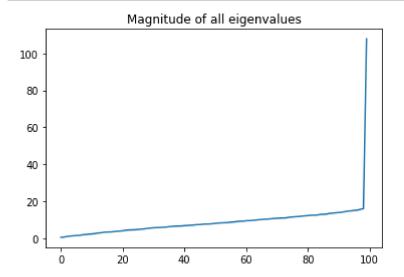
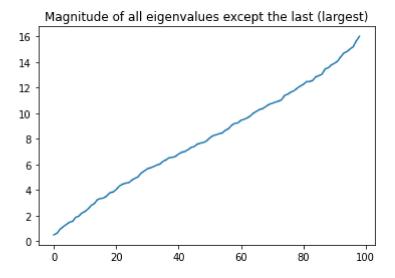
```
In [17]:
         import numpy as np
         import scipy
         import matplotlib.pyplot as plt
         import math
         print("Hello World")
         Hello World
In [18]: #Generating a random matrix
         M = np.random.rand(100,100)
         #Making it symmetric
         A = M + M.T
In [19]: #Finding the eigenvalues and eigenvectors for the matrix and sorting them in a
         eig = np.linalg.eigvals(A)
         idx = np.argsort(eig)
         eig = eig[idx]
In [20]: |print(eig[0:10])
         #Thus, the smallest eigenvalue is roughly - 8
         [-7.49050678 -7.37120721 -7.055183
                                              -6.85956998 -6.68348424 -6.51323141
          -6.44438292 -6.12730853 -6.0315539 -5.78679985]
In [21]: B = A + 8*np.eye(100,100)#Making a new, symmetric, positive eigenvalue matrix
In [22]: eigB = np.linalg.eigvals(B)
         idxB = np.argsort(eigB)
         eigB = eigB[idxB]
In [23]: #Thus, B has only positive eigenvalues and is symmetric
         print(eigB[0:10])
         print(eigB[90:])
         [0.50949322 0.62879279 0.944817
                                           1.14043002 1.31651576 1.48676859
          1.55561708 1.87269147 1.9684461 2.21320015]
                                     14.41607486 14.71230028 14.83131923
         [ 13.91333665 14.0829986
           15.03304404 15.20856989 15.66580165 16.01998977 107.9233933 ]
```

```
In [24]: #There's a sudden spike In Eigenvalues, plotting it here
plt.plot(eigB)
plt.title("Magnitude of all eigenvalues")
plt.show()
```



```
In [25]: #Plotting without the last eigenvalue
    plt.plot(eigB[0:99])
    plt.title("Magnitude of all eigenvalues except the last (largest)")
    plt.show()
```



```
In [26]: #Picking sigma to be uniformly distributed between 3 and 11
sigma = np.linspace(3,10,8)
print(sigma)
```

[3. 4. 5. 6. 7. 8. 9. 10.]

```
In [27]: #Generate the random vector
         s = np.random.normal(0, 1, 100)
         norm=math.sqrt(sum(s*s))
         S=s/norm
In [28]: #Creating Vi
         V = np.zeros((8,100))
         for i in range(8):
             V[i] = np.dot(np.linalg.inv(sigma[i]*np.eye(100) - B),S)
In [29]: #Creating the H and S matrices
         Hmatrix = np.ones((8,8))
         Smatrix = np.ones((8,8))
         for i in range(8):
             for j in range(8):
                 Hmatrix[i,j]= np.dot(V[i],np.dot(B,V[j]))
                 Smatrix[i,j] = np.dot(V[i],V[j])
         print(Hmatrix.shape)
         print(Smatrix.shape)
         (8, 8)
         (8, 8)
In [30]: eigvals, eigvecs = scipy.linalg.eigh(Hmatrix, b = Smatrix,eigvals_only=False)
In [31]: print("Eigenvalues from Filter Diagonalization: ", eigvals)
         print("Sigma values: ", sigma)
         Eigenvalues from Filter Diagonalization: [2.97397875 4.02097334 5.01184986
         6.0251865 6.99846498 8.08589617
          9.04534406 9.98174954]
         Sigma values: [ 3. 4. 5. 6. 7. 8. 9. 10.]
```

In [32]: print(eigB)
#List of all eigenvalues of the matrix

```
0.50949322
              0.62879279
                           0.944817
                                        1.14043002
                                                      1.31651576
 1.48676859
              1.55561708
                           1.87269147
                                        1.9684461
                                                      2.21320015
 2.33409813
              2.53729997
                           2.80837679
                                        2.95680091
                                                      3.25922472
 3.34021554
              3.38816053
                           3.55410613
                                        3.79996364
                                                      3.85173042
              4.31738901
                                        4.54294836
 4.04265721
                           4.46000193
                                                      4.57750192
4.77780494
              4.92057056
                           5.04191205
                                        5.33257578
                                                      5.49543185
 5.6716975
              5.747406
                           5.83842234
                                        5.96140596
                                                      6.0308938
 6.25402174
              6.37036122
                           6.536283
                                        6.56589273
                                                      6.6365373
 6.82406591
              6.95615608
                           7.02125754
                                        7.15593838
                                                      7.3391916
 7.40284814
              7.5954636
                           7.67454812
                                        7.72132611
                                                      7.85169538
 8.07765773
              8.2473632
                           8.32325089
                                        8.40800869
                                                      8.46606169
 8.67389203
              8.79739554
                           9.06372457
                                        9.21512087
                                                     9.25287449
 9.45537939
              9.52792909
                           9.64384484
                                        9.79478391
                                                     10.01711295
10.15687417
             10.30662349
                          10.37262679
                                       10.54202563
                                                     10.69756441
10.78239719
            10.87955119
                          10.95198356
                                       11.05817202
                                                     11.39881848
11.50375334
             11.65804613
                          11.76828614
                                       11.96212877
                                                     12.13119319
12.26519907
             12.47978768
                          12.48468629
                                       12.57611058
                                                     12.85830105
12.94486207
             13.06702238
                          13.46783201
                                       13.5627263
                                                     13.78665256
13.91333665
             14.0829986
                          14.41607486
                                       14.71230028
                                                     14.83131923
15.03304404
             15.20856989
                          15.66580165
                                       16.01998977 107.9233933 ]
```

Conclusions

Nature of Convergence

• One of the sigma's was 7, and the closest eigenvalue was 7.02, but the algorithm seemed to converge towards the next nearest - 6.956. Why?

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