

King County Dataset

Exploratory Data Analysis (EDA)

Last Week

- Introduction to Machine Learning
- Unsupervised / Supervised Learning
- Batch / Online Learning
- Instance-based / Model-based learning
- Underfitting and Overfitting
- Training, validation and test sets

End-to-End Machine Learning

Aurélien Géron, ***Hands-on-Machine Learning***

1. Look at the big picture
2. Get the data and set aside a test set
3. Discover and visualise the data to gain insights
4. Prepare the data for Machine Learning algorithms
5. Identify a suitable metric for evaluating the task
6. Select a model and train it
7. Fine-tune your model
8. Present your solution
9. Launch, monitor and maintain your system



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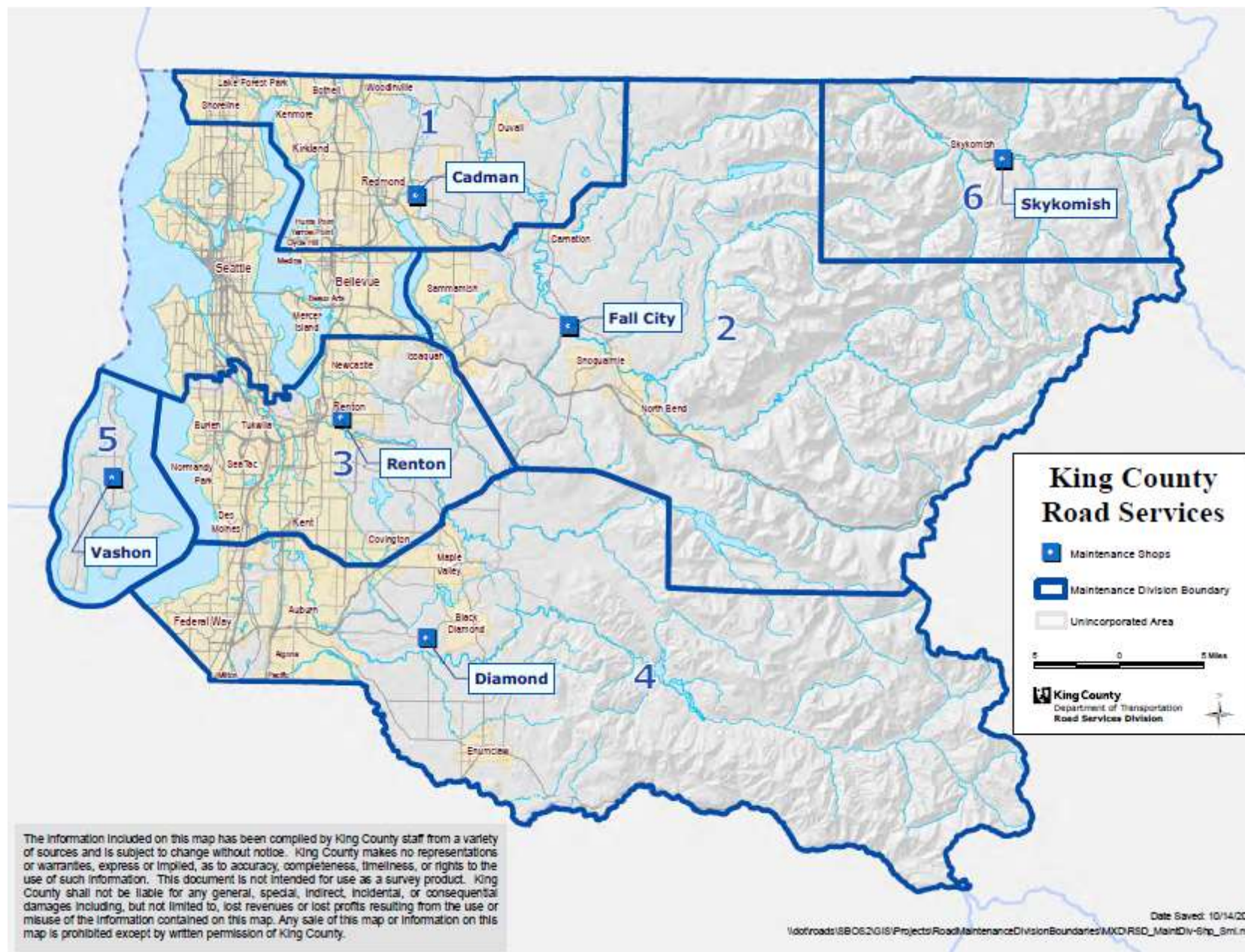
1. Frame the Problem

- We want to be able to **predict the price of houses** in King County, Washington, US.
- Questions for you to consider:
 - Is it supervised, unsupervised, or reinforcement learning?
 - Is it a classification task, a regression task or something else?
 - Should you use batch learning or online learning techniques?



King County, Washington, is the state's 11th largest county in size but its largest in population, with over 1.7 million people and 39 cities and towns.

Washington State
map courtesy US Census Bureau



2. Get the data

- Create a workspace (with enough storage space).
- Get the data
- Convert the data to a format you can easily manipulate (without changing the data itself).
- Ensure sensitive information is deleted or protected (e.g. anonymised)
- Check the size and type of data (time series, sample, geographical)
- **Sample a test set, put it aside**, and never look at it (no data snooping!)

2. Get the data

kings-county-housing-data

id	date	price	bedrooms	bathrooms	sqft_living	sqft_lot	floors	waterfront	view	condition	grade	sqft_above	sqft_basement	yr_built	yr_renovated	lat	long	sqft_living15	sqft_lot
7129300520	20141013T000000	221900.0	3	1.0	1180	5650	1.0	0	0	3	7	1180	0	1955	0	47.5112	-122.257	1340	56
6414100192	20141209T000000	538000.0	3	2.25	2570	7242	2.0	0	0	3	7	2170	400	1951	1991	47.721	-122.319	1690	76
5631500400	20150225T000000	180000.0	2	1.0	770	10000	1.0	0	0	3	6	770	0	1933	0	47.7379	-122.233	2720	80
2487200875	20141209T000000	604000.0	4	3.0	1960	5000	1.0	0	0	5	7	1050	910	1965	0	47.5208	-122.393	1360	50
1954400510	20150218T000000	510000.0	3	2.0	1680	8080	1.0	0	0	3	8	1680	0	1987	0	47.6168	-122.045	1800	75
7237550310	20140512T000000	1225000.0	4	4.5	5420	101930	1.0	0	0	3	11	3890	1530	2001	0	47.6561	-122.005	4760	1019
1321400060	20140627T000000	257500.0	3	2.25	1715	6819	2.0	0	0	3	7	1715	0	1995	0	47.3097	-122.327	2238	68
2008000270	20150115T000000	291850.0	3	1.5	1060	9711	1.0	0	0	3	7	1060	0	1963	0	47.4095	-122.315	1650	97
2414600126	20150415T000000	229500.0	3	1.0	1780	7470	1.0	0	0	3	7	1050	730	1960	0	47.5123	-122.337	1780	81
3793500160	20150312T000000	323000.0	3	2.5	1890	6560	2.0	0	0	3	7	1890	0	2003	0	47.3684	-122.031	2390	75
1736800520	20150403T000000	662500.0	3	2.5	3560	9796	1.0	0	0	3	8	1860	1700	1965	0	47.6007	-122.145	2210	89
9212900260	20140527T000000	468000.0	2	1.0	1160	6000	1.0	0	0	4	7	860	300	1942	0	47.69	-122.292	1330	60
114101516	20140528T000000	310000.0	3	1.0	1430	19901	1.5	0	0	4	7	1430	0	1927	0	47.7558	-122.229	1780	126
6054650070	20141007T000000	400000.0	3	1.75	1370	9680	1.0	0	0	4	7	1370	0	1977	0	47.6127	-122.045	1370	102

2. train_test_split ahead of EDA

```
1 from sklearn.model_selection import train_test_split
2 train_set, test_set = train_test_split(housing, test_size=0.2, random_state=42)
```

```
1 train_set.shape, test_set.shape
```

```
((17290, 21), (4323, 21))
```

3. Inspect the data to gain insights

- Study each attribute and its characteristics
 - Name
 - Type (categorical, int/float, bounded/unbounded, text, structured etc)
 - % of missing values
 - Noisiness and type of noise (stochastic, outliers, rounding errors etc)
 - Usefulness for the task
 - Type of distribution (Gaussian, uniform, logarithmic etc)
- For supervised learning tasks, identify the target attribute(s)

3. Inspect the data

Description of the features:

Here follows a detailed description of all the features (i.e. columns/variables) in the dataset.

- **id** - unique identifier for a house
- **date** - house was sold
- **price** - price, our prediction target
- **bedrooms** - number of Bedrooms/House
- **bathrooms** - number of bedrooms
- **sqft_living** - square footage of the home
- **sqft_lot** - square footage of the entire lot
- **floors** - total number of floors (levels) in house
- **waterfront** - house which has a view to a waterfront
- **view** - quality of view
- **condition** - how good the condition is (overall)
- **grade** - overall grade given to the housing unit, based on King County grading system
- **sqft_above** - square footage of house apart from basement
- **sqft_basement** - square footage of the basement
- **yr_built** - Built Year
- **yr_renovated** - Year when house was renovated
- **zipcode_group** - 9 groups aggregating some of the 70 zipcodes having similar characteristics
- **lat** - Latitude coordinate
- **long** - Longitude coordinate
- **sqft_living15** - The square footage of interior housing living space for the nearest 15 neighbours
- **sqft_lot15** - The square footage of the land lots of the nearest 15 neighbours

```
1 housing.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 21613 entries, 0 to 21612
Data columns (total 21 columns):
#   Column              Non-Null Count  Dtype
---  -
0   id                   21613 non-null  int64
1   date                 21613 non-null  object
2   price                21613 non-null  float64
3   bedrooms             21613 non-null  int64
4   bathrooms            21613 non-null  float64
5   sqft_living          21613 non-null  int64
6   sqft_lot             21613 non-null  int64
7   floors               21613 non-null  float64
8   waterfront           21613 non-null  int64
9   view                 21613 non-null  int64
10  condition            21613 non-null  int64
11  grade                21613 non-null  int64
12  sqft_above           21613 non-null  int64
13  sqft_basement        21613 non-null  int64
14  yr_built              21613 non-null  int64
15  yr_renovated         21613 non-null  int64
16  lat                   21613 non-null  float64
17  long                  21613 non-null  float64
18  sqft_living15        21613 non-null  int64
19  sqft_lot15           21613 non-null  int64
20  zipcode_group        21613 non-null  object
dtypes: float64(5), int64(14), object(2)
memory usage: 3.5+ MB
```

3. Exploratory Data Analysis (EDA)

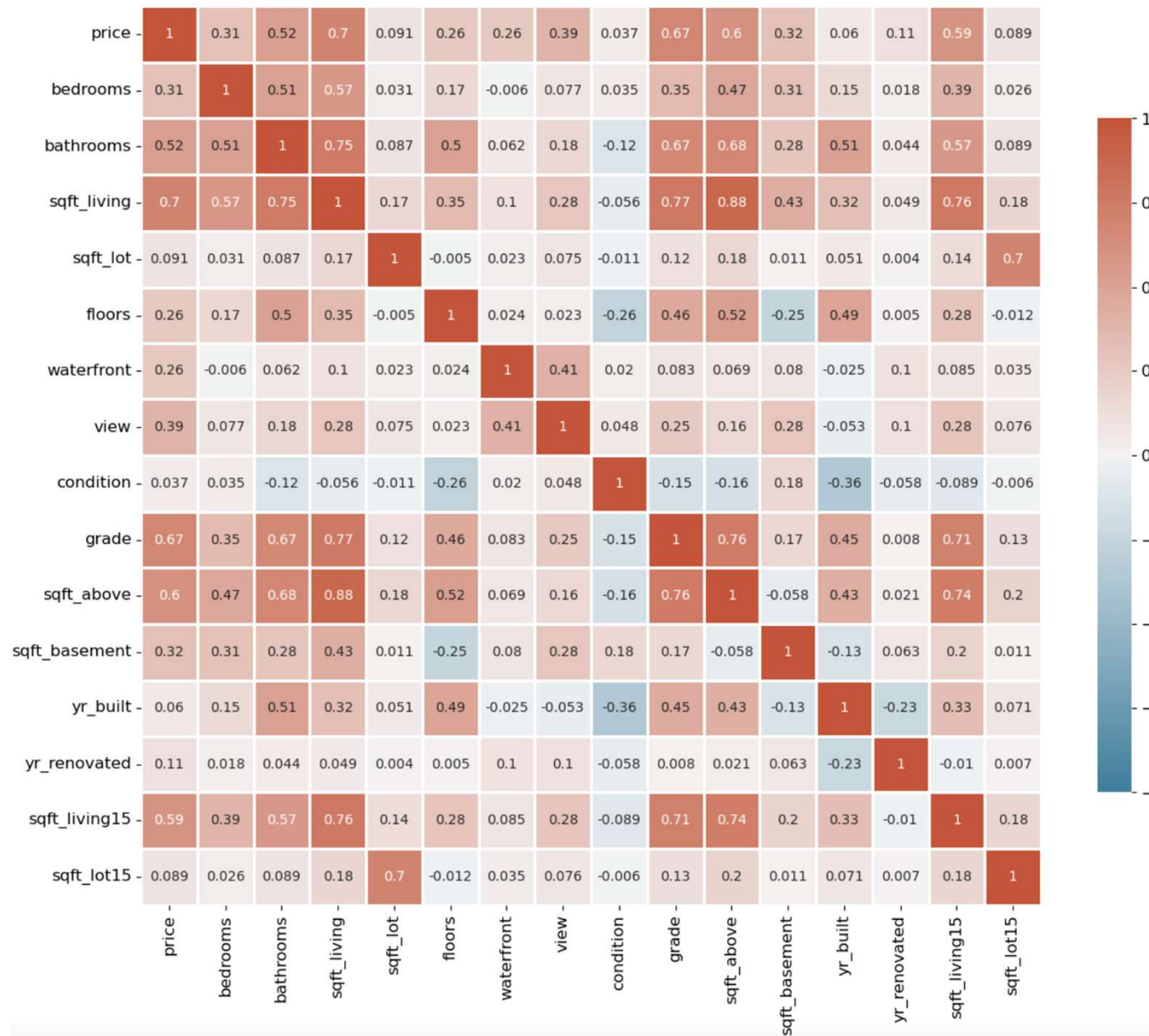
- Visualise the data
- Study the correlations between attributes
- Study how you would solve the problem manually
- Identify the promising transformations you may want to apply
- Identify extra data that would be useful to the investigation

Seaborn

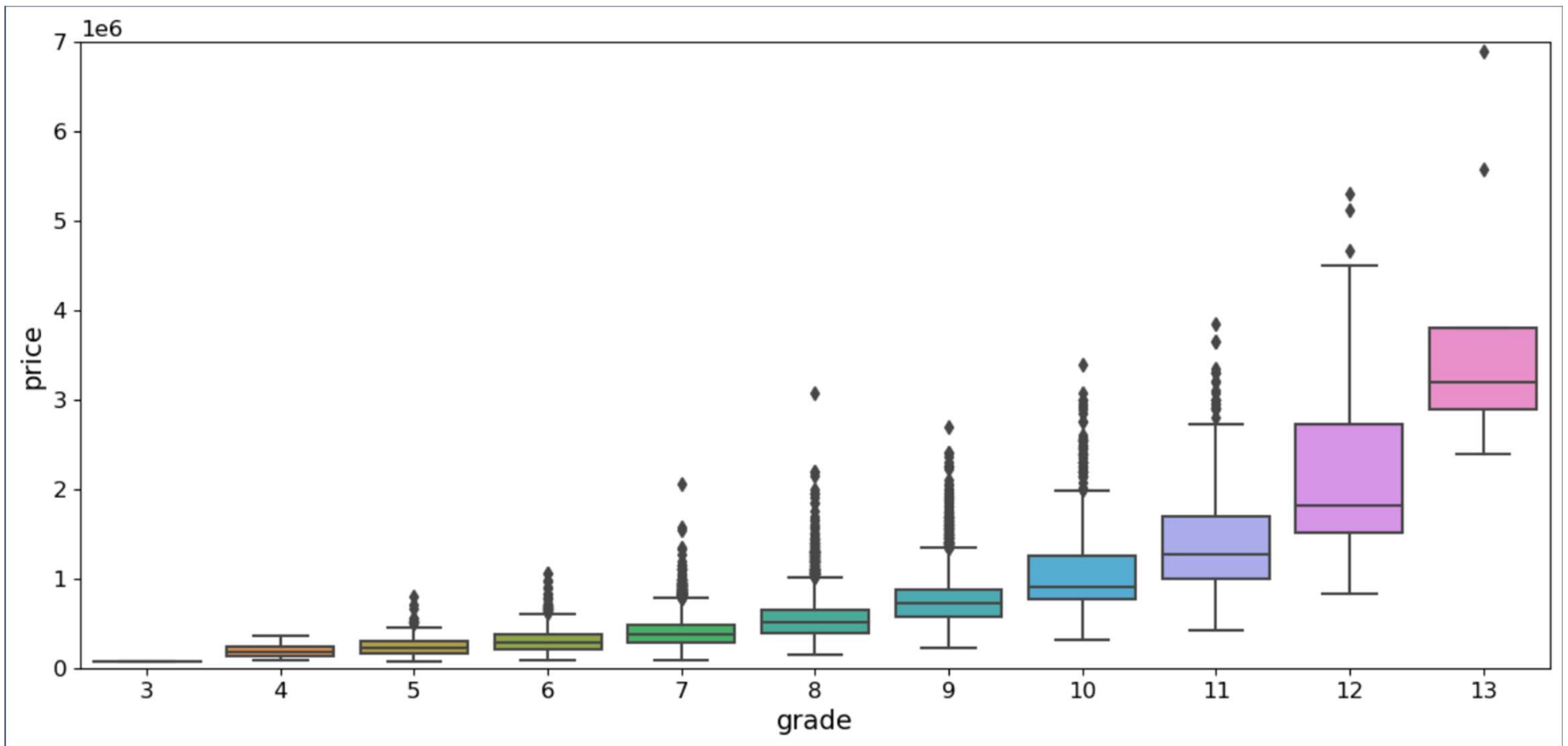
- High level data visualisation library based on matplotlib.
- Ideal for drawing informative statistical graphics
- Dataset-oriented API for examining relationships between multiple variables
- Convenient views onto the overall structure of complex datasets
- High-level abstractions for structuring multi-plot grids that let you easily build complex visualisations
- Concise control over matplotlib figure styling with several built-in themes

sns.heatmap

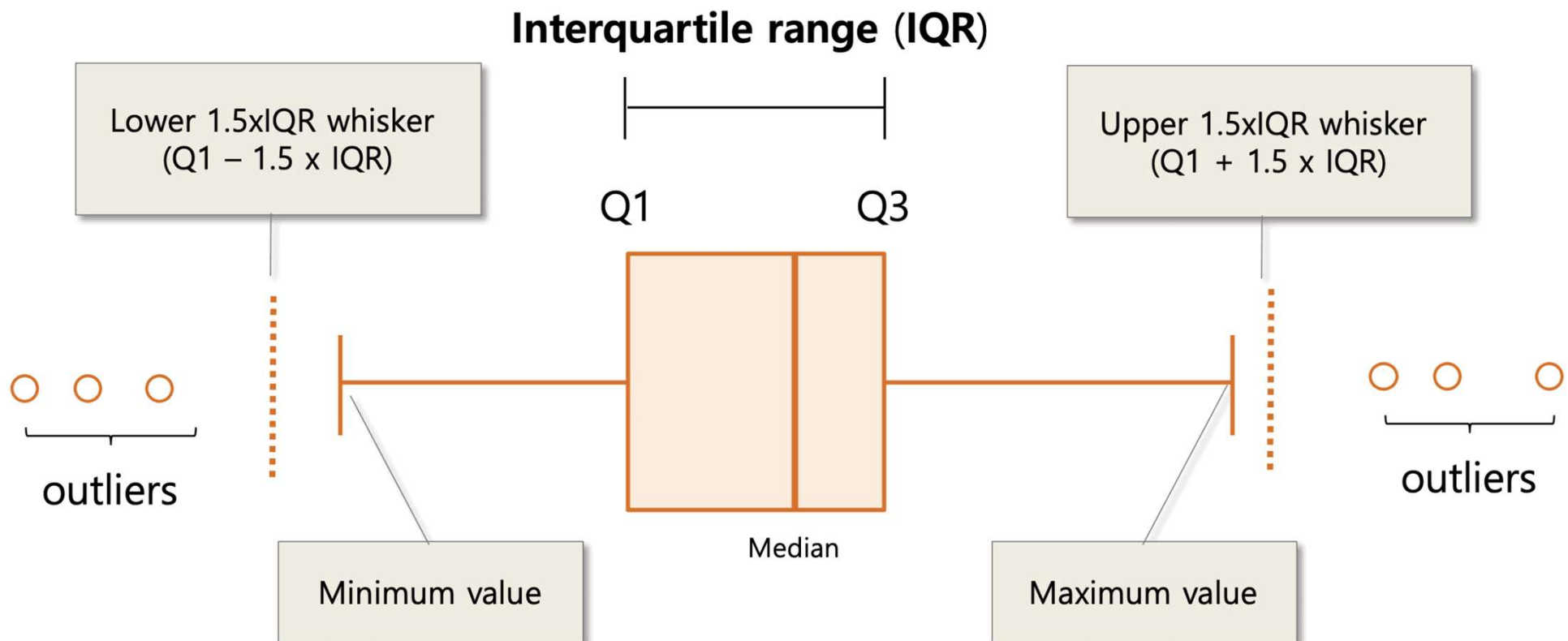
- Pearson's correlation coefficient drawn as a 'heat' map
- Colour coded for accessibility
- +1 = perfect positive correlation – **hotter!**
- -1 = perfect negative correlation – **cooler!**



sns.boxplot

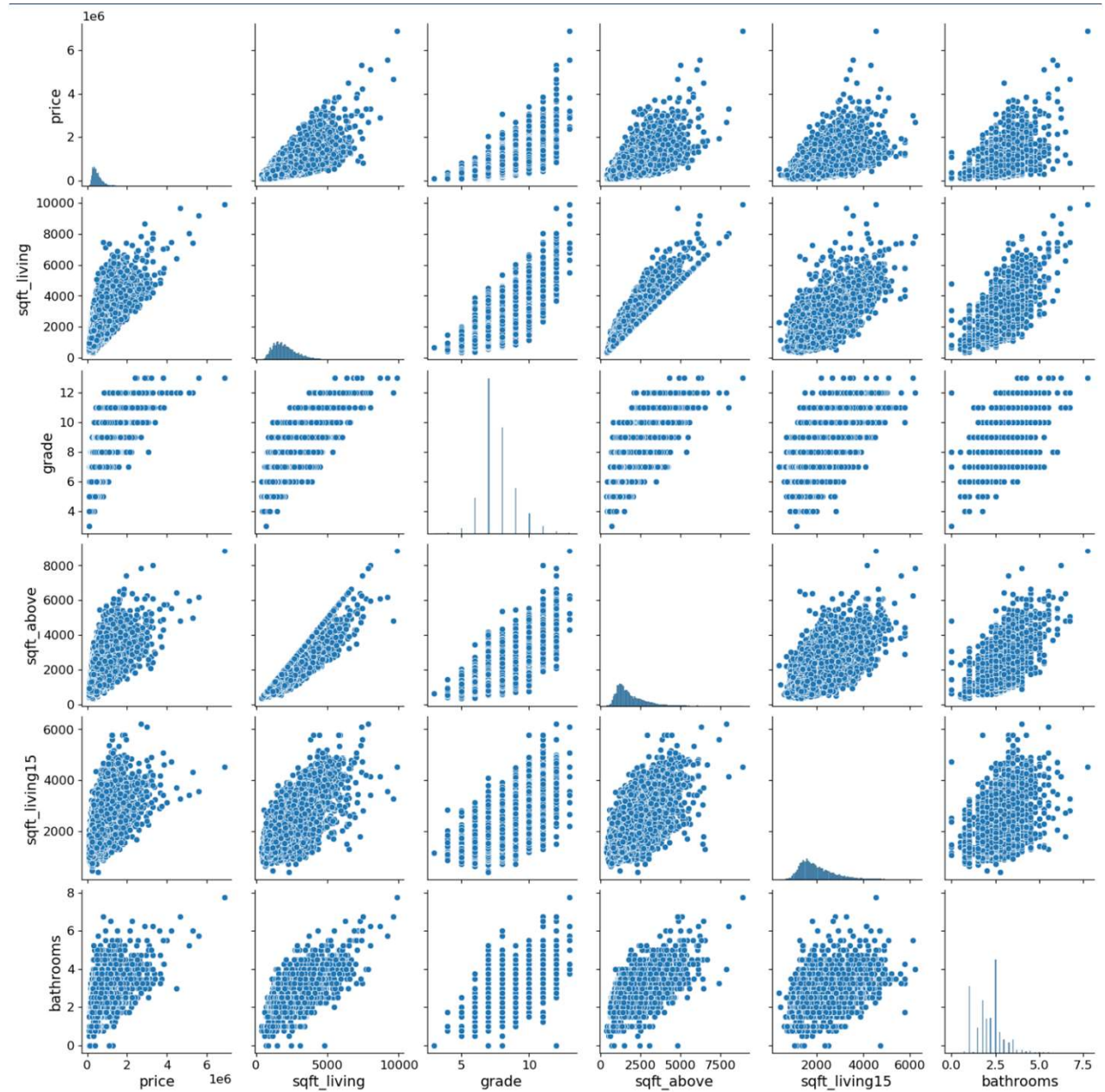


Boxplots and IQRs

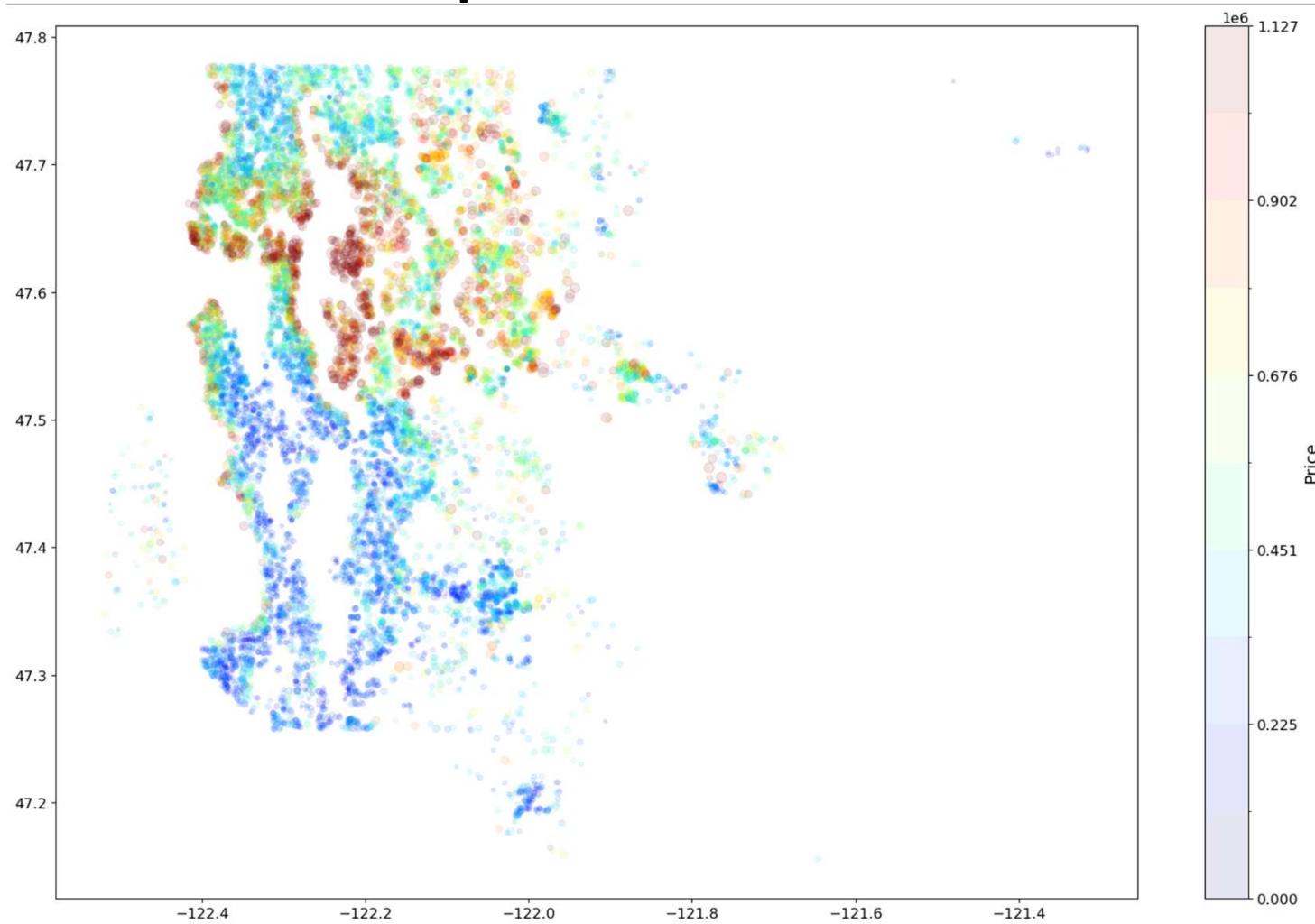


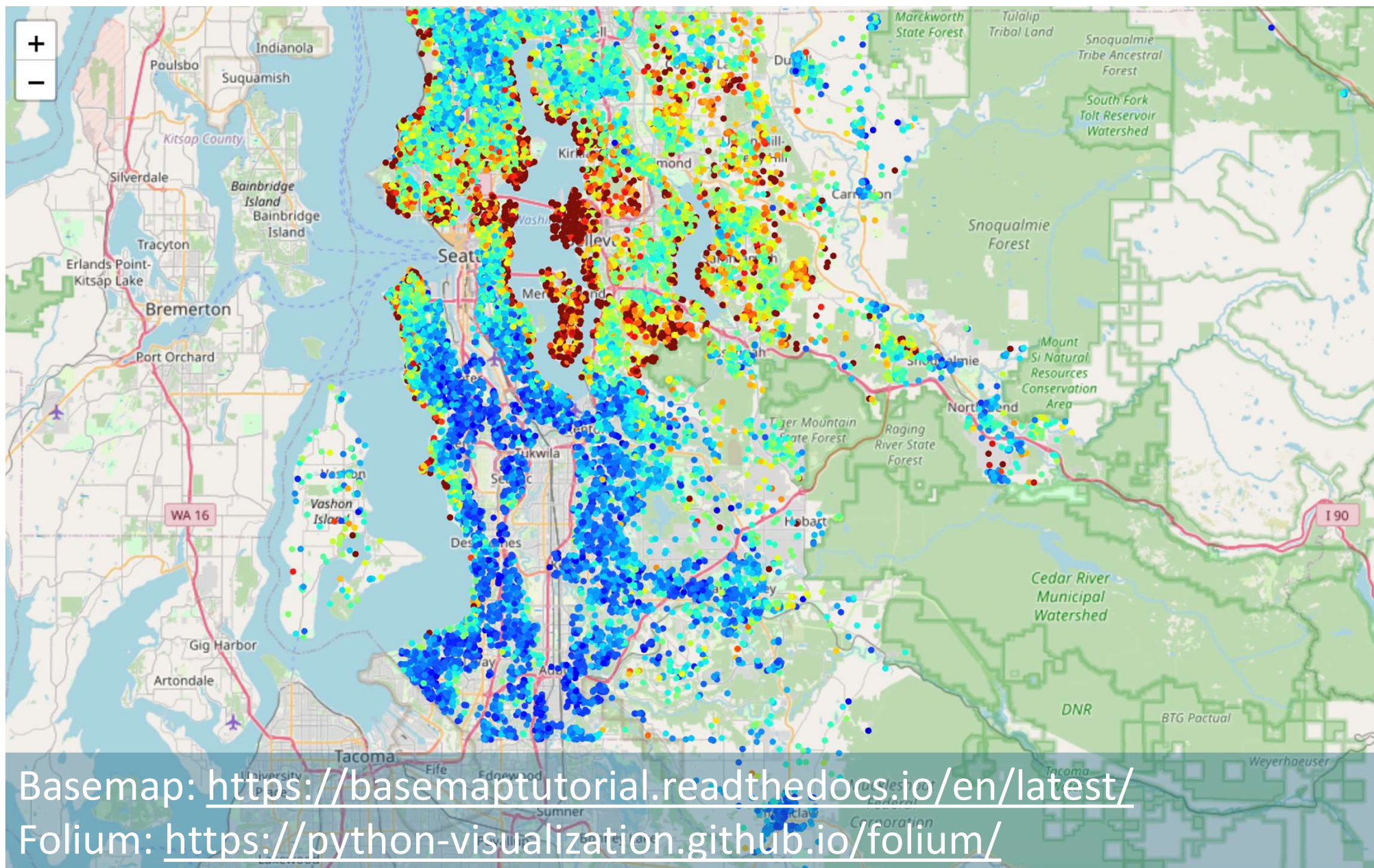
sns.pairplot

- We can get identify the distribution of data
- Linear relationships between variables



Example: colormap





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