

# To Get Started

- Paper sheet and online at:  
<http://www.eecs.qmul.ac.uk/~william/CAS-London-2020.html>
- Download sample notebooks and data
  - Create directory (N:\session3\fullname) and unzip notebooks and data
- Login to Google Colab: <https://colab.research.google.com/>
  - Create a new notebook
  - Use the file 'upload' menu to upload the 'example' notebook

# Introduction to Data Analysis

William Marsh



# How This Session Works

- Introduce concepts
- Practical work
  - Collaboration: *practice teaching!*
- Repeat
- Conclude

**Probably Not  
Enough Time**

# Outline

- Aims
- Introducing the Python notebook using Google Colab
- Part 1: the dataframe
- Part 2: transforming data and the pivot table
- Part 3: adding columns
- Conclusion and discussion

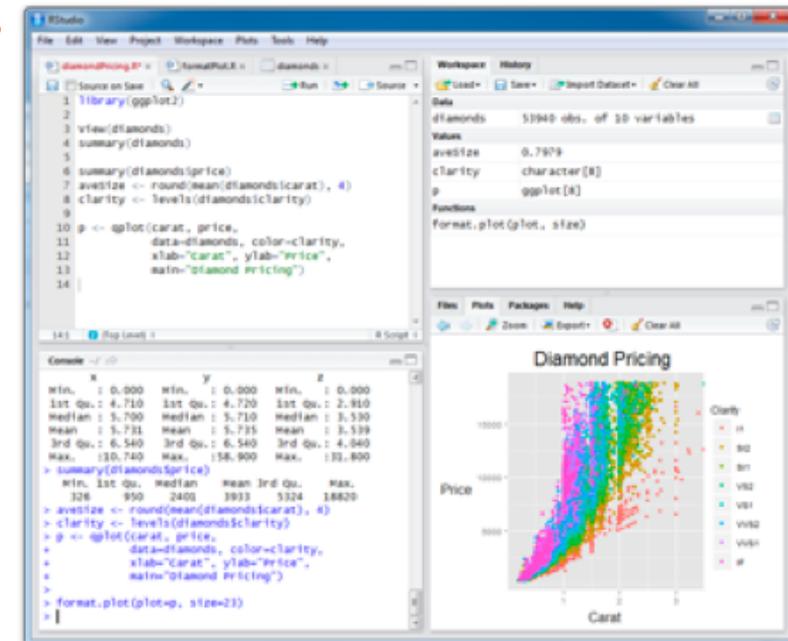
# Session Aims



# Aims

- Introduce Python Pandas
  - Popular library for data
  - Interactive notebook – Google's Colab
- Introduce key concepts
  - Dataframe
  - Filter and select
  - Pivot table
  - Visualisation

- Pandas is very complex
- Concepts common in other environments
  - Excel
  - RStudio
- Help develop pedagogy



# The Data Life Cycle

- Using data to answer a question
  - What is the problem?
  - .. do we have the data?
- Data analysis not just technical
- We focus on the technical



# Introducing the Interactive Notebook



# IPython Notebook

- IPython – interactive Python with graphics (2001)
- Jupyter – web based interface for IPython (2014)
  - Also supports other languages
- Google Colab
  - Hosted support for Jupyter on Google Drive
  - Better interface
- Concept: program as an executable document

New, upload, save

The screenshot shows the Google Colab interface for a notebook titled 'Example.ipynb'. The top navigation bar includes 'File', 'Edit', 'View', 'Insert', 'Runtime', 'Tools', and 'Help', with a status indicator 'All changes saved'. On the right, there are 'Comment', 'Share', and 'Settings' icons, along with a user profile picture. Below the navigation bar, there are resource usage indicators for RAM and Disk, and an 'Editing' mode indicator. The left sidebar contains a 'Table of contents' panel with a search icon and a list of sections: 'Introducing the Notebook', 'Cells in the Notebook', 'Write a simple program in the notebook', and 'Section'. The main content area is divided into cells. The first cell is a text cell containing the title 'Introducing the Notebook' and a paragraph of introductory text. The second cell is a text cell with the heading 'Cells in the Notebook' and a paragraph followed by a bulleted list. The third cell is a code cell containing a Python list comprehension, with its output displayed below. Annotations with blue boxes and lines point to various elements: 'New, upload, save' points to the top navigation bar; 'Text: written in 'markdown'' points to the first text cell; 'TOC: from 'sections'' points to the table of contents sidebar; 'Run button' points to the run icon in the code cell; and 'Code cell and output' points to the code and its output in the third cell.

Example.ipynb ☆

File Edit View Insert Runtime Tools Help All changes saved

Comment Share Settings

RAM  Disk  Editing

Table of contents

Introducing the Notebook

Cells in the Notebook

Write a simple program in the notebook

Section

+ Code + Text

## Introducing the Notebook

Google colab is a version of the 'iPython' notebook. This is an example note book. Edit this note book to learn how the note book system works.

### Cells in the Notebook

The notebook is a sequence if cells.

- A cell (such as this one) can contain text. This makes a data analysis a doocument.
- Other cells in the notebook contain code. The analysis document becomes executable.

The next cell contaions code. Run it!

```
[ ] x = [x for x in range(1,10)]  
x
```

```
[1, 2, 3, 4, 5, 6, 7, 8, 9]
```

Text: written in 'markdown'

TOC: from 'sections'

Run button

Code cell and output

# Practical Break

Please try the 'example' notebook



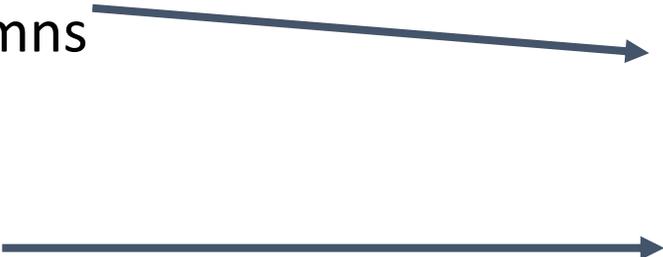
# Part 1: The Dataframe

Introducing the Python Pandas Library



# The Data Frame

- Header row
  - Shows the columns
- Rows
  - Shows individuals
- Tidy data
  - All columns have headings
  - All columns same 'type' (e.g. numbers)
  - No blanks



|   | Name  | Age | Team    |
|---|-------|-----|---------|
| 0 | John  | 24  | Arsenal |
| 1 | Mary  | 27  | Spurs   |
| 2 | Peter | 31  | Chelsea |

```
, Name, Age, Team  
0, John, 24, Arsenal  
1, Mary, 27, Spurs  
2, Peter, 31, Chelsea
```

Loaded  
from CSV

# The Data: Country of Birth

- Taken from 2011 census
  - 67,252 row
  - Example of ‘narrow’ or ‘tall’ data

| Area          | Age          | Sex     | Usual Residents | Birth Country | Birth Region |
|---------------|--------------|---------|-----------------|---------------|--------------|
| Tower Hamlets | Age 0 to 4   | Females | 3               | Ghana         | Africa       |
| Tower Hamlets | Age 5 to 9   | Females | 2               | Ghana         | Africa       |
| Tower Hamlets | Age 10 to 15 | Females | 4               | Ghana         | Africa       |

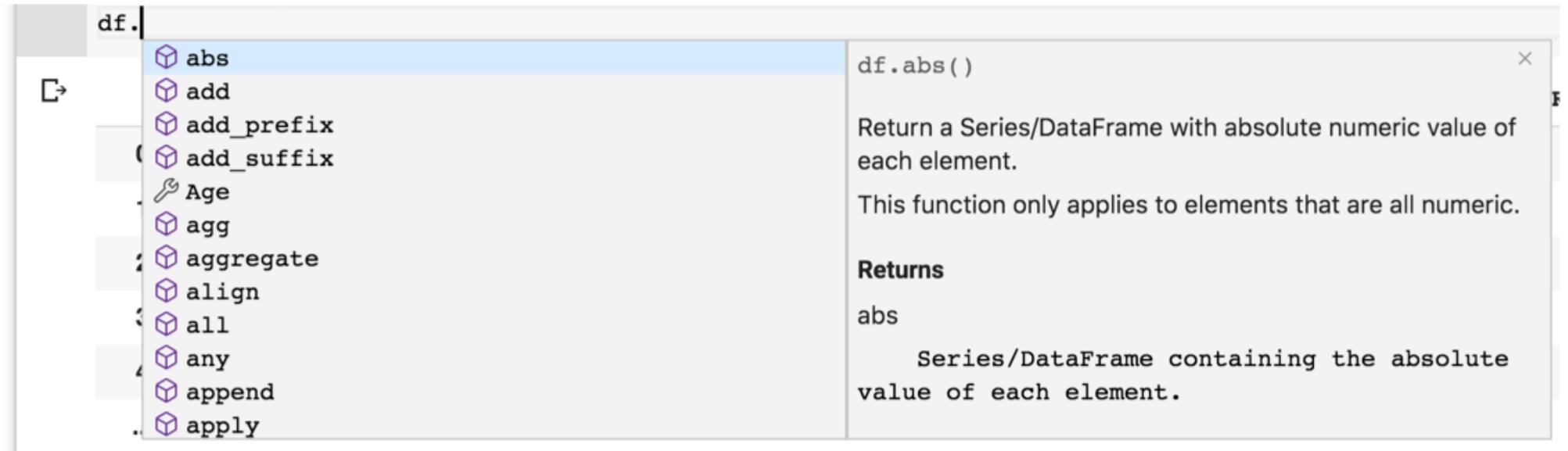
# Investigating the Data

- Task 1.1: Load the data to a dataframe
- Task 1.2: Look at values the unique values in each column
  - Answer some questions about the data

| Column          | Description                    |
|-----------------|--------------------------------|
| Area            | Includes London Boroughs       |
| Age             | The ages in a number of bands  |
| Sex             | Males and Females              |
| Usual Residents | An integer                     |
| BirthCountry    | A country                      |
| BirthRegion     | A region e.g. Africa or Europe |

# Getting Help

- Pop up help



The screenshot shows a Jupyter Notebook interface. On the left, a code cell contains the text `df.` followed by a list of pandas methods: `abs`, `add`, `add_prefix`, `add_suffix`, `Age`, `agg`, `aggregate`, `align`, `all`, `any`, `append`, and `apply`. The `abs` method is highlighted in blue. A pop-up help window is open on the right, displaying the following information:

```
df.abs()
```

Return a Series/DataFrame with absolute numeric value of each element.

This function only applies to elements that are all numeric.

**Returns**

```
abs
```

Series/DataFrame containing the absolute value of each element.

- Pandas documentation - <https://pandas.pydata.org/docs/>
  - API – many optional arguments
  - User guides

# Practical Break

## Part 1: 'Data analysis' notebook



# Part 2: Selecting, Transforming and visualising Data

- Data transformation: Tall to Wide
- The Pivot Table

# What Questions Can this Dataset Answer?

- How many people from the Americas live in Sutton?
  - Which Borough has the 'most' young people?
  - What age are people born outside the UK?
  - ...
- 
- *What additional information is missing?*



# Selecting and Transforming Data

- Selecting some data
  - Data for one borough
  - ... or one age group
  - ... or one country or region
- Use conditions
- *No loops*

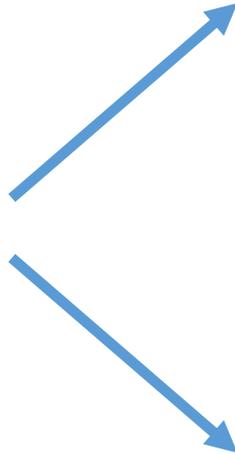
```
df[(df['Area'] == 'Tower Hamlets')]
```

```
df[(df['Area'] == 'Tower Hamlets') & (df['Age'] == 'Age 0 to 4')]
```

# The Pivot Table

- Transform data
- Origin in spreadsheets

| Person  | Genre   | Rating |
|---------|---------|--------|
| Andy    | Classic | Like   |
| Andy    | Jazz    | Hate   |
| Andy    | Folk    | Hate   |
| Bill    | Classic | Hate   |
| Bill    | Jazz    | Like   |
| Bill    | Folk    | Like   |
| Charlie | Classic | Like   |
| Charlie | Jazz    | Like   |
| Charlie | Folk    | Hate   |



| Person  | Genre   |      |      |
|---------|---------|------|------|
|         | Classic | Jazz | Folk |
| Andy    | Like    | Hate | Hate |
| Bill    | Hate    | Like | Like |
| Charlie | Like    | Like | Hate |

| Genre   | Person |      |         |
|---------|--------|------|---------|
|         | Andy   | Bill | Charlie |
| Classic | Like   | Hate | Like    |
| Jazz    | Hate   | Like | Like    |
| Folk    | Hate   | Like | Hate    |

# The Pivot Table

- Transform data

| Person  | Place  | Purpose | Visits |
|---------|--------|---------|--------|
| Andy    | Berlin | Hols    | 1      |
| Andy    | Berlin | Work    | 2      |
| Andy    | Paris  | Hols    | 2      |
| Andy    | NY     | Work    | 3      |
| Andy    | Madrid | Hols    | 1      |
| Bill    | Berlin | Work    | 4      |
| Bill    | Paris  | Work    | 3      |
| Charlie | Paris  | Hols    | 1      |
| Charlie | Rome   | Hols    | 1      |
| Charlie | Zurich | Hols    | 1      |

Aggregate over the places

| Purpose | Visits |      |         |
|---------|--------|------|---------|
|         | Andy   | Bill | Charlie |
| Hols    | 4      | 0    | 3       |
| Work    | 5      | 7    | 0       |

| Purpose | Visits |        |    |       |      |        |
|---------|--------|--------|----|-------|------|--------|
|         | Berlin | Madrid | NY | Paris | Rome | Zurich |
| Hols    | 1      | 0      | 3  | 3     | 1    | 1      |
| Work    | 6      | 1      | 0  | 3     | 0    | 0      |

Aggregate over the person

# Practical Break

Part 2: 'Data analysis' notebook



# Part 3: Adding Columns



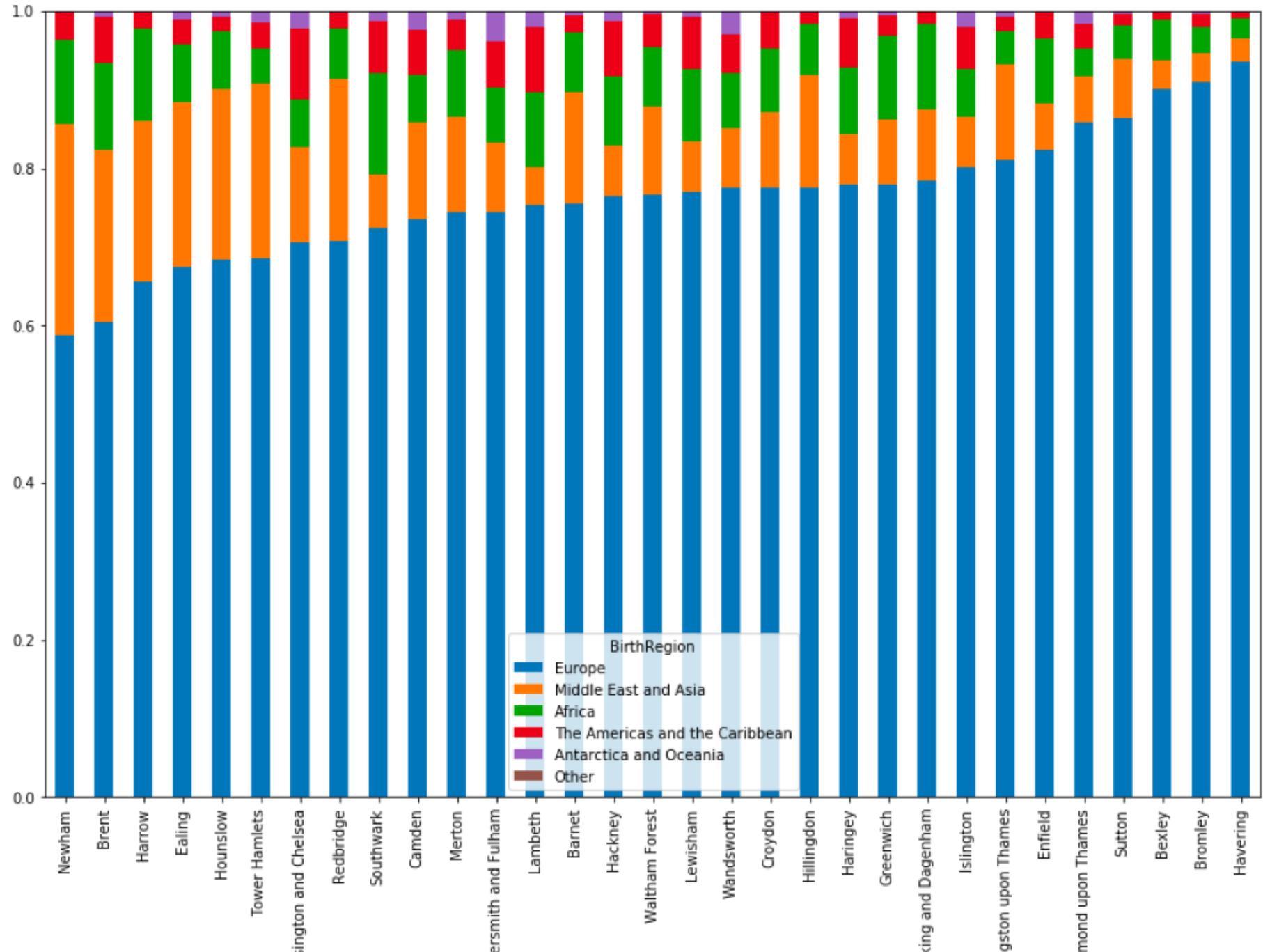
# The Problem of Comparing Boroughs

- Boroughs vary in population
- What does this question mean?

*Where in London do most people born in Spain live?*

- Need to transform the data into 'proportion'

# Proportion of Borough Born in Each Region



# Practical Break

## Part 3: 'Data analysis' notebook



# Summary and Discussion



# Summary

- The environment
  - Web-based Interactive Python
  - Does not have to be hosted
- The Pandas library
  - Comprehensive but complex
- Key concepts
  - Dataframe and selecting data
  - Transforming: pivot table
  - Visualizing: plot

## Discussion Questions

- What is emphasis of curriculum?
- Should we use a large data file?
- Is the complexity of Pandas manageable?
- Balance of technical versus 'interpretation' of data