M525: Final

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- 1. Here, we implement a Monte Carlo method for estimating a multidimensional Gaussian integral.
 - (i) We, consider J random variables distributed as

$$\begin{bmatrix} x \\ y \\ z \\ w \end{bmatrix}_{i} \sim \text{Normal}_{4} \begin{pmatrix} \begin{bmatrix} -2 \\ 0 \\ 1 \\ -1 \end{bmatrix}, \begin{bmatrix} 1 & \frac{4}{5} & 0 & 0 \\ \frac{4}{5} & 5 & -1 & 0 \\ 0 & -1 & 3 & 0 \\ 0 & 0 & 0 & 2 \end{bmatrix} \end{pmatrix} i = 1, \dots, J$$

To sample points from p(x, y, z, w), we first find the Cholesky decomposition of our covariance matrix S; this gives us a lower triangular matrix L such that $LL^{\top} = S$. Next, we generate J four dimensional points with components following the standard normal distribution. Next we multiply these points by our matrix L do give them the proper covariance. Lastly, we add our mean m to the points to properly translate them in space.

- (ii) We take J = 10000 and generate the data. See Figure 1.
- (iii) See Figure 1.
- (iv) Finally, we estimate the following integral using the method of Monte–Carlo integration:

$$I = \iiint_A (x - y)^2 p(x, y, z) dx dy dz,$$

where A is the interior of the ellipsoid defined by

$$\frac{(x+3)^2}{2} + \frac{y^2}{2} + \frac{z^2}{9} \le 1.$$

To accomplish this, we take all the points generated in our Monte-Carlo algorithm that fall within A; assume K points satisfy this condition. Note that our covariance matrix S gives us that p(x, y, z|w) = p(x, y, z) since the only entry for w is on the diagonal. Then we find

$$I \approx \frac{1}{J} \sum_{k=1}^{K} (x_k - y_k)^2 = 1.3492.$$

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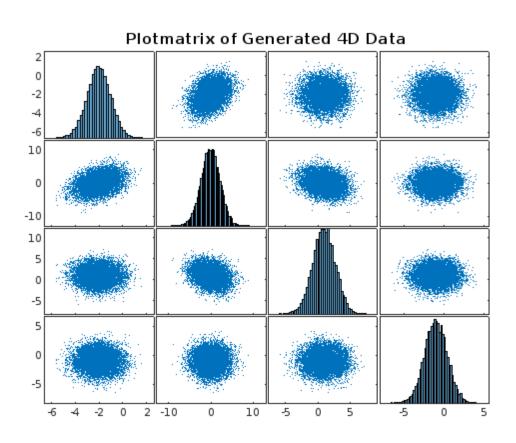


Figure 1: A visualization of 4 dimensional random data with distribution described in 1 (i).