modified_final

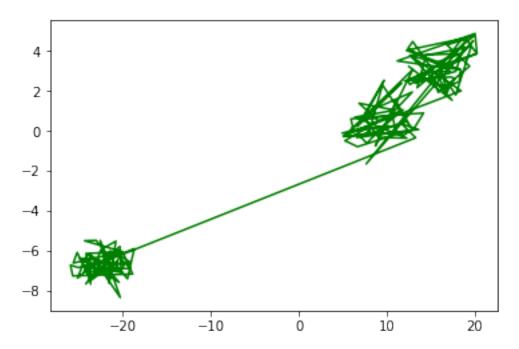
May 1, 2018

```
In [3]: %pylab inline
Populating the interactive namespace from numpy and matplotlib
In [4]: from sklearn.manifold import TSNE
In [5]: # Loading the iris datset
     from sklearn.datasets import load_iris
     iris = load_iris()
In [8]: iris.keys()
Out[8]: dict_keys(['data', 'target', 'target_names', 'DESCR', 'feature_names'])
In [9]: iris.target_names
Out[9]: array(['setosa', 'versicolor', 'virginica'], dtype='<U10')</pre>
In [10]: iris.feature_names
Out[10]: ['sepal length (cm)',
       'sepal width (cm)',
       'petal length (cm)',
       'petal width (cm)']
In [11]: iris.target
1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
           In [28]: from sklearn.manifold import TSNE
      X = \text{np.array}([[0, 0, 0], [0, 1, 1], [1, 0, 1], [1, 1, 1]])
      # X_embedded = TSNE(n_components=2).fit_transform(iris.data)
      X_embedded = TSNE(learning_rate = 100).fit_transform(iris.data)
      X_embedded.shape
```

```
Out[28]: (150, 2)
```

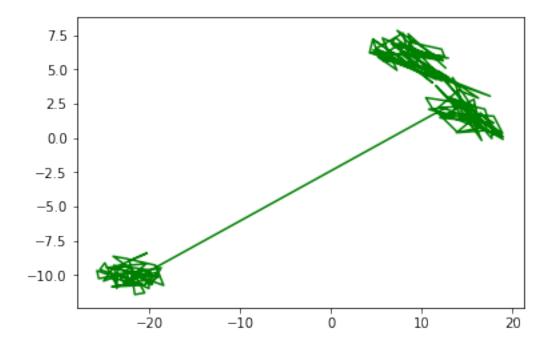
In [29]: plt.plot(X_embedded[:,0], X_embedded[:,1] , 'g')

Out[29]: [<matplotlib.lines.Line2D at 0x166b2f28358>]

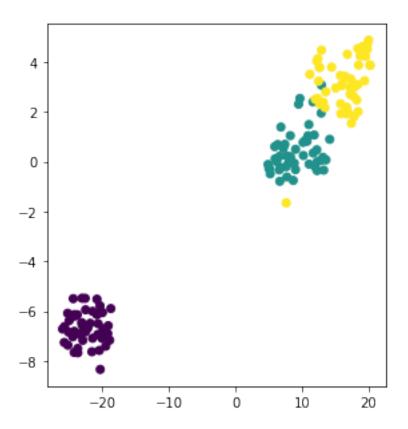


In [26]: plt.plot($X_{embedded[:,0]}$, $X_{embedded[:,1]}$, 'g')

Out[26]: [<matplotlib.lines.Line2D at 0x166b2ec8438>]



Out[38]: <matplotlib.collections.PathCollection at 0x166b326beb8>



```
In [89]: def find_good_size(Limits, windowSize):
             x_min = Limits[0][0]
             x_max = Limits[0][1]
             y_min = Limits[1][0]
             y_{max} = Limits[1][1]
             from math import floor, ceil
             # Clean the matrix
             x_min = floor(x_min)
             x_max = ceil(x_max)
             y_min = floor(y_min)
             y_max = ceil(y_max)
             # Printing the cleaend up values.
             print(x_min,x_max,y_min, y_max)
             # Fitting the size to handle the windows
             a = (x_max - x_min) \% windowSize
             b = (y_max - y_min) % windowSize
             if ((x max + a) \% windowSize == 0):
                 x_max = x_max + a
             else:
                 x_max = x_max + (windowSize - a)
             if ( (y_max + b) % windowSize == 0):
                 y_max = y_max + b
             else:
                 y_max = y_max + (windowSize - b)
             print("You Can use this matrix now.")
             print("X_min=",x_min)
             print("X_max=",x_max)
             print("Y_min=",y_min)
             print("Y_max=",y_max)
             print("And the Matrix Shape:", (x_max - x_min) , " * ", (y_max - y_min))
             return [[x_min,x_max],[y_min,y_max]]
In [90]: find_good_size(Limts,1)
-26 21 -9 5
You Can use this matrix now.
X \min = -26
X_max= 21
Y_min = -9
Y_max=5
And the Matrix Shape: 47 * 14
```

```
Out[90]: [[-26, 21], [-9, 5]]
In [66]: find_good_size(Limts,10)
-26 21 -9 5
You Can use this matrix now.
X_{min} = -26
X_max= 24
Y_min = -9
Y_max= 11
And the Matrix Shape: 50 * 20
Out[66]: [[-26, 24], [-9, 11]]
In [91]: def tell_windows(Limts, windowSize):
             g = find_good_size(Limts, windowSize)
             # Going Row Wise:
             count = 0
             for i in range(g[0][0],g[0][1], windowSize):
                 for j in range(g[1][0],g[1][1],windowSize):
                      count += 1
                     print("<Window:",count,">","X:[", i, ",", i +windowSize , "]" , "Y:[", j,
             print("Total Count:",count)
             ans = (g[0][1]-g[0][0])*(g[1][1]-g[1][0]) // (windowSize ** 2)
             print("Expected Count:",ans)
             print("OKAY:", count == ans )
In [93]: tell_windows([[0,3],[0,3]],1)
0 3 0 3
You Can use this matrix now.
X min= 0
X_{max}=3
Y min= 0
Y_{max}=3
And the Matrix Shape: 3 * 3
<Window: 1 > X:[ 0 , 1 ] Y:[ 0 , 1 ]
<Window: 2 > X:[ 0 , 1 ] Y:[ 1 , 2 ]
\{\text{Window: 3 > X:[0,1] Y:[2,3]}
\{\text{Window}: 4 > X: [1, 2] Y: [0, 1]
<Window: 5 > X:[ 1 , 2 ] Y:[ 1 , 2 ]
<Window: 6 > X:[ 1 , 2 ] Y:[ 2 , 3 ]
\langle Window: 7 \rangle X: [2, 3] Y: [0, 1]
<Window: 8 > X:[ 2 , 3 ] Y:[ 1 , 2 ]
\{\text{Window: 9 > X:[2,3]Y:[2,3]}
Total Count: 9
Expected Count: 9
OKAY: True
```

```
In [94]: tell_windows([[0,3],[0,3]],3)
0 3 0 3
You Can use this matrix now.
X min= 0
X \text{ max} = 3
Y_{min} = 0
Y_max=3
And the Matrix Shape: 3 * 3
\{\text{Window}: 1 > X: [0, 3] Y: [0, 3]
Total Count: 1
Expected Count: 1
OKAY: True
In [95]: tell_windows([[0,3],[0,3]],5)
0 3 0 3
You Can use this matrix now.
X_{min} = 0
X_{max} = 5
Y_{min} = 0
Y_max= 5
And the Matrix Shape: 5 * 5
\{\text{Window: 1 > X:[0,5]Y:[0,5]}
Total Count: 1
Expected Count: 1
OKAY: True
In [77]: Limts
Out[77]: [[-25.9487, 20.243372], [-8.339147, 4.883639]]
In [103]: # Okay So now We have a function which finds the windows we just now need to calcula
          def calEntropy(window,label):
               'Takes: [x_min, x_max],[y_min, y_max]'
               'returns entropy of the window'
              x_min, x_max , y_min, y_max = window
              # Set up entropy to be zero
              en = 0
              from math import log
              c = [0,0,0]
              # Calculate the number of samples in this window.
              \# X_{embedded}  stores the data.
              for i in range(150):
                  cx = X_embedded[i][0]
                  cy = X_embedded[i][1]
                   if x_min <= cx <=x_max and y_min <= cy <= y_max:</pre>
```

```
print("Found a sample of class:",label[i])
                      c[label[i]] += 1
              t = c[0] + c[1] + c[2]
              for i in c:
                  if i!= 0:
                      en += - ( i/t * log(i/t) )
              print("Found Entropy:",en)
              return en
In [109]: def tell_windows(Limts, windowSize, target):
              g = find_good_size(Limts,windowSize)
              # Going Row Wise:
              total_ent = 0
              count = 0
              for i in range(g[0][0],g[0][1],windowSize):
                  for j in range(g[1][0],g[1][1],windowSize):
                      count += 1
                      print("<Window:",count,">","X:[", i, ",", i +windowSize , "]" , "Y:[", j
                      total_ent += calEntropy([i,windowSize+i, j, windowSize+j], target)
              print("Total Count:",count)
              ans = (g[0][1]-g[0][0])*(g[1][1]-g[1][0]) // (windowSize ** 2)
              print("Expected Count:",ans)
              print("OKAY:", count == ans )
              print("----Done----",total_ent)
              return total ent
In [110]: tell_windows(Limts,3, add_noise(0))
Plotting the data with Noise => 0 %
Done!
-26 21 -9 5
You Can use this matrix now.
X_{min} = -26
X_max=22
Y min=-9
Y_max=6
And the Matrix Shape: 48 * 15
\{Window: 1 > X:[ -26 , -23 ] Y:[ -9 , -6 ]
Found a sample of class: 0
```

```
Found a sample of class: 0
Found Entropy: 0.0
\langle \text{Window}: 2 \rangle X: [-26, -23] Y: [-6, -3]
Found a sample of class: 0
Found a sample of class: 0
Found Entropy: 0.0
\langle \text{Window}: 3 > X: [ -26 , -23 ] Y: [ -3 , 0 ]
Found Entropy: 0
<Window: 4 > X:[ -26 , -23 ] Y:[ 0 , 3 ]
Found Entropy: 0
\{\text{Window: 5} > X: [-26, -23] Y: [3, 6]
Found Entropy: 0
\{\text{Window: 6} > X: [-23, -20] Y: [-9, -6]
Found a sample of class: 0
Found Entropy: 0.0
<Window: 7 > X:[ -23 , -20 ] Y:[ -6 , -3 ]
Found a sample of class: 0
Found Entropy: 0.0
<Window: 8 > X:[ -23 , -20 ] Y:[ -3 , 0 ]
Found Entropy: 0
```

```
\langle \text{Window: 9} \rangle X: [-23, -20] Y: [0, 3]
Found Entropy: 0
<Window: 10 > X:[ -23 , -20 ] Y:[ 3 , 6 ]
Found Entropy: 0
<Window: 11 > X:[ -20 , -17 ] Y:[ -9 , -6 ]
Found a sample of class: 0
Found Entropy: 0.0
\forall x \in X : [-20, -17] Y : [-6, -3]
Found a sample of class: 0
Found Entropy: 0.0
<Window: 13 > X:[ -20 , -17 ] Y:[ -3 , 0 ]
Found Entropy: 0
<Window: 14 > X:[ -20 , -17 ] Y:[ 0 , 3 ]
Found Entropy: 0
<Window: 15 > X:[ -20 , -17 ] Y:[ 3 , 6 ]
Found Entropy: 0
\langle \text{Window}: 16 \rangle X: [-17, -14] Y: [-9, -6]
Found Entropy: 0
<Window: 17 > X:[ -17 , -14 ] Y:[ -6 , -3 ]
Found Entropy: 0
\langle \text{Window}: 18 \rangle X: [-17, -14] Y: [-3, 0]
Found Entropy: 0
<Window: 19 > X:[ -17 , -14 ] Y:[ 0 , 3 ]
Found Entropy: 0
\langle \text{Window: 20} \rangle X: [-17, -14] Y: [3, 6]
Found Entropy: 0
\langle \text{Window}: 21 \rangle X: [-14, -11] Y: [-9, -6]
Found Entropy: 0
\{Window: 22 > X: [ -14 , -11 ] Y: [ -6 , -3 ] <math>\}
Found Entropy: 0
\langle \text{Window: } 23 \rangle \ \text{X:[} -14 \ , -11 \ ] \ \text{Y:[} -3 \ , 0 \ ]
Found Entropy: 0
<Window: 24 > X:[ -14 , -11 ] Y:[ 0 , 3 ]
Found Entropy: 0
\langle \text{Window: } 25 \rangle X: [-14, -11] Y: [3, 6]
Found Entropy: 0
\langle \text{Window: 26} \rangle \text{ X:[-11,-8] Y:[-9,-6]}
Found Entropy: 0
\langle \text{Window: } 27 \rangle \text{ X: [ } -11 \text{ , } -8 \text{ ] } \text{ Y: [ } -6 \text{ , } -3 \text{ ]}
Found Entropy: 0
<Window: 28 > X:[-11, -8] Y:[-3, 0]
Found Entropy: 0
<Window: 29 > X:[ -11 , -8 ] Y:[ 0 , 3 ]
```

```
Found Entropy: 0
<Window: 30 > X:[ -11 , -8 ] Y:[ 3 , 6 ]
Found Entropy: 0
\{\text{Window: } 31 > \text{X:} [-8, -5] \text{ Y:} [-9, -6] \}
Found Entropy: 0
\forall x = 1.5 \times 10^{-8} = 1.5 \times 
Found Entropy: 0
<Window: 33 > X:[ -8 , -5 ] Y:[ -3 , 0 ]
Found Entropy: 0
<Window: 34 > X:[ -8 , -5 ] Y:[ 0 , 3 ]
Found Entropy: 0
<Window: 35 > X:[ -8 , -5 ] Y:[ 3 , 6 ]
Found Entropy: 0
<Window: 36 > X:[ -5 , -2 ] Y:[ -9 , -6 ]
Found Entropy: 0
\langle \text{Window}: 37 \rangle X: [-5, -2] Y: [-6, -3]
Found Entropy: 0
\{\text{Window}: 38 > X: [-5, -2] Y: [-3, 0]
Found Entropy: 0
<Window: 39 > X:[ -5 , -2 ] Y:[ 0 , 3 ]
Found Entropy: 0
<Window: 40 > X:[ -5 , -2 ] Y:[ 3 , 6 ]
Found Entropy: 0
\{\text{Window: } 41 > X:[-2,1] \ Y:[-9,-6]
Found Entropy: 0
\{\text{Window}: 42 > X: [-2, 1] Y: [-6, -3]
Found Entropy: 0
<Window: 43 > X:[ -2 , 1 ] Y:[ -3 , 0 ]
Found Entropy: 0
<Window: 44 > X:[ -2 , 1 ] Y:[ 0 , 3 ]
Found Entropy: 0
<Window: 45 > X:[-2, 1] Y:[3, 6]
Found Entropy: 0
<Window: 46 > X:[1, 4] Y:[-9, -6]
Found Entropy: 0
<Window: 47 > X:[ 1 , 4 ] Y:[ -6 , -3 ]
Found Entropy: 0
<Window: 48 > X:[1, 4] Y:[-3, 0]
Found Entropy: 0
<Window: 49 > X:[1, 4] Y:[0, 3]
Found Entropy: 0
<Window: 50 > X:[ 1 , 4 ] Y:[ 3 , 6 ]
Found Entropy: 0
<Window: 51 > X:[ 4 , 7 ] Y:[ -9 , -6 ]
Found Entropy: 0
<Window: 52 > X:[ 4 , 7 ] Y:[ -6 , -3 ]
Found Entropy: 0
<Window: 53 > X:[4,7] Y:[-3,0]
```

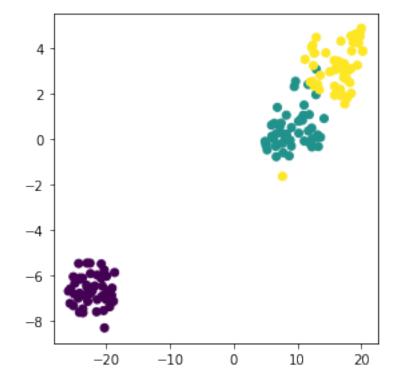
```
Found a sample of class: 1
Found Entropy: 0.0
<Window: 54 > X:[4,7] Y:[0,3]
Found a sample of class: 1
Found Entropy: 0.0
<Window: 55 > X:[ 4 , 7 ] Y:[ 3 , 6 ]
Found Entropy: 0
<Window: 56 > X:[ 7 , 10 ] Y:[ -9 , -6 ]
Found Entropy: 0
<Window: 57 > X:[ 7 , 10 ] Y:[ -6 , -3 ]
Found Entropy: 0
<Window: 58 > X:[ 7 , 10 ] Y:[ -3 , 0 ]
Found a sample of class: 1
Found a sample of class: 2
Found Entropy: 0.410116318288409
<Window: 59 > X:[ 7 , 10 ] Y:[ 0 , 3 ]
Found a sample of class: 1
Found Entropy: 0.0
<Window: 60 > X:[ 7 , 10 ] Y:[ 3 , 6 ]
Found Entropy: 0
<Window: 61 > X:[ 10 , 13 ] Y:[ -9 , -6 ]
Found Entropy: 0
<Window: 62 > X:[ 10 , 13 ] Y:[ -6 , -3 ]
```

```
Found Entropy: 0
<Window: 63 > X:[ 10 , 13 ] Y:[ -3 , 0 ]
Found a sample of class: 1
Found a sample of class: 1
Found a sample of class: 1
Found Entropy: 0.0
<Window: 64 > X:[10, 13] Y:[0, 3]
Found a sample of class: 1
Found a sample of class: 2
Found a sample of class: 2
Found a sample of class: 2
Found Entropy: 0.48257756517701206
<Window: 65 > X:[ 10 , 13 ] Y:[ 3 , 6 ]
Found a sample of class: 1
Found a sample of class: 2
Found Entropy: 0.37677016125643675
<Window: 66 > X:[ 13 , 16 ] Y:[ -9 , -6 ]
Found Entropy: 0
<Window: 67 > X:[ 13 , 16 ] Y:[ -6 , -3 ]
Found Entropy: 0
<Window: 68 > X:[ 13 , 16 ] Y:[ -3 , 0 ]
Found a sample of class: 1
Found Entropy: 0.0
<Window: 69 > X:[ 13 , 16 ] Y:[ 0 , 3 ]
Found a sample of class: 1
Found a sample of class: 1
Found a sample of class: 1
Found a sample of class: 2
Found a sample of class: 2
Found a sample of class: 2
```

```
Found a sample of class: 2
Found a sample of class: 2
Found a sample of class: 2
Found Entropy: 0.6365141682948128
\langle Window: 70 \rangle X: [13, 16] Y: [3, 6]
Found a sample of class: 2
Found a sample of class: 2
Found a sample of class: 2
Found Entropy: 0.0
<Window: 71 > X:[ 16 , 19 ] Y:[ -9 , -6 ]
Found Entropy: 0
<Window: 72 > X:[ 16 , 19 ] Y:[ -6 , -3 ]
Found Entropy: 0
<Window: 73 > X:[ 16 , 19 ] Y:[ -3 , 0 ]
Found Entropy: 0
<Window: 74 > X:[16, 19] Y:[0, 3]
Found a sample of class: 2
Found Entropy: 0.0
<Window: 75 > X:[ 16 , 19 ] Y:[ 3 , 6 ]
Found a sample of class: 2
Found Entropy: 0.0
\forall X: [19, 22] Y: [-9, -6]
Found Entropy: 0
\langle \text{Window}: 77 \rangle X: [19, 22] Y: [-6, -3]
Found Entropy: 0
<Window: 78 > X:[19, 22] Y:[-3, 0]
Found Entropy: 0
<Window: 79 > X:[ 19 , 22 ] Y:[ 0 , 3 ]
Found Entropy: 0
```

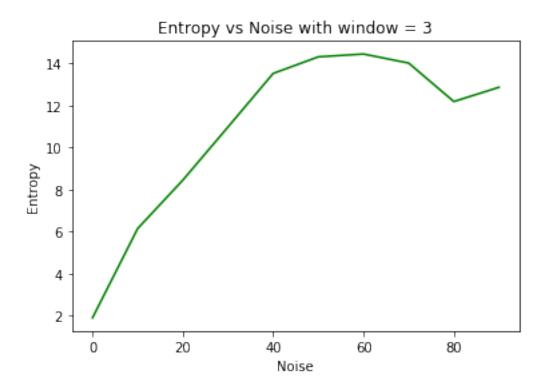
```
<Window: 80 > X:[ 19 , 22 ] Y:[ 3 , 6 ]
Found a sample of class: 2
Found Entropy: 0.0
Total Count: 80
Expected Count: 80
OKAY: True
----Done---- 1.9059782130166705
```

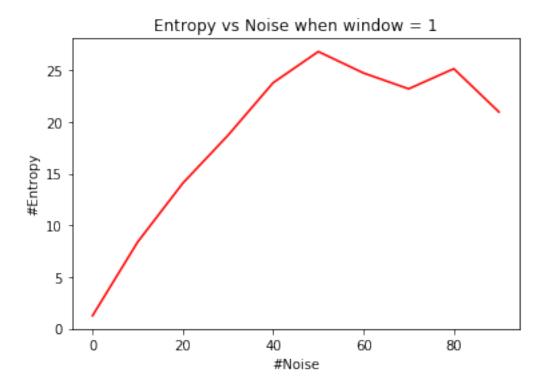
Out[110]: 1.9059782130166705



```
# 50 - 100 : class 1
              # 100 - 150 : class 2
              # We will change the labels in a class label randomly
              import random
              # Offset for percentage => (window size)* precentage
              offset = 50 * percentage // 100
              for i in range(0, 0 + offset ):
                  target[i] = random.choice([1,2])
              for i in range(50, 50 + offset):
                  target[i] = random.choice([0,2])
              for i in range(100,100 + offset):
                  target[i] = random.choice([0,1])
              # Got the new labeled data
              # Now we can plot the tsne with this.
              #print("Plotting the data with Noise => ", percentage , " %")
              figure(figsize = (20,10))
              subplot(242)
              scatter(X_embedded[:,0], X_embedded[:,1], c = target)
              print("Done!")
              return target
In [115]: def main(windowSize, noise):
              return tell_windows(Limts,windowSize,add_noise(noise))
In [ ]: xxxx = []
        yyyy = []
        for i in range(0,100,10):
            xxxx.append(i)
            yyyy.append(main(3,i))
In [118]: xxxx
Out[118]: [0, 10, 20, 30, 40, 50, 60, 70, 80, 90]
In [119]: yyyy
Out[119]: [1.9059782130166705,
           5.833869503703472,
           8.146882998185937,
           10.507330121040468,
           13.927450011307299,
           14.301825684532538,
           12.619612365784652,
           15.087570013154586,
           13.689747163330022,
           13.465812334559693]
In [120]: plt.plot(x,y,'g')
          plt.xlabel('Noise')
```

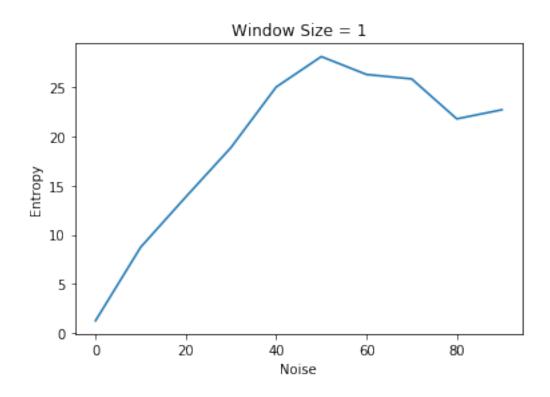
```
plt.ylabel('Entropy')
plt.title('Entropy vs Noise with window = 3')
plt.show()
```

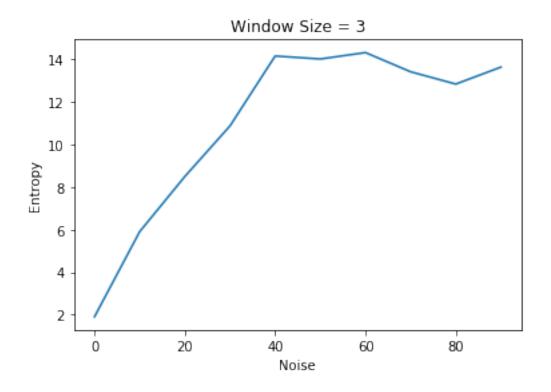


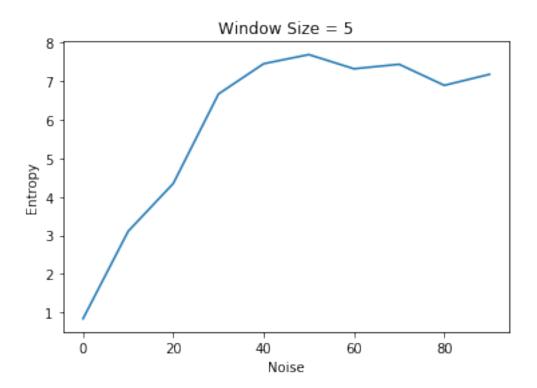


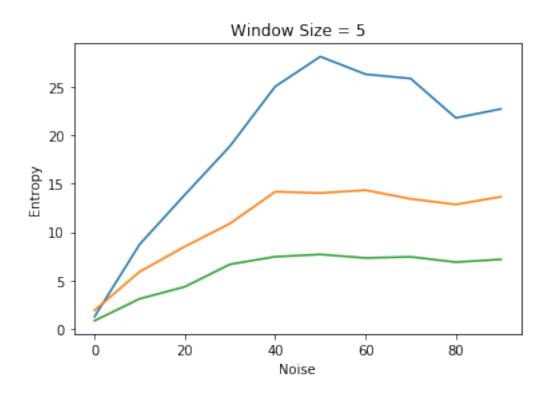
```
In []: dataSet= [[[],[]], [[],[]], [[],[]], [[],[]]]
       for window in [1,3,5]:
           for i in range(0,100,10):
               dataSet[window][0].append(i)
               dataSet[window][1].append(main(window,i))
In [126]: dataSet
Out[126]: [[[], []],
           [[0, 10, 20, 30, 40, 50, 60, 70, 80, 90],
           [1.2554823251787535,
            8.73851878067532,
            13.853681503172105,
             18.878635726235473,
            25.02940503282032,
             28.127269243771856,
             26.30719326841124,
            25.858124148322695,
            21.792007983531576,
            22.716655925390782]],
           [[], []],
           [[0, 10, 20, 30, 40, 50, 60, 70, 80, 90],
           [1.9059782130166705,
            5.906832364815045,
```

```
8.501479237313205,
             10.879963858860972,
             14.168158576926993,
             14.028571586277163,
             14.32908036737314,
             13.426159812900783,
             12.847992308871694,
             13.64677921343577]],
           [[], []],
           [[0, 10, 20, 30, 40, 50, 60, 70, 80, 90],
            [0.8377336454040922,
             3.1059009623027247,
             4.348245698116212,
             6.66788264811653,
             7.452498242994488,
             7.6894883073824065,
             7.321481995767062,
             7.438085109160104,
             6.891550087350982,
             7.177780753797773]]]
In [131]: print("The final plots are")
          def plotter(i):
              plt.plot(dataSet[i][0],dataSet[i][1])
              plt.title('Window Size = ' + str(i))
              plt.xlabel('Noise')
              plt.ylabel('Entropy')
The final plots are
In [130]: plotter(1)
          plotter(3)
          plotter(5)
```









In [133]: # Done !