

EBU5601 Data Design

Introduction to Data Analysis

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Learning Outcomes

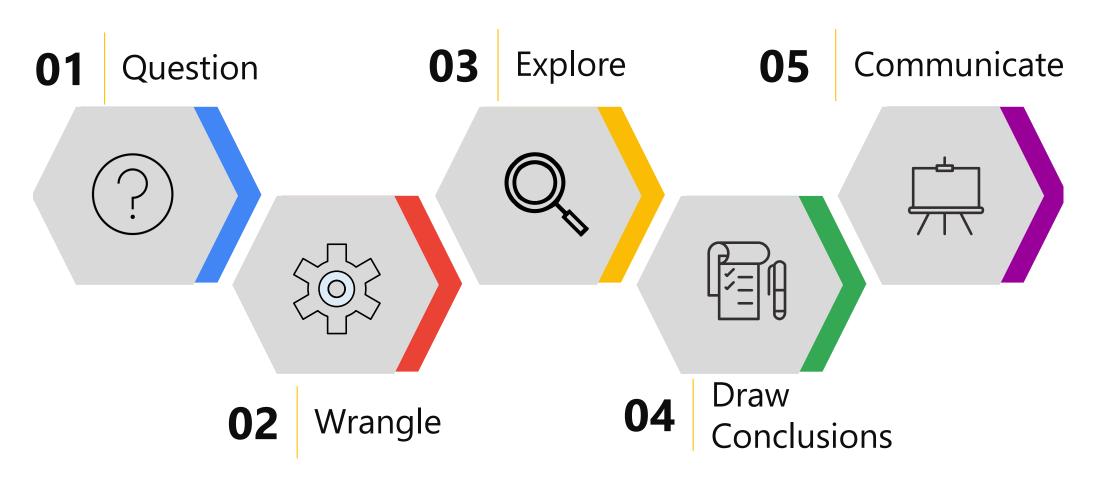
- The main outcomes are:
 - [LO1.1] Understand the main steps of the data analysis process
 - [LO1.2] Understand the format of the comma separated values (csv) file and be able to read csv files using pandas
 - [LO1.3] Be familiar with the basic techniques for assessing and cleaning data using pandas



DATA ANALYSIS OVERVIEW



The data analysis process can be organised into five steps:





Step 1: Ask Questions

- Ask questions about given data.
- Ask questions first and gather data based on that late

Step 2: Wrangle Data

- Gather the data you need to answer your questions.
 - Downloading files that are readily available
 - Getting data from an API or web scraping
 - Pulling data from existing databases
 - Combine data from multiple different formats.
- Assess the data to identify any problems in the data's quality or structure.
- Clean the data by modifying, replacing, or removing data to ensure that your dataset is of the highest quality and as well-structured as possible.

- Comma Separated Values (csv) File
 - A text file with a tabular structure that holds only raw data. It is easy to process manually using code such as python.

datetime	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	casual	registered	count
01/01/2011 00:00	1	0	0	1	9.84	14.395	81	0	3	13	16
01/01/2011 01:00	1	0	0	1	9.02	13.635	80	0	8	32	40
01/01/2011 02:00	1	0	0	1	9.02	13.635	80	0	5	27	32
01/01/2011 03:00	1	0	0	1	9.84	14.395	75	0	3	10	13
01/01/2011 04:00	1	0	0	1	9.84	14.395	75	0	0	1	1
01/01/2011 05:00	1	0	0	2	9.84	12.88	75	6.0032	0	1	1
01/01/2011 06:00	1	0	0	1	9.02	13.635	80	0	2	0	2
01/01/2011 07:00	1	0	0	1	8.2	12.88	86	0	1	2	3
01/01/2011 08:00	1	0	0	1	9.84	14.395	75	0	1	7	8
01/01/2011 09:00	1	0	0	1	13.12	17.425	76	0	8	6	14

Header

```
datetime, season, holiday, workingday, weather, temp, atemp, humidity, windspeed, casual, registered, count
01/01/2011 00:00,1,0,0,1,9.84,14.395,81,0,3,13,16
01/01/2011 01:00,1,0,0,1,9.02,13.635,80,0,8,32,40
4 01/01/2011 02:00,1,0,0,1,9.02,13.635,80,0,5,27,32
5 01/01/2011 03:00,1,0,0,1,9.84,14.395,75,0,3,10,13
6 01/01/2011 04:00,1,0,0,1,9.84,14.395,75,0,0,1,1
7 01/01/2011 05:00,1,0,0,2,9.84,12.88,75,6.0032,0,1,1
8 01/01/2011 06:00,1,0,0,1,9.02,13.635,80,0,2,0,2
9 01/01/2011 07:00,1,0,0,1,8.2,12.88,86,0,1,2,3
10 01/01/2011 08:00,1,0,0,1,9.84,14.395,75,0,1,7,8
11 01/01/2011 09:00,1,0,0,1,13.12,17.425,76,0,8,6,14
```



Step 3: Perform EDA (Exploratory Data Analysis)

- Explore and then augment the data to maximize the potential of your analyses, visualizations, and models.
- Exploring involves finding patterns in your data, visualizing relationships in your data, and building intuition about what you're working with.
- After exploring, you can do things like remove outliers and create better features from your data, also known as feature engineering.

Step 4: Draw conclusions (or make predictions)

- This step is typically approached with descriptive/inferential statistics or machine learning.
- In this module we mainly focus on drawing conclusions with descriptive/inferential statistics.

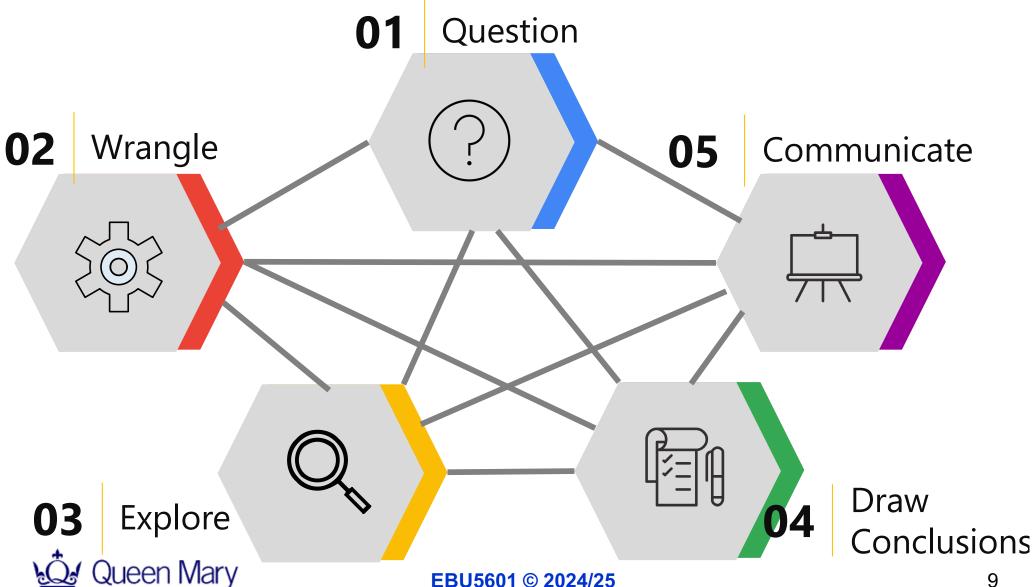


Step 5: Communicate Data Findings

- Justify and convey meaning in the insights you've found.
- If your end goal is to build a system, you usually need to share what you've built, explain how you reached design decisions, and report how well it performs.
- There are many ways to communicate your results: reports, slide decks, blog posts, emails, presentations, or even conversations.
 Data visualization will always be very valuable.



The data analysis process isn't always linear.



DATA ANALYSIS CASE STUDY



Demo: Bike Sharing Demand

- We'll use data from <u>Kaggle's Bike Sharing Demand</u> competition. (*The original dataset has been slightly modified for the purpose of this demonstration*).
- In this dataset, you are given hourly rental data spanning two years from the Capital Bikeshare program in Washington, D.C.
- In the original Kaggle competition, the participants were asked to use historical data to forecast bike rental demand.





Demo: Bike Sharing Demand

- There are 12 features in the dataset. Each feature is a column in the dataset.
- To get a decent prediction result, we must firstly get familiar with the data and conduct necessary pre-preprocess or feature engineering before we feed the data to the machine learning models.

	datetime	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	casual	registered	count
0	01/01/2011 00:00	1	0	0	1	9.84	14.395	81	0.0000	3	13	16
1	01/01/2011 01:00	1	0	0	1	9.02	13.635	80	0.0000	8	32	40
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3	01/01/2011 03:00	1	0	0	1	9.84	14.395	75	0.0000	3	10	13
4	01/01/2011 04:00	1	0	0	1	9.84	14.395	75	0.0000	0	1	1
10881	19/12/2012 19:00	4	0	1	1	15.58	19.695	50	26.0027	7	329	336
10882	19/12/2012 20:00	4	0	1	1	14.76	17.425	57	15.0013	10	231	241
10883	19/12/2012 21:00	4	0	1	1	13.94	15.910	61	15.0013	4	164	168
10884	19/12/2012 22:00	4	0	1	1	13.94	17.425	61	6.0032	12	117	129
10885	19/12/2012 23:00	4	0	1	1	13.12	16.665	66	8.9981	4	84	88
10886 ro	10886 rows × 12 columns											



Demo: Bike Sharing Demand

Step 1: Ask Questions

Q1: Given the above data on features that potentially influence the number of bikes rented each hour, what questions would be relevant to ask?

www.menti.com

Enter the code



Or use QR cod

	datetime	season	holiday	workingday	weather	temp	atemp	humidity	windspeed	casual	registered	count
0	01/01/2011 00:00	1	0	0	1	9.84	14.395	81	0.0000	3	13	16
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10885	19/12/2012 23:00	4	0	1	1	13.12	16.665	66	8.9981	4	84	88
10886 rows × 12 columns												



Demo: Bike Sharing Demand

Step 2: Data Wrangling

Q2: What potential problems do you see with this dataset that would need to be fixed before continuing with analysis?

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Enter the code



Data	columns (to	tal 12 columns):				1247 - 260 - 15	200 0	900			V 832	Avs (6) 507	927		
#	Column	Non-Null Count	Dtype		season	holiday	workingday	weather	temp	atemp	humidity	windspeed	casual	register	
				count	10886.000000	10886.000000	10886.000000	10886.000000	10848.000000	10848.000000	10886.000000	10886.000000	10886.000000	10886.0000	
0	datetime	10886 non-null	object												
1	season	10886 non-null	int64	mean	2.506614	0.028569	0.680875	1.418427	20.244801	23.674888	61.886460	12.799395	36.021955	155.5521	
2	holiday	10886 non-null	int64	-4-1	4.440474	0.400000	0.400450	0.000000	7,004050	0.004405	40.045000	0.404507	40.000.477	454,0000	
3	workingday	10886 non-null		int64	std	1.116174	0.166599	0.466159	0.633839	7.991858	8.891435	19.245033	8.164537	49.960477	151.0390
4	weather	10886 non-null		min	1.000000	0.000000	0.000000	1.000000	0.820000	0.760000	0.000000	0.000000	0.000000	0.0000	
5	temp	10848 non-null							********				*******	*****	
6	atemp	10848 non-null	float64	25%	2.000000	0.000000	0.000000	1.000000	13.940000	16.665000	47.000000	7.001500	4.000000	36.0000	
7	humidity	10886 non-null	int64	E00/			4000000	4 000000	00 500000	0404000		40.00000	47.00000	440.0000	
8	windspeed	10886 non-null	float64	50%	3.000000	0.000000	1.000000	1.000000	20.500000	24.240000	62.000000	12.998000	17.000000	118.0000	
9	casual	10886 non-null	int64	75%	4.000000	0.000000	1.000000	2.000000	26.240000	31.060000	77.000000	16.997900	49.000000	222.0000	
10	registered	10886 non-null	int64	, , , ,		0.000000		2.000000	2012 10000	011000000	77100000	10.007.000	10.00000		
11	count	10886 non-null	int64	max	4.000000	1.000000	1.000000	4.000000	200.000000	300.000000	100.000000	56.996900	367.000000	886.0000	



Demo: Bike Sharing Demand

Step 2: Data Wrangling

Deal with those missing values and apparently wrong values.

- 1. Remove the records/rows with missing values if not too many records/rows contain missing values or wrong values.
- 2. Fill in all missing values with some values we think might be close to the actual values, e.g., the average temperature of the days/hours around the day/hour that do not have temperature readings. We can also do the same thing to the records with unrealistic temperature readings.



Demo: Bike Sharing Demand

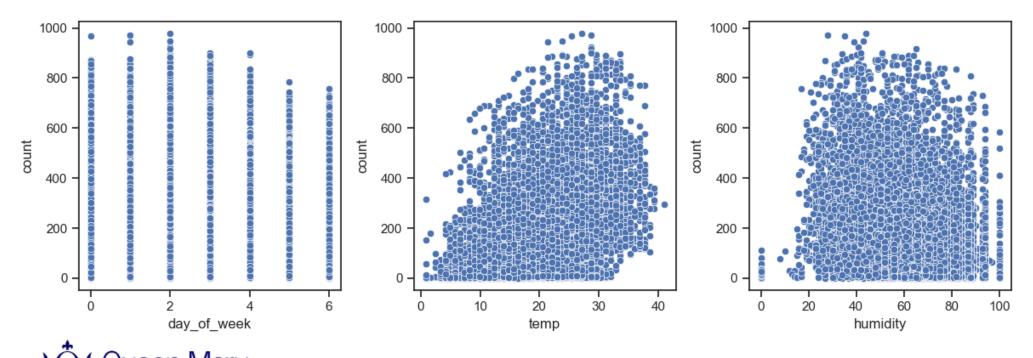
Step 3: Exploratory Data Analysis

Q3: Based on these scatterplots, which of these three features seems most helpful in predicting rental count?

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Enter the code

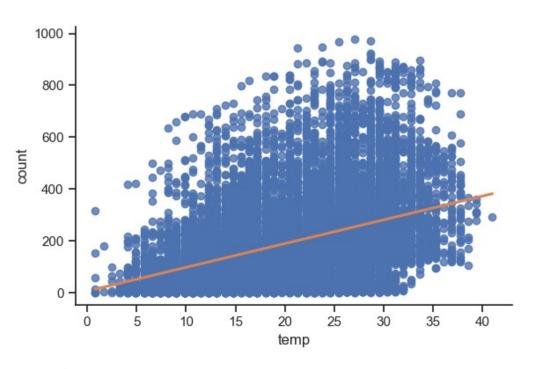




Demo: Bike Sharing Demand

Step 4: Draw Conclusions

Q4: Based on this graph of regressing bike rental count on the temperature, how many additional bikes do you think would be checked out if the temperature rose from 2°C to 30 °C?



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Enter the code

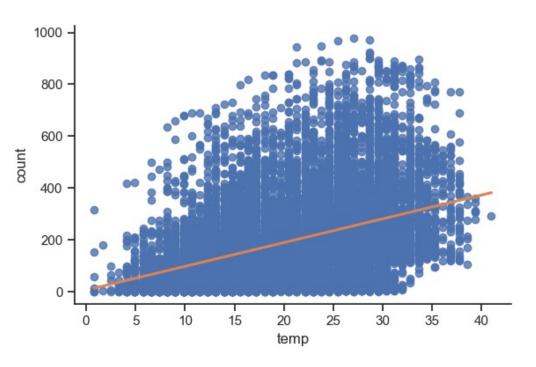




Demo: Bike Sharing Demand

Step 5: Communicate Data Findings

Q5: What would be valid methods of communicating your conclusions from the Bike Sharing data?





Enter the code





RECAP

Introduction to Data Analysis



Recap

Five Steps for Data Analysis

- Ask Questions
- Wrangle Data
- Perform Exploratory Data Analysis (EDA)
- Draw Conclusions
- Communicate Data Findings

Comma Separated Values (csv) File

Format

Techniques

- Read csv file
- Deal with missing values and anomalies



Questions

Use student forum on QM+ chao.shu@qmul.ac.uk xiaolanliu@qmul.ac.uk

