```
import matplotlib.pyplot as plt
from pandas import read csv
#####################
                         CONTROL PARAMETERS
###################################
zz = 2970 # length of dataset
           # downsampling
ds = 1
div = 20
           # no. of vertical divisions
Catch = 100 # no of least countinuous max sensors instance to be caught
as an anomaly
          # sequences of data to be used to calculate hidden markov
Cont = 5
model probability
#################
                                         training data
train filename = "000_Et_H_CO_n.csv"
############## testing data with 1.25 times
################################
#test filename = "000_Et_H_CO_n_7_1.30_1950_2150.csv"
#test filename = "000 Et H CO n 7 1.50 1950 2150.csv"
#test_filename = "032_Et_H_CO_n_3_1.25_1050_1230.csv"
#test_filename = "032_Et_H_CO_n_3_1.50_1050_1230.csv"
#test filename = "066 Et H CO n 3 1.25 2200 2450.csv"
#test filename = "066 Et H CO n 3 1.50 2200 2450.csv"
#test filename = "098 Et H CO n 2_1.30_1350_1550.csv"
#test filename = "098 Et H CO n 2 1.50 1350 1550.csv"
# train sensors : collectionn of individual sensor inputs into packets of
ds data each
train dataset = read csv(train filename)
train array = train dataset.values
a,b,c,d,e,f,g,h = [],[],[],[],[],[],[],[]
train sensors = [a,b,c,d,e,f,g,h]
count = 3
sum = 0
n=1
for i in train_sensors:
    while (n<zz and count<11):
        for k in range (ds*(n-1), ds*n):
           sum += train array[k][count]
        i.append(sum/ds)
       sum=0
       n += 1
    count+=1
    n=1
plt.plot(a)
plt.plot(b)
plt.plot(c)
plt.plot(d)
plt.plot(e)
plt.plot(f)
plt.plot(g)
plt.plot(h)
plt.legend(['1', '2', '3', '4','5','6','7','8'], loc='upper left')
plt.show()
print("Training data plot")
```

```
# test sensors : collectionn of individual sensor inputs into packets of
ds data each
test dataset = read csv(test filename)
test array = test dataset.values
aa,bb,cc,dd,ee,ff,gg,hh = [],[],[],[],[],[],[]
test sensors = [aa,bb,cc,dd,ee,ff,gg,hh]
count = 3
sum = 0
n=1
for i in test sensors:
    while (n<zz and count<11):
        for k in range(ds*(n-1), ds*n):
            sum += test_array[k][count]
        i.append(sum/ds)
        sum=0
        n+=1
    count+=1
    n=1
plt.plot(aa)
plt.plot(bb)
plt.plot(cc)
plt.plot(dd)
plt.plot(ee)
plt.plot(ff)
plt.plot(gg)
plt.plot(hh)
plt.legend(['1', '2', '3', '4','5','6','7','8'], loc='upper left')
plt.show()
print("Testing data plot")
###################
                                 min max fitting
###################
minimum, maximum, width = [],[],[]
for i in range(8):
    j = train sensors[i]
    k = test_sensors[i]
                               ###changes
    m = min(min(j), min(k))
    n = max(max(j), max(k))
   minimum.append(m)
    maximum.append(n)
    w = (n - m)/div
    width.append(w)
####################
                             Vertical disribution
########################
m, n, o, p, q, r, s, t = [], [], [], [], [], [], []
train values = [m,n,o,p,q,r,s,t]
mm,nn,oo,pp,qr,rr,ss,tt = [],[],[],[],[],[],[],[]
test values = [mm,nn,oo,pp,qr,rr,ss,tt]
for i in range(8):
    j = train values[i]
    k = test values[i]
    w = width[i]
```

```
m = minimum[i]
    for l in train sensors[i]:
        for n in range(div+1):
            if l >= (m + n*w) and l < (m + (n+1)*w):
                j.append(n)
                break
                                         ###changes
    for l in test sensors[i]:
        for n in range(div+1):
            if l >= (m + n*w) and l < (m + (n+1)*w):
                k.append(n)
                break
plt.plot(train values[0])
plt.plot(train_values[1])
plt.plot(train values[2])
plt.plot(train values[3])
plt.plot(train_values[4])
plt.plot(train values[5])
plt.plot(train values[6])
plt.plot(train values[7])
plt.legend(['1', '2', '3', '4','5','6','7','8'], loc='upper left')
plt.show()
print("Vertical Assignment of Training data plot")
plt.plot(test values[0])
plt.plot(test_values[1])
plt.plot(test values[2])
plt.plot(test values[3])
plt.plot(test_values[4])
plt.plot(test_values[5])
plt.plot(test values[6])
plt.plot(test values[7])
plt.legend(['1', '2', '3', '4','5','6','7','8'], loc='upper left')
plt.show()
print("Vertical Assignment of Testing data plot")
#######################
                           Variation in data streams
#########################
var stream = []
for l in range(zz-10):
    diff array = [] # difference of sensor values b/w training & testing
per instance
    for i in range(8):
        j = train_values[i]
        k = test_values[i]
        diff array+=[abs(j[l] - k[l])]
                                          ###changes
    var stream += [diff array]
ga,gb,gc,gd,ge,gf,gg,gh = [],[],[],[],[],[],[],[]
gaps = [ga,gb,gc,gd,ge,gf,gg,gh]
for i in range(8):
    for j in var stream:
        gaps[i].append(j[i])
plt.plot(ga)
plt.plot(gb)
plt.plot(gc)
plt.plot(gd)
```

```
plt.plot(ge)
plt.plot(gf)
plt.plot(gg)
plt.plot(gh)
plt.legend(['1', '2', '3', '4','5','6','7','8'], loc='upper left')
plt.show()
print("Training And Testing Differnce")
                          *****
########################
                                      Markov Model ******
#############################
#print(var stream)
####################
                       Emmision matrix probability assignment
#################
emm prob = []
sensor max list = []
for i in var stream:
    g sum=0
    for j in i:
        g sum += j
    emm prob.append(8*max(i) - g sum) #prob of being an anomaly
    for k in range(8):
        if max(i) == i[k]:
            sensor max list.append(k)
#### normalisation
z = max(emm prob)
for i in range(len(emm prob)):
    emm prob[i] = emm prob[i]/z
############ Observation data with "sensor max list" and "emm prob"
combined
sequence = []
Observation = []
obs = []
seq = []
for i in range(len(sensor max list)):
    obs.append([sensor max list[i],emm prob[i]])
    seq.append(i)
    if i+1 == len(sensor max list) or sensor max list[i] !=
sensor max list[i+1]:
        Observation +=[obs]
        sequence += [seq]
        obs = []
        seq = []
#print (Observation)
#print(sequence) #tracks
###############
                  transition matrix and initial state matrix
pi = 0.5
A = [1, 0.5]
##############
                   probability of observation for each block(100 instance
each) of data
P = []
k = Cont
for i in Observation:
    if len(i)>Catch: #significant deviation considered an anomaly
        n = len(i)//k # n = len(i)-Catch #cont=5 Block of data
        Ps.append( i[0][0] )
```

```
for c in range(n):
             p = (i[c*k][1] * pi)
             for j in range (1, k):
                                                  #cont
                 p = p * (i[c*k+j][1]) * A[0]
             Ps.append(p)
         P += [Ps]
#print(P)
#################
                      ******* Final Selection *******
##################
A_count = []  # Location Track of Anomaly
A_intensity = []  # No of Countinious Anomalous Blocks
A_sensor = []  # Anomalous Sensor Number (0-7)
A_prob = []  # Probability of anomalous sequence
for i in P:
    A sensor.append(i[0])
    A intensity.append(len(i))
    c = len(i)
    s = 0
    for j in range (1,c):
        s += i[j]
    A prob += [ s/(c-1) ]
for i in range(len(sequence)):
    k = sequence[i][0]+2 # margin
    if len(sequence[i])>Catch:
        A count += [k]
###################
                                   Print it out
print("Seq. Track Sensor no. Intensity Probability of Anomaly")
for i in range(len(A sensor)):
    print(" %-5d %-7d %-12d %-14d %-12.10f"%
(i+1,A count[i],A sensor[i],A intensity[i],A prob[i]))
```