

#### Introduction

- Business Problem: As we encountered pandemic situation, small business including restaurants are having trouble with keeping their business. However, I believe that we will get through this situation and will be able to overcome this situation. When we are able to travel around the world after the COVID-19 situation, it would be the great time for people to go around the world freely. So, analysis of 'Michelin one-star Restaurants would be the one of the ways we could pick to try some great food and enjoy sightseeing around the world.
- Data Source: https://www.kaggle.com/jackywang529/michelin-restaurants
- Methodology: This analysis was conducted mainly United Kingdom, California, New York City, and South Korea. United Kingdom, California, New York City have a lot of Michelin star restaurants. I also put South Korea as well since I want to show how their Michelin one-star restaurants are placed in capital. I visualized with mainly using pandas and folium.

### Read the One-Star Michelin Restaurants from csv file

#### 1.2 Read the One-Star Michelin Reastaurants from csv file

```
df = pd.read_csv('one-star-michelin-restaurants.csv')
df_counts = df["region"].value_counts()
df_counts
United Kingdom
                 138
California
                   69
New York City
Hong Kong
                   44
                  34
Singapore
Denmark
Thailand
South Korea
                   19
                  18
Chicago
                  18
Taipei
                  16
Sweden
Ireland
Washington DC
Austria
                  11
Macau
                  10
Sao Paulo
Norway
Finland
Croatia
Rio de Janeiro
Hungary
Greece
Poland
Czech Republic
Name: region, dtype: int64
```

## United Kingdom's Michelin One-star Restaurants list

#### 1.3 United Kingdom's Michelin One Star Restaurants

df\_onestar\_uk = df[df["region"] == "United Kingdom"].copy()
df\_onestar\_uk

:											
		name	year	latitude	longitude	city	region	zipCode	cuisine	price	url
	411	Loch Bay	2019	57.514870	-6.571140	Waternish	United Kingdom	IV55 8GA	Modern cuisine	NaN	https://guide.michelin.com/gb/en/highland/wate
	412	Braidwoods	2019	55.694740	-4.742500	Dalry	United Kingdom	KA24 4LN	Classic cuisine	NaN	https://guide.michelin.com/gb/en/north-ayrshir
	413	Eipic	2019	54.595898	-5.932242	Belfast	United Kingdom	BT1 6PF	Modern cuisine	NaN	https://guide.michelin.com/gb/en/belfast/belfa
	414	OX	2019	54.598910	-5.921980	Belfast	United Kingdom	BT1 3LA	Modern British	NaN	https://guide.michelin.com/gb/en/belfast/belfa
	415	The Peat Inn	2019	56.278610	-2.884580	Peat Inn	United Kingdom	KY15 5LH	Classic cuisine	NaN	https://guide.michelin.com/gb/en/fife/peat-inn
	544	The Sportsman	2019	51.343920	0.958850	Seasalter	United Kingdom	CT5 4BP	Modern British	NaN	https://guide.michelin.com/gb/en/kent/seasalte
	545	West House	2019	51.115210	0.642150	Biddenden	United Kingdom	TN27 8AH	Modern British	NaN	https://guide.michelin.com/gb/en/kent/biddende
	546	Fordwich Arms	2019	51.295284	1.126187	Fordwich	United Kingdom	CT2 0DB	Modern cuisine	NaN	https://guide.michelin.com/gb/en/kent/fordwich
	547	Samphire	2019	49.184880	-2.105700	Saint Helier/Saint- Hélier	United Kingdom	JE2 4TQ	Modern cuisine	NaN	https://guide.michelin.com/gb/en/saint-helier/
	548	Bohemia	2019	49.181225	-2.102417	Saint Helier/Saint- Hélier	United Kingdom	JE2 4UH	Modern cuisine	NaN	https://guide.michelin.com/gb/en/saint-helier/

138 rows × 10 columns

## Cleaning the dataset to visualize

```
uk_name = df_onestar_uk.groupby(["city", "cuisine"])["name"].count()
uk_name
city
             cuisine
             Modern cuisine
Anstruther
             Modern cuisine
Ascot
             Modern cuisine
Bagshot
             Modern cuisine
Baslow
Bath
             Modern cuisine
             Modern cuisine
Waternish
Westminster Modern British
Whitebrook
             Modern British
Winchester Modern cuisine
Winteringham Modern cuisine
Name: name, Length: 128, dtype: int64
cols_uk = df_onestar_uk.columns
cols_uk
Index(['name', 'year', 'latitude', 'longitude', 'city', 'region', 'zipCode',
       'cuisine', 'price', 'url'],
      dtype='object')
cols_code_uk = cols_uk[cols_uk.str.contains("zipCode|url")]
cols_code_uk
Index(['zipCode', 'url'], dtype='object')
df_onestar_uk = df_onestar_uk.drop(cols_code_uk, axis=1)
df_onestar_uk
```

## Getting latitude, longitude to visualize in folium

uk

```
uk_long = df_onestar_uk["longitude"].mean()
uk_lat = df_onestar_uk["latitude"].mean()
print('Latitude and longitude values of United Kingdom are {}, {}.'.format(uk_lat, uk_long))

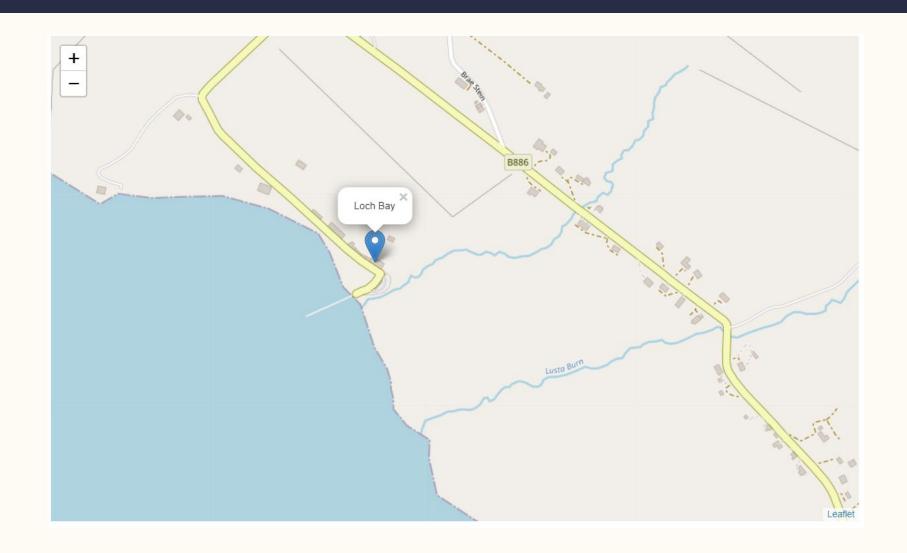
Latitude and longitude values of United Kingdom are 52.11519519565217, -1.3619043362318835.

import folium
uk = folium.Map(location=[52.11519519565217, -1.3619043362318835])
```

#### Getting a list of restaurants in United Kingdom with latitude and longitude

```
for i in df onestar uk.index:
    popup = df onestar uk.loc[i, "name"] + "-" + df onestar uk.loc[i, "cuisine"]
     lat = df onestar uk.loc[i, "latitude"]
     long = df_onestar_uk.loc[i, "longitude"]
    print(popup, lat, long)
 Loch Bay-Modern cuisine 57.51486999999995 -6.57114
 Braidwoods-Classic cuisine 55.69474 -4.7425
 Eipic-Modern cuisine 54.595898 -5.932242400000001
 0X-Modern British 54.59891 -5.9219800000000005
 The Peat Inn-Classic cuisine 56.27861 -2.8845797
 Kitchin-Modern cuisine 55.976967 -3.1728419999999997
 Martin Wishart-Modern cuisine 55.97551999999996 -3.1701900000000000
 Number One-Modern cuisine 55.95298000000004 -3.18966
 21212-Creative 55.95674 -3.18018
 The Cellar-Modern cuisine 56.222126 -2.6963209999999997
 Forest Side-Modern British 54.46396 -3.0157998
 HRiSHi-Modern cuisine 54.35013000000001 -2.91172
 Rogan & Co-Creative British 54.20074 -2.95464
 House of Tides-Modern cuisine 54.967580000000005 -1.61127
 Sosban & The Old Butchers-Modern cuisine 53,225925 -4,163006
 Northcote-Modern British 53.80933 -2.4476
 Yorke Arms-Modern cuisine 54.13435 -1.8183095
 Fraiche-Creative 53.38183000000001 -3.04307
 White Swan-Modern British 53,83916 -2,25097
Tuddun Han-Classic cuisine 52 92/05 -2 ///229
 import folium
 j = folium.Map(location=[57.51486999999995, -6.57114], zoom_start=12)
 folium.Marker([57.51486999999995. -6.57114], popup="Loch Bay"),add to(i)
```

# Output-one of the restaurants using a folium



# Using folium to visualize all the one-star Michelin restaurants in the United Kingdom

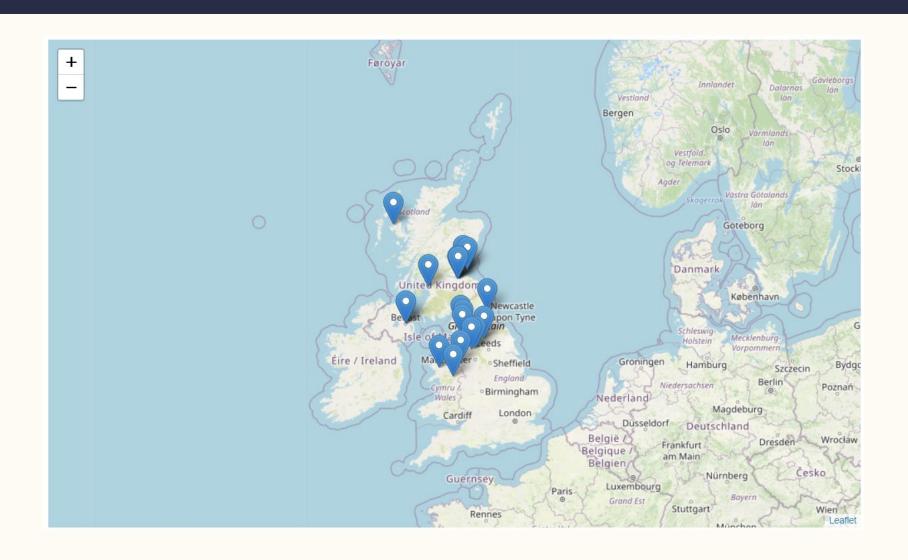
```
j = folium.Map(location=(lat, long), zoom_start=5)

for i in df_onestar_uk.index[:20]:
    popup = df_onestar_uk.loc[i, "name"] + "-" + df_onestar_uk.loc[i, "cuisine"]
    lat = df_onestar_uk.loc[i, "latitude"]
    long = df_onestar_uk.loc[i, "longitude"]

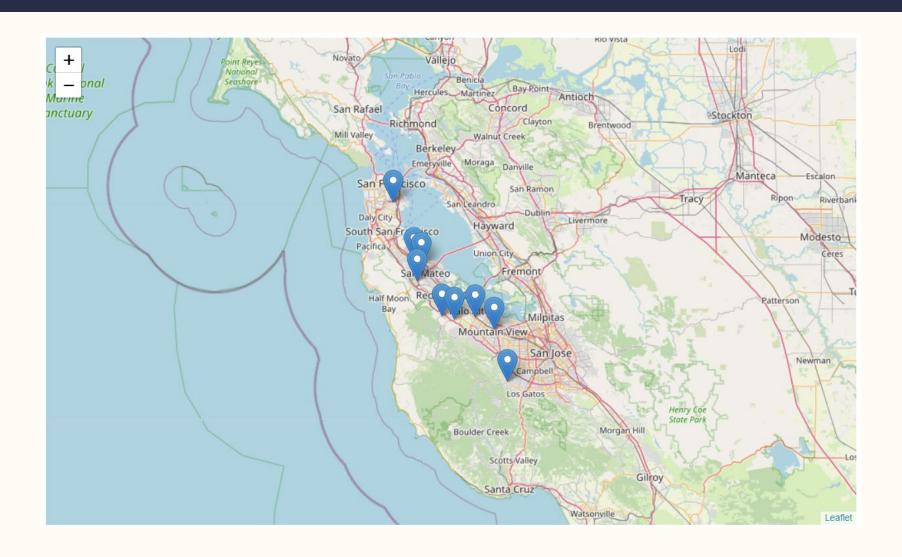
folium.Marker([lat, long], popup=popup).add_to(j)

j
```

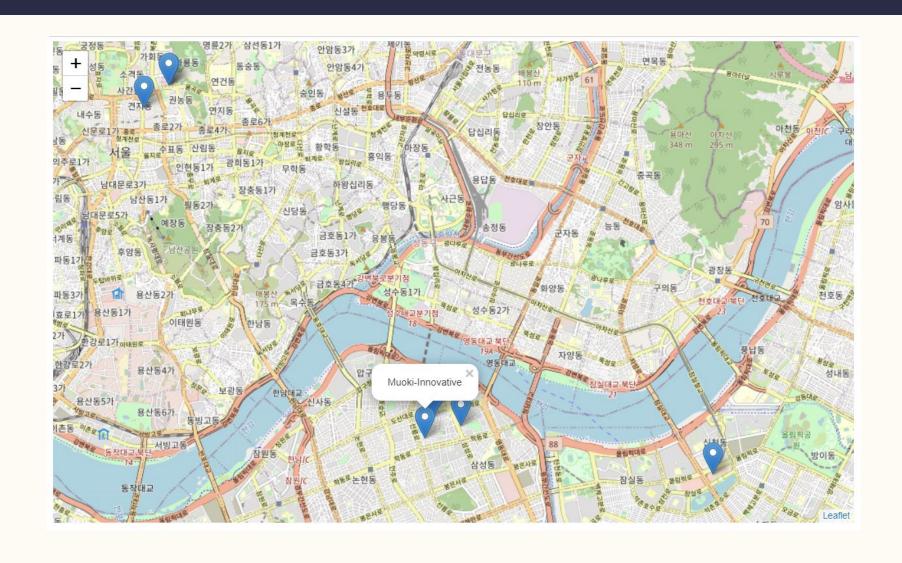
## **United Kingdom-Output**

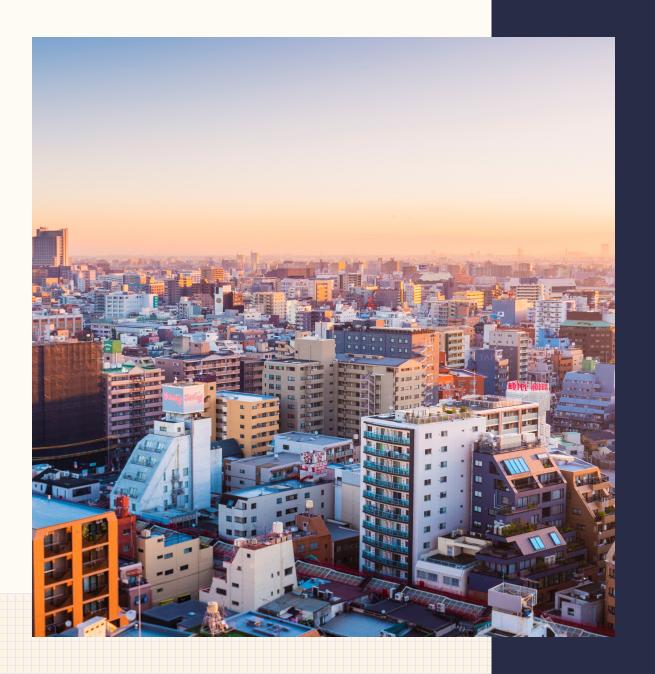


## California-Output



## **South Korea-Output**





#### **Result & Conclusion**

- Result: Top 3 region(United Kingdom, California, and New York City) are having their Michelin onestar restaurants in the city. However, as you can see at the United Kingdom case, they are also having Michelin one-star restaurants in the suburb area. So, if Michelin star restaurants are placed in the suburb area, it would bring more tourists in certain county and make more profit in those area.
- Conclusion: After looking at this analysis, it would be helpful for people who want to travel and try some new food. Also, it would be helpful for city developer to know this information since great restaurants could bring more tourists and profit.