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# Assignment 08

Yuval S. Katz  
204025258

1) a- ex8-sec1.png + ass-1.py.

b- PC1 dimension spans 2 main clusters around  $\{0, -30\}$  and  $\{40, -10\}$ .  
with 2 centers -  $\rightarrow$  prob.  $(0, 0.5)$

PC2 dimension spans data around  $\{0, -15\}$  with no clear division.

c- PC1 explains 58.3%.  
PC2 explains 37.2%. } of the variance, together 95.5%  $\Rightarrow$  of the variance in the data can be explained using PC1+2

(out of 20)

1-95.5% remains unexplained by this representation

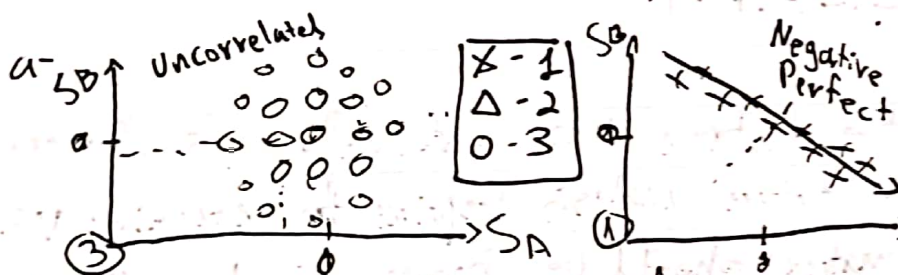
d- Vector 13 has max-var, with entropy of 2.84 bits on avg to pass information

PC1 explains has high variance, with 2.83

I used 10 bins for the dist. calculations. Since PC1 gets most of the attribute of the data orthogonally to the features, it uses a similar dist.

to represent the data, as do vector / feature 13. thus, the same (or at least very close) entropy, (regardless of bin size, just to represent the data to a sufficient level)

2)



$$\begin{aligned} \text{COV}_1(S_A, S_B) &= -1 \\ \text{COV}_2(S_A, S_B) &= 0.5 \\ \text{COV}_3(S_A, S_B) &= 0 \end{aligned}$$

b- ass-2.py  
Simulation of values with 1-3 cov:

① redundant! the same parameter	② $\lambda_1 = 1.5$ $\lambda_2 = 0.5$ $\lambda_3 = 0.5$	③ redundant! $\lambda_1 = 1$ $\lambda_2 = 0$ $\lambda_3 = 0$
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c- The first (\*) is redundant because  $S_B$  doesn't add information beyond  $S_A$  information. (Cov=-1). it sufficient to look at either  $S_A/S_B$  alone!

The last (Cov=0,0) display the highest redundancy because  $S_A$  and  $S_B$  are unrelated (=uncorrelated) and hence, adds a lot of information.

3) a - see file ex8-ass3.py + png.

b - on the fig

c - on the fig - } After 60 iterations  
centroids      red      black      purple      blue      green  
                  $(-7, -3)$   $(-5, 5)$   $(0, -7)$   $(5, -7)$   $(6, 3)$

4) EM GMM.

a - see file ass\_4.py + png

b -  $\mu_i, \sigma_i \Rightarrow \mu_1, \sigma_1, \mu_2, \sigma_2, \mu_3, \sigma_3, \mu_4, \sigma_4$   
                  $k=i$        $(4.35, 2.8)$   $(-1.05, 0.87)$   $(4.02, 5.07)$   $(3.9, 4.83)$

c - on fig

d - on fig ass\_4-3.png

e) That depends on the distance matrix. Visually,  $k=3$  clusters fit the data well, but future analysis requires to determine vs.  $k=4$ , and a distance matrix should be chosen. (Euclidean / Manhattan, ....)  
for example, we can make a DBSCAN and look for outliers...