

4SeriouslySexyStudents-Assignment#4

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1.1 EX#1.2

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In [1]: import numpy as np
import matplotlib.pyplot as plt

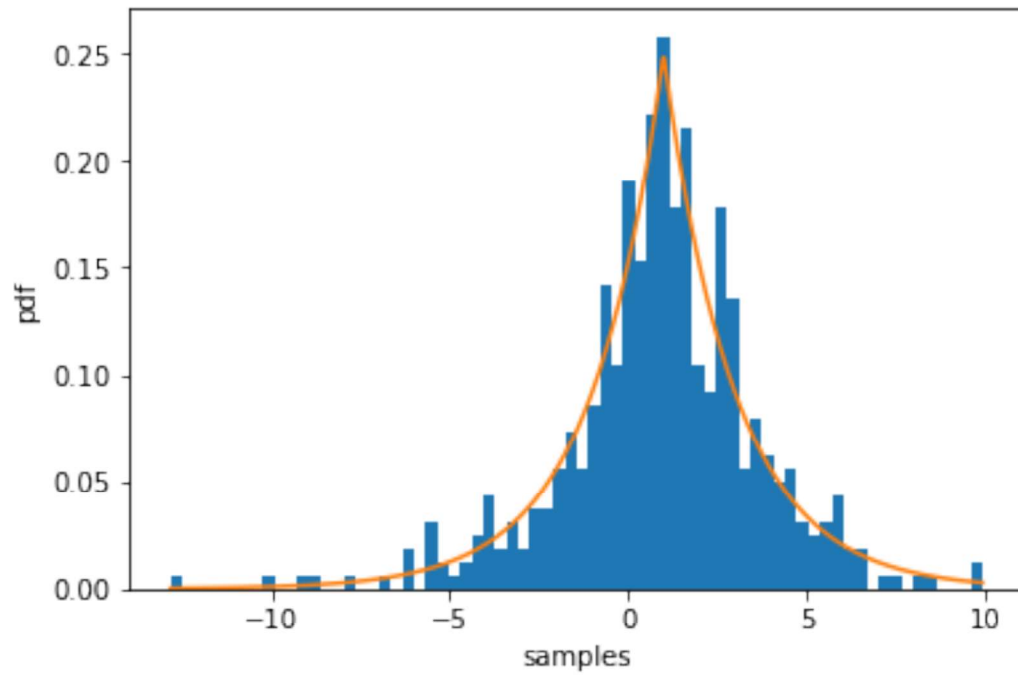
In [24]: def p(x, m=1, b=2):
    return 1 / (2 * b) * np.exp(-(np.abs(x - m) / b))

def f_inverse(y, m=1, b=2):
    return np.where(y >= 0.5, m - b * np.log(2 - 2 * y), m + b * np.log(2 * y))

samples = np.asarray([f_inverse(y) for y in np.random.random(500)])

x = np.linspace(samples.min(), samples.max(), 700)
pdf = p(x)

plt.hist(samples, bins=70, density=True)
plt.plot(x, pdf)
plt.xlabel('samples')
plt.ylabel('pdf')
plt.show()
```



Q#1

$$p_X(x) = \frac{1}{2b} \cdot e^{\frac{-(x-\mu)}{b}}$$

$$= \begin{cases} \frac{1}{2b} \cdot e^{\frac{-(x-\mu)}{b}} & x \geq \mu \\ \frac{1}{2b} \cdot e^{\frac{x-\mu}{b}} & x < \mu \end{cases}$$

\Rightarrow compute CDF

$$F_X(x) = \begin{cases} \int_{-\infty}^{\mu} \frac{1}{2b} \cdot e^{\frac{x-\mu}{b}} + \int_{\mu}^x \frac{1}{2b} \cdot e^{\frac{-(x-\mu)}{b}} & x \geq \mu \\ \int_{-\infty}^x \frac{1}{2b} \cdot e^{\frac{x-\mu}{b}} & x < \mu \end{cases} = \frac{1}{2}(1) - \frac{1}{2}e^{\frac{-(x-\mu)}{b}} + \frac{1}{2}$$
$$= \frac{1}{2} [e^{\frac{x-\mu}{b}} - 0]$$

$$= \begin{cases} 1 - \frac{1}{2} e^{\frac{-(x-\mu)}{b}} & x \geq \mu \\ \frac{1}{2} e^{\frac{x-\mu}{b}} & x < \mu \end{cases}$$

\Rightarrow Compute Inverse CDF

$$\textcircled{a} y = 1 - \frac{1}{2} e^{\frac{-(x-\mu)}{b}} \Rightarrow x = -b \cdot \ln(2-2y) + \mu$$

$$by = \frac{1}{2} e^{\frac{x-\mu}{b}} \Rightarrow x = b \cdot \ln(2y) + \mu$$

$$F_X^{-1}(x) = \begin{cases} \mu - b \cdot \ln(2-2y) & y \geq 0.5 \\ \mu + b \cdot \ln(2y) & y < 0.5 \end{cases}$$

Q #2

(a) $P_x(x) = e^{-x}$

$$u(x) = e^{-x} \Rightarrow X(u) = -\ln(y)$$

$$y = e^{-x} \Rightarrow x = -\ln(y)$$

$$P_u(x)(u) = e^{\ln(u)} \cdot \left| \frac{\partial X(u)}{\partial u} \right|$$

$$= \frac{u}{u} = 1$$

(b) $u_1 = \sqrt{2\ln x_1} \cdot \cos(2\pi x_2)$, $u_2 = \sqrt{2\ln x_1} \cdot \sin(2\pi x_2)$

$$\frac{u_2}{u_1} = \frac{\sin(2\pi x_2)}{\cos(2\pi x_2)} = \tan(2\pi x_2)$$

$$\Rightarrow x_2 = \tan^{-1}\left(\frac{u_2}{u_1}\right) * \frac{1}{2\pi}$$

$$u_1^2 + u_2^2 = (-2 \ln x_1) (\cos^2(2\pi x_2) + \sin^2(2\pi x_2))$$

$$x_1 = e^{\frac{u_1^2 + u_2^2}{-2}} \quad \text{--- (1)}$$

$$J = \begin{bmatrix} \frac{\partial x_1}{\partial u_1} & \frac{\partial x_1}{\partial u_2} \\ \frac{\partial x_2}{\partial u_1} & \frac{\partial x_2}{\partial u_2} \end{bmatrix} = \begin{bmatrix} -u_1 \cdot e^{\frac{u_1^2 + u_2^2}{-2}} & -u_2 \cdot e^{\frac{u_1^2 + u_2^2}{-2}} \\ \frac{u_2}{2\pi(u_1^2 + u_2^2)} & \frac{-u_1}{2\pi(u_1^2 + u_2^2)} \end{bmatrix}$$

$$\frac{\partial x_1}{\partial u_1} = -u_1 \cdot e^{\frac{u_1^2 + u_2^2}{-2}}$$

$$\frac{\partial x_1}{\partial u_2} = -u_2 \cdot e^{\frac{u_1^2 + u_2^2}{-2}}$$

$$\frac{\partial x_2}{\partial u_1} = \frac{u_2}{2\pi(u_1^2 + u_2^2)}$$

$$\frac{\partial x_2}{\partial u_2} = \frac{-u_1}{2\pi(u_1^2 + u_2^2)}$$

$$\det J = \frac{\partial x_1}{\partial u_1} \cdot \frac{\partial x_2}{\partial u_2} - \frac{\partial x_1}{\partial u_2} \cdot \frac{\partial x_2}{\partial u_1}$$

$$= \frac{u_1^2 \cdot e^{\frac{u_1^2 + u_2^2}{-2}}}{2\pi(u_1^2 + u_2^2)} + \frac{u_2^2 \cdot e^{\frac{u_1^2 + u_2^2}{-2}}}{2\pi(u_1^2 + u_2^2)}$$

$$= \frac{1}{2} \pi \cdot e^{\frac{u_1^2 + u_2^2}{-2}}$$

In Gaussian: $\hat{q} = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{\frac{-(x-\mu)^2}{2\sigma^2}}$

$$= \frac{1}{\sqrt{2\pi}} \cdot e^{\frac{-x^2}{2}}$$

$$\sigma = 1$$

$$\mu = 0$$