

# **Python APIs: Day 3**

Data Boot Camp

Lesson 6.3



### **Class Objectives**

By the end of today's class, you will be able to:



Use the Geoapify API to obtain information about geographic areas.



Use the Census API to get population counts, average income, and poverty rates of cities.



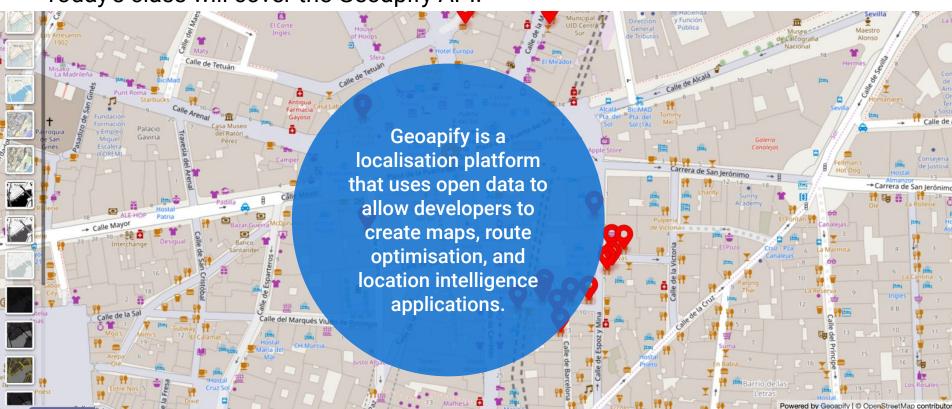
Visually represent banking and income data by using GeoViews.



**Geoapify API** 

### **Geoapify API**

Today's class will cover the Geoapify API.



## Obtaining an API Key

#### **Instructions**

Visit <a href="https://www.geoapify.com/">https://www.geoapify.com/</a>, and click "LOG IN / SIGN UP".

In the "Sign in" page, register a new account or sign in with Google or Facebook.

Set up a project and assign a name.

Copy and safely store you API key.

You can can use this API free of charge to make up to 3,000 API calls.

Review the API documentation page: <a href="https://apidocs.geoapify.com/">https://apidocs.geoapify.com/</a>.



Geoapify Geocode

#### Google Geocode



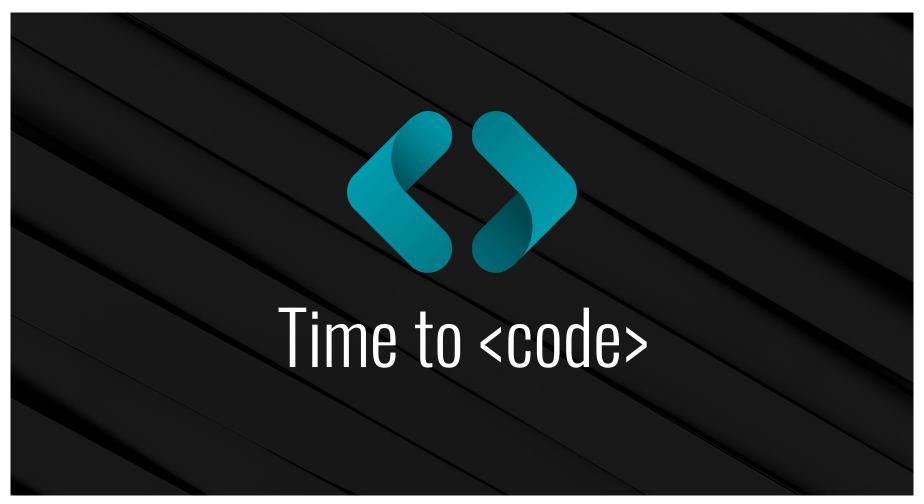
Geoapify API geocoding feature converts addresses into latitudinal and longitudinal coordinates.



This process is known as **geocoding**.



Many applications require locations to be formatted in terms of latitude and longitude.





**Geoapify Places** 





## **Activity: Geoapify Drills**

In this activity, you will generate code that makes calls to both the Geoapify Places and Geoapify Geocoding endpoints.

Suggested Time:





**Exploring Nearest Restaurants in Madrid** 

#### **Another Way to Traverse JSONs**

The Pandas method iterrows() returns an index number and the contents of each row.

 Rows can be accessed using row['column label'].

With each iteration, the keyword is overwritten.

The get() method retrieves results

- If the result exists, the value is retrieved.
- If not, then "None" is stored.
- Similar to try-except.

The try-except clause is used with loc to store the responses.

```
# Iterate through the types of DataFrame
for index, row in types df.iterrows():
    # Get the ethnicity type from the current DataFrame's row
    ethnicity = types_df.loc[index, "ethnicity"]
    # Add the current ethnicity type to the parameters
    params["categories"] = f"catering.restaurant.{ethnicity}"
    # Make an API request using the params dictionary
    restaurant = requests.get(base_url, params=params)
    # Convert the API response to JSON format
    restaurant = restaurant.json()
    # Grab the first restaurant from the results and store the details in the DataFrame
    try:
        types df.loc[index. "name"] = restaurant["features"][0]["properties"]["name"]
        types_df.loc[index, "address"] = restaurant["features"][0]["properties"]["address_line2"]
        types_df.loc[index, "distance"] = int(restaurant["features"][0]["properties"]["distance"])
    except (KevError, IndexError):
        # If no restaurant is found, set the restaurant name as "No restaurant found".
        types_df.loc[index, "name"] = "No restaurant found"
        # Set the distance column value to np.nan to allow sorting values
        types_df.loc[index, "distance"] = np.nan
    # Log the search results
    print(f"nearest {types df.loc[index, 'ethnicity']} restaurant: {types df.loc[index, 'name']}")
```







# Activity: Exploring Airports in Australia

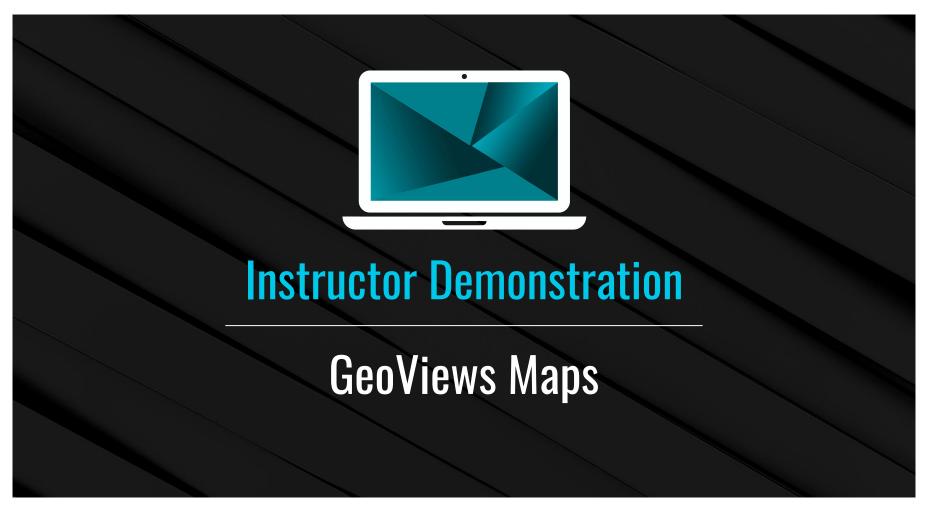
In this activity, you will obtain information for the airports in major Australian cities.

Suggested Time:

## **Activity: Exploring Airports in Australia**

Instructions	With airports.ipynb as your starting point, use the Geoapify Geocoding API, the Geoapify Places API, and Python to create a script that retrieves information of some Australian airports in each of the cities found in Cities.csv.
	Your final notebook file should contain each of the following headers: Lat, Lon, Airport Name, IATA Name, Airport Address, Distance, Website
Hints	You will need to obtain the latitude (lat) and longitude (lon) of each airport prior to sending it through the Geoapify Places API to obtain the information.
	When using the Geoapify Places API, make sure to use the "airport" category to ensure that the data received is for an airport in the city.
	Use a try-except to identify airports for which there are missing data







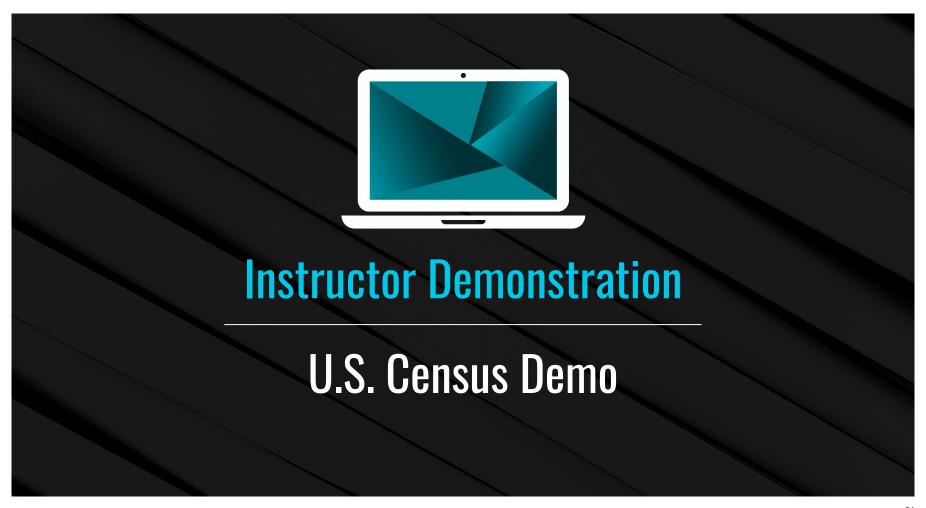


# **Activity: Australian Airports Map**

In this activity, you will create a map based on airport information.

Suggested Time:





#### U.S. Census API

#### Instructions:



Obtain an API key from <a href="http://www.census.gov/developers/">http://www.census.gov/developers/</a>.



Run pip install census in your environment.



The wrapper provides an easy way to retrieve data from the 2013 Census based on zip code, state, district, or county.



Each census field (for example, Poverty Count, Unemployment Count) is denoted with a label like B201534\_10E.



The results are then returned as a list of dictionaries, which can be immediately converted into a DataFrame.

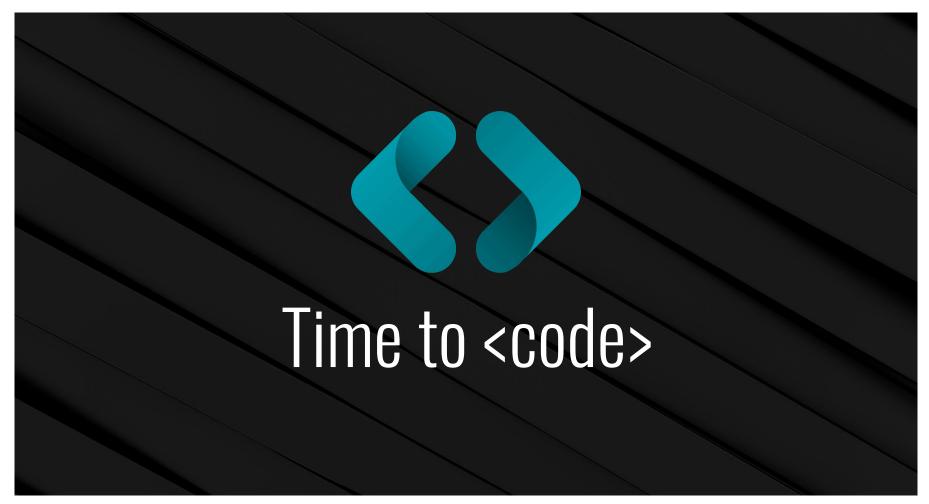
#### How We'll Use the U.S. Census API



Poverty Rate is divided by Total Population to evaluate Poverty Rate.

The U.S. Census does not explicitly calculate Poverty Rate.

```
census data = c.acs5.get(("NAME", "B19013 001E", "B01003 001E", "B01002 001E",
                          "B19301 001E".
                          "B17001 002E"), {'for': 'zip code tabulation area:*'})
# Convert to DataFrame
census_pd = pd.DataFrame(census_data)
# Column Reordering
census pd = census pd.rename(columns={"B01003 001E": "Population",
                                      "B01002 001E": "Median Age",
                                      "B19013 001E": "Household Income",
                                      "B19301 001E": "Per Capita Income",
                                      "B17001 002E": "Poverty Count",
                                      "NAME": "Name", "zip code tabulation area": "Zipcode"})
# Add in Poverty Rate (Poverty Count / Population)
census pd["Poverty Rate"] = 100 * \
   census pd["Poverty Count"].astype(
        int) / census pd["Population"].astype(int)
# Final DataFrame
census_pd = census_pd[["Zipcode", "Population", "Median Age", "Household Income",
                       "Per Capita Income", "Poverty Count", "Poverty Rate"]]
```





## **Activity: Banking Deserts**

In this activity, you will create a data visualisation to understand how prominent the banking desert phenomenon truly is.

Suggested Time:

#### **Activity: Banking Deserts Heatmap**

#### **Instructions**

Using GeoViews, create a poverty rate map:

- Use the "Poverty Rate" column to set the point's size. Recall using the `scale` parameter to modify the size appearance.
- Use the "Zipcode" column to set the point's colour.
- Read the HoloViews documentation and learn how you can use the <a href="hover\_cols">hover\_cols</a> parameter to add additional information to the tooltip of a point. Add the "Address" and the "Bank Count" columns.

Print the summary statistics for Unemployment, Bank Count, and Population.

Create a scatter plot with linear regression for Bank Count vs. Unemployment Rate.

Plot the data points.

Plot the linear regression line.

Print the R<sup>2</sup> value.

Write a sentence describing your findings. Were they what you expected? What other factors could be at play?

