```
In [1]:
    #Import of the main libraries for DataViz
    import pandas as pd
    import matplotlib.pyplot as plt
    %matplotlib inline
    import seaborn as sns
    from scipy import stats as stats
    print("Setup Complete")
```

Setup Complete

Quick Analysis for Covid Evolution in France

Goal

The main goal is to visualize the covid 19 pandemic in france using jupyter Notebook.

Sources

I used french gouverment open data website here. The data is for the whole France.

```
In [2]:
# definition of the file path
file_path_indicateurs = '/table-indicateurs-open-data-france.csv'

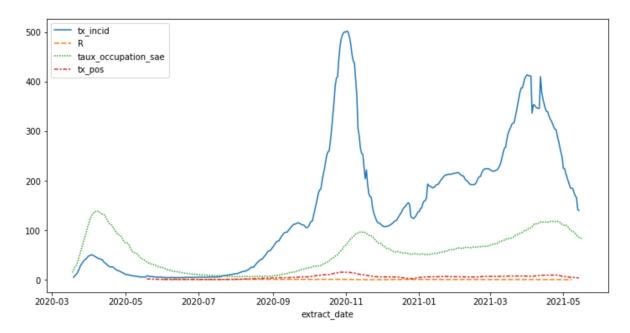
#I put the data into a pandas dataframe
data_indicateurs = pd.read_csv(file_path_indicateurs)
data_indicateurs['extract_date'] = pd.to_datetime(data_indicateurs['extract_data_indicateurs.set_index('extract_date', inplace=True)
#I read the head of the dataframe to have an understanding of data structur_data_indicateurs.head()
```

Out [2]: tx_incid R taux_occupation_sae tx_pos

```
extract date
2020-03-18
                                         15.2
                                               NaN
                NaN NaN
2020-03-19 5.570823 NaN
                                         198
                                               NaN
2020-03-20 8.403950 NaN
                                               NaN
                                         25.6
2020-03-21 11.013409 NaN
                                         28.7
                                               NaN
2020-03-22 13.600502 NaN
                                         33.1
                                               NaN
```

Quick graph of the data in function of time

```
In [3]:
#Plotting data in function of time
ax1 = plt.figure(figsize=(12,6))
ax1 = sns.lineplot(data=data_indicateurs)
```



Clearly the main phases of the epidemic can be seen with the incidence rate evolution trough time.

Looking for relation between variables

we delete missing values to make better correlatins between variables. Then we can ask ourselves what are the relation among outputs.

```
In [4]: #Deleting missing values
    data_indicateurs2 = data_indicateurs.dropna()
    data_indicateurs2.reset_index(inplace=True)
    data_indicateurs2.head()
```

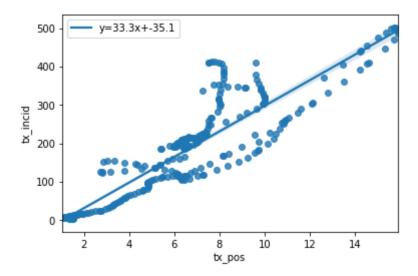
Out[4]:		extract_date	tx_incid	R	taux_occupation_sae	tx_pos
	0	2020-06-01	5.197050	0.84	25.1	1.501300
	1	2020-06-02	5.147881	0.84	24.1	1.520426
	2	2020-06-03	5.067422	0.83	23.3	1.507729
	3	2020-06-04	5.048052	0.84	22.4	1.522279
	4	2020-06-05	5.000373	0.85	21.1	1.526863

Use of Linear Regression

I use linear regression to any sort of correlation between the variables.

```
In [5]:
#Linear regression between tx_pos and tx_incid with equation of the regress
slope, intercept, r_value, p_value, std_err = stats.linregress(data_indicat

ax2 = sns.regplot(x="tx_pos", y="tx_incid", data=data_indicateurs2,
    line_kws={'label':"y={0:.1f}x+{1:.1f}".format(slope,intercept)})
ax2.legend()
```



Here we can see a strong relation between the incidence rate and the positivity rate.

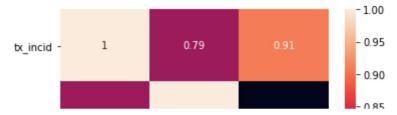
Correlation between variables

```
In [6]: #Creating a dataframe with only the numerics values
   dataframe = pd.DataFrame(data_indicateurs2, columns=['tx_incid', 'taux_occu
   dataframe.head()
```

Out[6]:		tx_incid	taux_occupation_sae	tx_pos
	0	5.197050	25.1	1.501300
	1	5.147881	24.1	1.520426
	2	5.067422	23.3	1.507729
	3	5.048052	22.4	1.522279
	4	5.000373	21.1	1.526863

```
In [7]:
         #creating the correlation matrix
         matrix = dataframe.corr()
         print (matrix)
                              tx incid taux occupation sae
                                                                tx pos
        tx incid
                              1.000000
                                                    0.794302
                                                              0.908775
        taux occupation sae
                             0.794302
                                                    1.000000
                                                              0.662972
                              0.908775
                                                    0.662972
        tx pos
                                                              1.000000
In [8]:
         sns.heatmap(data=matrix, annot=True)
```

Out[8]: <matplotlib.axes._subplots.AxesSubplot at 0x7ffa1158b950>



With this correlation matrix we can see that all variables are correlated. This is normal, because they all monitor the pandemic.

They also show that we don't need all of this to monitor the epidemic, but only the ones the less correlated; like: incidence rate and occupation rate.

Indeed if we take the positivity rate we can make good approximations based on the incidence rate. The incidence rate being more contrasted across time it will show more phases or informations than a flat line; Even if the incidence rate is a rate.

Conclusion

Better Analysis we'll take broader data and make more precise analysis. As a quick analysis it as shows that we can make some statement in a few times based on data freely at disposal.

On the other hand we can also critic the sources that are already formated. That was the job of this quick examination.