Programming Assignment 2: Linear regression

As we know, the COVID-19 pandemic has been disrupting people's life around the world. To get it under control, a crucial aspect is to be able to accurate forecast the spread of the disease, which can be helpful as a planning tool for policymakers, clinicians, and public health officers to deal with this crisis. In this notebook,we will try to do some forecasting on the covid-19 epidemic progression using machine learning. We will use a dataset based on the COVID-19 Data Repository at John Hopkins university.

```
In [99]: import pandas as pd
    from datetime import datetime, timedelta
    import matplotlib.pyplot as plt
    from matplotlib.dates import DateFormatter
    import numpy as np
    from math import sqrt
    import sys
In [100]: print (sys.version)

3.8.1 (tags/v3.8.1:1b293b6, Dec 18 2019, 23:11:46) [MSC v.1916 64 bit (AMD6 4)]
```

First, let us load the data.

Out[101]:

	date	Country	hospitalized_with_symptom	Intensive_care	Total_hospitalized	Home_I
0	2020-02- 24T18:00:00	ITA	101	26	127	
1	2020-02- 25T18:00:00	ITA	114	35	150	
2	2020-02- 26T18:00:00	ITA	128	36	164	
3	2020-02- 27T18:00:00	ITA	248	56	304	
4	2020-02- 28T18:00:00	ITA	345	64	409	
201	2020-09- 12T17:00:00	ITA	1951	182	2133	
202	2020-09- 13T17:00:00	ITA	2042	187	2229	
203	2020-09- 14T17:00:00	ITA	2122	197	2319	
204	2020-09- 15T17:00:00	ITA	2222	201	2423	
205	2020-09- 16T17:00:00	ITA	2285	207	2492	

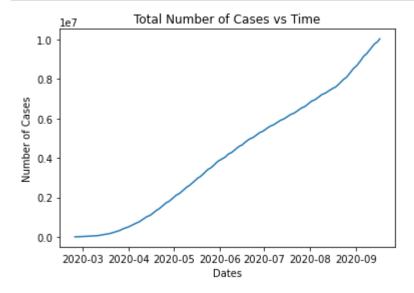
206 rows × 13 columns

```
In [102]: # change the date format
dates = data['date']
date_format = [pd.to_datetime(d) for d in dates]
```

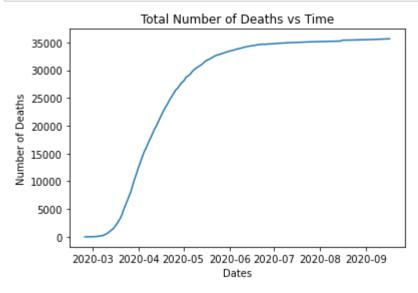
Data Visualization

Task P1: complete the following **three** visualization graphs that show the trend of the epidemic progression. Copy them to the solution file.

Graph 1: plot the total number of people tested for the entire period of the dataset. Your X axis will be the dates ("Dates") and Y-axis will be the total number of cases ("People tested") over the period of time.



Graph 2: plot the total number of deaths for the entire period. Your X axis will be the dates ("Dates") and Y-axis will be the total number of death cases("Deaths") over the period of time.



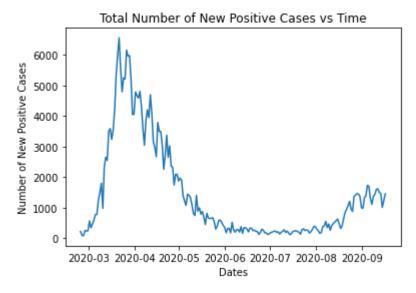
Graph 3: plot the total number of infected cases for the entire period. Your X axis will be the dates ("Dates") and Y-axis will be the total number of infected cases ('New_positive_cases') over the period of time.

```
In [105]: ### STUDENT: Start of Code ###
    plt.title("Total Number of New Positive Cases vs Time")
    plt.xlabel("Dates")
    plt.ylabel("Number of New Positive Cases")
    New_positive_cases = data['New_positive_cases']

    plt.plot(date_format, New_positive_cases)

    plt.show()

### End of code ####
```



As we can see that our data has different ranges of values for every feature and this can cause problems in our model, so here we will normalize our data (ignoring the categorical variables) so that our data is scaled between 0 and 1. The downside, however, is that the numbers are no longer interpretable. To interpret it, you need to multiply back by the scaling factor.

IMPORTANT: From now on, we will work with the normalized features to build the regression model. However, in **Task P8**, you need to convert the number back to the actual units.

```
In [106]: data_list = data_orig.columns.values.tolist()

for i in data_list[-11:]:
    data[[i]]=(data_orig[i]-data_orig[i].min())/(data_orig[i].max()-data_orig[i].min())

data
```

Out[106]:

date	Country	hospitalized_with_symptom	Intensive_care	Total_hospitalized	Home_I
2020-02- 24T18:00:00	ITA	0.000000	0.000000	0.000000	(
2020-02- 25T18:00:00	ITA	0.000450	0.002227	0.000700	(
2020-02- 26T18:00:00	ITA	0.000934	0.002474	0.001125	(
2020-02- 27T18:00:00	ITA	0.005085	0.007422	0.005384	(
2020-02- 28T18:00:00	ITA	0.008440	0.009401	0.008577	(
2020-09- 12T17:00:00	ITA	0.063994	0.038595	0.061015	(
2020-09- 13T17:00:00	ITA	0.067142	0.039832	0.063935	(
2020-09- 14T17:00:00	ITA	0.069909	0.042306	0.066673	(
2020-09- 15T17:00:00	ITA	0.073368	0.043295	0.069836	(
2020-09- 16T17:00:00	ITA	0.075547	0.044780	0.071935	(
	2020-02- 24T18:00:00 2020-02- 25T18:00:00 2020-02- 26T18:00:00 2020-02- 27T18:00:00 2020-09- 13T17:00:00 2020-09- 14T17:00:00 2020-09- 15T17:00:00 2020-09-	2020-02- 25T18:00:00 ITA 2020-02- 25T18:00:00 ITA 2020-02- 26T18:00:00 ITA 2020-02- 27T18:00:00 ITA 2020-02- 28T18:00:00 ITA 2020-09- 12T17:00:00 ITA 2020-09- 13T17:00:00 ITA 2020-09- 14T17:00:00 ITA 2020-09- 15T17:00:00 ITA	2020-02- 24T18:00:00 ITA 0.000000 2020-02- 25T18:00:00 ITA 0.000450 2020-02- 26T18:00:00 ITA 0.000934 2020-02- 27T18:00:00 ITA 0.005085 2020-02- 28T18:00:00 ITA 0.008440 2020-09- 12T17:00:00 ITA 0.063994 2020-09- 14T17:00:00 ITA 0.069909 12020-09- 15T17:00:00 ITA 0.075547	2020-02- 25T18:00:00	2020-02- 25T18:00:00

206 rows × 13 columns

Calculate the feature matrix

The following is a function that accepts a list of feature names (e.g. ['Total_Hospitalized', 'People_tested']) and an target feature e.g. ('Deaths') and returns two things:

- 1. A numpy matrix whose columns are the desired features plus a column with a constant value 1, which is also known as the 'intercept'.
- 2. A numpy array that contains the values of the target output.

```
In [107]: def get_numpy_data(data_frame, features, output):
              # select the columns of data Frame given by the features list into the var
          iable features sframe which will include the constant
              # Convert the features_frame into a numpy matrix
              # assign the column of data frame associated with the output to the array
           output array
              # convert the array into a numpy array by first converting it to a list
              # return feature matrix, output array
              data frame['constant'] = 1 # here we are adding a constant column
              # add the column 'constant' to the front of the features list.
              features = ['constant'] + features
              # select the columns of data Frame given by the features list into the var
          iable features sframe which will include the constant)
              features frame = data frame[features]
              # Convert the features_frame into a numpy matrix
              feature matrix = features frame.to numpy()
              # print ("feature matrix:", feature matrix)
              # assign the column of data frame associated with the output to the array
           output array
              output array = data frame[output]
              # convert the array into a numpy array by first converting it to a list
              output array = output array.to numpy()
              return(feature matrix, output array)
```

For dates, we need to convert them into a sequence of numbers. We now add a new column to our dataframe corresponding to the number of days since the start of the dataset.

Test the above function for a particular input and output feature.

```
In [109]: (example_features, example_output) = get_numpy_data(data, ['Days'], 'New_posit
    ive_cases')
    print (example_features[0,:])
    print (example_output[0])

[1 1]
    0.022071307300509338
```

In [111]:

Predict the outputs with given regression weights

Suppose we had the weights [1,1] corresponding to the features [1,100], to compute the predicted output, we can simply take the dot product between them, so the output is 1*1+1*100=101. Now, let's create the data with

```
In [110]: (test_features, output) = get_numpy_data(data, ['Days'], 'People_tested')
```

Task P2: Complete the following function 'predict' output'. Copy the the outputs of the code to the solution file.

def predict output(feature matrix, weights):

```
# Inputs:
              # feature matrix: a numpy matrix containing the features as columns (inclu
          ding the intercept),
                                and each row corresponds to a data point
              # weights: a numpy array for the corresponding regression weights (includi
          ng the intercept)
              # Output:
              # a numpy array that contains the predicted outputs (according to the prov
          ided weights)
              # for all the data points in the feature matrix
              # STUDENT: Start of code ####
              features = np.array(feature matrix)
              regress weights = np.array(weights)
              return np.dot(features, regress weights)
              ## end of code
In [112]: # Copy the outputs of this code to the solution file
          my weights = np.array([1., 1.])
          test_predictions = predict_output(example_features, my_weights)
          print ("(normalized) prediction at day 5: ", test_predictions[5])
          print ("(normalized) prediction at day 20 ", test predictions[20])
          (normalized) prediction at day 5: 7.0
          (normalized) prediction at day 20 22.0
```

Compute the derivative

We will now compute the derivative of the regression cost function:

$$L_D(w) = rac{1}{n} \sum_{i=1}^n (y_i - w \cdot x_i)^2,$$

where $x_i \in \mathbb{R}^d$ is the input feature of dimension d, $y_i \in \mathbb{R}$ is the output response, and $w \in \mathbb{R}^d$ is the regression weights.

Task P3: Complete the function 'weight_derivative' to calculate the derivative of the cost function with respect to regression weights w, i.e., $\frac{\partial}{\partial w}L_D(w)$. Note that this should be a d dimensional vector. Also copy the output of the code for the test example to the solution file.

```
In [113]: def weight derivative(weights, feature matrix, labels):
              # Input:
              # weights: weight vector w, a numpy vector of dimension d
              # feature matrix: numpy array of size n by d, where n is the number of dat
          a points, and d is the feature dimension
              # labels: true labels y, a numpy vector of dimension d
              # Output:
              # Derivative of the regression cost function with respect to the weight w,
          a numpy array of dimension d
              ## STUDENT: Start of code ###
              error = np.subtract( labels, np.dot(feature matrix, weights ) )
              derivative = -2/len(labels)*np.dot(np.transpose(feature matrix), error)
                print("Size of Weights:", weights.shape )
                print("Size of features:", feature_matrix.shape )
                print("Size of Labels:", labels.shape )
              return derivative
              # End of code ###
```

```
In [114]: # NOTE: copy the output to the solution file.

    (example_features, example_output) = get_numpy_data(data, ['Days'], 'People_te sted')
    my_weights = np.array([0., 0.]) # this makes all the predictions 0
    derivative = weight_derivative(my_weights, example_features, example_output)
    print (derivative)
```

-0.82103242 -120.60087518]

Gradient descent algorithm

Here, we will write a function to perform gradient descent algorithm on the lineare regression cost. Given an initial point, we will update the current weights by moving in the negative gradient direction to minimize the cost function. Thus, in each iteration we obtain the updated weight w_{t+1} from the current iterate w_t as follows:

$$w_{t+1} = w_t - hrac{\partial}{\partial w} L_D(w_t),$$

where h is the 'step size' that is the amount by which we move in the negative gradient direction.

We stop when we are sufficiently close to the optimum (where gradient is the zero vector) by checking the condition with respect to the magnitude (length) of the gradient vector:

$$\|rac{\partial}{\partial w}L_D(w_t)\|_2 \leq \epsilon,$$

where ϵ is the 'tolerance' parameter.

Task P4: Complete the code section to perform the gradient decent in the function regression_gradient_descent . Copy the code to the solution file.

```
In [115]: def regression gradient descent(feature matrix, labels, initial weights, step
          size, tolerance):
              # Gradient descent algorithm for linear regression problem
              # Input:
              # feature_matrix: numpy array of size n by d, where n is the number of dat
          a points, and d is the feature dimension
              # labels: true labels y, a numpy vector of dimension d
              # initial weights: initial weight vector to start with, a numpy vector of
           dimension d
              # step size: step size of update
              # tolerance: tolerace epsilon for stopping condition
              # Output:
              # Weights obtained after convergence
              converged = False
              weights = np.array(initial weights) # current iterate
              while not converged:
                  i += 1
                  # STUDENT: Start of code: your impelementation of what the gradient de
          scent algorithm does in every iteration
                  # Refer back to the update rule listed above: update the weight
                  derivative = weight_derivative(weights, feature_matrix, labels);
                  weights = weights - step size*derivative
                  # Compute the gradient magnitude:
                  gradient magnitude = np.linalg.norm( weight derivative(weights, featur
          e matrix, labels) )
                  # Check the stopping condition to decide whether you want to stop the
           iterations
                  if gradient magnitude <= tolerance: # STUDENT: check the stopping c</pre>
          ondition here
                      converged = True
                  # End of code
                   print ("Iteration: ",i,"gradient_magnitude: ", gradient_magnitude) # f
          or us to check about convergence
              return(weights)
```

Use gradient descent for linear regression

Let's test the gradient descent algorithm for linear regression with a single feature ('Day'). Here we are using first 180 days' data as our training data.

```
In [116]: #train_data
train_data = data[:180]
```

Task P5: Specify the initial_weights, step_size, and tolerance for the function regression_gradient_descent . Copy the outputs of the code to the solution file.

```
In [117]:
          simple features = ['Days']
          my_output = 'People_tested'
          # Use get numpy data method to calculate the feature matrix and output.
          (simple_feature_matrix, output) = get_numpy_data(train_data, simple_features,
          my_output)
          #Initialize the weights, step size and tolerance
          # Start of code
          #STUDENT: Specify the initial_weights, step_size, and tolerance
          initial weights = [1.0, 1.0]
          step\_size = 7e-8
          tolerance = 1.2e4
          # end of code
          # Use the regression_gradient_descent function to calculate the gradient decen
          t and store it in the variable 'final_weights'
          final_weights = regression_gradient_descent(simple_feature_matrix, output, ini
          tial_weights, step_size, tolerance)
          # end of code
          print ("Here are the final weights after convergence:")
          print (final weights)
```

<ipython-input-107-8b2916f5b454>:9: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s table/user_guide/indexing.html#returning-a-view-versus-a-copy data_frame['constant'] = 1 # here we are adding a constant column

```
Iteration:
            1 gradient magnitude:
                                    21840.81707866705
Iteration:
            2 gradient magnitude:
                                    21807.51575965956
            3 gradient magnitude:
Iteration:
                                    21774.265216126485
Iteration:
            4 gradient magnitude:
                                    21741.06537064902
Iteration:
            5 gradient magnitude:
                                    21707.91614592638
Iteration:
            6 gradient magnitude:
                                    21674.81746477566
            7 gradient magnitude:
Iteration:
                                    21641.769250131623
Iteration:
            8 gradient magnitude:
                                    21608.77142504654
Iteration:
            9 gradient magnitude:
                                    21575.82391269002
            10 gradient magnitude:
Iteration:
                                     21542.92663634879
            11 gradient magnitude:
Iteration:
                                     21510.079519426563
Iteration:
            12 gradient magnitude:
                                     21477.282485443844
            13 gradient magnitude:
Iteration:
                                     21444.53545803773
Iteration:
            14 gradient magnitude:
                                     21411.838360961778
Iteration:
            15 gradient magnitude:
                                     21379.191118085775
            16 gradient magnitude:
Iteration:
                                     21346.593653395597
            17 gradient magnitude:
Iteration:
                                     21314.045890993017
Iteration:
            18 gradient magnitude:
                                     21281.547755095533
Iteration:
               gradient magnitude:
                                     21249.099170036196
Iteration:
            20 gradient magnitude:
                                     21216.70006026343
Iteration:
            21 gradient magnitude:
                                     21184.350350340832
            22 gradient magnitude:
Iteration:
                                     21152.04996494705
            23 gradient magnitude:
Iteration:
                                     21119.79882887556
Iteration:
            24 gradient magnitude:
                                     21087.59686703451
            25 gradient magnitude:
Iteration:
                                     21055.444004446545
            26 gradient magnitude:
Iteration:
                                     21023.340166248632
Iteration:
               gradient magnitude:
                                     20991.28527769187
            28 gradient magnitude:
                                     20959.279264141343
Iteration:
Iteration:
               gradient magnitude:
                                     20927.32205107593
Iteration:
            30 gradient magnitude:
                                     20895.41356408814
Iteration:
            31 gradient magnitude:
                                     20863.55372888393
            32 gradient magnitude:
Iteration:
                                     20831.74247128252
            33 gradient magnitude:
Iteration:
                                     20799.979717216258
Iteration:
            34 gradient magnitude:
                                     20768.26539273041
Iteration:
            35 gradient magnitude:
                                     20736.599423983014
            36 gradient magnitude:
Iteration:
                                     20704.98173724469
Iteration:
            37 gradient magnitude:
                                     20673.412258898476
            38 gradient magnitude:
Iteration:
                                     20641.89091543965
Iteration:
               gradient magnitude:
                                     20610.417633475583
            40 gradient magnitude:
Iteration:
                                     20578.99233972552
            41 gradient magnitude:
Iteration:
                                     20547.614961020485
            42 gradient magnitude:
Iteration:
                                     20516.28542430301
Iteration:
            43 gradient magnitude:
                                     20485.003656627076
Iteration:
            44 gradient magnitude:
                                     20453.769585157825
            45 gradient magnitude:
Iteration:
                                     20422.583137171507
            46 gradient magnitude:
Iteration:
                                     20391.44424005523
Iteration:
            47 gradient magnitude:
                                     20360.352821306817
            48 gradient magnitude:
Iteration:
                                     20329.308808534643
Iteration:
               gradient magnitude:
                                     20298.312129457467
            50 gradient magnitude:
Iteration:
                                     20267.36271190423
            51 gradient magnitude:
Iteration:
                                     20236.46048381396
            52 gradient magnitude:
Iteration:
                                     20205.605373235514
                                     20174.797308327477
Iteration:
            53 gradient magnitude:
Iteration:
            54 gradient magnitude:
                                     20144.036217357967
Iteration:
            55 gradient magnitude:
                                     20113.322028704482
            56 gradient magnitude:
Iteration:
                                     20082.654670853706
            57 gradient magnitude:
Iteration:
                                     20052.03407240138
```

```
Iteration:
            58 gradient magnitude:
                                     20021.46016205212
Iteration:
            59 gradient magnitude:
                                     19990.932868619224
Iteration:
            60 gradient magnitude:
                                     19960.45212102455
            61 gradient magnitude:
                                     19930.017848298332
Iteration:
            62 gradient magnitude:
                                     19899.629979579015
Iteration:
            63 gradient magnitude:
Iteration:
                                     19869.28844411305
Iteration:
            64 gradient magnitude:
                                     19838.993171254828
Iteration:
            65 gradient magnitude:
                                     19808.744090466414
            66 gradient magnitude:
Iteration:
                                     19778.54113131744
Iteration:
            67 gradient magnitude:
                                     19748.38422348492
            68 gradient magnitude:
Iteration:
                                     19718.273296753105
               gradient magnitude:
Iteration:
            69
                                     19688.208281013285
Iteration:
            70 gradient magnitude:
                                     19658.189106263653
Iteration:
            71 gradient magnitude:
                                     19628.215702609137
Iteration:
            72 gradient magnitude:
                                     19598.28800026125
            73 gradient magnitude:
Iteration:
                                     19568.405929537894
Iteration:
            74 gradient magnitude:
                                     19538.56942086323
Iteration:
            75 gradient magnitude:
                                     19508.7784047675
            76 gradient magnitude:
Iteration:
                                     19479.03281188685
Iteration:
            77 gradient magnitude:
                                     19449.33257296321
Iteration:
            78 gradient magnitude:
                                     19419.677618844125
               gradient magnitude:
Iteration:
            79
                                     19390.06788048253
            80 gradient magnitude:
Iteration:
                                     19360.503288936685
Iteration:
            81 gradient magnitude:
                                     19330.98377536993
Iteration:
            82 gradient_magnitude:
                                     19301.50927105059
Iteration:
            83 gradient magnitude:
                                     19272.07970735178
Iteration:
            84 gradient magnitude:
                                     19242.69501575124
            85 gradient magnitude:
Iteration:
                                     19213.355127831197
Iteration:
            86 gradient magnitude:
                                     19184.059975278215
Iteration:
            87 gradient magnitude:
                                     19154.809489882977
               gradient_magnitude:
Iteration:
                                     19125.603603540203
Iteration:
               gradient magnitude:
                                     19096.442248248433
            90 gradient magnitude:
Iteration:
                                     19067.325356109905
            91 gradient magnitude:
Iteration:
                                     19038.25285933037
Iteration:
            92 gradient magnitude:
                                     19009.224690218958
Iteration:
            93 gradient magnitude:
                                     18980.240781187997
            94 gradient magnitude:
                                     18951.301064752897
Iteration:
Iteration:
            95 gradient magnitude:
                                     18922.405473531922
Iteration:
            96 gradient magnitude:
                                     18893.55394024609
Iteration:
            97 gradient magnitude:
                                     18864.74639771902
            98 gradient_magnitude:
Iteration:
                                     18835.982778876747
Iteration:
            99 gradient magnitude:
                                     18807.26301674756
Iteration:
            100 gradient magnitude:
                                      18778.58704446187
Iteration:
            101 gradient magnitude:
                                      18749.954795252062
Iteration:
            102 gradient magnitude:
                                      18721.366202452296
Iteration:
            103 gradient magnitude:
                                      18692.821199498412
Iteration:
            104 gradient_magnitude:
                                      18664.319719927713
Iteration:
            105 gradient magnitude:
                                      18635.861697378845
                                      18607.447065591656
Iteration:
            106 gradient magnitude:
Iteration:
                gradient magnitude:
                                      18579.075758407
Iteration:
            108 gradient magnitude:
                                      18550.747709766623
Iteration:
            109 gradient magnitude:
                                      18522.462853712976
            110 gradient magnitude:
Iteration:
                                      18494.221124389092
            111 gradient magnitude:
Iteration:
                                      18466.022456038405
Iteration:
            112 gradient magnitude:
                                      18437.866783004625
Iteration:
            113 gradient magnitude:
                                      18409.754039731557
Iteration:
            114 gradient magnitude:
                                      18381.684160762976
```

```
Iteration:
            115 gradient magnitude:
                                      18353.657080742418
Iteration:
            116 gradient magnitude:
                                      18325.672734413143
Iteration:
            117 gradient_magnitude:
                                      18297.731056617842
Iteration:
            118 gradient magnitude:
                                      18269.83198229859
            119 gradient magnitude:
Iteration:
                                      18241.975446496646
            120 gradient_magnitude:
Iteration:
                                      18214.161384352312
Iteration:
            121 gradient magnitude:
                                      18186.389731104784
Iteration:
            122 gradient magnitude:
                                      18158.660422092016
Iteration:
            123 gradient magnitude:
                                      18130.97339275053
Iteration:
            124 gradient magnitude:
                                      18103.328578615303
            125 gradient magnitude:
Iteration:
                                      18075.7259153196
            126 gradient magnitude:
                                      18048.165338594838
Iteration:
Iteration:
            127 gradient magnitude:
                                      18020.646784270404
Iteration:
            128 gradient magnitude:
                                      17993.17018827355
Iteration:
            129 gradient magnitude:
                                      17965.735486629208
            130 gradient magnitude:
Iteration:
                                      17938.34261545987
Iteration:
            131 gradient magnitude:
                                      17910.991510985405
            132 gradient magnitude:
Iteration:
                                      17883.68210952294
Iteration:
            133 gradient magnitude:
                                      17856.414347486694
Iteration:
            134 gradient magnitude:
                                      17829.18816138785
Iteration:
            135 gradient magnitude:
                                      17802.00348783438
Iteration:
            136 gradient magnitude:
                                      17774.86026353092
Iteration:
            137 gradient magnitude:
                                      17747.75842527861
Iteration:
            138 gradient magnitude:
                                      17720.69790997496
Iteration:
            139 gradient_magnitude:
                                      17693.678654613675
Iteration:
            140 gradient magnitude:
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localhost:8888/nbconvert/html/Assignment2_linear_regression.ipynb?download=false

Task P6: Use the learned weights to predict 'Peopletested' in the last three weeks in the dataset. Copy the predictions to the solution file, and calculate the test error $\$ \frac{1}{n{\mathbf{tst}}}\sum{i=1}^{n}{\mathbf{tst}}} (y_i^{\mathbf{tst}})^2, \ where $\$ \frac{1}{n{\mathbf{tst}}}\sum{i=1}^{n}{\mathbf{tst}}} i^{\mathbf{tst}}} i^{\mathbf{tst}} is the predicted label.

```
In [118]: | # Create the test data
          test data = data.iloc[-21:]
          (test simple feature matrix, test output) = get numpy data(test data, simple f
          eatures, my_output)
          test predictions = predict output(test simple feature matrix, final weights)
          print (test predictions)
          [102.59928352 103.14553658 103.69178964 104.2380427 104.78429576
           105.33054882 105.87680188 106.42305494 106.969308
                                                                107.51556106
           108.06181412 108.60806718 109.15432024 109.7005733 110.24682637
           110.79307943 111.33933249 111.88558555 112.43183861 112.97809167
           113.52434473]
          <ipython-input-107-8b2916f5b454>:9: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame.
          Try using .loc[row indexer,col indexer] = value instead
          See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s
          table/user guide/indexing.html#returning-a-view-versus-a-copy
            data frame['constant'] = 1 # here we are adding a constant column
In [121]: # Calculate the test error
          # STUDENT: Start of code
          test error = (1/len(test predictions))* np.sum(np.square( np.subtract(test dat
          a.People tested, test predictions)))
          print(test error)
          #end of code
```

11490.99138344532

Linear regression using multiple features

Here, we will be considering multiple input features (Intensive_care, New_positive_cases, Days) to predict the People tested in the future.

Task P7: Specify the initial_weights, step_size, and tolerance for the function regression_gradient_descent . Print the outputs of the code.

```
model features = ['Intensive care', 'New positive cases', 'Days']
In [127]:
          my_output = 'People_tested'
          #call the get nupy data method to calculate the feature matrix and output. Sto
          re them in the variables "multi feature matrix" & "output"
          (multi_feature_matrix, output) = get_numpy_data(data, model_features, my_outpu
          t)
          # Initialize the weights, step size and tolerance
          # STUDENT: Start of code
          # STUDENT: Specify the initial_weights, step_size, and tolerance
          initial_weights = [ 1.0, 1.0, 1.0, 1.0 ]
          step size = 7e-8
          tolerance = 1.2e4
          # end of code
          weight_2 = regression_gradient_descent(multi_feature_matrix, output, initial_w
          eights, step_size, tolerance)
          print ("Here are the final weights after convergence:")
          print (weight 2)
```

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Iteration:
            1 gradient magnitude:
                                    28577.826184864996
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            2 gradient magnitude:
                                    28520.816323869836
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Here are the final weights after convergence:
[0.99572553 0.99950436 0.99941219 0.41591547]
```

Task P8: Use the learned weights to predict 'People_tested' in the last three weeks in the dataset. Find the value of the model predictions on the 10th day of the forecasting period. Also print the actual number of people tested on that particular day. Copy the predictions to the solution file, and calculate the test error. Note: here we are asking you to report the number before normalization. So you need to convert the prediction back to the unit of people.

```
In [128]:
          (test feature matrix, test output) = get numpy data(test data, model features,
          my output)
          test predictions 2 = predict output(test feature matrix, weight 2)
          #Prediction for the 10th day of the forecasting period.
          print (test predictions 2[10])
          #Convert the normalized data back to original figures using the same min-max n
          ormalization
          prediction_10th_day = test_predictions_2[10] * (data_orig['People_tested'].max
          () - data_orig['People_tested'].min()) + data_orig['People_tested'].min()
          print ("Model prediction of the 10th day:",int(prediction 10th day))
          # Get the actual number of people tested from our test data on 10 th day of fo
          recasting period.
          actual_people_tested = data_orig["People_tested"].iloc[190]
          print ("Actual number of people tested on the 10th day:",actual people tested)
          82.72965202021162
          Model prediction of the 10th day: 830628809
          Actual number of people tested on the 10th day: 8725909
          <ipython-input-107-8b2916f5b454>:9: SettingWithCopyWarning:
          A value is trying to be set on a copy of a slice from a DataFrame.
          Try using .loc[row_indexer,col_indexer] = value instead
          See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/s
          table/user guide/indexing.html#returning-a-view-versus-a-copy
            data frame['constant'] = 1 # here we are adding a constant column
In [129]: # Calculate the test error
          # STUDENT: Start of code
          test error = (1/len(test predictions 2))* np.sum(np.square( np.subtract(test d
          ata.People_tested, test_predictions_2)))
          print(test error)
          # end of code
```

6701.844086335183

Explore on your own

Now that you have tried two models for predictions, in this section, you can explore on your own an aspect of the problem that interests you. Here are some examples:

- What features or what combination of features are most predictive for 'People tested'?
- · How does tolerance for convergence affect prediction errors?
- · How does step size affect prediction errors?
- How can we use validation to select the set of features to improve prediction?

Report your question of investigation, as well as your results/interpretation in the solution file.

```
In [138]: def regression gradient descent 2(feature matrix, labels, initial weights, ste
          p size, tolerance):
              # Gradient descent algorithm for linear regression problem
              # Input:
              # feature matrix: numpy array of size n by d, where n is the number of dat
          a points, and d is the feature dimension
              # labels: true labels y, a numpy vector of dimension d
              # initial_weights: initial weight vector to start with, a numpy vector of
              # step size: step size of update
              # tolerance: tolerace epsilon for stopping condition
              # Output:
              # Weights obtained after convergence
              converged = False
              weights = np.array(initial weights) # current iterate
              while not converged:
                  i += 1
                  # STUDENT: Start of code: your impelementation of what the gradient de
          scent algorithm does in every iteration
                  # Refer back to the update rule listed above: update the weight
                  derivative = weight derivative(weights, feature matrix, labels);
                  weights = weights - step_size*derivative
                  # Compute the gradient magnitude:
                  gradient magnitude = np.linalg.norm( weight derivative(weights, featur
          e matrix, labels) )
                  # Check the stopping condition to decide whether you want to stop the
           iterations
                   if gradient magnitude <= tolerance: # STUDENT: check the stopping c</pre>
          ondition here
                      converged = True
                  # End of code
                  #print ("Iteration: ",i,"gradient magnitude: ", gradient magnitude) #
           for us to check about convergence
              return(weights)
```

```
In [141]: # Explore an aspect of the model that interests you
          ### STUDENT: Start of code
          simple features = ['Days']
          my output = 'People tested'
          # Use get numpy data method to calculate the feature matrix and output.
          (simple feature matrix, output) = get numpy data(train data, simple features,
          my output)
          #Initialize the weights, step size and tolerance
          # Start of code
          #STUDENT: Specify the initial_weights, step_size, and tolerance
          initial weights = [1.0, 1.0]
          step size = 7e-8
          tolerance 1 = 0.8e4
          tolerance 2 = 1.0e4
          tolerance_3 = 1.2e4
          tolerance 4 = 1.4e4
          # end of code
          # Use the regression gradient descent function to calculate the gradient decen
          t and store it in the variable 'final_weights'
          final_weights_1 = regression_gradient_descent_2(simple feature matrix, output,
          initial weights, step size, tolerance 1)
          final weights 2 = regression gradient descent 2(simple feature matrix, output,
          initial_weights, step_size, tolerance_2)
          final weights 3 = regression gradient descent 2(simple feature matrix, output,
          initial weights, step size, tolerance 3)
          final weights 4 = regression gradient descent 2(simple feature matrix, output,
          initial_weights, step_size, tolerance_4)
          # end of code
          #print ("Here are the final weights after convergence with tolerance of 0.8e
          4:")
          #print (final weights 1)
          #print ("Here are the final weights after convergence with tolerance of 1.0e
          4:")
          #print (final weights 2)
          #print ("Here are the final weights after convergence with tolerance of 1.2e
          4:")
          #print (final weights 3)
          #print ("Here are the final weights after convergence with tolerance of 1.4e
          4:")
          #print (final_weights_4)
          test data = data.iloc[-21:]
          (test simple feature matrix, test output) = get numpy data(test data, simple f
          eatures, my output)
          test predictions 1 = predict output(test simple feature matrix, final weights
          1)
          test_predictions_2 = predict_output(test_simple_feature_matrix, final_weights_
          2)
          test predictions 3 = predict output(test simple feature matrix, final weights
          3)
```

```
test predictions 4 = predict output(test simple feature matrix, final weights
4)
test error = (1/len(test predictions 1))* np.sum(np.square( np.subtract(test d
ata.People tested, test predictions 1)))
print("Error for tolerance of 0.8e4:", test_error)
test error = (1/len(test predictions 2))* np.sum(np.square( np.subtract(test d
ata.People tested, test predictions 2)))
print("Error for tolerance of 1.0e4:", test_error)
test error = (1/len(test predictions 3))* np.sum(np.square( np.subtract(test d
ata.People tested, test predictions 3)))
print( "Error for tolerance of 1.2e4:", test_error)
test error = (1/len(test predictions 4))* np.sum(np.square( np.subtract(test d
ata.People_tested, test_predictions_4)))
print("Error for tolerance of 1.4e4:", test error)
#end of code
### End of code
```

```
Error for tolerance of 0.8e4: 5067.1321292306275
Error for tolerance of 1.0e4: 7969.441808885362
Error for tolerance of 1.2e4: 11490.99138344532
Error for tolerance of 1.4e4: 15670.65863747879

<ipython-input-107-8b2916f5b454>:9: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy data_frame['constant'] = 1 # here we are adding a constant column
```

My question of investigation was how does tolerance for convergence affect the test error. I used 4 different tolerance values ranging from 0.8e4 to 1.4e4. As I increased the tolerance value, the error also increased as well. It seems like the smaller the tolerance I give it, the less room for error there is for the model to make.