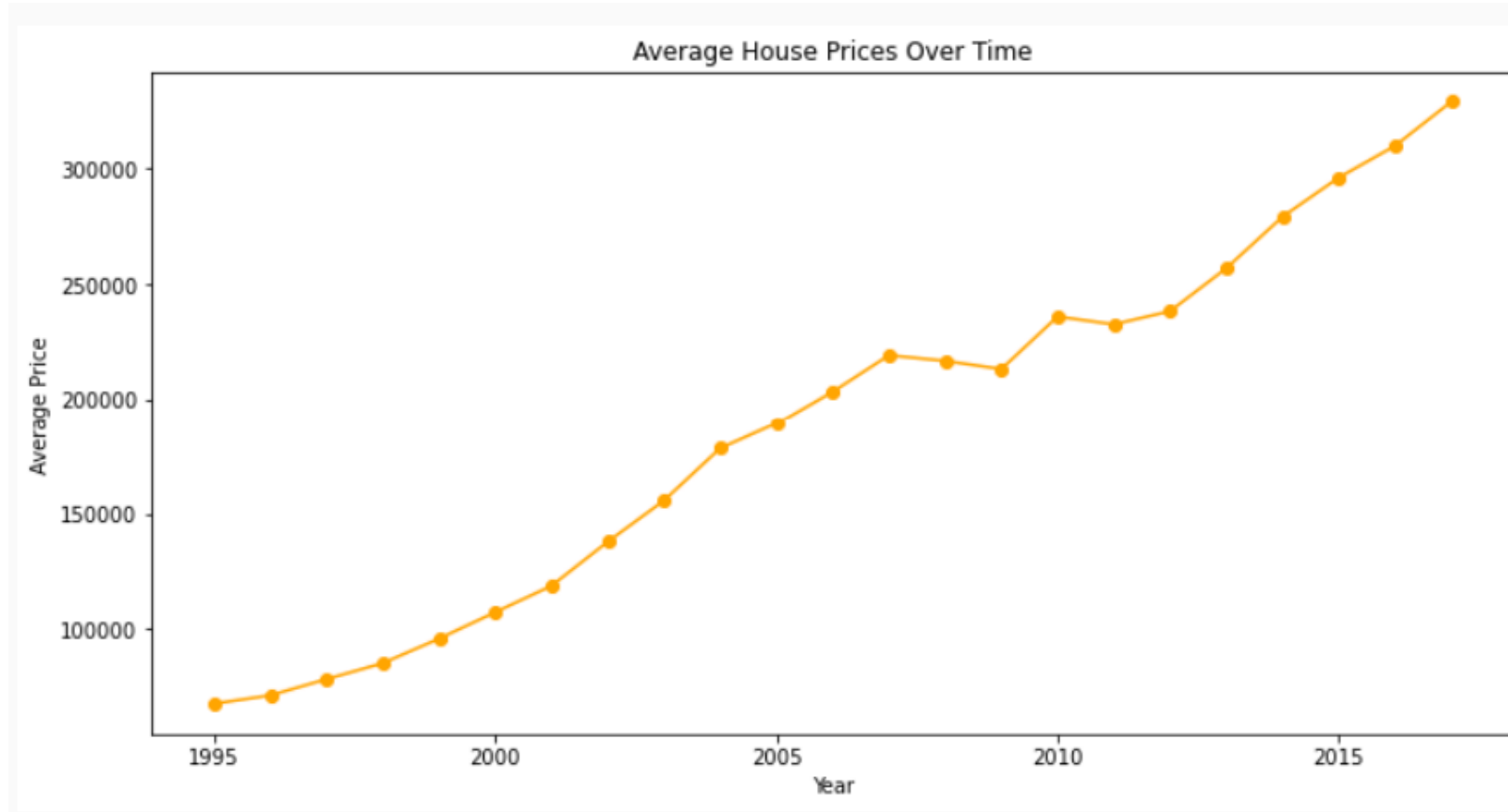
3. KEY INSIGHTS

4. INVESTMENT IDEAS

5. NEXT STEPS: PREDICTIVE ANALYSIS

3. Key Insights (2/9)

Overall Sales price trend analysis



- Avg Sales Price follows upward trend throughout the years (1995-2017)

Fig.1

3. Key Insights (3/9)

Overall Sales transactions trend analysis

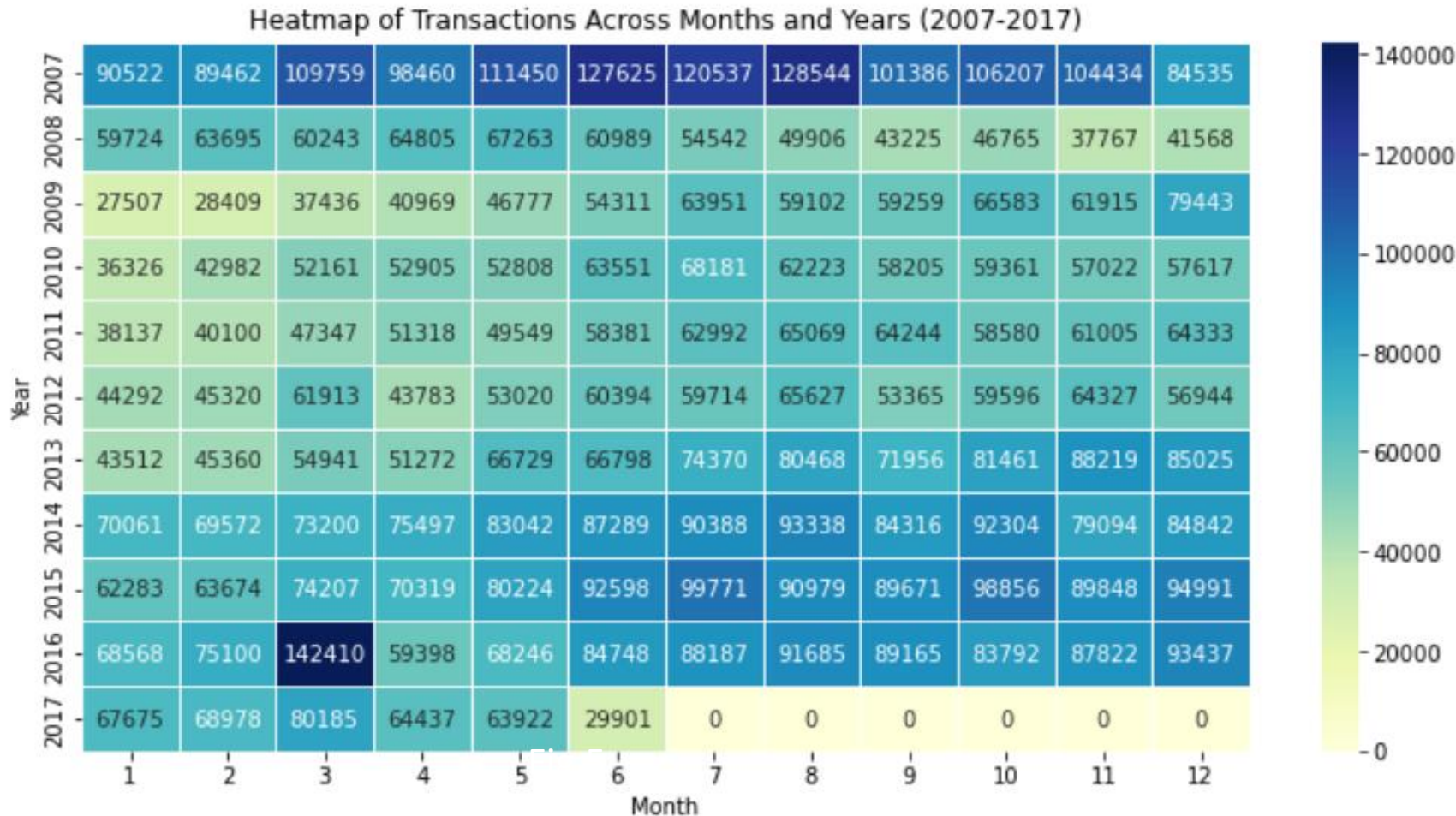


Fig.2

- High in 2007
- Drops from 2008
- Recovery starts from 2013
- Becomes stable in 2014, 2015, 2016
- Low in Jan and Feb
- Good from mid year

3. Key Insights (4/9)

Property analysis 1/2

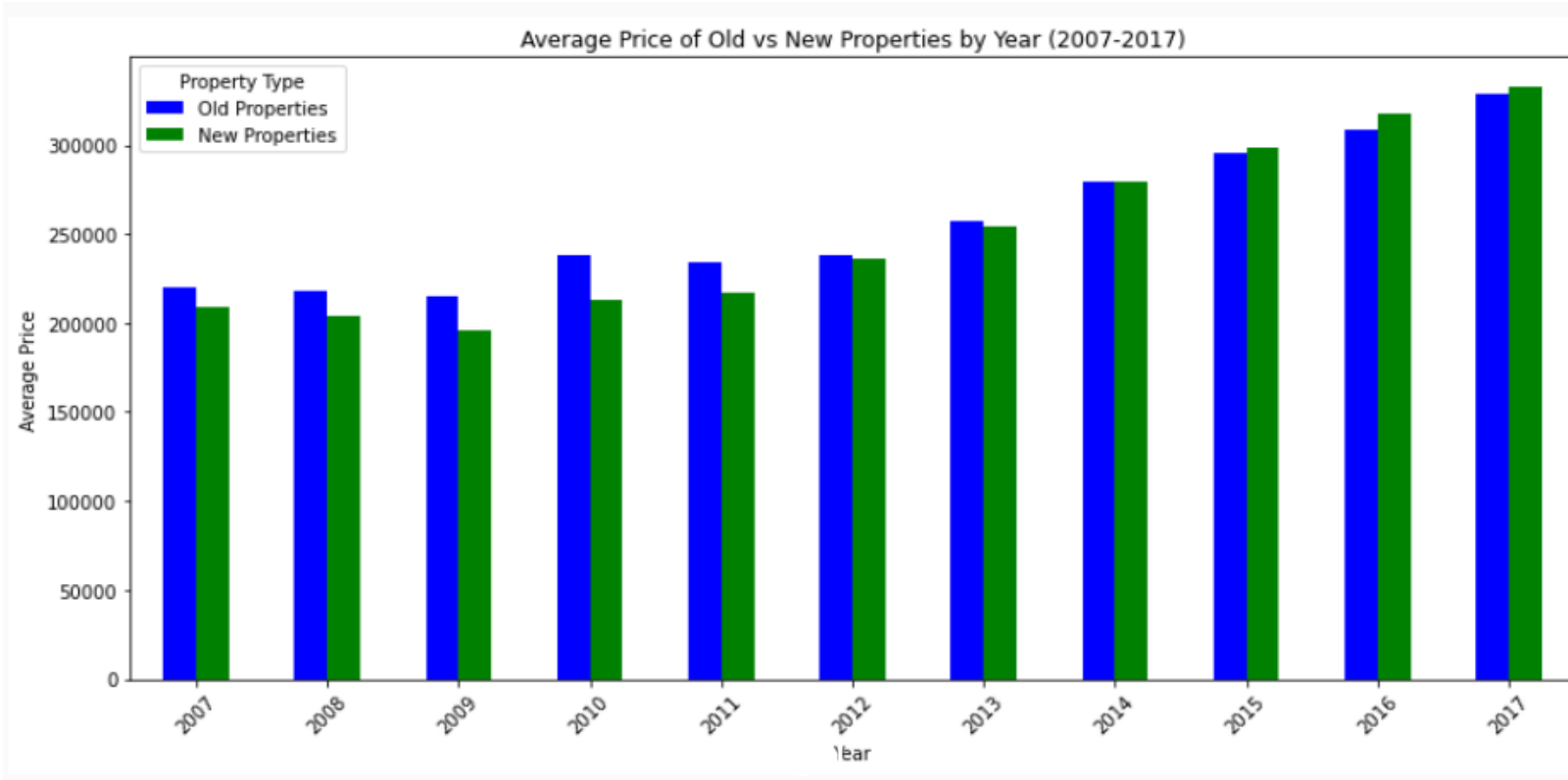


Fig.3

- 2007 – 2013 – Avg price of new property < Avg price of old property
- 2014 – Avg price of new property equals Avg price of old property
- 2015 – 2017 – Avg price of new property > Avg price of old property

3. Key Insights (5/9)

Property analysis 2/2

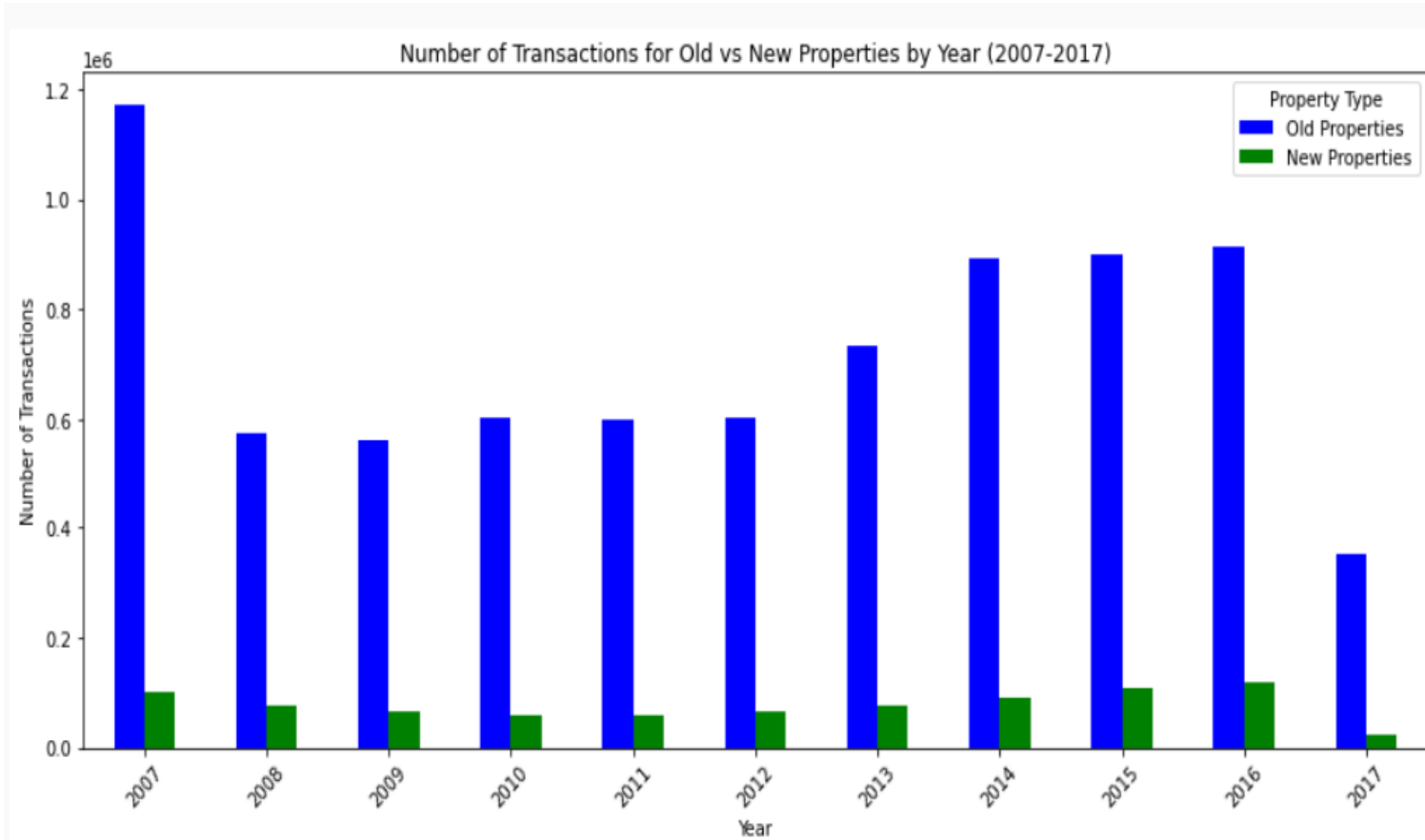


Fig.4

- The number of transactions of old properties beats the number of transactions of new properties

3. Key Insights (6/9)

Property type analysis 1/2

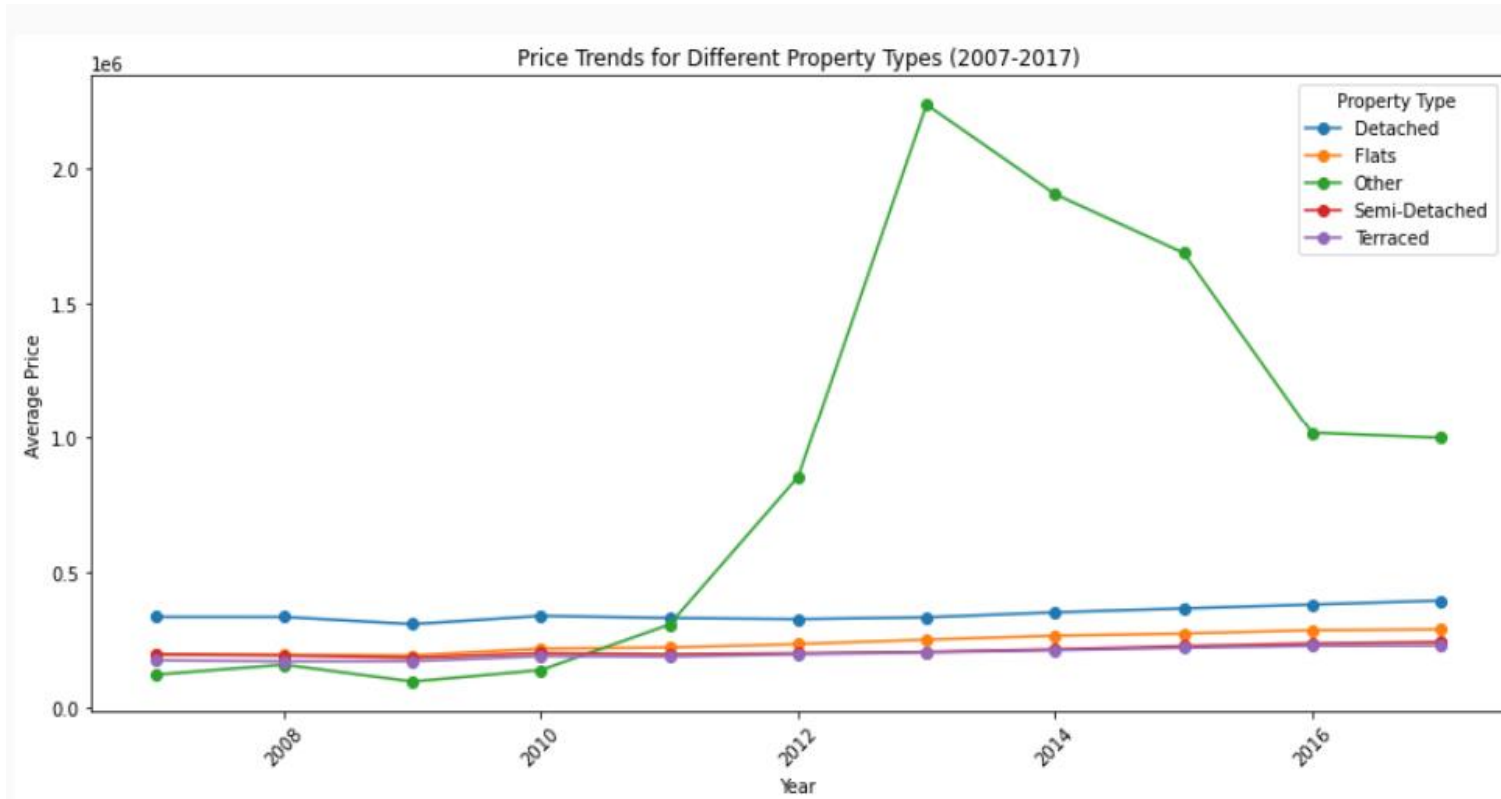


Fig.5

- The avg price of ` Other ` property type is high from 2011 onwards
- The avg price of ` Terraced ` property is the lowest through 2007-2017

3. Key Insights (7/9)

Property type analysis 2/2

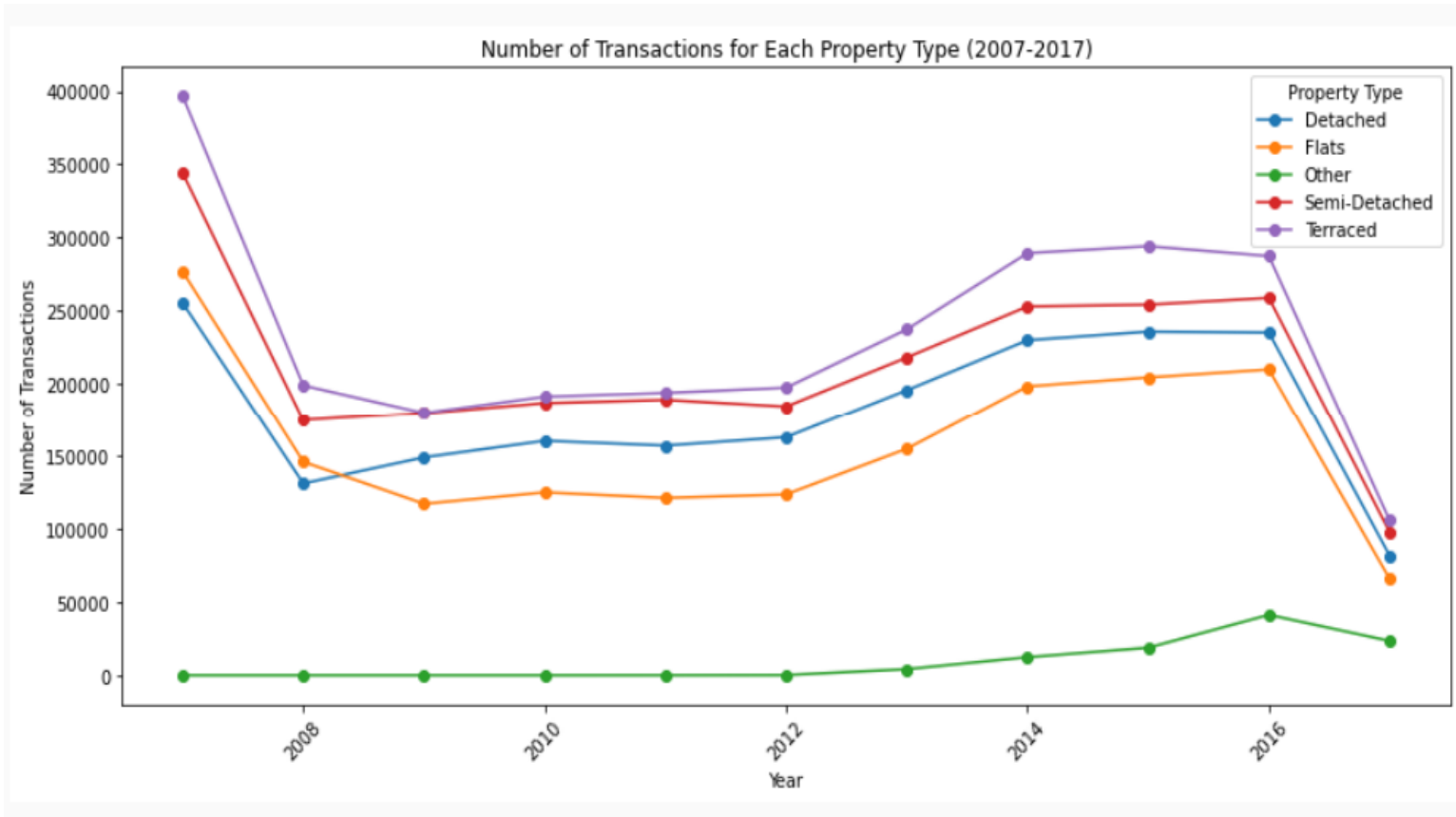


Fig.6

- Number of transactions is the lowest for `Other` property type.
- Number of transactions remains high for `Terraced` property type through 2007-2017

3. Key Insights (8/9)

Sales transactions trend analysis 1/2

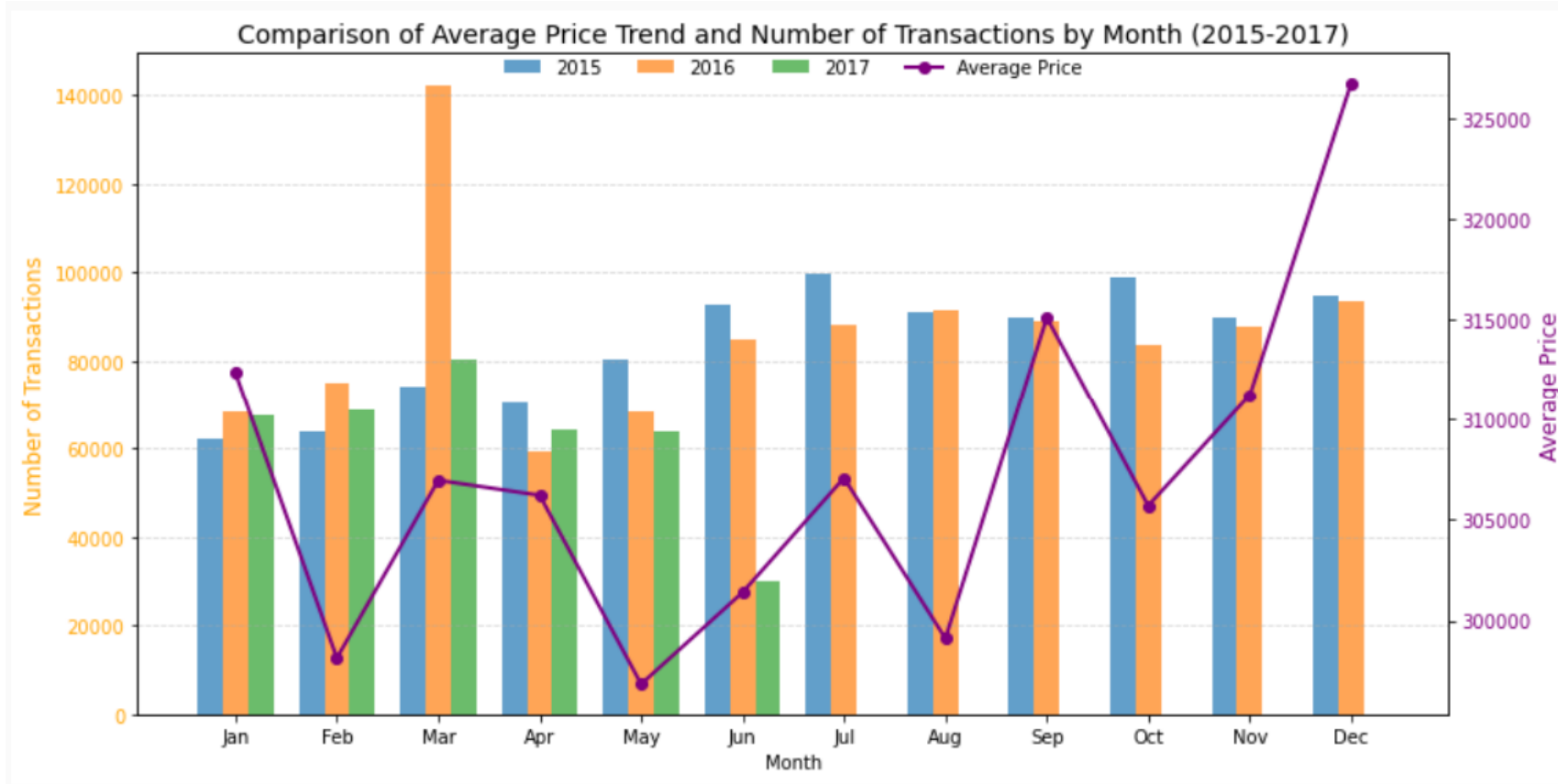


Fig.7

- Prices drop the most during February and May
- Prices attain it's peak during September and December

3. Key Insights (9/9)

Sales transactions trend analysis 2/2

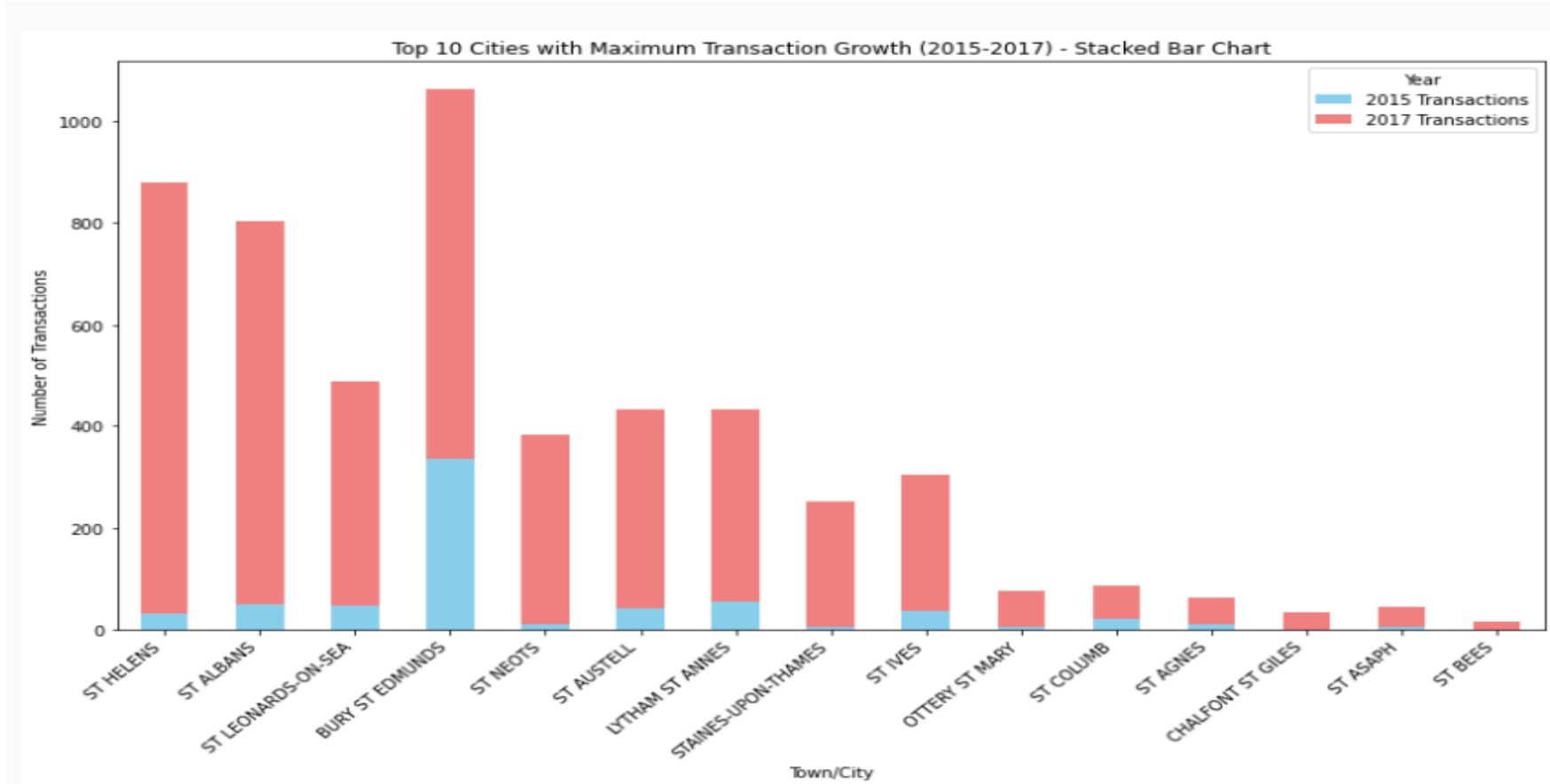


Fig.8

- 14 out of 15 cities belong to England
- 2017's Transaction: 6-month performance surpasses entire 2015

Should I invest on old or new property?

Invest in old property

- Enter new market
- Rent for steady income
- Modernize, sell high & improve ROI

Develop marketing strategies to increase sales of acquired new property

Which type of property to focus on?

- Focus more on high transaction volume properties ensuring consistent cash flow
- Keep a small, targeted portfolio of "Other" properties (for long term growth)

What is the right time to invest in properties?

- Consider buying properties in Feb & May to maximize returns when prices surge (e.g., March, July, September & December)

In which cities to invest ?

- Cities with high transaction growth (from England) Eg. Bury St Edmunds, St Helens, St Albans

5. Next Steps: Predictive Analysis (Phase - 2)

