

Erasmus Mundus JMD Nuclear Physics

Computational and Numerical Physics Exercises

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First of all, Import all Packages for use and Future Use

```
In [1]: %matplotlib inline
import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
import scipy
import math
import random
np.warnings.filterwarnings('ignore')
from mpl_toolkits import mplot3d
```

Exercise 1 ¶

```
In [2]: '''
This portion of code reads the file of the weather for future use and will show the first 5 rows.
'''

df = pd.read_csv('T_Agrinio_EM.csv', header = 0, index_col = 0)
df.head()

#to ensure all the data points were read
#df.head(136)
```

Out[2]:

	t1	t2	t3	t4	t5	t6	t7	t8	t9	t10	t11	t12
year												
1961	8.56	9.18	10.86	14.62	18.13	21.98	25.97	25.33	21.75	17.61	13.15	10.70
1962	8.48	8.94	10.21	13.93	17.34	22.34	26.51	26.01	21.45	17.65	13.10	9.33
1963	9.30	9.85	10.75	13.45	17.71	22.24	25.52	25.44	22.35	16.57	13.41	8.56
1964	8.67	8.42	10.49	14.27	17.52	21.87	25.25	25.19	22.32	17.99	13.52	9.03
1965	7.59	8.05	10.68	13.45	17.64	22.75	25.90	25.53	22.26	16.99	13.31	10.42

Exercise 1.1: Determine the Minimum and Maximum Temperature for the Month of March for this Given Set

```
In [3]: '''
First, to create a list from the table above of all the values in march, which is given by t3.
By using Pandas the values in the table are z values which would be organized in a 1x136 array,
So will have to append those values to a list to easily acquire the maximum value.
'''

z = df[['t3']][:136].values
List = []
for n in range(136):
    List.append(float((z[n])))
    n = n+1

print('The maximum March temperature from the dataset is:',np.max(List),'\n' 'The minimum March te
mperature from the dataset is: ',np.min(List))
print('The average March Temperature from the dataset is:', '%.2f'%np.average(List))

The maximum March temperature from the dataset is: 14.44
The minimum March temperature from the dataset is:  9.95
The average March Temperature from the dataset is: 11.90
```

Exercise 1.2: Check the info on the numpy mean function and compute the average monthly temperatures and the average annual temperatures for the given dataset (Hint: axis option)

```
In [4]: '''
To save Scrolling, I have # the line of code to show the info on the mean function
to check it. simply un-# it.
'''

#np.info(np.mean)

Out[4]: '\nTo save Scrolling, I have # the line of code to show the info on the mean function\nto check i
t. simply un-# it.\n'
```

```
In [47]: '''
This calculates the average temperature of each year by taking the array of all values,
which are separated by each year and calculates the mean of the column which is equal to
the average temperature over that year...
'''

np.array(df)
z = np.array(df)
AnnualAverage = []
for n in range(136):
    l = '%.2f' %(np.average(z[n]))
    AnnualAverage.append(float(l))
    n = n+1

print(AnnualAverage)

'''
Expanding on the above code by transposing the array we acquire an array with lists of the monthly
average temperatures

'''

TransposedZ = z.T #this transposes the array
MonthlyAverage = []

for n in range(12):
    l = '%.2f'%(np.mean(TransposedZ[n]))
    MonthlyAverage.append(l)
    n= n+1

'''
We see that the average March temperature is the same as 1.1 so we know these values are correct.
'''

print(MonthlyAverage)
print('The average March Temperature from the dataset is:',MonthlyAverage[2])

[16.49, 16.27, 16.26, 16.21, 16.21, 15.95, 16.53, 16.25, 16.2, 16.24, 15.7, 15.96, 16.39, 16.09,
16.25, 16.34, 16.03, 16.41, 16.25, 16.5, 16.23, 15.71, 16.19, 16.01, 16.19, 16.25, 16.37, 16.18,
16.44, 16.53, 16.52, 16.93, 17.12, 16.36, 16.8, 16.64, 16.69, 16.99, 16.79, 16.83, 16.7, 17.02, 1
6.8, 16.91, 16.81, 16.52, 16.79, 16.88, 17.36, 17.01, 16.76, 16.93, 17.26, 16.84, 16.63, 17.39, 1
7.3, 17.06, 17.34, 16.75, 16.77, 17.18, 16.98, 17.29, 17.42, 17.0, 17.18, 17.31, 17.3, 17.52, 17.
77, 17.87, 17.91, 17.78, 18.02, 17.71, 18.06, 17.62, 17.84, 17.93, 17.81, 17.64, 18.33, 18.04, 1
7.72, 18.48, 18.18, 18.43, 18.51, 18.49, 18.09, 18.55, 18.71, 18.55, 18.74, 18.83, 18.71, 18.43,
18.55, 18.83, 18.52, 18.71, 18.77, 18.71, 18.85, 19.3, 18.88, 18.98, 18.87, 19.02, 18.94, 19.19,
18.88, 18.77, 19.12, 19.33, 19.26, 19.42, 19.47, 19.94, 19.64, 19.62, 19.45, 19.44, 19.27, 19.68,
19.36, 19.94, 20.02, 20.09, 19.95, 19.58, 19.72, 19.99, 19.97, 19.89]
['9.58', '9.94', '11.90', '14.89', '19.22', '24.04', '27.64', '27.63', '23.93', '18.84', '14.18',
'10.70']
The average March Temperature from the dataset is: 11.90
```

Exercise 1.3: Get the maximum and minimum monthly temperatures for year 2000 and at what months did these temperatures occur

```
In [6]: '''
By looking at the list, it's easy to see that the year 2000 will occur at line 39,
... craig remember to create a loop to locate this line instead though
'''

Y2000List = []
p = df.head(136)[39:40].values
print(p)
print('The lowest Temperature for the year 2000 is:',np.min(p))
print('The maximum Temperature for the year 2000 is:', np.max(p))

[[ 9.18 10.56 11.81 14.32 17.55 22.93 25.56 25.73 21.72 17.47 14.54 10.54]]
The lowest Temperature for the year 2000 is: 9.18
The maximum Temperature for the year 2000 is: 25.73
```

Exercise 2

Exercise 2.1: Plot the monthly and annual difference between max and min temperatures as a function of the month (1-12) and the year (1961-2096), respectively

```

In [7]: '''
We first have to evaluate the maximum and minimum temperatures of each month and each year.
This expands on the work done in exercise 1.3 which evaluated the maximum and minimum of a single
year,
we'll want to create a loop that will determine the maximum and minimum, take the difference and add
that difference to a list. For each month and each year
'''

'''
This block of code sets up the empty lists and bounds for the graphs
'''
x = []
f = []
for n in range(1,13):
    x.append(n)
    n=n+1

for n in range(1961,2097):
    f.append(n)
    n=n+1

Y_diff = []
M_diff = []

TransposedZ = z.T

for n in range(136):
    '''
    This loop determines the difference between the lowest and highest temperature of each
    year and appends them to a list to be used for the annual difference as a function of the year
    '''
    minn = np.min(z[n])
    maxx = np.max(z[n])

    Diff = maxx - minn
    Y_diff.append(float('%0.2f'%Diff))

for n in range(12):
    '''
    This loop determines the difference between the lowest and highest temperature of each month
    over the 136 years and appends them to a list to be used in the monthly differences as a function
    of the month
    '''
    mminn = np.min(TransposedZ[n])
    mmaxx = np.max(TransposedZ[n])

    Dif = mmaxx - mminn
    M_diff.append(float('%0.2f'%Dif))

'''
Plotting Code, first chunk is the monthly average and second chunk is yearly average
'''

fig, axarr = plt.subplots(nrows = 1, ncols = 2, figsize = (16,8))

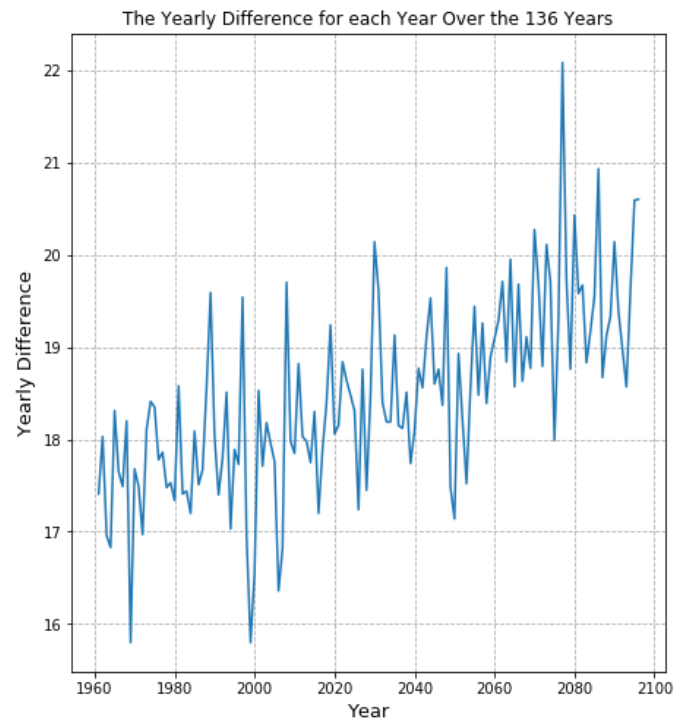
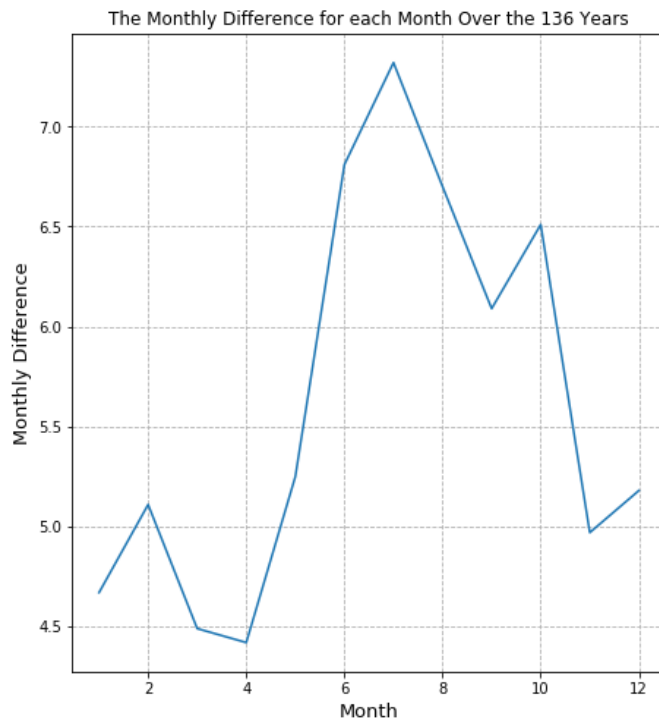
axarr[0].plot(x,M_diff)
axarr[0].grid(linestyle = 'dashed')
axarr[0].set_xlabel('Month', fontsize=13)
axarr[0].set_ylabel('Monthly Difference', fontsize= 13)
axarr[0].set_title('The Monthly Difference for each Month Over the 136 Years')

#axarr[1].set_yticks(range(11),['10','11','12','13','14','15','16','17','18','19','20'])

```

```
#axarr[1].set_ylim(10,30)
axarr[1].grid(linestyle='dashed')
axarr[1].set_xlabel('Year', fontsize=13)
axarr[1].set_ylabel('Yearly Difference', fontsize= 13)
axarr[1].set_title('The Yearly Difference for each Year Over the 136 Years')
axarr[1].plot(f,Y_diff)
```

Out[7]: [



Exercise 2.2: Plot the standard deviation of the monthly and annual temperatures as a function of the month (1-12) and the year (1961-2096), respectively

```
In [8]: '''
We know numpy has a function for standard deviation so we jut have to integrate it into a loop and
append it to a list for graphing
'''
StdForYears = []

for n in range(136):
    '''
    This loop evaluates the standard deviation for each year and appends it to a list
    '''
    StdForYears.append(float('%0.2f'%np.std(z[n])))
    n=n+1
print(StdForYears[:12])

StdForMonths = []
for n in range(0,12):
    '''
    This list evaluates the standard deviation for each month and appends it to a list
    '''
    StdForMonths.append(float('%0.2f'%np.std(TransposedZ[n])))
    n=n+1
print(StdForMonths[:12])

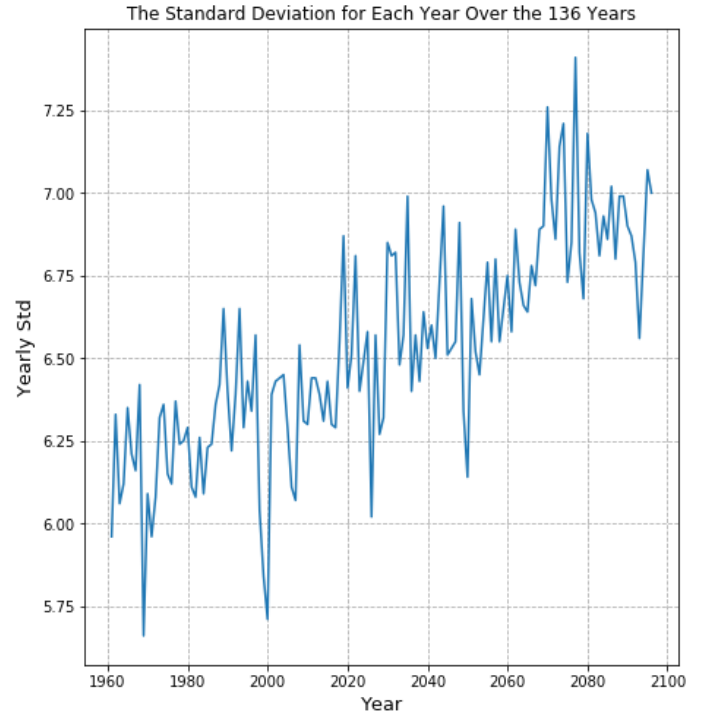
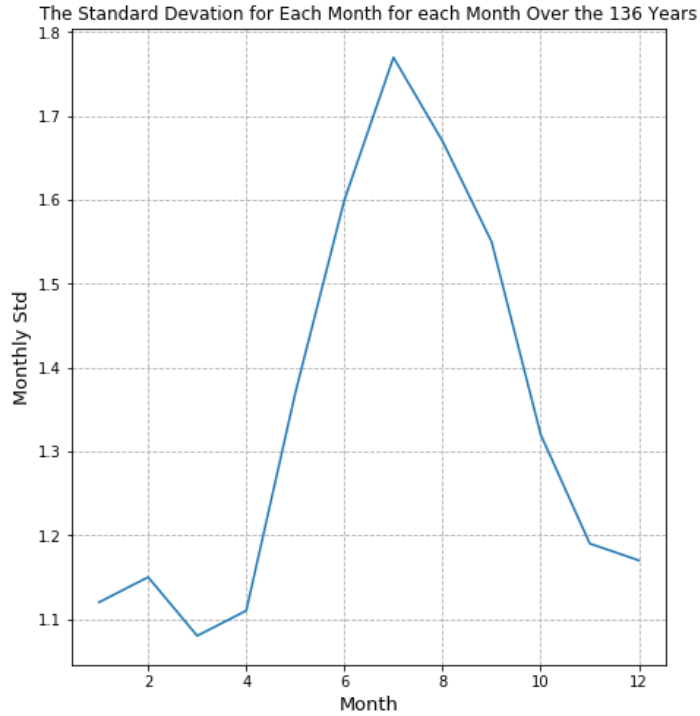
[5.96, 6.33, 6.06, 6.12, 6.35, 6.21, 6.16, 6.42, 5.66, 6.09, 5.96, 6.08]
[1.12, 1.15, 1.08, 1.11, 1.37, 1.6, 1.77, 1.67, 1.55, 1.32, 1.19, 1.17]
```

```
In [9]: fig, axarr = plt.subplots(nrows = 1, ncols = 2, figsize = (16,8))

axarr[0].plot(x,StdForMonths)
axarr[0].grid(linestyle = 'dashed')
axarr[0].set_xlabel('Month', fontsize=13)
axarr[0].set_ylabel('Monthly Std', fontsize= 13)
axarr[0].set_title('The Standard Deviation for Each Month for each Month Over the 136 Years')

axarr[1].grid(linestyle='dashed')
axarr[1].set_xlabel('Year', fontsize=13)
axarr[1].set_ylabel('Yearly Std', fontsize= 13)
axarr[1].set_title('The Standard Deviation for Each Year Over the 136 Years')
axarr[1].plot(f,StdForYears)
```

Out[9]: [



Exercise 2.3: Prepare a plot with six panels (arranged as you wish) which depicts the annual dependence of the average monthly temperature for meteorological Spring (Mar, Apr, May) and Fall (Sep, Nov, Dec) seasons

In [10]:

```
'''  
Alot prettier chunk of code that performs the same tasks as above  
'''
```

```
fig, axarr = plt.subplots(nrows=2, ncols=3, figsize=(16,6), sharex=True, sharey=True)  
fig.tight_layout(h_pad=2, w_pad=2)
```

```
# axis labels outside of loops
```

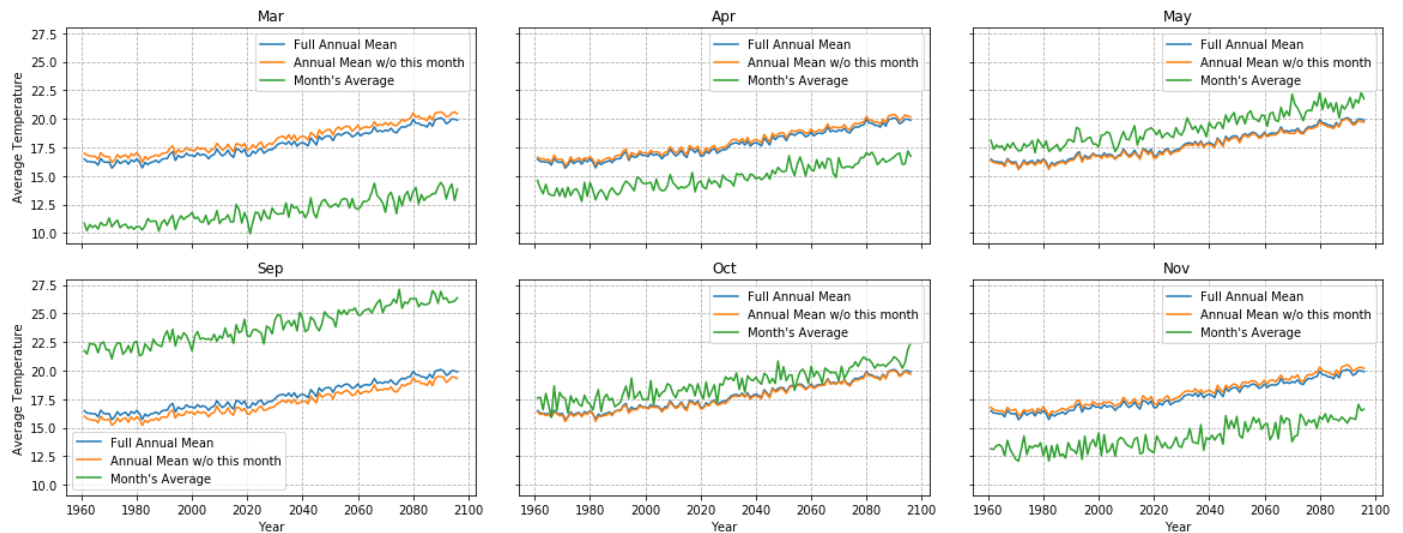
```
axarr[0][0].set_ylabel("Average Temperature")  
axarr[1][0].set_ylabel("Average Temperature")  
axarr[1][0].set_xlabel("Year")  
axarr[1][1].set_xlabel("Year")  
axarr[1][2].set_xlabel("Year")
```

```
# For the Spring Months:
```

```
for m, monthName in enumerate(["Mar", "Apr", "May"]):  
    axarr[0][m].set_title(monthName)  
    axarr[0][m].plot(df.mean(axis=1), label="Full Annual Mean")  
    axarr[0][m].plot(df.drop('t'+str(m+3), axis=1).mean(axis=1), \  
                      label="Annual Mean w/o this month")  
    axarr[0][m].plot(df['t'+str(m+3)], label="Month's Average")  
    axarr[0][m].legend()  
    axarr[0][m].grid(linestyle='dashed')
```

```
# For the Fall Months:
```

```
for m, monthName in enumerate(["Sep", "Oct", "Nov"]):  
    axarr[1][m].set_title(monthName)  
    axarr[1][m].plot(df.mean(axis=1), label="Full Annual Mean")  
    axarr[1][m].plot(df.drop('t'+str(m+9), axis=1).mean(axis=1), \  
                      label="Annual Mean w/o this month")  
    axarr[1][m].plot(df['t'+str(m+9)], label="Month's Average")  
    axarr[1][m].legend()  
    axarr[1][m].grid(linestyle='dashed')
```



Exercise 3

Exercise 3.1: Prepare a loop such that given when it is given as an input a string variable it produces as output another string variable equal to the first string in reversed order, e.g. $a = "abcd"$ and $reversed a = "dcba"$


```
In [11]: def reverse(a):
'''
    I create a reverse function incase it is needed for later..
    this function reads through any string when put in a = ''
    and then will create a new string where it will append the read character
    to the list so its like appending a new value to the start of a list rather
    than the end.
'''
    Reversedata = ''
    for char in a:

        Reversedata = char + Reversedata

    return Reversedata
print(Reverse(a))
```

```
In [12]: a = 'abcd'
reverse(a)
```

```
Out[12]: 'dcba'
```

Exercise 3.2: Use a loop to convert the string "Testing loops and strings..." into another string, changing spaces into "_" (underscore).

```
In [13]: test = 'Testing loops and Strings'
reversedtest = ''
for char in test:
    if char == ' ':
        reversedtest = reversedtest + '_'
    else:
        reversedtest = reversedtest + char

print(reversedtest)

Testing_loops_and_Strings
```

Exercise 3.3: Given the list AA = [1.0, -2.0, 3.0, 5.5, 0.3] and considering that these are the values of the coefficients for a polynomial $P_x = AA[0] + AA[1]x + \dots + AA[n]x^n$, prepare a loop that computes the polynomial value for a given independent variable x value. For example, in the given case $P_x = 7.8$ for $x = 1$

```
In [14]: '''
We Know:
    AA = [1.0,-2.0,3.0,5.5,0.3]
    x = input()
    P_x = AA[0]+AA[1]x+...+AA[n]x**n
'''
def functionPoly(x):
    AA = [1.0,-2.0,3.0,5.5,0.3]
    Plist = []

    for n in range(5):
        P_x = AA[n]*x**n
        Plist.append(P_x)

    #print(Plist) #unhashtag this line to see the sum of each intermediate step, which shows the a
    ffect
                                # of that polynomial term
    print(np.sum(Plist))
```

```
In [15]: functionPoly(1)
```

```
7.8
```

Exercise 3.4: Compute for the loaded temperature dataset the average seasonal temperatures and gives as a results a Python list with these temperatures

```
In [16]: def SeasonalAvgTemp():

    list1 = []
    SeasonalTemp = []
    Months = ['January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', \
              'September', 'October', 'November', 'December']
    for m, monthname in enumerate(Months):
        MonthMean = '%.2f'%np.mean(df['t'+str(m+1)])
        list1.append(float(MonthMean))

    WintAvg = (list1[11]+list1[0]+list1[1])/3
    SeasonalTemp.append(float('%.2f'%WintAvg))
    SpringAvg = (list1[2]+list1[3]+list1[4])/3
    SeasonalTemp.append(float('%.2f'%SpringAvg))
    SummerAvg = (list1[5]+list1[6]+list1[7])/3
    SeasonalTemp.append(float('%.2f'%SummerAvg))
    AutumnAvg = (list1[8]+list1[9]+list1[10])/3
    SeasonalTemp.append(float('%.2f'%AutumnAvg))

    print(SeasonalTemp)
```

```
In [17]: SeasonalAvgTemp()
#this list is Winter, Spring, Summer, August

[10.07, 15.34, 26.44, 18.98]
```

Exercise 3.5: Build a code that, for a given month and considering the loaded temperature dataset, finds the mean and minimum temperature values for the selected month and then it builds a Python list with the years that have an average temperature for the selected month that is less than the average value of the minimum and mean temperature

```
In [18]: def MonthMinMean():

    List123 = [] #list of each months min and mean values
    List321 = [] #List of each months average value of min and mean values
    input1 = str(input())# input
    Months = ['January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', \
              'September', 'October', 'November', 'December']
    AR = [] #list for months that have less than the average value of min and mean temperatures
    Years = list(range(1961,2097))

    for m, MonthName in enumerate (Months):
        '''
        Evaluates the min, mean and the average of the two to append to a list for each month
        '''
        MonthMin = np.min(df['t'+str(m+1)])
        MonthMean = float('%.2f'%np.mean(df['t'+str(m+1)]))
        element = MonthMin, MonthMean
        List123.append(element)

        MinMeanAvg = (MonthMin+MonthMean)/2
        List321.append(MinMeanAvg)

    if str(input1) in Months:
        '''
        Takes the user input for what month they want to evaluate
        '''
        mn = Months.index(str(input1))
        lp = df['t'+str(mn+1)]

        for y,n in enumerate(lp):
            '''
            Appends years where the months are lower than the average of the min and mean
            '''
            if n < List321[mn]:
                AR.append(y+1961)

    print(AR)
```

```
In [19]: MonthMinMean()

June
[1961, 1962, 1963, 1964, 1966, 1967, 1968, 1969, 1970, 1971, 1972, 1974, 1975, 1976, 1979, 1981,
1982, 1983, 1984, 1985, 1986, 2001, 2006, 2020, 2023]
```

Exercise 4

Exercise 4.1: Prepare a file with a function called `fence` such that given two strings as arguments: `string_1 = "aaa"` and `string_2 = "bbb"`, the output is `aaa_bbb_aaa`. In the same file define a second function called `outer` such that given a string returns another string made up of just the first and last characters of its input. Therefore if the input is `Betis` the function output should be `Bs`. Include in both cases a docstring with a brief function description and an example. Load the functions from the file and check what is the output of this statement: `print outer(fence('carbon', '+'))`.

```
In [20]: def fence(arg1, arg2):
'''
    this function takes two arguments, combines the statements into the
    required output and then prints it
'''

fencestatement = arg1+'_'+arg2+'_'+arg1
#print(fencestatement)
return fencestatement

def outer(function, pos1,pos2):
'''
    This function takes the function of fence and prints the first and last letters of the doc string
    produced by fence so for fence(carbon, +) we expect to see 'cn'
'''

out = fence(pos1,pos2)
print(str(out[0])+str(out[-1]))
```

```
In [21]: outer(fence, 'carbon', '+')
```

cn

Exercise 4.2: Gaussian distributed data are frequently normalized to have a mean value equal to zero and a standard deviation equal to one subtracting the actual mean value and dividing by the standard deviation of the dataset. Making use of the mean and std NumPy methods, define a function that takes as an argument a data vector, a new mean value, and a new standard deviation value and transforms the original set of data to a new set with a the new mean as its average value and with a dispersion given by the new standard deviation value. By default the function should standardize the data to mean = 0 and sdev = 1.

The gaussian distribution equation is given as:

$$P(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

Here we know σ is denoting sdev, μ is denoting mean and x is the mean of the data set

```
In [22]: def GuassianDist(datavector,mean=0,sdev=1):

fig, axarr = plt.subplots(nrows=1, ncols=2, figsize=(16,6))
count, bins, ignored = axarr[0].hist(datavector, 100, density=True)
P = 1/(sdev * np.sqrt(2 * np.pi)) * np.exp( - (bins - mean)**2 / (2 * sdev**2) )
axarr[0].plot(bins,P)
axarr[0].grid(linestyle='dashed')
axarr[0].set_title('Original Set of Data with Inputed Mean and Sdev')

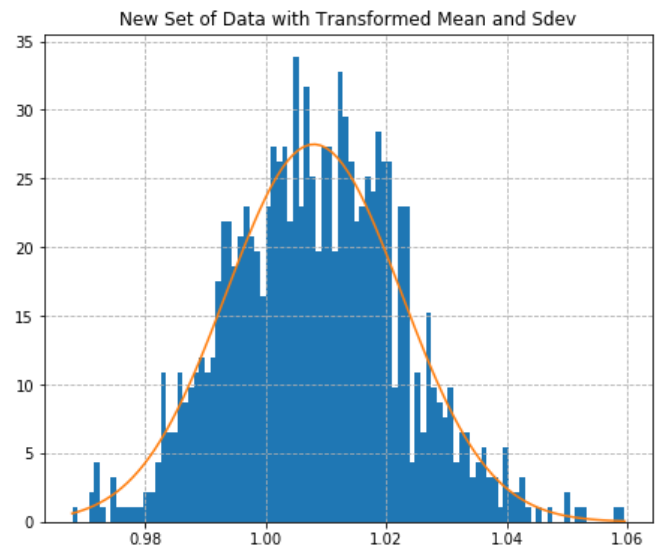
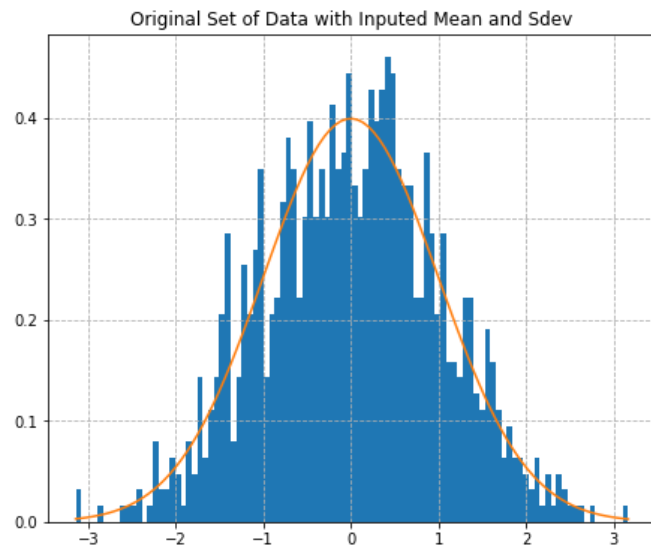
'''
    now to make new data, based off a linear transform; y=mx+b, where x is the random data, Y is the
    new Data, m and b are the adjusting factors
'''

m = np.average(datavector)
b = np.std(datavector)

new_mean = m*mean+b
new_std = np.abs(m*sdev)

newcount, newbins, ignored = axarr[1].hist(np.random.normal(new_mean,new_std,1000),100, density=True)
P1 = 1/(new_std * np.sqrt(2 * np.pi)) * np.exp( - (newbins - new_mean)**2 / (2 * new_std**2) )
axarr[1].plot(newbins,P1)
axarr[1].set_title('New Set of Data with Transformed Mean and Sdev')
axarr[1].grid(linestyle = 'dashed')
```

```
In [23]: GuassianDist(np.random.normal(0,1,1000))
```



Exercise 4.3: Define a function that reads out temperature data from the sample Cyprus dataset and prepare graphics. Prepare a function with helpful docstring and comments that for given list of file names prepares a plot with three columns for each data file: the first including the max, min and mean monthly temperatures, the second the max, min, and mean annual temperatures, and the third depicting the monthly temperatures for all years.

```

In [24]: def Monthtemps(filename): #use T_Agrinio_EM
'''
    This function creates three seperate plots;
        1. Max, Min and Mean of monthly temperautes
        2. Max, min and mean of annual temperatures
        3. all months temps for all years
    it acquires the data for this from the file used in exercies 1 and 2
'''
#read the filename
df = pd.read_csv((filename)+'.csv',header = 0, index_col = 0)
df.head()

Months = ['January','February','March','April','May','June','July','August',\
          'September','October','November','December']

#this block of code creates the array of graphs with the required columns columns
fig, axarr = plt.subplots(nrows=2, ncols=3, figsize=(16,6), sharex=False, sharey=False)
fig.tight_layout(h_pad=1, w_pad=2)
#Once you figure out what is meant by monthly temperatures for all years can change nrows = 3

'''
Setting up parameters for the requested plots
'''
#First row
axarr[0][0].set_ylabel('Temperature')
axarr[0][0].set_xlabel('Month Max')
axarr[0][1].set_xlabel('Month Min')
axarr[0][2].set_xlabel('Month Mean')

#second row
axarr[1][0].set_ylabel('Temperature')
axarr[1][0].set_xlabel('Annual Max')
axarr[1][1].set_xlabel('Annual Min')
axarr[1][2].set_xlabel('Annual Mean')

'''
Loop to Evaluate and plot the requested problem
'''
for m in range(3):
    if m == 0:
        #maximum
        axarr[0][m].plot(np.max(df))
        axarr[1][m].plot(np.max(df.T))
    elif m ==1:
        #minumum
        axarr[0][m].plot(np.min(df))
        axarr[1][m].plot(np.min(df.T))
    else:
        #mean
        axarr[0][m].plot(np.mean(df))
        axarr[1][m].plot(np.mean(df.T))

    axarr[0][m].grid(linestyle='dashed') #grids to make plots look nicer
    axarr[1][m].grid(linestyle='dashed')

'''
    I think this is what is asked for the monthly temperatures for all years? I'm not sure... I do
    n't know if it's
        asking for this or if it's asking for the max, min and mean of month temperatues of all years
        which would work
        in the axarr method as it's asking for 3 more plots so it keeps the 3x3 shape and fills all. w
        hile if I add
            the code below to the axarr loop, it would produce the same plot in [3][0] and leave the other
            2 empty and would
            be very hard to see the graph due to sharing the width with the other two plots... the plot is
            hard to see
            when it takes up a full line
'''

```

```
data = []
data2 = []
```

```
for m in range(1961,2097):
    data.append(df.T[m])

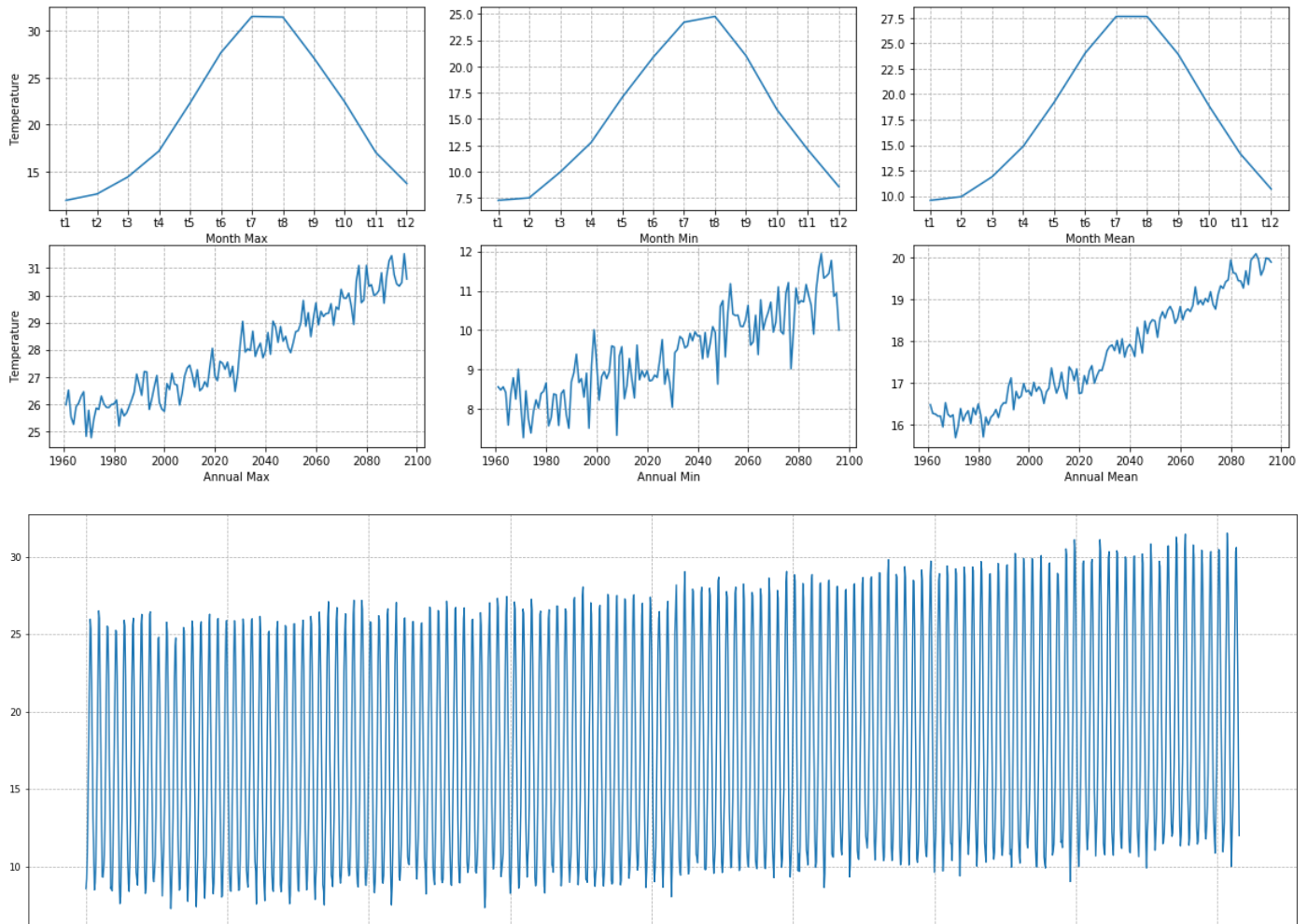
for n in range(136):
    for l in range(12):
        data2.append(data[n][l])
```

```
plt.figure(figsize=(24,8))
plt.plot(data2)
plt.grid(linestyle='dashed')
'''
```

*the x axis is wack for this graph as a result of plotting a list, where the zero position is j anuary 1961, fist position is february 1961, so the x axis is zero to 136*12 not the years*

```
plt.tick_params(axis='x', which='both', bottom=False, top=False, labelbottom=False)
```

In [25]: Monthtemps('T_Agrinio_EM')



Exercise 4.4: Write a function that generates a random password. The password should have a random length of between 10 and 12 random characters from positions 33 to 122 in the ASCII table. Your function will not take any parameters and will return the password as its only result. Make another function that checks if the password has at least two lowercase, two uppercase, and two digit characters and check how many times you need to run the original function to obtain a compliant password.

```
In [26]: def randpass():
    randpass1 = ''

    asc = []
    for n in range(0,int(np.random.random_integers(10,12,1))):
        asc.append(int(np.random.random_integers(33,122,1)))

    for n in asc:
        randpass1 = randpass1+str(chr(n))
    return(randpass1)

def passwordcheck(randpass1,counts=0):
    l,u,d, count = 0,0,0,counts
    lowercase = []
    uppercase = []
    twodigits = []
    for char in randpass1:
        lowercase.append(char.islower())
        uppercase.append(char.isupper())
        twodigits.append(char.isdigit())

    for m in range(0,len(lowercase)):
        if lowercase[m] == True:
            l = l+1
    for m in range(0,len(uppercase)):
        if uppercase[m] == True:
            u = u+1
    for m in range(0,len(twodigits)):
        if twodigits[m] == True:
            d = d+1
    if l&u&d >= 2:
        print(randpass1,'is good and took:',counts,' randomisations until this password is achieved')
    else:
        newpassword = randpass()
        count = count +1
        passwordcheck(newpassword,count)
```

```
In [27]: passwordcheck(randpass())
```

```
77-QkKzSKC,R is good and took: 31 randomisations until this password is achieved
```

Exercise 5

Exercise 5.1: The NIST Digital Library of Mathematical Functions (DLMF) is a very useful site, where you can find an updated and expanded version of the well-known reference *Handbook of Mathematical Functions* of Abramowitz and Stegun. Define a function to compute the Bessel function of the first kind of integer index from the series 10.2.2 in the DLMF, add a docscript and plot the functions of order 0, 1, and 2 in the interval of x between 0 and 10.

The Bessel function in question is:

$$J_v(x) = \left(\frac{1}{2}x\right)^v \sum_{k=0}^{\infty} (-1)^k \frac{\left(\frac{1}{4}x^2\right)^k}{k! \Gamma(v+k+1)}$$

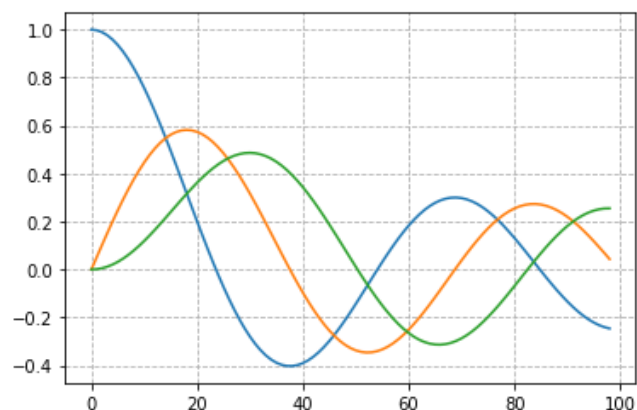

```
In [28]: '''
        want a function to evaluate this function for different v for 0,1 and 2 and x being in the interval of 0 and 10
        '''

def BesselFunction(x,v,k):
    J = 0
    coeff = (0.5*x)**v

    for k in range(k):
        J += (-1)**k*(0.25*x*x)**k/(math.factorial(k)*(math.gamma(v+k+1)))

    return coeff*J
```

```
In [29]: for n in range(3):
        v0 = BesselFunction(x=np.linspace(0,10,99),v=n,k=99)
        plt.plot(v0)
        plt.grid(linestyle = 'dashed')
```



Exercise 5.2: Define and test a function that estimates the value of the special constant π by generating N pairs of random numbers in the interval -1 and 1 and checking how many of the generated number fall into a circumference of radius 1 and centered in the origin. Improve the function showing in a graphical output the square, the circumference and the points inside and outside the circumference with different colors.

```
In [30]: def PiEstimate(nPairs):
        test_points = []
        in_circle = []

        for point in range(nPairs):
            #get test point coords
            x = random.uniform(-1,1)
            y = random.uniform(-1,1)
            test_points.append((x,y))

            if (np.sqrt(x*x+y*y)) <=1:
                in_circle.append((x,y))

        pi = 4*len(in_circle)/nPairs

        return test_points, pi
```

```

In [31]: '''Circle area: pi r^2 = pi
So square area: 4r^2 = 4
Use NM Rule and count how many in circle vs total (in square)
so n_circle/n_square = pi/4 thus 4*n_circle/n_square = pi
'''

points, pi = PiEstimate(9999)
points_plot = list(zip(*points))

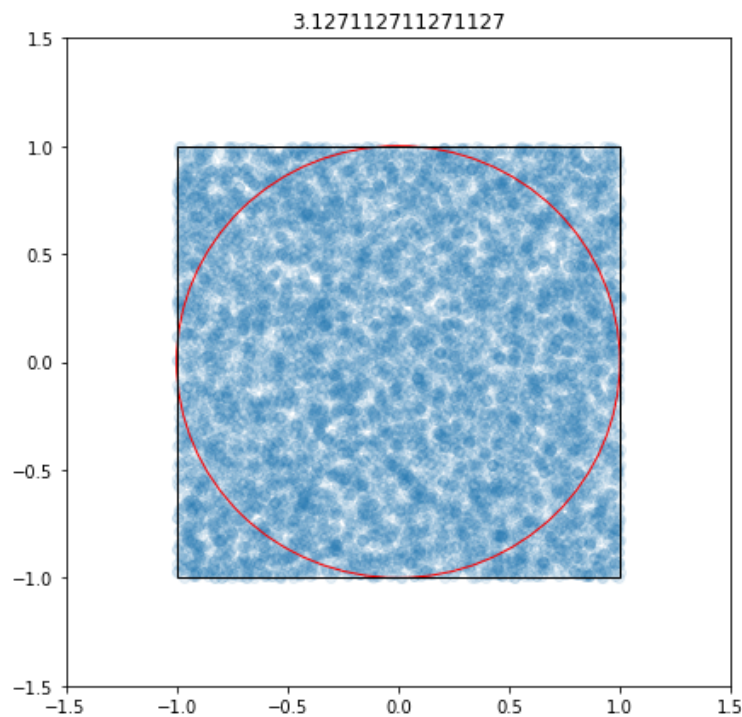
fig = plt.figure(figsize=(7,7))
ax = plt.gca()

plt.scatter(points_plot[0], points_plot[1], alpha = 0.1)
circle = plt.Circle((0,0),radius = 1, fill = False, ec = 'r')
ax.add_artist(circle)
square = plt.Rectangle((-1,-1),2,2,fill = False)
ax.add_artist(square)

ax.set_xlim(-1.5,1.5)
ax.set_ylim(-1.5,1.5)
ax.set_title(pi,fontsize=12)

```

Out[31]: Text(0.5, 1.0, '3.127112711271127')



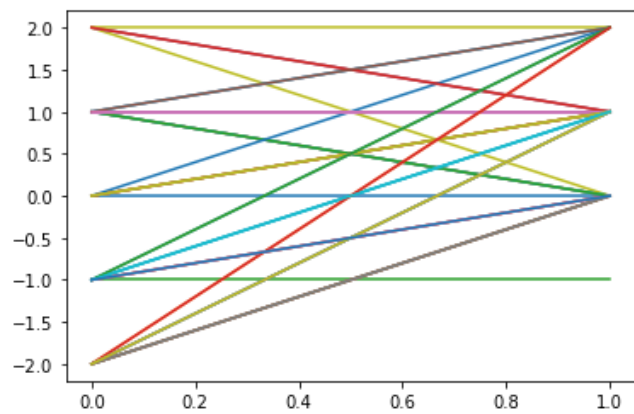
Exercise 5.3: The aim of this exercise is to generate a set of two-dimensional random walks, plot their trajectories and look at the end point distribution. The random walks considered always begin at the origin and take N_{step} random steps of unit or zero size in both directions in the x and y axis. For a total number of N_w walks: 1. Compute the trajectories and save the final point of all them. 2. Plot a sample of these random walks in the plane. 3. Plot all the final points together. 4. Compute the average final distance from the origin of the system. 5. Plot a histogram with the values of the distance to the origin.

```
In [32]: #PART 1
def randwalk(nsteps):
    '''
    takes a random walk in nsteps
    '''
    x = 0
    y = 0
    path = [(0,0)]
    for n in range(nsteps):
        #to randomly decide where to go
        direc = random.choice([(0,1),(0,-1),(-1,0),(1,0),(0,0)])
        x += direc[0]
        y += direc[1]
        path.append((x,y))

    return path
```

```
In [33]: #Part 2
allwalks = []
Nw = 1000
for walk in range(Nw):
    allwalks.append(randwalk(50))
#Example of walk path
for i in range(1):
    plot_xy_lists = list(zip(*allwalks[i]))

    plt.plot(plot_xy_lists)
```



```

In [34]: #Part 3 & 4
final_points = []
origindist = []
fig, axarr = plt.subplots(nrows=1, ncols=2, figsize=(16,6))

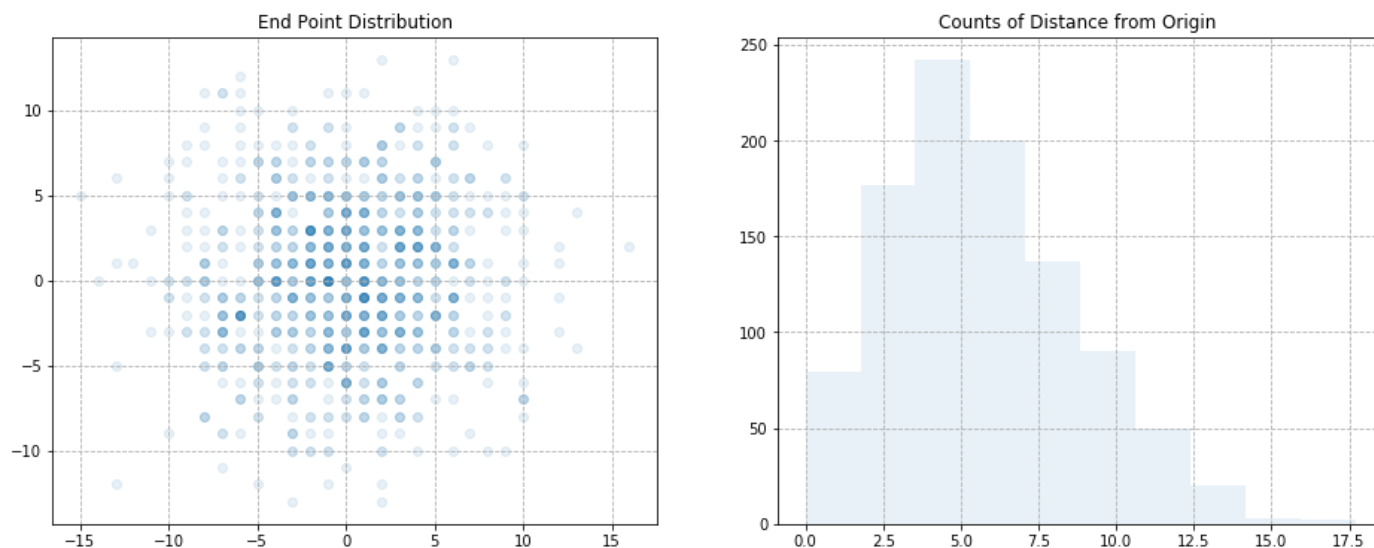
for walk in allwalks:
    '''
    Loop to plot the end point distribution and the distance from the origin
    '''
    final_points.append(walk[-1]) #Determines the final path
    origindist.append(np.linalg.norm(walk[-1])) #Calculates the distance from the origin (sqrt(x^2
+y^2)

end_points_plot = list(zip(*final_points))

axarr[0].scatter(end_points_plot[0],end_points_plot[1],alpha = 0.1)
axarr[0].grid(linestyle = 'dashed')
axarr[0].set_title("End Point Distribution")
axarr[1].hist(origindist, alpha = 0.1)
axarr[1].grid(linestyle = 'dashed')
axarr[1].set_title('Counts of Distance from Origin ')

```

Out[34]: Text(0.5, 1.0, 'Counts of Distance from Origin ')



Exercise 5.4: The Julia set is an important concept in fractal theory. Given a complex number a , a point z in the complex plane is said to be in the filled-in Julia set of a function $f(z) = z^2 + a$ if the iteration of the function over the point does not finish with the point going to infinity. It can be proved that if at some iterate of a point under $f(z)$ the result has a module larger than 2 and larger than the module of a , this point will finish going to infinity. Build and plot the filled-in Julia sets for $f(z)$ with $a = (-0.5, 0), (0.25, -0.52), (-1, 0), (-0.2, 0.66)$ in the interval of $-1 < \text{Re}(z), \text{Im}(z) < 1$ and consider that the point belongs to the set once the previous condition has not been accomplished after $Niter = 100$. Hint: You can make use of the NumPy meshgrid and the PyPlot pplot functions for displaying the filled-in Julia sets.

```

In [45]: '''
Things you need in function:
    a is complex number and z is a point in complex plane
    1. Input: pairs of complex values of a and range for z, which forms nxn grid
    2. The equation:  $z^{**2} + a$ 
    3. Iterate it over a meshgrid(values of z) and if it doesn't go to inf (module > 2 & > a.... While or If ?)
        then it is not in Julia
    4. perform iteration over Niter = 100? - while loop
    5. How do you make it so if it's not in the set be a colour on the pyplot...

Outcome: 4 plots for the different a
'''

def juliaset(Niter):

    #set up parameters
    counter = 100
    z = np.zeros((1000,1000))
    x,y = np.meshgrid(np.linspace(-1,1,1000),np.linspace(-1,1,1000))
    a = [complex(-0.5,0),complex(0.25,-0.52),complex(-1,0),complex(-0.2,0.66)]
    fig, axarr = plt.subplots(nrows=1, ncols=4, figsize=(16,6))

    for n, cmplx in enumerate(a):
        '''
        loop to acquire the real and complex values for each complex and the module
        '''

        a_re = cmplx.real
        a_im = cmplx.imag
        amodule = np.sqrt(a_re*a_re+a_im*a_im)

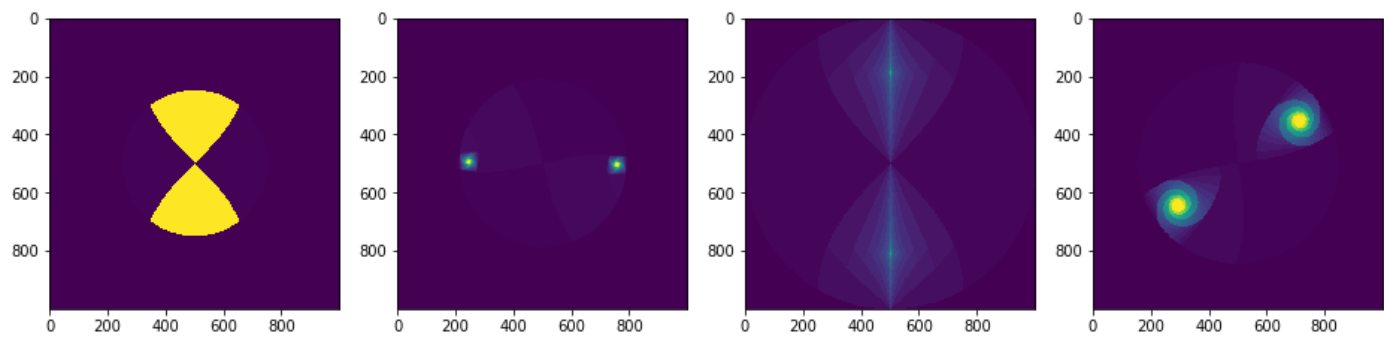
        for ll in range(len(x)):
            for mm in range(len(y)):
                '''
                double loop to evaluate every point along the meshgrid and evaluate if the point is
                in the julia set through a while loop
                '''
                l = ll/1000*2-1
                m = mm/1000*2-1
                point = complex(l,m)

                counter = 0
                while abs(point) < 2 and abs(point) < amodule and counter < 100:
                    point = point**2+cmplx
                    counter += 1

                ratio = counter/100
                z[ll,mm] = ratio
            axarr[n].imshow(z)

```

```
In [46]: juliaset(100)
```



Exercise 6

Exercise 6.1: Read the different files with Cyprus towns temperatures provided in the TData folder and build a dataframe combining all the information. The columns should be the year and months and you can distinguish between data for the different towns adding an extra column with the town name.

Hint: the function concat can be very useful in this case.

```
In [37]: Locations = ['Agrinio', 'Alexandroupolis', 'Alicante', 'Antalya', 'Araxos']

#for n, name in enumerate(Locations):
#    df+str(n+1) = pd.read_csv('T_'+str(name)+'_EM.csv', header =0, index_col = 0)

df1 = pd.read_csv('T_Agrinio_EM.csv',header = 0, index_col = 0)
df2 = pd.read_csv('T_Alexandroupolis_EM.csv',header = 0, index_col = 0)
df3 = pd.read_csv('T_Alicante_EM.csv',header = 0, index_col = 0)
df4 = pd.read_csv('T_Antalya_EM.csv',header = 0, index_col = 0)
df5 = pd.read_csv('T_Araxos_EM.csv',header = 0, index_col = 0)

totaldata = pd.concat([df1,df2,df3,df4,df5], keys = ['Agrinio', 'Alexandroupolis', 'Alicante', 'Antalya', 'Araxos'])
totaldata.columns = ['January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', 'September', 'October', 'November', 'December']
'''
To see if they did combine will see the first 137 rows since it should show 2096, for Agrinio and
then
year 1961 for Alexandroupolis
'''
totaldata.head(137)
```

Out[37]:

		January	February	March	April	May	June	July	August	September	October	November	December
year													
Agrinio	1961	8.56	9.18	10.86	14.62	18.13	21.98	25.97	25.33	21.75	17.61	13.15	10.70
	1962	8.48	8.94	10.21	13.93	17.34	22.34	26.51	26.01	21.45	17.65	13.10	9.33
	1963	9.30	9.85	10.75	13.45	17.71	22.24	25.52	25.44	22.35	16.57	13.41	8.56
	1964	8.67	8.42	10.49	14.27	17.52	21.87	25.25	25.19	22.32	17.99	13.52	9.03
	1965	7.59	8.05	10.68	13.45	17.64	22.75	25.90	25.53	22.26	16.99	13.31	10.42

	2093	11.77	12.62	13.84	16.02	21.69	25.75	29.75	30.34	25.94	20.21	15.72	12.95
	2094	11.74	10.86	14.31	16.07	21.39	27.20	30.47	30.39	26.01	20.71	17.04	13.71
	2095	10.94	11.73	12.86	17.19	22.29	26.55	31.53	29.83	26.08	21.86	16.50	12.29
	2096	10.00	12.20	13.87	16.73	21.76	25.84	30.33	30.60	26.37	22.40	16.62	11.99
Alexandroupolis	1961	5.41	6.40	8.75	13.61	17.28	21.87	25.97	24.51	20.25	14.78	10.62	7.86

137 rows × 12 columns

Exercise 6.2: A different way to combine the Cyprus towns temperature data provided in the TData folder is to build a dataframe whose index is a hierarchical one, with the year and month and there are as many columns as towns, labeled with the

town names.

Hint: The concat and unstack functions can be helpful in this

```
In [38]: pd.DataFrame.unstack(totaldata)
```

Out[38]:

	January										...	December									
year	1961	1962	1963	1964	1965	1966	1967	1968	1969	1970	...	2087	2088	2089	2090	2091	2092	2093	2094	2095	2096
Agrinio	8.56	8.48	9.30	8.67	7.59	8.86	9.16	8.25	9.01	8.09	...	11.46	12.76	12.61	13.74	12.56	12.26	12.96	12.76	12.56	12.26
Alexandroupolis	5.41	4.49	5.86	5.51	4.48	5.55	6.47	4.54	5.57	4.91	...	7.51	10.06	9.60	10.84	9.30	8.93	9.60	10.84	9.30	8.93
Alicante	11.55	10.72	11.04	11.29	10.74	10.63	11.66	11.00	10.28	10.41	...	14.44	14.30	14.55	15.36	14.81	15.00	14.55	15.36	14.81	15.00
Antalya	9.63	9.41	9.75	9.77	9.03	10.28	10.29	9.66	9.37	9.43	...	13.55	14.32	13.99	14.97	14.16	13.37	13.99	14.97	14.16	13.37
Araxos	10.32	10.17	10.96	10.25	9.30	10.53	10.90	9.88	10.69	9.83	...	13.12	14.46	14.39	15.34	14.31	14.07	14.39	15.34	14.31	14.07

5 rows × 1632 columns

Exercise 6.3: Compute the correlation matrix between the temperatures in the provided

Cyprus cities from the previous exercise dataframe.

```
In [39]: pd.DataFrame.corr(totaldata)
```

Out[39]:

	January	February	March	April	May	June	July	August	September	October	November	December
January	1.000000	0.935128	0.903183	0.788183	0.508684	0.262493	0.238966	0.373861	0.663833	0.815572	0.891838	0.880541
February	0.935128	1.000000	0.938298	0.818324	0.544468	0.301938	0.263190	0.391466	0.686567	0.833171	0.892563	0.854624
March	0.903183	0.938298	1.000000	0.891873	0.658987	0.435398	0.381539	0.508778	0.769744	0.878337	0.904261	0.845019
April	0.788183	0.818324	0.891873	1.000000	0.845778	0.668932	0.619504	0.702199	0.876089	0.906807	0.851131	0.738539
May	0.508684	0.544468	0.658987	0.845778	1.000000	0.901518	0.888242	0.910603	0.905837	0.789029	0.642709	0.476581
June	0.262493	0.301938	0.435398	0.668932	0.901518	1.000000	0.952397	0.935087	0.820377	0.623037	0.439847	0.267272
July	0.238966	0.263190	0.381539	0.619504	0.888242	0.952397	1.000000	0.950431	0.803802	0.588385	0.401430	0.225486
August	0.373861	0.391466	0.508778	0.702199	0.910603	0.935087	0.950431	1.000000	0.886231	0.699904	0.536664	0.363640
September	0.663833	0.686567	0.769744	0.876089	0.905837	0.820377	0.803802	0.886231	1.000000	0.903181	0.791883	0.636976
October	0.815572	0.833171	0.878337	0.906807	0.789029	0.623037	0.588385	0.699904	0.903181	1.000000	0.914714	0.779918
November	0.891838	0.892563	0.904261	0.851131	0.642709	0.439847	0.401430	0.536664	0.791883	0.914714	1.000000	0.846277
December	0.880541	0.854624	0.845019	0.738539	0.476581	0.267272	0.225486	0.363640	0.636976	0.779918	0.846277	1.000000

Exercise 6.4: We provide as an example data set the file 'meteodat.csv' with an excerpt of data with a 10 minute frequency from an automated meteorological station with a span of two months (Jan and Feb 2014). Data are comma separated and you can read them using pd.read_csv. The first column is the date and the second the time. Transform this data into a dataframe with an index of datetime, compute dataframes with downsampling to hourly and daily average values for temperature (Tout), pressure (Pressure), relative humidity (H out), wind speed (Wind Speed), and dew point temperature (Dew point); sum of rainfall (Rain); and maximum and minimum values of temperature (Tmax and Tmin) and show the correlation between these variables at hourly and daily scales.

```
In [40]: import datetime

odf = pd.read_csv('meteodat.csv',header = 0, index_col = False)
odf.head()
#odf.head(50000) #To find out there is 8495 rows of trash
```

Out[40]:

	date	time	Tout	Tmax	Tmin	H out	Dew point	Wind Speed	Wind Direction	Wind Speed Hi	...	Wind Chill	Heat Index	THM Index	Pressure	Rain	Rain Max	Tin	H in
0	01/01/14	0:10	15.1	15.1	15.1	85	12.6	9.7	SSW	17.7	...	14.7	15.1	14.7	1023.0	0.0	0.0	16.4	56
1	01/01/14	0:20	15.1	15.1	15.1	85	12.6	8.0	SSW	14.5	...	14.9	15.1	14.9	1022.9	0.0	0.0	16.6	56
2	01/01/14	0:30	15.1	15.1	15.1	86	12.7	8.0	SSW	16.1	...	14.9	15.1	14.9	1022.8	0.0	0.0	16.6	56
3	01/01/14	0:40	14.9	15.1	14.9	86	12.6	8.0	SSW	14.5	...	14.8	14.9	14.8	1022.7	0.0	0.0	16.6	56
4	01/01/14	0:50	14.9	14.9	14.8	88	12.9	6.4	SSW	12.9	...	14.9	14.9	14.9	1022.6	0.0	0.0	16.6	56

5 rows × 21 columns


```

In [41]: '''
To be able to downsample into days and hours we need to combine the date and time columns into
a date:time format in the shape d/m/y/h:m and then reappend it to a new table so we can
easily perform the proposed computations
'''

#need to transform the year from /14 to /2014
date = []
for n in range(len(odf)):
    date.append(odf.iloc[n]['date'][0:-2]+str(20)+odf.iloc[n]['date'][-2:])
time = odf['time']

#create new data frame so we can combine date and time easily
dictio = {'date':date,'time':time}
newodf = pd.DataFrame(dictio)
date_time = newodf[['date','time']].agg('/'.join,axis=1)

#Format it into the correct form so we can easily downsample
stamps = []
for n in range(len(date_time)):
    stamps.append(datetime.datetime.strptime(date_time.iloc[n], '%d/%m/%Y:%H:%M'))

index = pd.DatetimeIndex(stamps)
print(type(index))

#new dataset ready for downsampling
new1 = odf.copy()
new1 = new1.set_index(index)
new1= new1.drop(columns = ['date','time'])
new1.head()

```

```
<class 'pandas.core.indexes.datetimes.DatetimeIndex'>
```

Out[41]:

	Tout	Tmax	Tmin	H out	Dew point	Wind Speed	Wind Direction	Wind Speed Hi	Wind Direction Hi	Wind Chill	Heat Index	THM Index	Pressure	Rain	Rain Max	Tin	H in	D P In
2014-01-01 00:10:00	15.1	15.1	15.1	85	12.6	9.7	SSW	17.7	SW	14.7	15.1	14.7	1023.0	0.0	0.0	16.4	56	
2014-01-01 00:20:00	15.1	15.1	15.1	85	12.6	8.0	SSW	14.5	SSW	14.9	15.1	14.9	1022.9	0.0	0.0	16.6	56	
2014-01-01 00:30:00	15.1	15.1	15.1	86	12.7	8.0	SSW	16.1	SW	14.9	15.1	14.9	1022.8	0.0	0.0	16.6	56	
2014-01-01 00:40:00	14.9	15.1	14.9	86	12.6	8.0	SSW	14.5	SSW	14.8	14.9	14.8	1022.7	0.0	0.0	16.6	56	
2014-01-01 00:50:00	14.9	14.9	14.8	88	12.9	6.4	SSW	12.9	SSW	14.9	14.9	14.9	1022.6	0.0	0.0	16.6	56	

```

In [42]: #Daily Downsample
Daily = new1.resample('d').agg({'Tout':'mean','Pressure':'mean','H out':'mean','Wind Speed':'mean',
, \
'Dew point':'mean','Rain':'sum','Tmax':'max','Tmin':'min'})
Daily.head()

```

Out[42]:

	Tout	Pressure	H out	Wind Speed	Dew point	Rain	Tmax	Tmin
2014-01-01	14.391608	1020.513287	91.881119	3.664336	13.077622	1.2	15.6	11.7
2014-01-02	15.670139	1019.097917	93.250000	8.112500	14.585417	1.4	16.4	14.3
2014-01-03	16.106250	1021.252778	94.027778	7.278472	15.124306	0.8	18.3	14.8
2014-01-04	15.131250	1022.018750	88.013889	9.936111	13.119444	1.0	17.4	12.2
2014-01-05	11.514583	1025.277083	88.375000	2.171528	9.620139	0.2	16.1	7.3

```
In [43]: #Hourly Downsample
hourly = new1.resample('H').agg({'Tout': 'mean', 'Pressure': 'mean', 'H out': 'mean', 'Wind Speed': 'mean', \
                                'Dew point': 'mean', 'Rain': 'sum', 'Tmax': 'max', 'Tmin': 'min'})
hourly.head()
```

Out[43]:

	Tout	Pressure	H out	Wind Speed	Dew point	Rain	Tmax	Tmin
2014-01-01 00:00:00	15.020000	1022.800000	86.000000	8.020000	12.680000	0.0	15.1	14.8
2014-01-01 01:00:00	14.800000	1022.150000	88.833333	5.866667	12.983333	0.4	14.9	14.6
2014-01-01 02:00:00	14.266667	1021.616667	92.833333	4.533333	13.116667	0.4	14.6	14.2
2014-01-01 03:00:00	14.450000	1021.266667	94.166667	4.266667	13.516667	0.2	14.6	14.3
2014-01-01 04:00:00	14.550000	1021.150000	94.833333	6.416667	13.750000	0.2	14.7	14.4

```
In [44]: '''
To show the correlation between the daily and hourly we just follow similar
steps through 6.1, 6.2 & 6.3
'''
total_d = pd.concat([hourly, Daily], keys = ['hourly', 'daily'])
pd.DataFrame.unstack(total_d)
pd.DataFrame.corr(total_d)
```

Out[44]:

	Tout	Pressure	H out	Wind Speed	Dew point	Rain	Tmax	Tmin
Tout	1.000000	0.055500	-0.214522	0.452574	0.727271	-0.025571	0.964104	0.968241
Pressure	0.055500	1.000000	-0.015641	-0.260977	0.039140	-0.172102	0.057483	0.047483
H out	-0.214522	-0.015641	1.000000	-0.258015	0.511729	0.083671	-0.235816	-0.177183
Wind Speed	0.452574	-0.260977	-0.258015	1.000000	0.219287	0.068920	0.433846	0.439407
Dew point	0.727271	0.039140	0.511729	0.219287	1.000000	0.033165	0.681619	0.724961
Rain	-0.025571	-0.172102	0.083671	0.068920	0.033165	1.000000	0.051132	-0.104424
Tmax	0.964104	0.057483	-0.235816	0.433846	0.681619	0.051132	1.000000	0.873714
Tmin	0.968241	0.047483	-0.177183	0.439407	0.724961	-0.104424	0.873714	1.000000