

Appendix: Emma Luk (Mobile: 07974 522 805): A Selection of my data science work

1A: Deep Learning with TensorFlow Long Short-Term Memory (LSTM) Neural Network for Stock Market Predictions with Python



Figure 1: Predicted stock prices and Actual Stock Prices

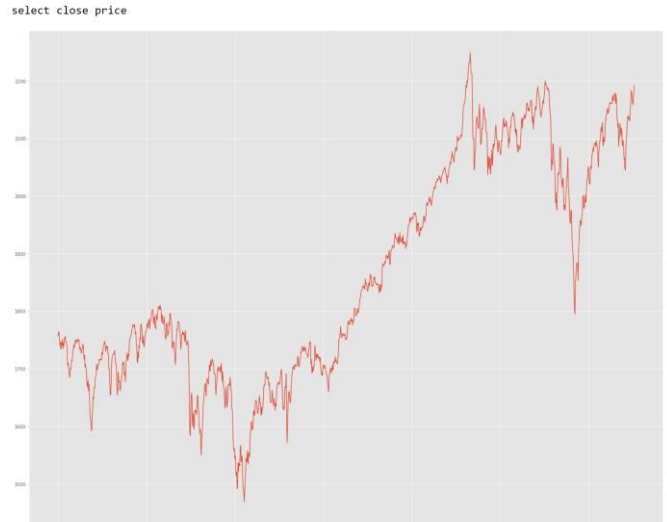
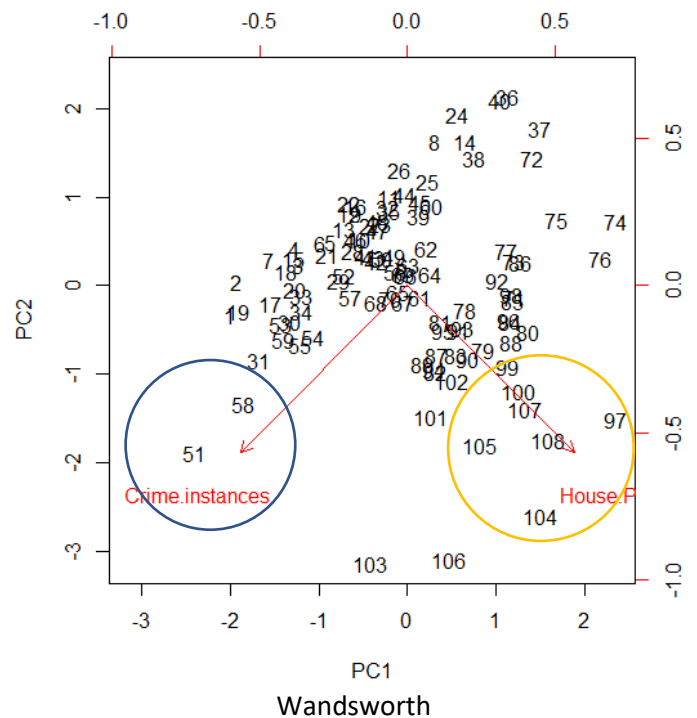
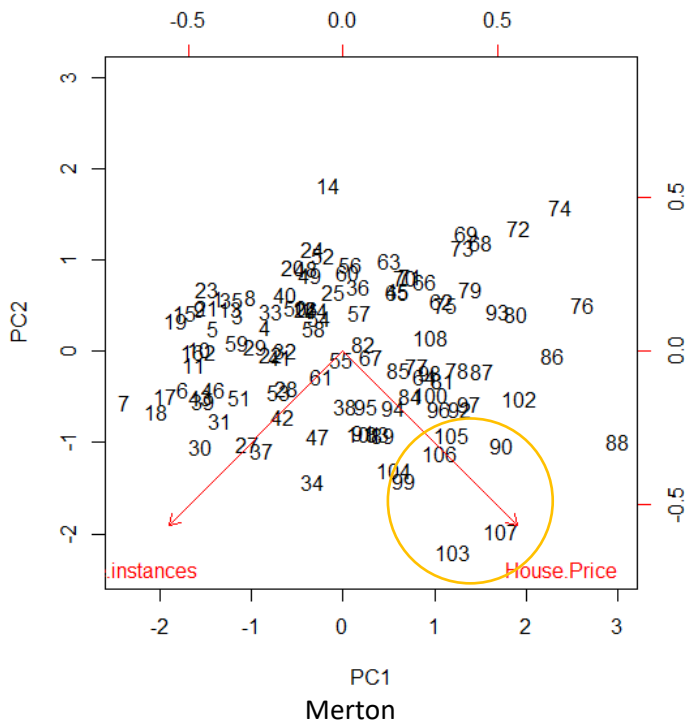


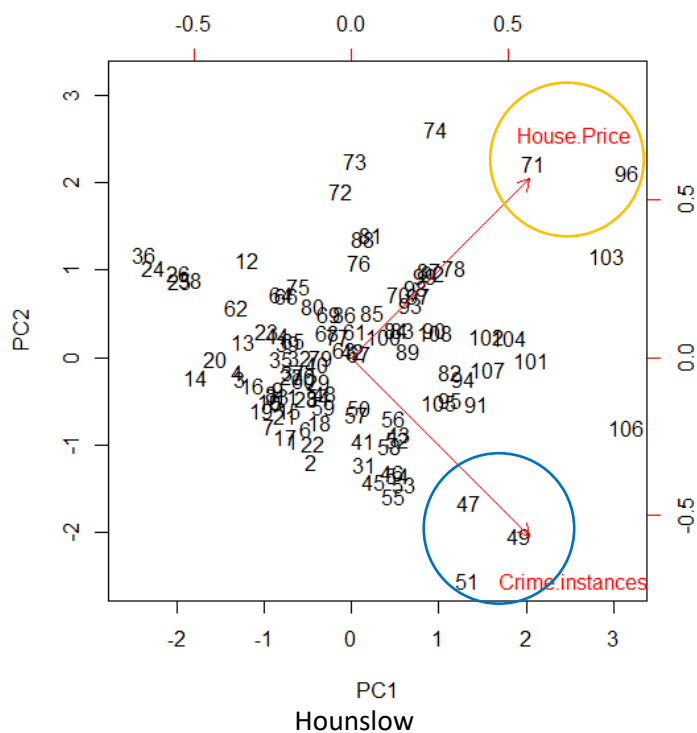
Figure 2: Plot of 'Close' for Global Equity Income sector price history

1B: Analysing House Prices and Crime in R

Main analysis Unsupervised learning: Principal Component Analysis (PCA):

- The aim was to investigate how current house prices were affected by recent crime levels in London Boroughs
- Performed data cleaning, transformation, manipulation and conducted Principal Component Analysis (PCA): this was the process by which compute principal components and used them for **better understanding of the data**. PCA is considered an **unsupervised machine learning** method because it involves only a set of feature variables and no associated response variable. PCA also serves as a useful tool for **exploratory analysis** and **data visualisation**





In Wandsworth:

House 51: £ 538,999.3

House 104: £ 973,938.4

In Hounslow:

House 71: £689,162.6

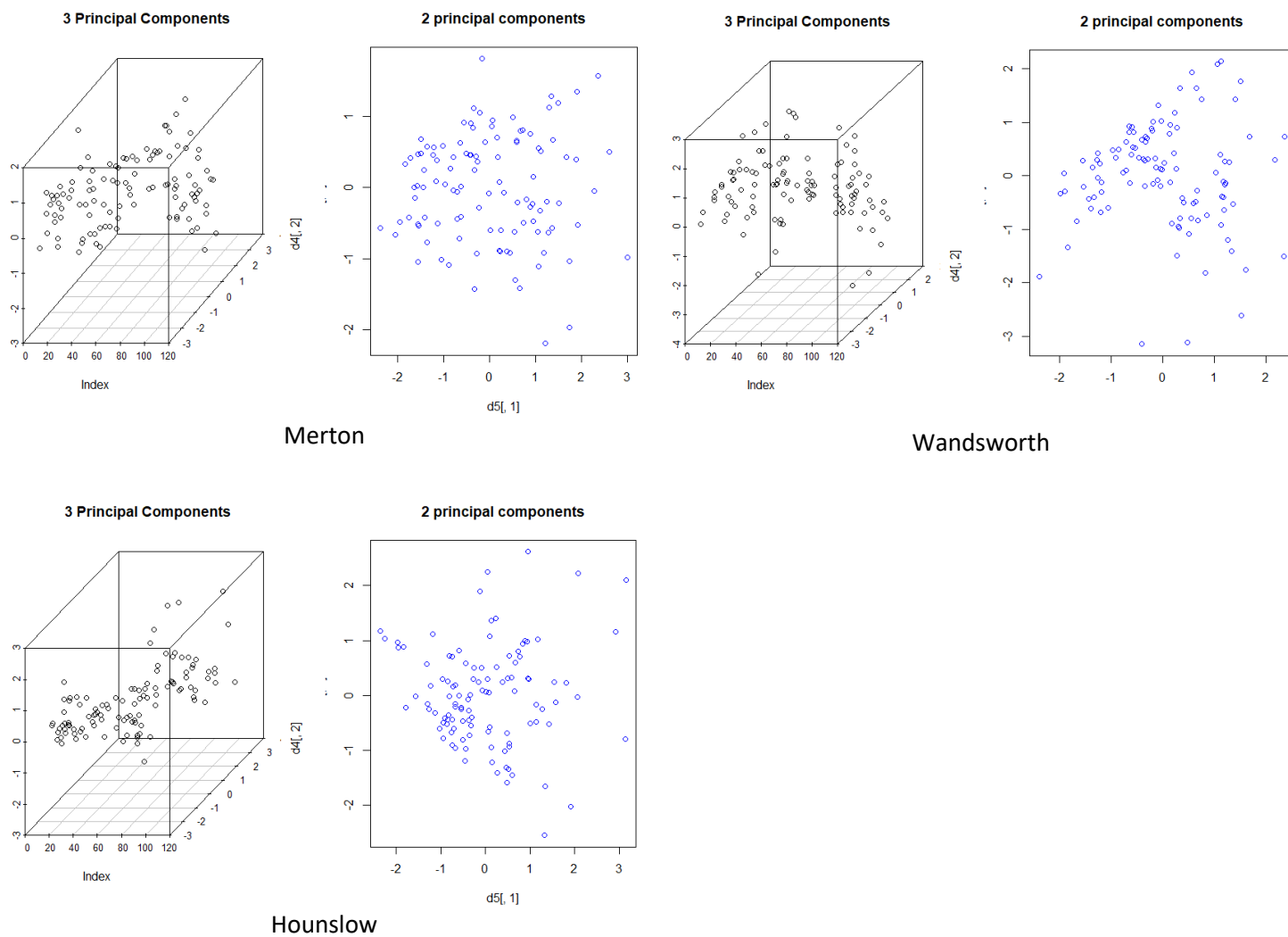
House 49: £396,623.0

For PCA:

House 104 & 71 data points are near House Price point. They are more expensive.

House 51 and 49 data points are near Crime point. They are less expensive. This is an indication that crime has some effect on house prices.

Figure 3: Plot of the first two principal components (PC1 & PC2) for Hounslow, Merton & Wandsworth



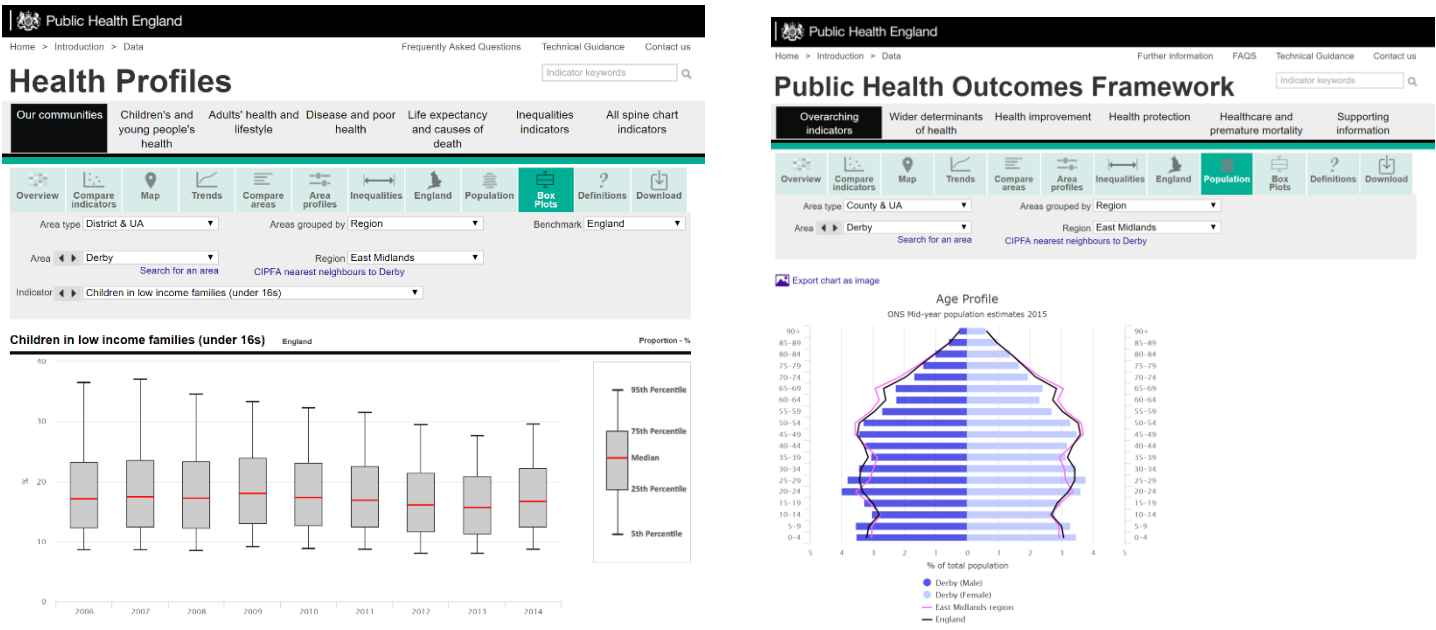
Left: Figure 4: 3D scatterplot using three principal components

Right: Figure 5: 2D scatterplot using two principal components

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<h3>Mercedes-Benz A-Class Hatchback</h3> <h4>A140 Avantgarde Sdr 2001 (CAP: 21228)</h4> <p>Introduction Date: 08/06/01 Discontinued Date: Status: Current Comment</p> <table border="1"> <thead> <tr> <th>Profile</th> <th>Options</th> <th>Mileage</th> <th>Factors</th> <th>Comparisons</th> <th>Returns</th> <th>Payments</th> <th>Derivatives</th> </tr> </thead> </table> <p>Net Cost: £13,490 List Price: £14,200</p> <table border="1"> <thead> <tr> <th rowspan="2">Year</th> <th colspan="2">Current @10k p.a.</th> <th colspan="2">Proposed @10k p.a.</th> <th rowspan="2">@10000 p.a.</th> <th rowspan="2">CAP Clean @10k p.a.</th> <th rowspan="2">CAP Monitor @10k p.a.</th> <th rowspan="2">GG Trade @10k p.a.</th> </tr> <tr> <th>Value</th> <th>% of Net Cost</th> <th>Value</th> <th>% of Net Cost</th> </tr> </thead> <tbody> <tr><td>1</td><td>8650</td><td>64%</td><td>8650</td><td>64%</td><td>8650</td><td>9300</td><td>10025</td><td>n/a</td></tr> <tr><td>2</td><td>7275</td><td>54%</td><td>7275</td><td>54%</td><td>7277</td><td>7900</td><td>8300</td><td>n/a</td></tr> <tr><td>3</td><td>6200</td><td>46%</td><td>6200</td><td>46%</td><td>6206</td><td>6775</td><td>6775</td><td>n/a</td></tr> <tr><td>4</td><td>5275</td><td>39%</td><td>5275</td><td>39%</td><td>5272</td><td>n/a</td><td>5650</td><td>n/a</td></tr> <tr><td>5</td><td>4310</td><td>32%</td><td>4310</td><td>32%</td><td>4311</td><td>n/a</td><td>4775</td><td>n/a</td></tr> <tr><td>6</td><td>3540</td><td>26%</td><td>3540</td><td>26%</td><td>3542</td><td>n/a</td><td>n/a</td><td>n/a</td></tr> <tr><td>7</td><td>2860</td><td>21%</td><td>2860</td><td>21%</td><td>2856</td><td>n/a</td><td>n/a</td><td>n/a</td></tr> </tbody> </table> <div> <div>Variance from Core Model</div> <div>Core Model First Year Value: £7,875</div> <div>Link to Core Model by: <input type="radio"/> Percentage <input checked="" type="radio"/> Amount</div> <div>This Model First Year Value: £ <input type="text" value="8650"/> <input type="text" value="9.84"/> % £ <input type="text" value="775"/></div> <div>Variance Comment: <input type="text"/></div> </div> <div> <div>Mileage Adjustment Factor (per 1000 miles)</div> <div>Current: 0.6 Proposed: <input type="text" value="0.6"/></div> </div> <div> <div>Factors Applied to this Derivative</div> <div>None</div> </div>									Profile	Options	Mileage	Factors	Comparisons	Returns	Payments	Derivatives	Year	Current @10k p.a.		Proposed @10k p.a.		@10000 p.a.	CAP Clean @10k p.a.	CAP Monitor @10k p.a.	GG Trade @10k p.a.	Value	% of Net Cost	Value	% of Net Cost	1	8650	64%	8650	64%	8650	9300	10025	n/a	2	7275	54%	7275	54%	7277	7900	8300	n/a	3	6200	46%	6200	46%	6206	6775	6775	n/a	4	5275	39%	5275	39%	5272	n/a	5650	n/a	5	4310	32%	4310	32%	4311	n/a	4775	n/a	6	3540	26%	3540	26%	3542	n/a	n/a	n/a	7	2860	21%	2860	21%	2856	n/a	n/a	n/a
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1E. Examples of data visualisation from when I worked on the ‘Fingertips’ website for Public Health England (Government Agency): On the left are boxplots depicting the percentage of children in low income families in the East Midlands between the years 2006 and 2014; on the right is a bar graph with negative stack depicting the proportion of males and females of different age groups in the East Midlands region.



1F Industry Project: How Robots are making Farming Profitable
Weather Data Analytics Using Hadoop

- Leading the big data flow of the application starting from data ingestion from upstream to HDFS, processing and analysing the data in HDFS and data visualisation in R & JavaScript

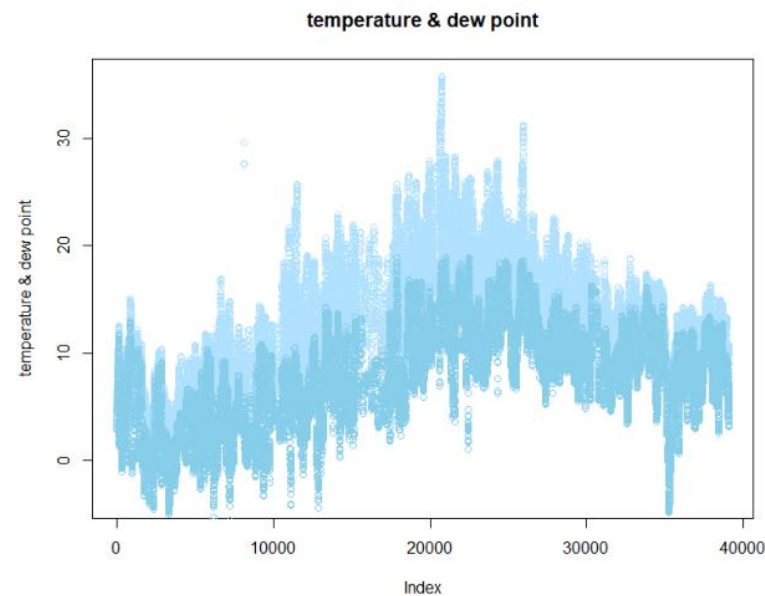
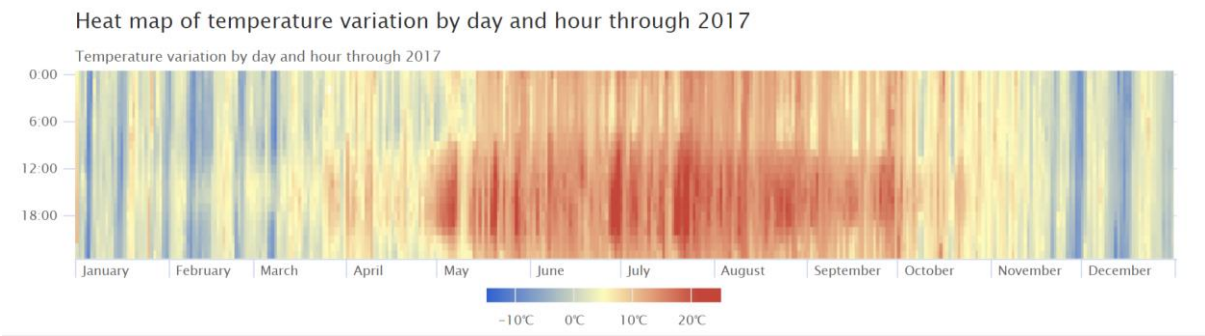
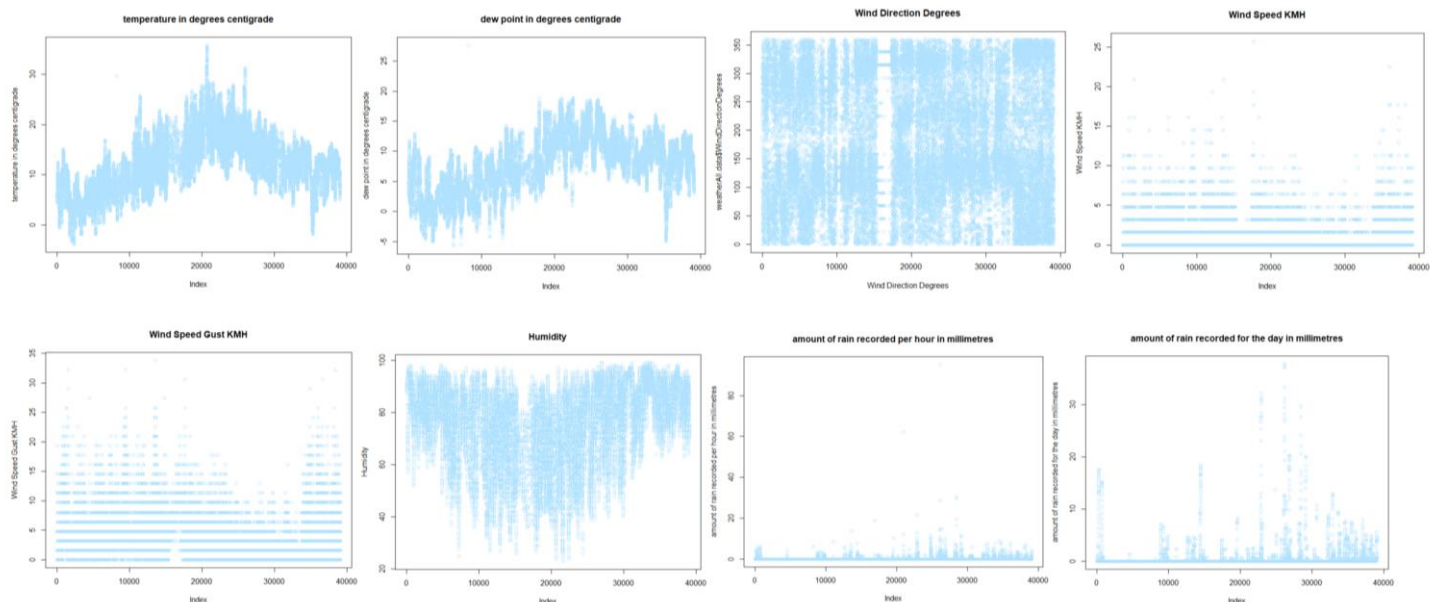


Figure 6: Scatter plots with two level
Added two level first level is dew point and second level is temperature in R



1G: Data Visualisation in Microsoft Power BI & Tableau

- The aim was to create Executive Dashboard, tracked and reported on business metrics & the KPIs
- This dashboard included key performance, top ten products, top performing cities and top performing city, customer reviews and sales by Month.

Microsoft Power BI:

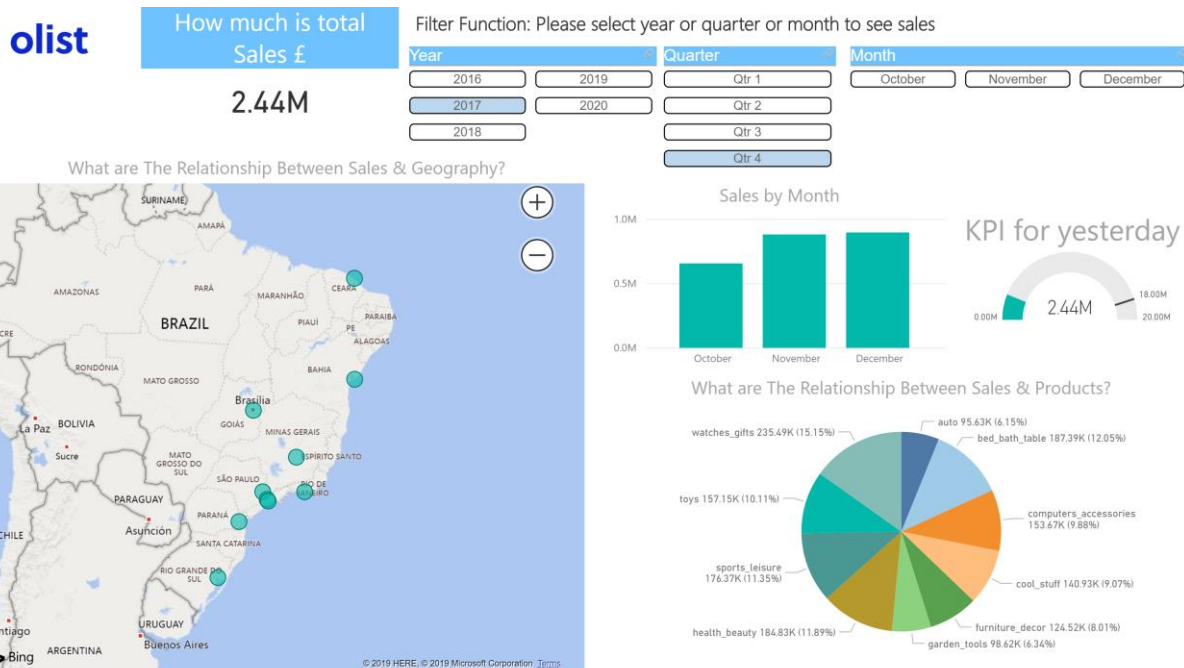


Tableau:

