## MA323: Lab 5 Report

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## $\mathbf{Q}\mathbf{1}$

We know that:

$$\Sigma = \begin{pmatrix} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{pmatrix}$$

We observe the given  $\Sigma$  matrix and it is deductible that  $\sigma_1 = 3$ ,  $\sigma_2 = 2$  and  $\rho = a$ Given that the covariance matrix  $(\Sigma)$  is positive definite, we know that we can generate bivariate normal distribution  $(X_1, X_2)$  by using the following formulae:

$$\begin{aligned} \mathbf{X}_{1} &= \mu_{1} + \sigma_{1} Z_{1} \\ \mathbf{X}_{2} &= \mu_{2} + \rho \sigma_{2} Z_{1} + \sqrt{1 - \rho^{2}} \sigma_{2} Z_{2} \end{aligned}$$

where  $\mathbb{Z}_1$  and  $\mathbb{Z}_2$  are independent standard normal variables.

For the cases of a = -1 and a = 1, the covariance matrix is positive semi-definite. The expression  $x^T \Sigma x$  expands to  $(3x_1 - 2x_2)^2$  and  $(3x_1 + 2x_2)^2$  for a = -1 and a = 1 respectively. This means that the join distribution function for the variables  $(X_1, X_2)$  won't exist.

However with a correlation value of  $\pm 1$ , there is complete dependence between the two variables, i.e. knowing one variable means that we know the second variable for sure.

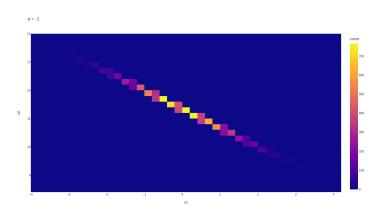
For perfect correlation  $(\rho = \pm 1)$ , we can generate  $X_2$  from  $X_1$  using :

$$x_2 = \mu_2 + \rho \frac{\sigma_2}{\sigma_1} (x_1 - \mu_1)$$

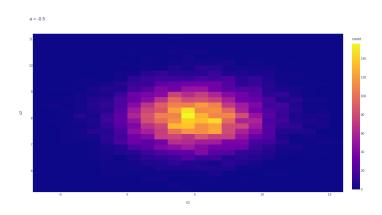
So, for these two cases, we can generate one of the variables normally and obtain the second one deterministically.

I used Box-Muller method to generate two independent standard normal variables in my code. Here are the obtained 2D histograms or density heatmaps:

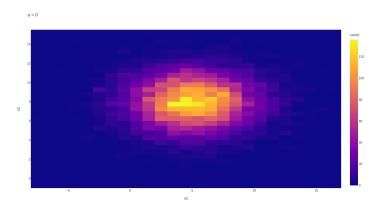
• a = -1



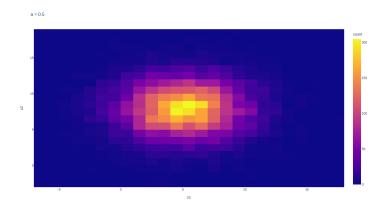
• a = -0.5



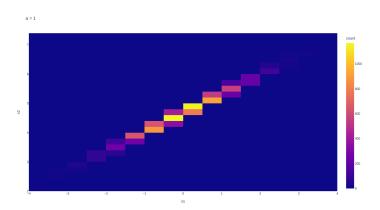
 $\bullet \ a=0$ 



• a = 0.5

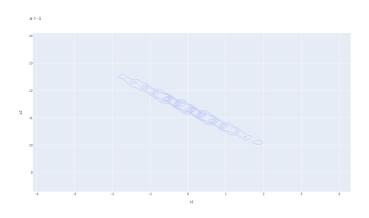


 $\bullet$  a=1

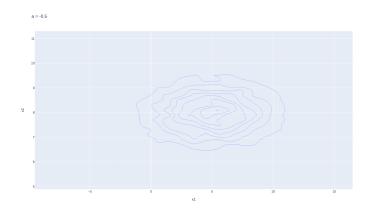


Here are the contour plots:

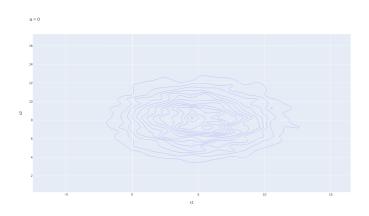
 $\bullet$  a=-1



• a = -0.5



• a = 0



• a = 0.5

