

ESMA 5015: Asignacion #1a

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Problem 1

Demuestre como generar una variable aleatoria $Cauchy(a, b)$ a partir de una variable aleatoria $u \sim unif(0, 1)$.

1. Proceso matematico

$$\begin{aligned}
 f(x) &= \frac{1}{\pi} \frac{b}{(x-a)^2 + b^2} \\
 F_X(x) &= \int_{-\infty}^x f(x) dx \\
 F_X(x) &= \int_{-\infty}^x \frac{2}{\pi} \frac{b}{(x-a)^2 + b^2} \\
 F_X(x) &= \frac{1}{\pi} \arctan\left(\frac{x-a}{b}\right) + \frac{1}{2} \\
 u &= \frac{1}{\pi} \arctan\left(\frac{x-a}{b}\right) + \frac{1}{2} \\
 u - \frac{1}{2} &= \frac{1}{\pi} \arctan\left(\frac{x-a}{b}\right) \\
 \pi\left(u - \frac{1}{2}\right) &= \arctan\left(\frac{x-a}{b}\right) \\
 \tan\left(\pi\left(u - \frac{1}{2}\right)\right) &= \frac{x-a}{b} \\
 b \tan\left(\pi\left(u - \frac{1}{2}\right)\right) + a &= x \\
 x &= a + b \tan\left(\pi\left(u - \frac{1}{2}\right)\right)
 \end{aligned}$$

2. algorithms

- (a) Generar $u \sim U(0, 1)$
- (b) Definir $x = a + b \tan(\pi(u - \frac{1}{2}))$

3. Graphics y implementacion

```

1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 rng = np.random.default_rng(seed=787)
5 x = rng.uniform(0,1,1000000)
6 a,b = 0,1
7
8 data = b * np.tan(np.pi*(x-0.5)) + a
9 data = data[(data>-25) & (data<25)]

```

Listing 1: Generacion de variable aleatoria Cauchy

```

1 s = rng.standard_cauchy(1000000)
2
3 s = s[(s>-25) & (s<25)]
4
5 plt.hist([data, s], bins=100, label=['Simulated Cauchy', 'Actual Cauchy'])
6 plt.legend(loc='upper right')
7 plt.show()

```

Listing 2: Grafica de la variable aleatoria Cauchy

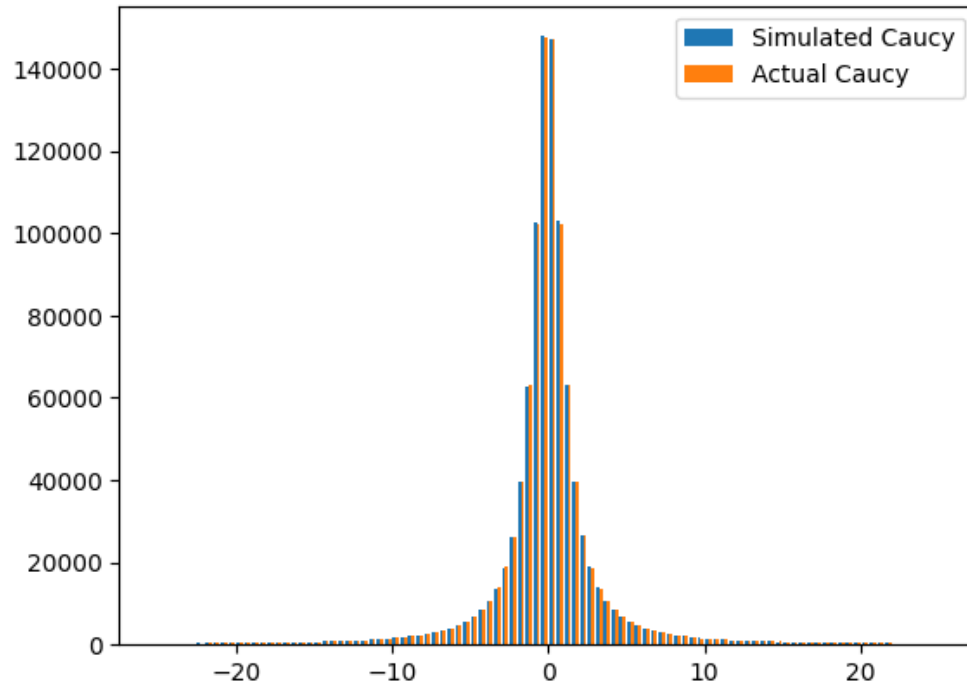


Figure 1: Histograma de la variable aleatoria Cauchy

Problem 2

Demuestre como generar una variable aleatoria $X \sim DoubleExp(\mu, \beta)$ a partir de una variable aleatoria $u \sim uni f(0, 1)$.

1. Proceso Matematico

$$\begin{aligned}
 f(x) &= \frac{1}{\beta} e^{-\frac{x-\mu}{\beta}} + e^{-\frac{x-\mu}{\beta}} \\
 F_X(x) &= \int_{-\infty}^x f(x) dx \\
 F_X(x) &= \int_{-\infty}^x \frac{1}{\beta} e^{-\frac{x-\mu}{\beta}} + e^{-\frac{x-\mu}{\beta}} \\
 F_X(x) &= e^{-e^{-\frac{x-\mu}{\beta}}} \\
 u &= e^{-e^{-\frac{x-\mu}{\beta}}} \\
 \log(-\log(u)) &= -\frac{x-\mu}{\beta} \\
 \beta \log(-\log(u)) + \mu &= x \\
 x &= \beta \log(-\log(u)) + \mu
 \end{aligned}$$

2. algorithms

- (a) Generar $u \sim U(0, 1)$
- (b) Definir $x = \beta \log(-\log(u)) + \mu$

3. Graphics

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3
4 rng = np.random.default_rng(seed=787)
5 x = rng.uniform(0,1,1000000)
6 mu, beta = 0, 0.1
7
8 data = beta*(-np.log(-np.log(x))+mu)
9 data = data[(data>-1) & (data<1)]
```

Listing 3: Generacion de variable aleatoria DoubleExp

```
1 s = rng.gumbel(mu, beta, 1000000)
2 s = s[(s>-1) & (s<1)]
3
4
5 plt.hist([data, s], bins=100, label=['Simulated Gumbel ', 'Actual Gumbel'])
6 plt.legend(loc='upper right')
7 plt.show()
```

Listing 4: Grafica de la variable aleatoria DoubleExp

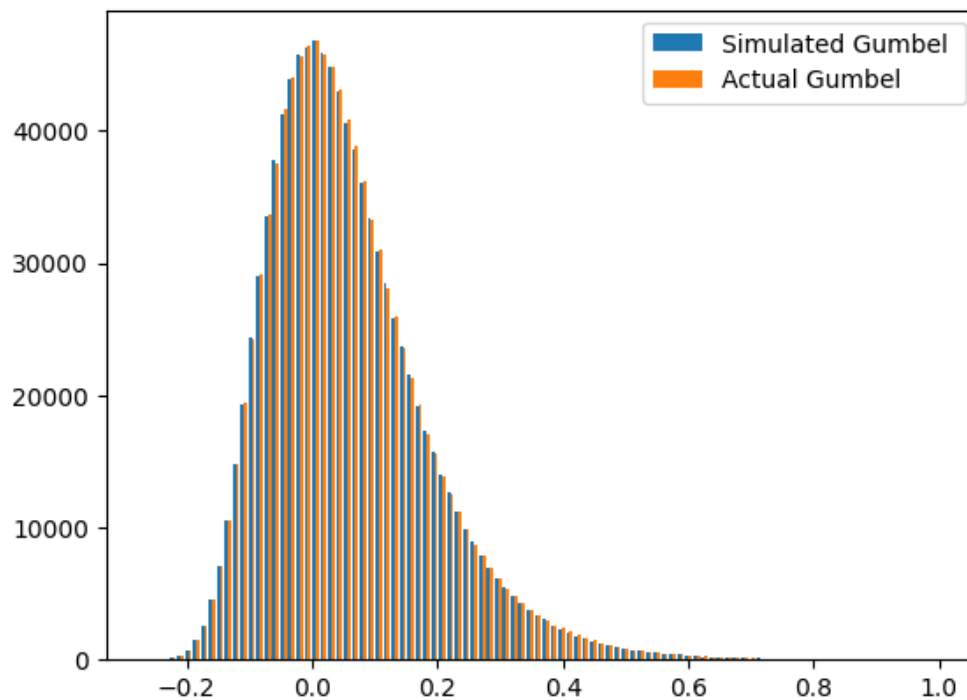


Figure 2: Histograma de la variable aleatoria DoubleExp