

**Step 1: Importing Libraries(Using Darts for timeseries forecasting)**

```
In [1]: from darts import TimeSeries
```

**Step 2: Importing covid cases data from covid19h (Source: John Hopkins) from 1st November and relevant data for analysis**

```
In [2]: # Importing data obtained from Data Acquisition Team (This data contains 40% o
data = pd.read_csv('data_sa_new.csv')
```

```
In [3]:
```

```
In [4]: from datetime import datetime
from covid19dh import covid19
x, src = covid19(countries, start = datetime(2021,11,1), end = "2022-03-04")
```

C:\Users\Aditya\anaconda3\lib\site-packages\covid19dh\\_cite.py:85: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.

```
src = src.append(sources[
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```

```
references = references.append(src)
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```

**Step3: Data Preparation for Analysis**

```
In [5]: # Dividing data based on the country
import numpy as np
countries=list(data["Countries"].unique())
countries_data=[data.loc[data["Countries"] == i] for i in countries]
start, end = list(countries_data[0].day)[0],list(countries_data[0].day)[-1]
dt_ran = pd.date_range(start=start,end=end,freq="D")
t_index=pd.DatetimeIndex(dt_ran)
countries_data = [i.set_index("day").reindex(t_index).fillna(0).reset_index()
for i in countries_data:
    pop_d = [i for i in list(i['Population Density'].unique()) if i!=0]
    dest_cnt = [i for i in list(i['dest cou'].unique()) if i!=0]
    i['Population Density'] = i['Population Density'].replace(0,pop_d[0]).values
    i['dest cou'] = i['dest cou'].replace(0,dest_cnt[0]).values
```

```
In [6]: # Mapping covid cases to each country
import datetime
for cnt in countries_data:
    cases=[]
    population = []
    for index,i in list(cnt.iterrows()):
        start = i["index"]+datetime.timedelta(days=2)
        end = i["index"]+datetime.timedelta(days=1)
        y = x[x["iso_alpha_2"]==i["dest cou"]]
        date_range = y[(y["date"]>=end) & (y["date"]<=start)]
        confirmed = list(date_range["confirmed"])
        if(len(confirmed)>1):
            cases.append(confirmed[1]-confirmed[0])
        else:
            cases.append(cases[-1])
        population.append(list(y["population"])[0])
    cnt["cases"] = cases
```

C:\Users\Aditya\anaconda3\lib\site-packages\pandas\core\ops\array\_ops.py:73: FutureWarning: Comparison of Timestamp with datetime.date is deprecated in order to match the standard library behavior. In a future version these will be considered non-comparable. Use 'ts == pd.Timestamp(date)' or 'ts.date() == date' instead.

result = libops.scalar\_compare(x.ravel(), y, op)

C:\Users\Aditya\anaconda3\lib\site-packages\pandas\core\ops\array\_ops.py:73: FutureWarning: Comparison of Timestamp with datetime.date is deprecated in order to match the standard library behavior. In a future version these will be considered non-comparable. Use 'ts == pd.Timestamp(date)' or 'ts.date() == date' instead.

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C:\Users\Aditya\anaconda3\lib\site-packages\pandas\core\ops\array\_ops.py:73:

In [7]:

**Step4: Conversion to time series**

```
In [8]: series_main = [TimeSeries.from_dataframe(i, 'index', ['cases', "Seats"], fill_mi
```

**Step5: Standardization of Data**

```
In [9]: from darts.dataprocessing.transformers import Scaler
        from sklearn.preprocessing import MinMaxScaler
        scaler = MinMaxScaler()
        scaler_cases = [Scaler(scaler) for i in series_main]
        scaler_seats = [Scaler(scaler) for i in series_main]
        series_cases = [y.fit_transform(i["cases"]) for y,i in zip(scaler_cases,serie
```

```
In [10]:
```

**Step6: Applying Algorithms on data****LSTM**

```
In [11]: from darts.models import BlockRNNModel
# Model takes minimum 10 days and forecasts maximum 20 days of future data
model = BlockRNNModel(input_chunk_length=10,output_chunk_length=20,model="LST
```

```
2022-08-26 21:11:18 pytorch_lightning.utilities.rank_zero INFO: GPU available: False, used: False
2022-08-26 21:11:18 pytorch_lightning.utilities.rank_zero INFO: TPU available: False, using: 0 TPU cores
2022-08-26 21:11:18 pytorch_lightning.utilities.rank_zero INFO: IPU available: False, using: 0 IPUs
2022-08-26 21:11:18 pytorch_lightning.utilities.rank_zero INFO: HPU available: False, using: 0 HPUs
2022-08-26 21:11:18 pytorch_lightning.callbacks.model_summary INFO:
```

	Name	Type	Params
0	criterion	MSELoss	0
1	train_metrics	MetricCollection	0
2	val_metrics	MetricCollection	0
3	rnn	LSTM	2.9 K
4	fc	Sequential	1.0 K

```
-----
3.9 K    Trainable params
0        Non-trainable params
3.9 K    Total params
0.032    Total estimated model params size (MB)
```

```
Epoch 99: 100%                               34/34 [00:01<00:00, 24.90it/s, loss=0.0168, train_loss=0.0218]
```

```
2022-08-26 21:13:27 pytorch_lightning.utilities.rank_zero INFO: `Trainer.fit` stopped: `max_epochs=100` reached.
```

```
Out[11]: <darts.models.forecasting.block_rnn_model.BlockRNNModel at 0x1fc4cb9cc70>
```

### Step7: Inverse Standardization

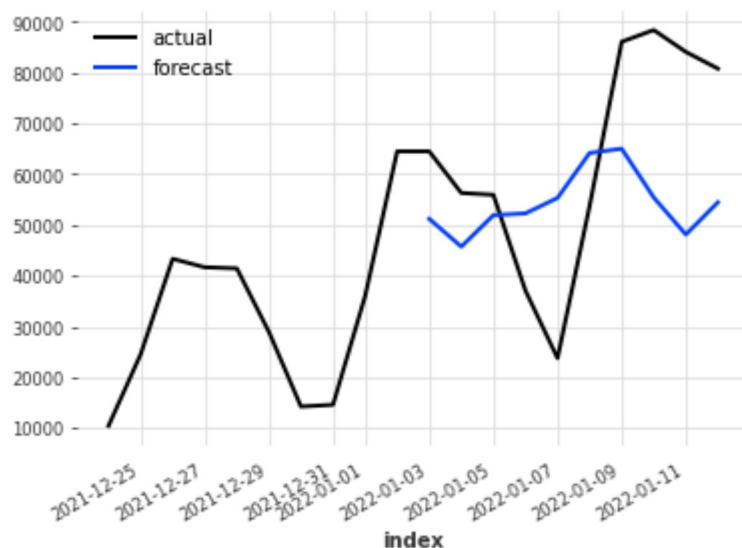
```
In [12]: pred = model.predict(n=10,series=series[4][53:63])

inverse_series_cases = [scaler_cases[index].inverse_transform(i["cases"][:94])
inverse_series_seats = [scaler_seats[index].inverse_transform(i["Seats"][:94])
inverse_pred_cases = scaler_cases[4].inverse_transform(pred["cases"])
inverse_pred_seats = scaler_seats[4].inverse_transform(pred["Seats"])
```

```
Predicting DataLoader 0: 100%
```

```
1/1 [00:00<00:00, 20.24it/s]
```

```
In [13]: # Plotting actual against forecast for a random country
import matplotlib.pyplot as plt
inverse_series_cases[4][53:73]["cases"].plot(label="actual")
```



**Step8: Calculating error percentage  $((\text{actual}-\text{forecast}) \times 100 / \text{actual})$**

```
In [15]: from darts.metrics import mae
forecast_data = []
mape_nos = []
inverse_series_cases = [scaler_cases[index].inverse_transform(i["cases"][:94])
for index,i in enumerate(countries):
    pred = model.predict(n=11,series=series[index][53:73])
    inverse_pred_cases = scaler_cases[index].inverse_transform(pred["cases"])
    print(i)
    print("MAE for cases = {:.2f}".format(mae(inverse_series_cases[index],inverse_pred_cases)
    mae_no = mae(inverse_series_cases[index],inverse_pred_cases)
    mape_no = mae_no/(int(inverse_series_cases[index].pd_dataframe().max())+1)
    mape_nos.append(mape_no)
    print(inverse_series_cases[index].pd_dataframe().max())
    forecast_df = inverse_pred_cases.pd_dataframe().reset_index()
    forecast_df["Country"] = i
```

Predicting DataLoader 0: 100%

1/1 [00:00<00:00, 55.60it/s]

United Arab Emirates (the)  
MAE for cases = 1235.45  
component  
cases 3116.0  
dtype: float64

Predicting DataLoader 0: 100%

1/1 [00:00<00:00, 58.86it/s]

Belgium  
MAE for cases = 21726.92  
component  
cases 76034.0  
dtype: float64

Predicting DataLoader 0: 100%

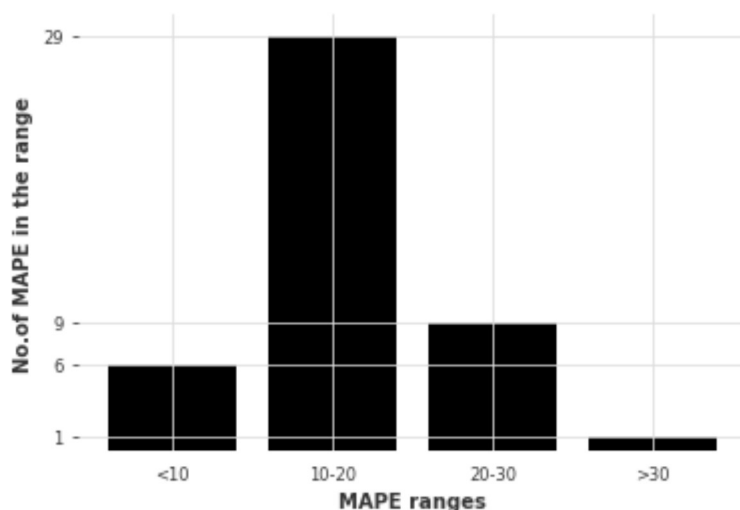
1/1 [00:00<00:00, 50.02it/s]

**Step9: Plotting the errors of country as bar plots**

```
In [16]: first=len([i for i in mape_nos if i<10])
second=len([i for i in mape_nos if i>10 and i<30])
third = len([i for i in mape_nos if i>30 and i<50])
fourth = len([i for i in mape_nos if i>50])

weights = [1,2,3,4]
bars_list = [first,second,third,fourth]

import matplotlib.pyplot as plt
x = [1, 2, 3, 4]
ax1 = plt.subplot(1,1,1)
ax1.set_xticks(x)
ax1.set_yticks(bars_list)
plt.bar(x,bars_list)
ax1.set_xticklabels(["<10", "10-20", "20-30", ">30"])
ax1.set_yticklabels(bars_list)
plt.ylabel("No.of MAPE in the range")
plt.xlabel("MAPE ranges")
```



```
In [17]:
```

**Step10: Finding Incidence Rate (no. of. cases in the country/(population of the country \* time frame of consideration))**

```
In [18]:
```

```
In [19]: pop_density = []
for row in class_data.iterrows():
    pop_density.append(list(data[data["Countries"]==row[1]["Country"]]["Popula

class_data["Population"] = pop_density
```

```
In [20]:
```

```
In [21]: class_data["Incidence"] = class_data["cases"]*1000000/class_data["Population"]
```

**Step11: One hot encoding for Country names**

```
In [22]: one_hot = pd.get_dummies(class_data["Country"])
data_sa = class_data.drop("Country", axis=1)
```

**Step12: Normalization of data**

```
In [23]: from sklearn import preprocessing

x = np.array(data_sa["Incidence"]).reshape(-1,1) #returns a numpy array
min_max_scaler = preprocessing.StandardScaler()
x_scaled = min_max_scaler.fit_transform(x)
x_scaled = [i[0] for i in x_scaled]
```

```
In [24]: # deleting unused columns
del data_sa["index"]
del data_sa["Population"]
```

**Step13: Clustering using K means clustering**

```
In [25]: from dtaidistance import dtw,clustering
from dtaidistance.clustering import kmeans
model = kmeans.KMeans(k=3)
series_clu = data_sa.to_numpy()
```

```
2022-08-26 21:15:21 be.kuleuven.dtai.distance WARNING: Warning: loading libra
ry to link with numpy returned an error
2022-08-26 21:15:21 be.kuleuven.dtai.distance WARNING: numpy.ndarray size cha
nged, may indicate binary incompatibility. Expected 96 from C header, got 88
from PyObject
40%|██████████████████████████████████████████████████████████████████████████|
| 4/10 [01:08<01:42, 17.05s/it]
```

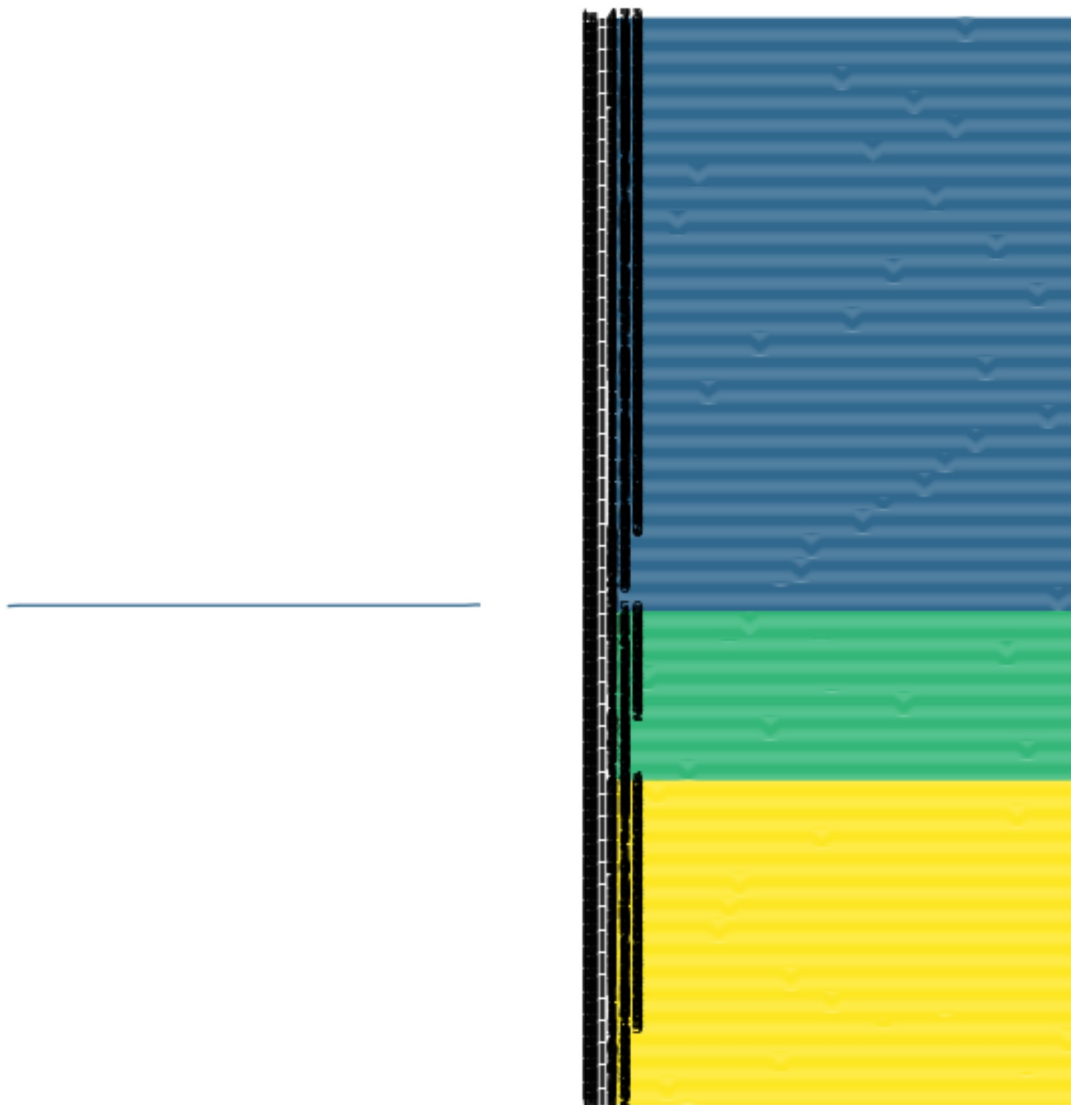
```
In [26]:
```

```
In [27]: for i in cluster_idx.keys():
         for j in cluster_idx[i]:
```



```
In [28]: # Plotting the clusters
import matplotlib.pyplot as plt
fig, ax = plt.subplots(nrows=1, ncols=2, figsize=(10, 10))
show_ts_label = lambda idx: "ts-" + str(idx)
model.plot(axes=ax, show_ts_label=show_ts_label,
           show_tr_label=True, ts_label_margin=-10,
```

```
Out[28]: (None, array([<AxesSubplot:>, <AxesSubplot:>], dtype=object))
```



```
In [29]:
```

```
In [30]: class_data=class_data.groupby("Country").agg({"Clusters":pd.Series.mode,
```

In [31]: `for i in range(3):`

```
['Argentina', 'Austria', 'Bahrain', 'Croatia', 'Cyprus', 'Czech Republic (the)', 'Germany', 'Greece', 'Ireland', 'Italy', 'Sweden', 'United Kingdom of Great Britain and Northern Ireland (the)', 'United States of America (the)']  
['Australia', 'Belgium', 'Denmark', 'France', 'Netherlands (the)', 'Spain', 'Switzerland']  
['Algeria', 'Bangladesh', 'Brazil', 'Canada', 'Egypt', 'Hong Kong', 'India', 'Kazakhstan', 'Korea (the Republic of)', 'Lebanon', 'Malaysia', 'Malta', 'Morocco', 'Pakistan', 'Philippines (the)', 'Poland', 'Qatar', 'Romania', 'Russian Federation (the)', 'Saudi Arabia', 'Singapore', 'South Africa', 'Turkey', 'Ukraine', 'United Arab Emirates (the)']
```

***Step14: Export the results into excel file for visualization***

In [32]: