## Exercise 2

In this exercise, we will plot COVID hospital admissions in the UK from March 2020 to February 2023, and you will fit an exponential curve to understand the rise in hospital admissions in a three week period from late December 2021 to early January 2022.

The data is in a file called hospital\_cases\_2023-02-16.csv (comma-separated-values format). It was downloaded from the official website for UK COVID-19 data.

```
# If you are running this on Google Colab, uncomment and run the
following lines; otherwise ignore this cell
# from google.colab import drive
# drive.mount('/content/drive')
import math
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
/var/folders/ds/gvg7h1s93x16vvczwp8vr83m0000gn/T/
ipykernel 57903/188978642.py:4: DeprecationWarning:
Pyarrow will become a required dependency of pandas in the next major
release of pandas (pandas 3.0),
(to allow more performant data types, such as the Arrow string type,
and better interoperability with other libraries)
but was not found to be installed on your system.
If this would cause problems for you,
please provide us feedback at
https://github.com/pandas-dev/pandas/issues/54466
  import pandas as pd
```

We will use pandas, a library for data analysis in Python to load and view the data. Pandas uses a structure called a *data frame* to represent and manipulate data tables. All the required commands are included here, so you won't need to learn Pandas for this exercise. But if you are interested in learning more, this is a good place to start.

The command pd. read\_csv loads the data onto a data frame. We have used the .head() command to display the top 3 rows of the data frame.

We can also display a random sample of rows from the data frame using .sample(), or the last few rows using .tail().

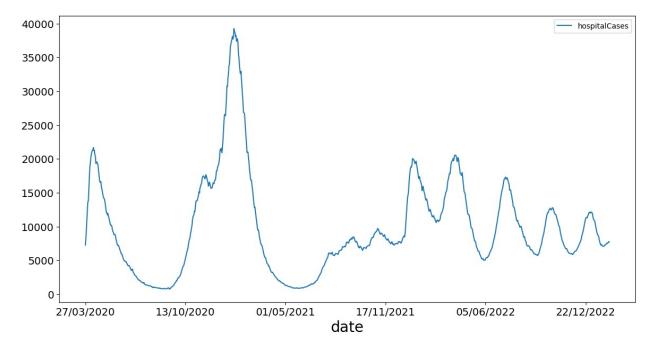
```
df_hosp.tail(3)

date hospitalCases
1045 05/02/2023 7647
1046 06/02/2023 7795
1047 07/02/2023 7737
```

You can plot one column against another by just using their column names. Let us plot the hospital Cases column versus date.

```
plt.rcParams['figure.figsize'] = [14, 7]
plt.rcParams['axes.titlesize'] = 20
plt.rcParams['axes.labelsize'] = 20
plt.rcParams['xtick.labelsize'] = 14
plt.rcParams['ytick.labelsize'] = 14

df_hosp.plot(x='date', y='hospitalCases')
plt.show()
```



Observe the sharp increase in hospital admissions corresponding to each wave; also notice that the peaks after mid-2021 are smaller (due to the vaccines). We now extract the rows spanning a three week period starting 22 December 2021 (when Omicron first spread in the UK) into a data frame called  $\frac{df_{part}}{df}$ 

```
df_part = df_hosp[635:656]
df part.head(2)
                 hospitalCases
           date
635
     22/12/2021
                           8400
                           8436
636
     23/12/2021
df part.tail(2)
                 hospitalCases
           date
654
     10/01/2022
                          20065
                          19967
655
     11/01/2022
```

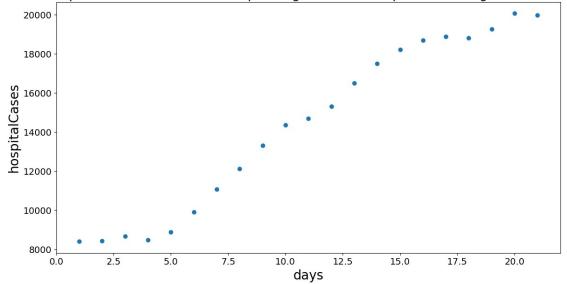
We now convert the hospitalCases column to a numpy array yvals

```
yvals = np.array(df_part['hospitalCases'])
N = np.size(yvals)
xvals = np.linspace(1,N,N) #an array containing the values 1,2...,N
```

## 2a) Plot the data yvals vs xvals in a scatterplot

```
# Your code for scatterplot here
plt.scatter(xvals, yvals)
plt.title("COVID hospital admissions in the UK spanning a three week
period starting 22 December 2021")
plt.xlabel("days")
plt.ylabel('hospitalCases')
plt.show()
```

COVID hospital admissions in the UK spanning a three week period starting 22 December 2021



## 2b) Fit an exponential model to the data

From our knowledge of how the virus spreads, we know that the number of infections, hospital admissions etc. should (roughly) follow an exponential curve. We would therefore like to fit a model of the form  $y = c_1 e^{c_2 x}$ , where y is the number of admissions on day x.

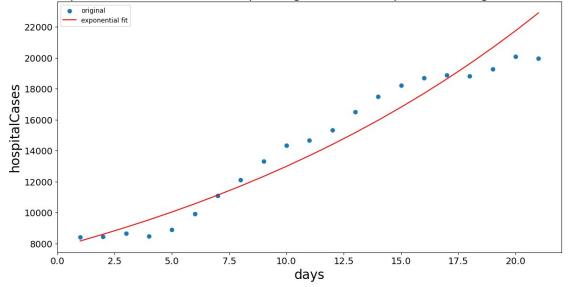
Note that this is a linear model on a log-scale for y. That is,  $\log y = \log c_1 + c_2 x$ .

- Fit a linear model for  $\log (yvals)$  vs xvals, and print the values of  $c_1$  and  $c_2$
- Plot the fit  $y = c_1 e^{c_2 x}$  along with the scatterplot of the data

```
# your code here
def exponentialreg(xvals, yvals):
    # Your code here
    # The function should return the the coefficient vector beta, the
fit, and the vector of residuals
    log yvals = np.log(yvals)
    lsg matrix = np.column stack((xvals**0, xvals))
    lsq matrix T = lsq matrix.T
    optimal params =
np.linalg.inv(lsq matrix T.dot(lsq matrix)).dot(lsq matrix T).dot(log
yvals)
    log yvals pred = lsq matrix.dot(optimal params.T)
    yvals pred = np.exp(log yvals pred)
    residuals = yvals - yvals_pred
    c1 = np.exp(optimal params[0])
    c2 = optimal params[1]
    return (c1,c2), yvals pred, residuals
best_fit_params, y_vals_pred, residuals = exponentialreg(xvals, vvals)
print(f"c1: {best fit params[0]} | c2: {best fit params[1]}")
SSE = np.linalg.norm(residuals)**2
var = np.var(vvals)
R2 = 1 - SSE/(var * yvals.shape[-1])
print(f"SSE: {SSE} | R2: {R2}")
# Your code for scatterplot here
plt.scatter(xvals, yvals, label="original")
plt.plot(xvals, y_vals_pred, label="exponential fit", color="red")
plt.title("COVID hospital admissions in the UK spanning a three week
period starting 22 December 2021")
plt.xlabel("days")
plt.ylabel('hospitalCases')
plt.legend()
plt.show()
```

```
c1: 7754.297564390236 | c2: 0.051565659399377045
SSE: 29200845.657093737 | R2: 0.9234924160353204
```

COVID hospital admissions in the UK spanning a three week period starting 22 December 2021



## 2c) Estimate the weekly growth rate in hospital admissions (in %) over this period

*Hint*: According to the model, admissions increase every 7 days by a factor of  $(c_1e^{c_2(x+7)})I(c_1e^{c_2x})=e^{7c_2}$ .

```
# compute and print weekly growth rate (in %)
weekly_growth_rate = np.exp(7 * best_fit_params[1])
perentage_weekly_growth_rate = (weekly_growth_rate - 1) * 100
print("weekly growth rate in hospital admissions (in %)
=",perentage_weekly_growth_rate, "%")
weekly growth rate in hospital admissions (in %) = 43.470552026769305
%
```