

# SANS2021-22-HW1: Review of probability and Linear Algebra

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## 1 Discrete random variables

Generate  $M = 10000$  samples of discrete random variables with the following distributions:

- Bernoulli with  $p = p\{X = 1\} = 3/4$ ,
- Binomial with  $n = 10$  and  $p = 3/4$ ,
- Geometric with  $p = 3/4$  ( $X$  taking values in  $\{0, 1, \dots\}$ ),
- Poisson with  $\lambda = 1$ ,

For the all cases plot the graphic  $X(m)$  for  $m = 1 \dots 1000$  and the histogram.

## 2 Continuous random variables

Generate = 10000 samples of continuous random variables with the following distributions:

- Uniform in  $[0, 1]$ ,
- Exponential with  $\lambda = 1$ ,
- Gaussian with  $\mu = 1, \sigma^2 = 1$
- Gaussian with  $\mu = 1, \sigma^2 = 5$

For the three cases plot the graphic  $X(m)$  for  $m=1 \dots 1000$  and the histogram.

### 3 Law Large Numbers, LLN

Generate  $N = 100$  sets of  $M = 1000$  samples of random variables following Bernoulli(  $p = pX = 1 = 3/4$ ), Uniform in  $[0, 1]$ , Exponential with  $\lambda = 1$ , and Gaussian with  $\mu = 1$ ,  $\sigma^2 = 1$  distributions.

For every distribution, use the different generated sets for generate  $M=1000$  samples of the random variable  $X^* = \frac{\sum_{k=1}^N X_k}{N}$ . For the all cases, plot the graphic  $X(m)$  for  $m = 1 \dots 1000$  and the histogram.

### 4 Central Limit Theorem, CLT

Generate  $N = 100$  sets of  $M = 1000$  samples of random variables following Bernoulli(  $p = pX = 1 = 3/4$ ), Uniform in  $[0, 1]$ , Exponential with  $\lambda = 1$ , and Gaussian with  $\mu = 1$ ,  $\sigma^2 = 1$  distributions.

For each distribution, generate now  $N = 100$  sets of  $M = 1000$  samples of the normalized random variables  $Z = \frac{X - \mathbb{E}(X)}{\sigma}$ .

For every distribution, use the different generated sets for generate  $M = 1000$  samples of the random variable  $X^{**} = \frac{\sum_{k=1}^N Z_k}{\sqrt{N}}$ . For the all cases, plot the graphic  $Z(m)$  for  $m = 1 \dots 1000$  and the histograms.

Check that the CLT holds.

### 5 Multivariate Gaussian

Generate  $M = 10000$  samples from a multivariate gaussian distribution with  $\mu = (3, 3)$  and  $\Sigma = \begin{bmatrix} \frac{5}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{5}{2} \end{bmatrix}$ . Plot the scatterplot for  $(x_1, x_2)$ . Find the eigenvalues and eigenvectors of  $\Sigma$  and relate them with the axis of the ellipse that appears in the plot. Use a python `numpy.linalg.eig()` that returns a tuple (eigvals, eigvecs) where eigvals is a 1D NumPy array of complex numbers giving the eigenvalues of , and eigvecs is a 2D NumPy array with the corresponding eigenvectors in the columns.

### 6 Subspaces, eigenvalues and eigenvectors

Find the subspaces  $Col(A)$ ,  $Ker(A)$ ,  $Col(A^t)$ ,  $Ker(A^t)$  for the matrices:  $A = \begin{bmatrix} 2 & 2 \\ 3 & 3 \end{bmatrix}$ ,  $B = \begin{bmatrix} 2 & 1 \\ 1 & 3 \end{bmatrix}$ ,  $C = \begin{bmatrix} 2 & -2 \end{bmatrix}$ ,  $D = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$ . Plot these subspaces.

For matrices  $A$  and  $B$ , find the eigenvalues and eigenvectors using `numpy.linalg.eig()`. Plot the scatter plot of the vectors in the unit circle  $x(n) = (\cos(n\frac{2\pi}{50}), \sin(n\frac{2\pi}{50}))$

for  $n = 0..49$ . Plot also the scatter plot for the vectors  $Ax(n)$  and  $Bx(n)$ . Relate the obtained plots with the eigenvectors and eigenvalues of the matrices.

## 7 Orthogonal, symmetric and positive definite matrices

Using `numpy.linalg.eig()` find the eigenvectors and eigenvalues of:  $A = \begin{bmatrix} \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \end{bmatrix}$ ,

$$B = \begin{bmatrix} 3 & 5 \\ 5 & 2 \end{bmatrix}, C = \begin{bmatrix} 3 & 1 \\ 1 & 2 \end{bmatrix}.$$