

Inconsolata

Metodo de la Trasformada Inversa

Curso: Temas Selectos I: O25 LAT4032 1

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Abstract

This document provides a template for reports in the "AI in Financial Services" course, using EB Garamond for prose and Libertinus Math for formulas. It includes a cover page, abstract, table of contents, and sample sections for math and text. Additional content demonstrates tables, code, and references.

Contents

Overview

This document is a minimal example using EB Garamond for prose and `libertinust1math` for formulas. Links lik are active.

```
[1]: import math
import random
from collections import Counter

import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[2]: color = sns.color_palette("muted")
np.random.shuffle(color)
sns.set(style="whitegrid", context="paper", palette=color)
sns.color_palette()
```

```
[3]: [(0.5843137254901961, 0.4235294117647059, 0.7058823529411765),
(0.5490196078431373, 0.3803921568627451, 0.23529411764705882),
(0.8627450980392157, 0.49411764705882355, 0.7529411764705882),
(0.2823529411764706, 0.47058823529411764, 0.8156862745098039),
(0.41568627450980394, 0.8, 0.39215686274509803),
(0.4745098039215686, 0.4745098039215686, 0.4745098039215686),
(0.8392156862745098, 0.37254901960784315, 0.37254901960784315),
(0.9333333333333333, 0.5215686274509804, 0.2901960784313726),
(0.8352941176470589, 0.7333333333333333, 0.403921568627451),
(0.5098039215686274, 0.7764705882352941, 0.8862745098039215)]
```

Distribución Uniforme (a, b)

Sea $U \sim \text{Unif}(0, 1)$. Si $X \sim \text{Unif}(a, b)$, entonces su función de distribución acumulada es:

$$F_X(x) = \frac{x-a}{b-a} \mathbf{1}_{[a,b]}(x) + \mathbf{1}_{(b,\infty)}(x)$$

Encontrando la inversa:

$$\begin{aligned} F_X(x) = u &\iff \frac{x-a}{b-a} = u, \\ &\iff x-a = (b-a)u, \\ &\iff x = a + (b-a)u. \end{aligned}$$

Entonces:

$$F_X^{-1}(u) = a + (b-a)u.$$

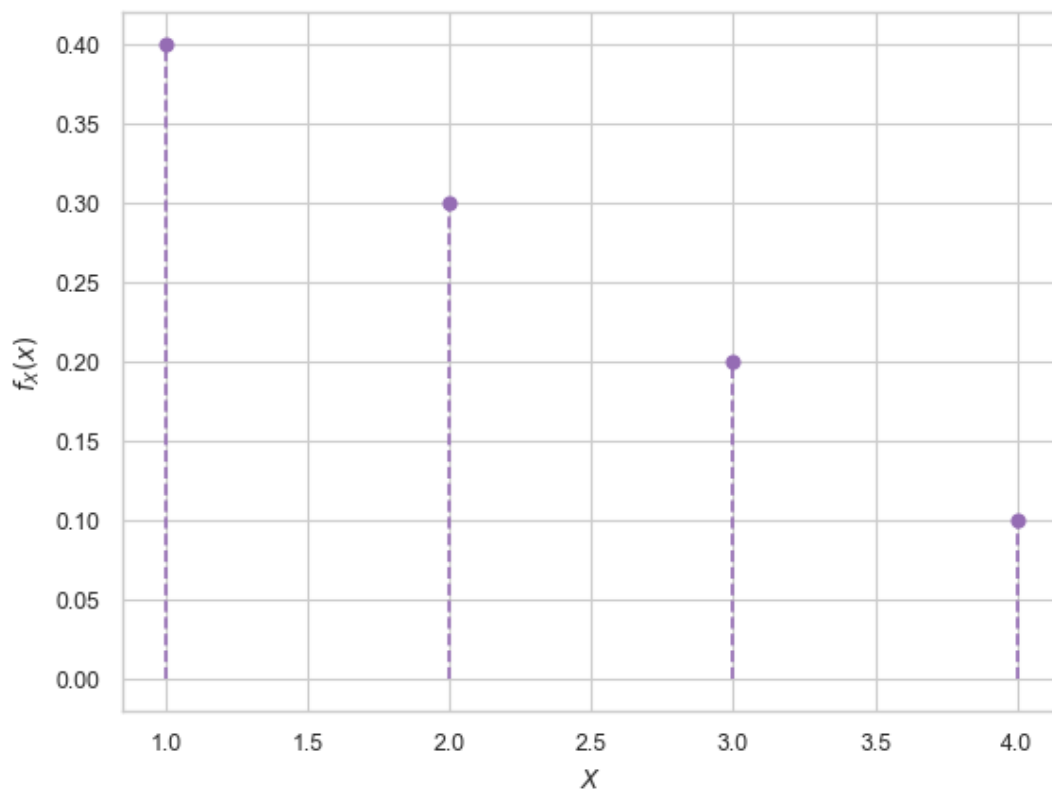
Distribución Discreta

a)

$$f_X(x) = P(X = x) = \begin{cases} 0.4, & x = 1, \\ 0.3, & x = 2, \\ 0.2, & x = 3, \\ 0.1, & x = 4, \\ 0, & \text{en otro caso} \end{cases}$$

```
[3]: x_vals = [1, 2, 3, 4]
pmf = [0.4, 0.3, 0.2, 0.1]

plt.vlines(x_vals, 0, pmf, linestyles='--')
plt.plot(x_vals, pmf, 'o')
plt.xlabel('$X$')
plt.ylabel('$f_X(x)$')
plt.show()
```

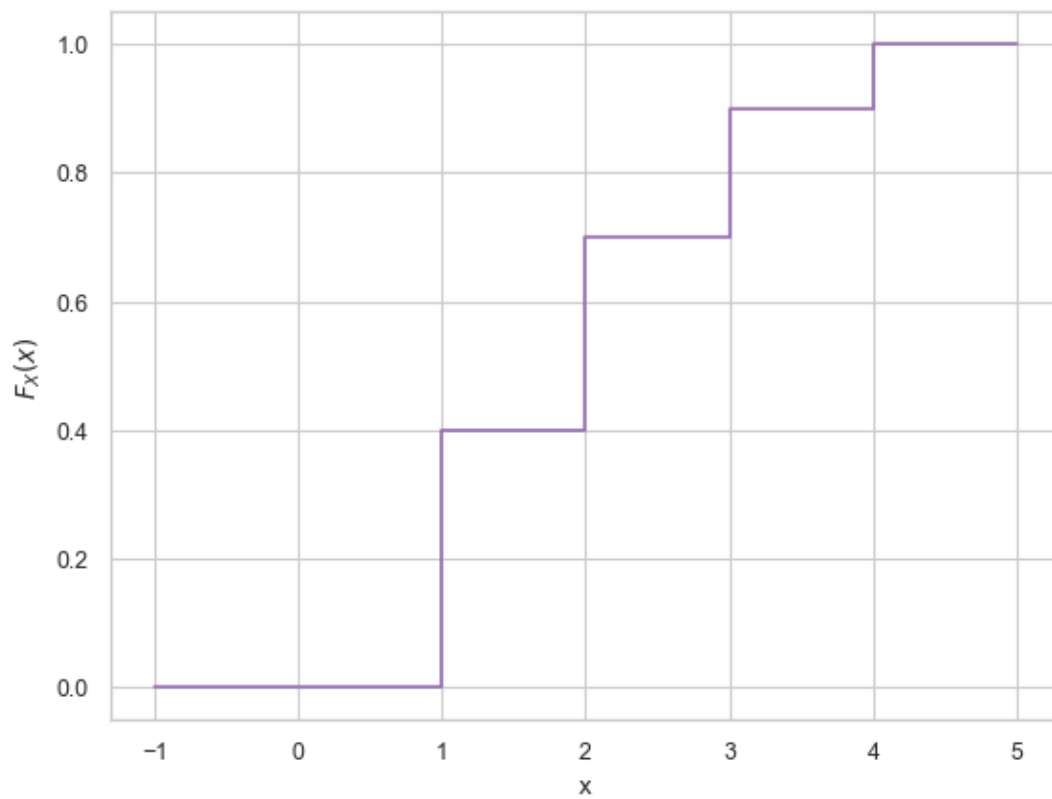


b)

$$F_X(x) = \mathbb{P}(X \leq x) = \begin{cases} 0, & x < 1, \\ 0.4, & 1 \leq x < 2, \\ 0.7, & 2 \leq x < 3, \\ 0.9, & 3 \leq x < 4, \\ 1, & x \geq 4. \end{cases}$$

```
[4]: x_cdf = [-1, 1, 2, 3, 4, 5]
F_cdf = [0, 0.4, 0.7, 0.9, 1, 1]

plt.step(x_cdf, F_cdf, where='post')
plt.xlabel('x')
plt.ylabel('$F_X(x)$')
plt.show()
```



c)

Sea $U \sim \text{Unif}(0, 1)$.

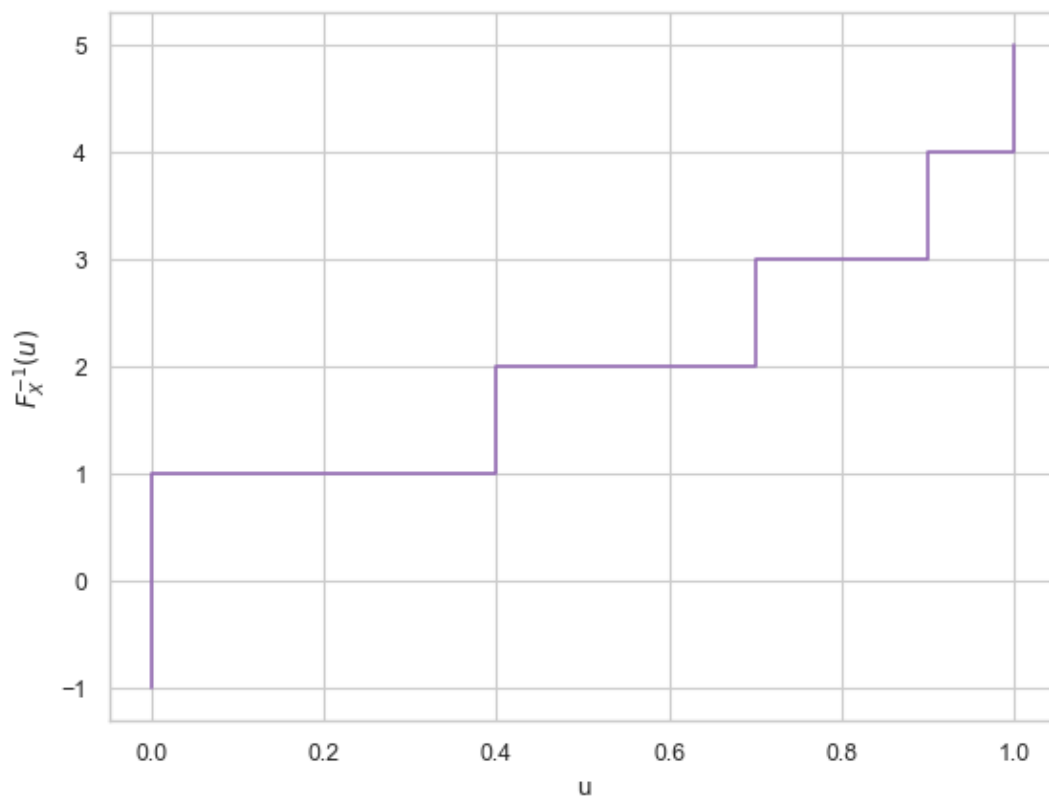
$$\begin{aligned}
 F_X(x) = u &\iff u \in (0, 0.4] \Rightarrow x = 1, \\
 &\iff u \in (0.4, 0.7] \Rightarrow x = 2, \\
 &\iff u \in (0.7, 0.9] \Rightarrow x = 3, \\
 &\iff u \in (0.9, 1] \Rightarrow x = 4.
 \end{aligned}$$

Entonces:

$$F_X^{-1}(u) = \begin{cases} 1, & 0 < u \leq 0.4, \\ 2, & 0.4 < u \leq 0.7, \\ 3, & 0.7 < u \leq 0.9, \\ 4, & 0.9 < u \leq 1. \end{cases}$$

```
[5]: u_vals = [0, 0, 0.4, 0.7, 0.9, 1]
F_inv_vals = [-1, 1, 2, 3, 4, 5]

plt.step(u_vals, F_inv_vals, where="post")
plt.xlabel("u")
plt.ylabel("$F_X^{-1}(u)$")
plt.show()
```

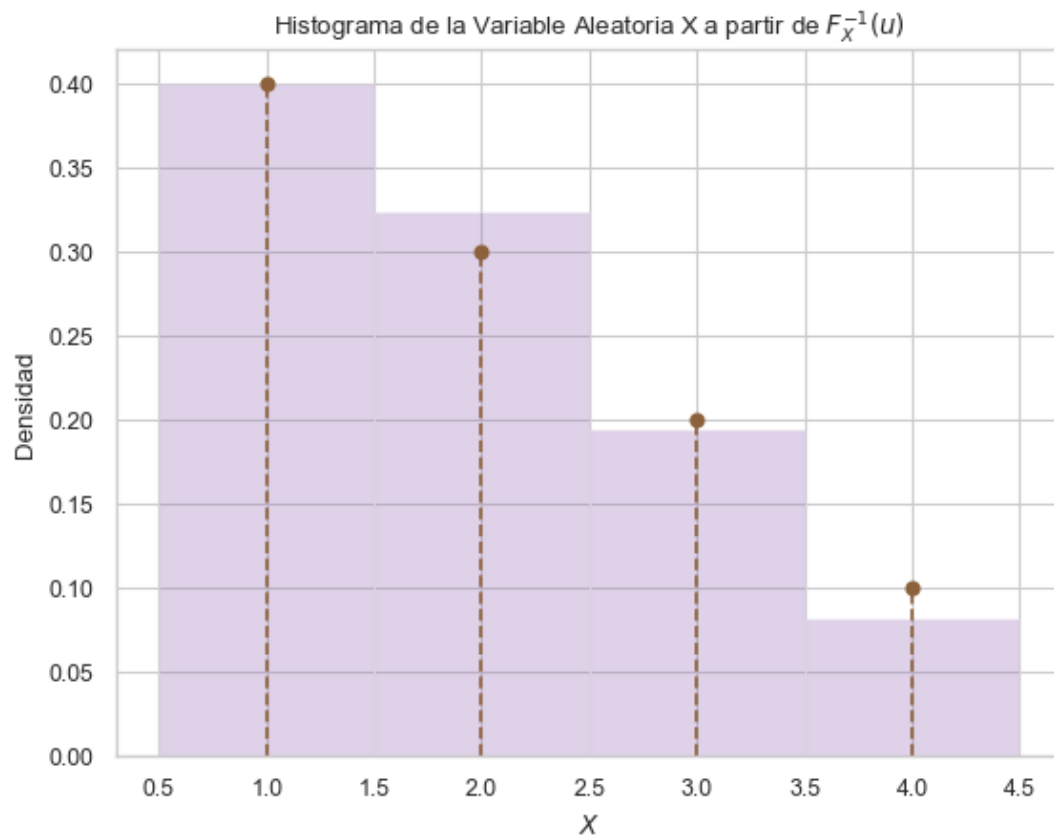


d)

```
[6]: def F_inv(u):  
    if u <= 0.4:  
        return 1  
    elif u <= 0.7:  
        return 2  
    elif u <= 0.9:  
        return 3  
    else:  
        return 4  
  
u_samples = np.random.uniform(0, 1, 500)  
x_samples = [F_inv(u) for u in u_samples]
```

e)

```
[7]: plt.hist(x_samples, bins=np.arange(0.5, 5.5, 1), density=True, alpha=0.3)
plt.vlines(x_vals, 0, pmf, linestyles='--', color=color[1])
plt.plot(x_vals, pmf, 'o', color=color[1])
plt.xlabel("$X$")
plt.ylabel("Densidad")
plt.title("Histograma de la Variable Aleatoria X a partir de  $F_X^{-1}(u)$ ")
plt.show()
```



Exponencial $\text{Exp}(\lambda)$

Sea $U \sim \text{Unif}(0, 1)$. Si $X \sim \text{Exp}(\lambda)$, entonces su función de distribución acumulada es:

$$F_X(x) = (1 - e^{-\lambda x}) \mathbf{1}_{[0, \infty)}(x)$$

Encontrando la inversa:

$$\begin{aligned} F_X(x) = u &\iff 1 - e^{-\lambda x} = u, \\ &\iff e^{-\lambda x} = 1 - u, \\ &\iff -\lambda x = \ln(1 - u), \\ &\iff x = -\frac{1}{\lambda} \ln(1 - u). \end{aligned}$$

Entonces:

$$F_X^{-1}(u) = -\frac{1}{\lambda} \ln(1 - u).$$

Por lo tanto, con $U \sim \text{Unif}(0, 1)$,

$$X = F_X^{-1}(U) = -\frac{1}{\lambda} \ln(1 - U) \sim \text{Exp}(\lambda).$$

Weibull (r, λ)

Sea $U \sim \text{Unif}(0, 1)$, con $r > 0$ y $\lambda > 0$. Si $X \sim \text{Weibull}(r, \lambda)$, entonces su funcion de distribucion acumulada es:

$$F_X(x) = (1 - e^{-(\lambda x)^r}) \mathbf{1}_{[0, \infty)}(x)$$

Encontrando la inversa:

$$\begin{aligned} F_X(x) = u &\iff 1 - e^{-(\lambda x)^r} = u, \\ &\iff e^{-(\lambda x)^r} = 1 - u, \\ &\iff -(\lambda x)^r = \ln(1 - u), \\ &\iff (\lambda x)^r = -\ln(1 - u), \\ &\iff x = \frac{1}{\lambda} [-\ln(1 - u)]^{1/r}. \end{aligned}$$

Entonces:

$$F_X^{-1}(u) = \frac{1}{\lambda} [-\ln(1 - u)]^{1/r}.$$

Por lo tanto, con $U \sim \text{Unif}(0, 1)$:

$$X = F_X^{-1}(U) = \frac{1}{\lambda} [-\ln(1 - U)]^{1/r} \sim \text{Weibull}(r, \lambda).$$

Cauchy (a, b)

Sea $U \sim \text{Unif}(0, 1)$. Si $X \sim \text{Cauchy}(a, b)$, entonces su función de distribución acumulada es:

$$F_X(x) = \frac{1}{\pi} \arctan\left(\frac{x-a}{b}\right) + \frac{1}{2}, \quad x \in \mathbb{R}, \quad b > 0.$$

Encontrando la inversa:

$$\begin{aligned} F_X(x) = u &\iff \frac{1}{\pi} \arctan\left(\frac{x-a}{b}\right) + \frac{1}{2} = u, \\ &\iff \arctan\left(\frac{x-a}{b}\right) = \pi\left(u - \frac{1}{2}\right), \\ &\iff \frac{x-a}{b} = \tan\left(\pi\left(u - \frac{1}{2}\right)\right), \\ &\iff x = a + b \tan\left(\pi\left(u - \frac{1}{2}\right)\right). \end{aligned}$$

Entonces:

$$F_X^{-1}(u) = a + b \tan\left(\pi\left(u - \frac{1}{2}\right)\right).$$

Por lo tanto, $X = F_X^{-1}(U) = a + b \tan\left(\pi\left(U - \frac{1}{2}\right)\right) \sim \text{Cauchy}(a, b)$.

Pareto I (a, b)

Sea $U \sim \text{Unif}(0, 1)$, con $a > 0$ y $b > 0$. Si $X \sim \text{Pareto I}(a, b)$, entonces su función de distribución acumulada es:

$$F_X(x) = (1 - (b/x)^a) \mathbf{1}_{[b, \infty)}(x)$$

Encontrando la inversa:

$$\begin{aligned} F_X(x) = u &\iff 1 - \left(\frac{b}{x}\right)^a = u, \\ &\iff \left(\frac{b}{x}\right)^a = 1 - u, \\ &\iff \frac{b}{x} = (1 - u)^{1/a}, \\ &\iff x = b(1 - u)^{-1/a}. \end{aligned}$$

Entonces:

$$F_X^{-1}(u) = b(1 - u)^{-1/a}.$$

Por lo tanto,

$$X = F_X^{-1}(U) = b(1 - U)^{-1/a} \sim \text{Pareto I}(a, b).$$

Mínimo $X_{(1)} = \min\{X_1, \dots, X_n\}$

Sea $U \sim \text{Unif}(0, 1)$ y $X_{(1)} := \min\{X_1, \dots, X_n\}$ con X_i i.i.d. de CDF F .

$$F_{X_{(1)}}(x) = \mathbb{P}(X_{(1)} \leq x) = 1 - \mathbb{P}(X_1 > x, \dots, X_n > x) = 1 - (1 - F(x))^n.$$

Encontrando la inversa:

$$\begin{aligned} F_{X_{(1)}}(x) = u &\iff 1 - (1 - F(x))^n = u, \\ &\iff (1 - F(x))^n = 1 - u, \\ &\iff 1 - F(x) = (1 - u)^{1/n}, \\ &\iff F(x) = 1 - (1 - u)^{1/n}, \\ &\iff x = F^{-1}(1 - (1 - u)^{1/n}). \end{aligned}$$

Entonces:

$$F_{X_{(1)}}^{-1}(u) = F^{-1}(1 - (1 - u)^{1/n}), \quad 0 < u < 1.$$

Mixta $X = \min\{Y, M\}$ con $Y \sim \text{Exp}(\lambda)$

Caracteriza la cdf, localiza la masa en $x = M$ y construye F^{-1} con caso discreto/continuo. Implementa `sample_min_exp_M(lam, M, N)`.

```
[8]: # TODO: implementa aquí
```

Mixta $X = \max\{Y, M\}$ con $Y \sim \text{Exp}(\lambda)$

Sea $Y \sim \text{Exp}(\lambda)$ y $M > 0$. Defina $X = \max\{Y, M\}$.

CDF

$$F_X(x) = \mathbb{P}(X \leq x) = \begin{cases} 0, & x < M, \\ 1 - e^{-\lambda x}, & x \geq M, \end{cases}$$

con salto en $x = M$ de tamaño $1 - e^{-\lambda M}$.

Inversa (cuantil generalizado) $F_X^{-1}(u) = \inf\{x : F_X(x) \geq u\}$

Sea $U \sim \text{Unif}(0, 1)$.

$$F_X^{-1}(u) = \begin{cases} M, & 0 < u \leq 1 - e^{-\lambda M}, \\ -\frac{1}{\lambda} \ln(1 - u), & 1 - e^{-\lambda M} < u < 1, \end{cases}$$

y además $F_X^{-1}(0^+) = M$ y $F_X^{-1}(1) = +\infty$.

Para muestrear: si $U \leq 1 - e^{-\lambda M}$ devuelve M ; en caso contrario devuelve $-\frac{1}{\lambda} \ln(1 - U)$.

c) **Probar** $F(F^{-1}(u)) \geq u$, $0 < u < 1$. Defina el cuantil generalizado $F^{-1}(u) := \inf\{x : F(x) \geq u\}$. Sea $S_u = \{x : F(x) \geq u\}$. Para todo $\varepsilon > 0$ existe $x_\varepsilon \in S_u$ con $x_\varepsilon \leq F^{-1}(u) + \varepsilon$. Entonces

$$F(F^{-1}(u) + \varepsilon) \geq u.$$

Por derecha-continuidad,

$$F(F^{-1}(u)) = \lim_{\varepsilon \downarrow 0} F(F^{-1}(u) + \varepsilon) \geq u.$$

Verificación para $X = \max\{Y, M\}$. Con $p_0 := 1 - e^{-\lambda M}$,

$$F^{-1}(u) = \begin{cases} M, & 0 < u \leq p_0, \\ -\frac{1}{\lambda} \ln(1 - u), & p_0 < u < 1. \end{cases}$$

Luego $F(F^{-1}(u)) = F(M) = p_0 \leq u$ si $u \leq p_0$, y $F(-\frac{1}{\lambda} \ln(1 - u)) = u$ si $u > p_0$.

d) **Probar** $F^{-1}(F(x)) \leq x$ cuando $0 < F(x) < 1$. Tome $u = F(x)$. El conjunto $S_u = \{t : F(t) \geq u\}$ contiene a x (trivialmente $F(x) \geq u$). Por lo tanto

$$F^{-1}(F(x)) = \inf S_u \leq x.$$

La igualdad se da cuando F es continua en x .

Verificación para $X = \max\{Y, M\}$. Si $x > M$,

$$F^{-1}(F(x)) = -\frac{1}{\lambda} \ln(1 - (1 - e^{-\lambda x})) = x.$$

Si $x = M$, $F^{-1}(F(M)) = F^{-1}(p_0) = M = x$.

e) **Generación por transformada inversa.** Con $p_0 := 1 - e^{-\lambda M}$ y $U \sim \text{Unif}(0, 1)$:

$$X = \begin{cases} M, & U \leq p_0, \\ -\frac{1}{\lambda} \ln(1 - U), & U > p_0. \end{cases}$$

Es todo.

Variable con CDF por tramos

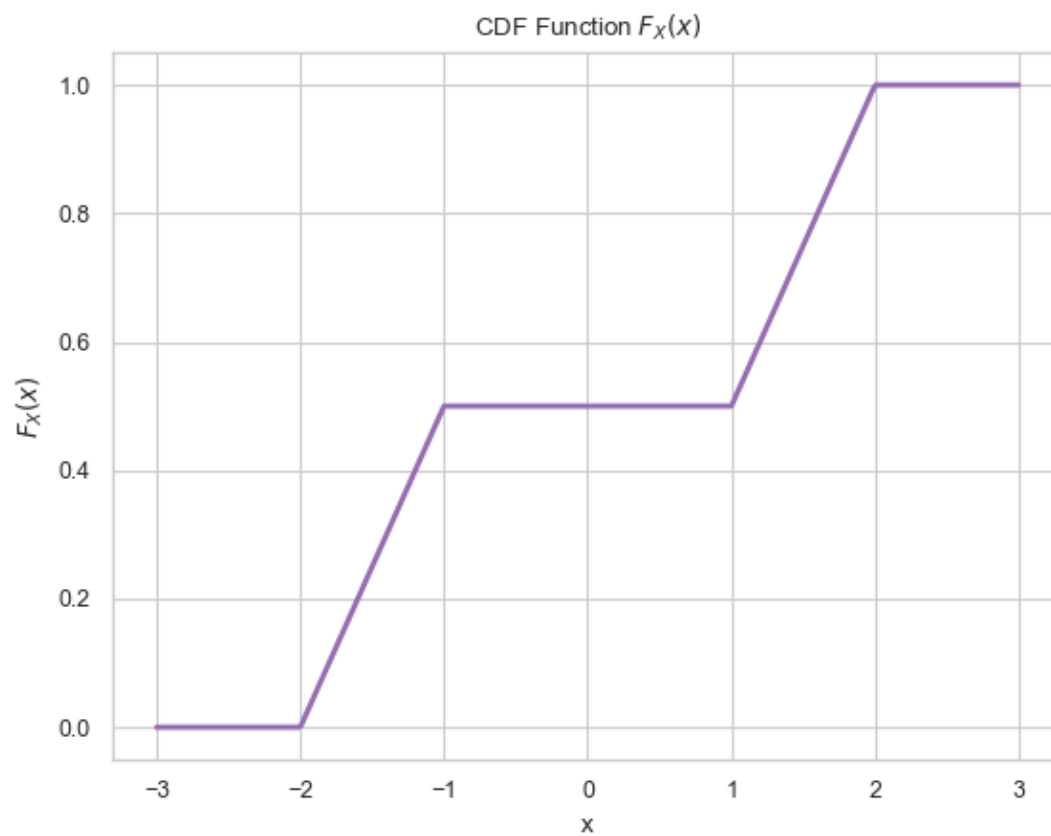
a)

$$F_X(x) = \begin{cases} 0, & x \leq -2, \\ \frac{x+2}{2}, & -2 < x < -1, \\ \frac{1}{2}, & -1 \leq x < 1, \\ \frac{x}{2}, & 1 \leq x < 2, \\ 1, & x \geq 2. \end{cases}$$

```
[23]: # Define a fine grid of x values
x = np.linspace(-3, 3, 400)

# Define the piecewise function for F_X(x)
F_x = np.piecewise(x,
                    [x <= -2, (x > -2) & (x < -1), (x >= -1) & (x < 1), (x >= 1) &
                    ↪(x < 2), x >= 2],
                    [0, lambda x: (x + 2) / 2, 0.5, lambda x: x / 2, 1])

# Plot the CDF
plt.plot(x, F_x, lw=2, color='blue')
plt.xlabel('x')
plt.ylabel('$F_X(x)$')
plt.title('CDF Function $F_X(x)$')
plt.grid(True)
plt.show()
```



b)

Sea $U \sim \text{Unif}(0, 1)$.

Encontrando la inversa:

- Para $x \leq -2, u \in \{0\}$: $F_X^{-1}(0) = -2$ (convención en el extremo).
- Para $-2 < x < -1, u \in (0, \frac{1}{2})$:

$$\begin{aligned} F_X(x) = u &\iff \frac{x+2}{2} = u, \\ &\iff x = 2u - 2. \end{aligned}$$

- Para $-1 \leq x < 1, u \in \{\frac{1}{2}\}$: meseta $\Rightarrow F_X^{-1}(\frac{1}{2}) = -1$ (borde izquierdo).
- Para $1 \leq x < 2, u \in (\frac{1}{2}, 1)$:

$$\begin{aligned} F_X(x) = u &\iff \frac{x}{2} = u, \\ &\iff x = 2u. \end{aligned}$$

- Para $x \geq 2, u \in \{1\}$: $F_X^{-1}(1) = 2$.

Entonces:

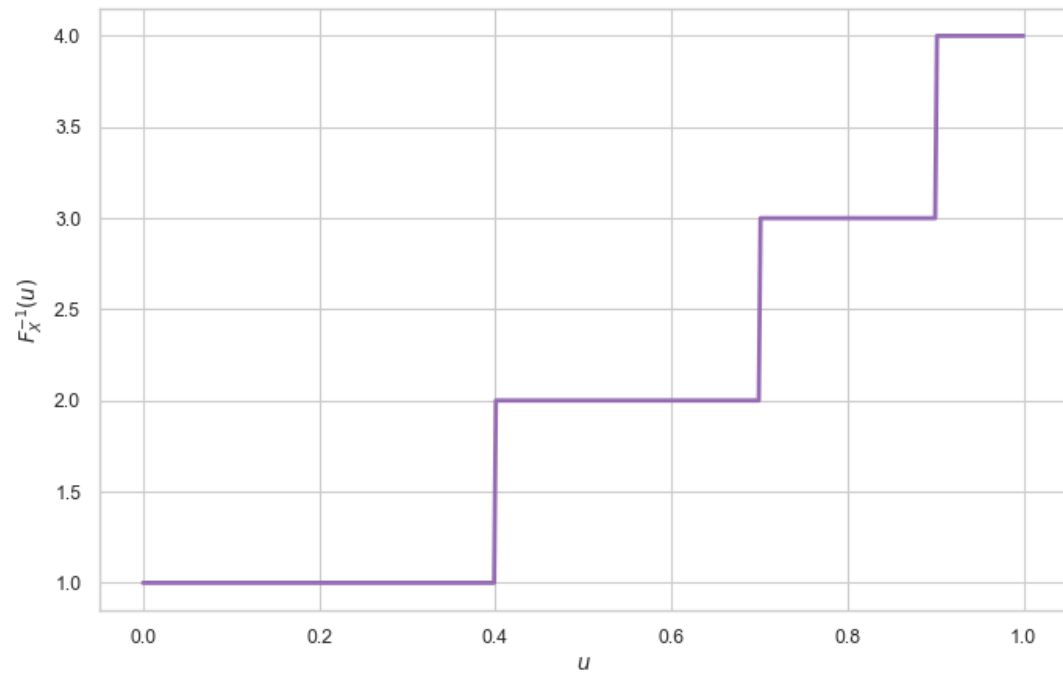
$$F_X^{-1}(u) = \begin{cases} -2, & u = 0, \\ 2u - 2, & 0 < u < \frac{1}{2}, \\ -1, & u = \frac{1}{2}, \\ 2u, & \frac{1}{2} < u < 1, \\ 2, & u = 1. \end{cases}$$

```
[11]: # Create a fine grid of u values in [0,1] ensuring u=0 and u=1 are included
u = np.linspace(0, 1, 500)

def F_inv_func(u):
    if u <= 0.4:
        return 1
    elif u <= 0.7:
        return 2
    elif u <= 0.9:
        return 3
    else:
        return 4

F_inv = np.vectorize(F_inv_func)(u)

plt.figure(figsize=(8, 5))
plt.plot(u, F_inv, color='red', lw=2)
plt.xlabel('$u$')
plt.ylabel('$F_X^{-1}(u)$')
plt.show()
```

Bernoulli (p) desde $U(0, 1)$

Sea $U \sim \text{Unif}(0, 1)$ y $0 < p < 1$. Defina

$$X = \mathbf{1}_{(0,p]}(U) = \begin{cases} 1, & U \leq p, \\ 0, & U > p. \end{cases}$$

$$\mathbb{P}(X = 1) = \mathbb{P}(U \leq p) = p, \quad \mathbb{P}(X = 0) = \mathbb{P}(U > p) = 1 - p,$$

usando que $\mathbb{P}(U = p) = 0$. Por tanto $X \sim \text{Bernoulli}(p)$.

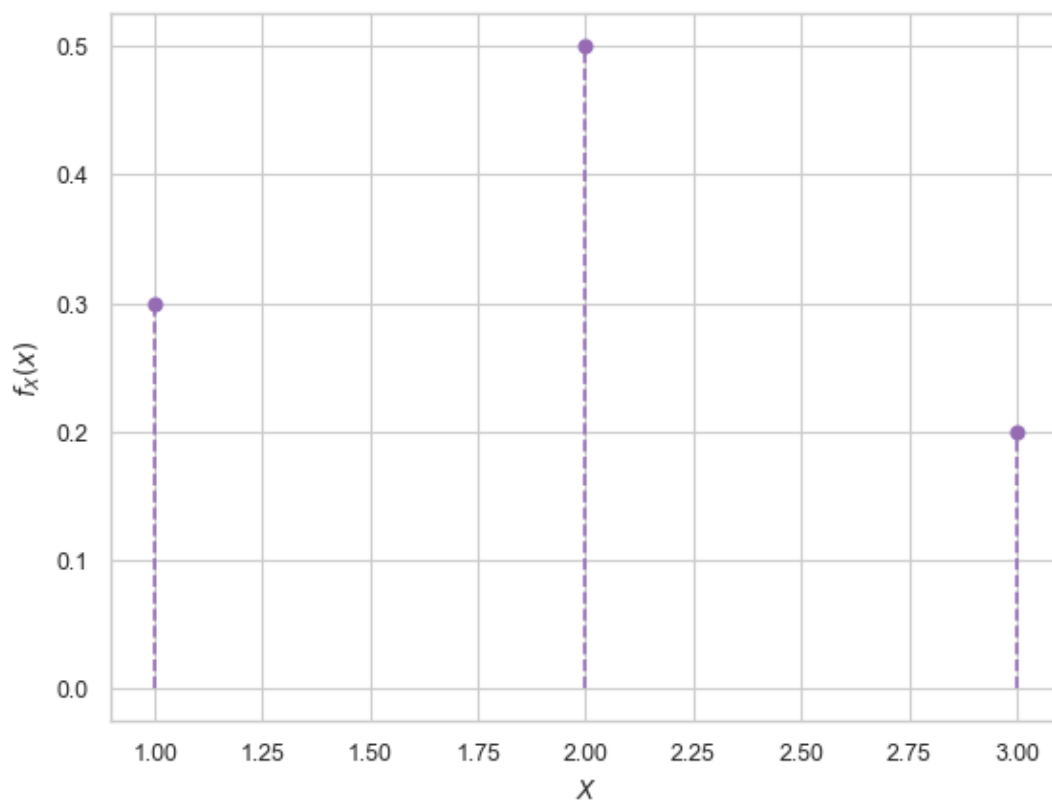
Variable aleatoria discreta

a)

$$f_X(x) = \begin{cases} 0.3 & x = 1, \\ 0.5, & x = 2, \\ 0.2, & x = 3, \\ 0, & \text{en otro caso.} \end{cases}$$

```
[12]: x_vals = [1, 2, 3]
      pmf = [0.3, 0.5, 0.2]

      plt.vlines(x_vals, 0, pmf, linestyle='--')
      plt.plot(x_vals, pmf, 'o')
      plt.xlabel('$X$')
      plt.ylabel('$f_X(x)$')
      plt.show()
```

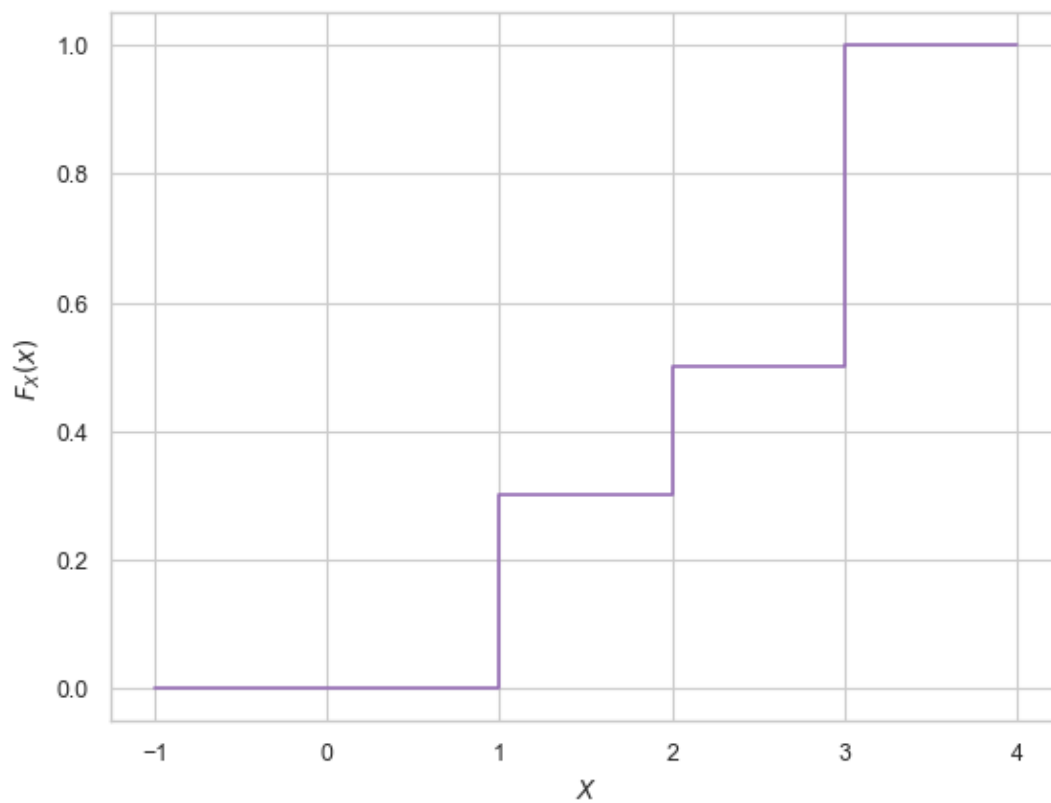


b)

$$F_X(x) = \mathbb{P}(X \leq x) = \begin{cases} 0, & x < 1, \\ 0.3, & 1 \leq x < 2, \\ 0.8, & 2 \leq x < 3, \\ 1, & x \geq 3. \end{cases}$$

```
[13]: x_cdf = [-1, 1, 2, 3, 4]
      F_cdf = [0, 0.3, 0.5, 1, 1]

      plt.step(x_cdf, F_cdf, where='post', linestyle='-')
      plt.xlabel('$X$')
      plt.ylabel('$F_X(x)$')
      plt.show()
```



c)

Sea $U \sim \text{Unif}(0, 1)$.

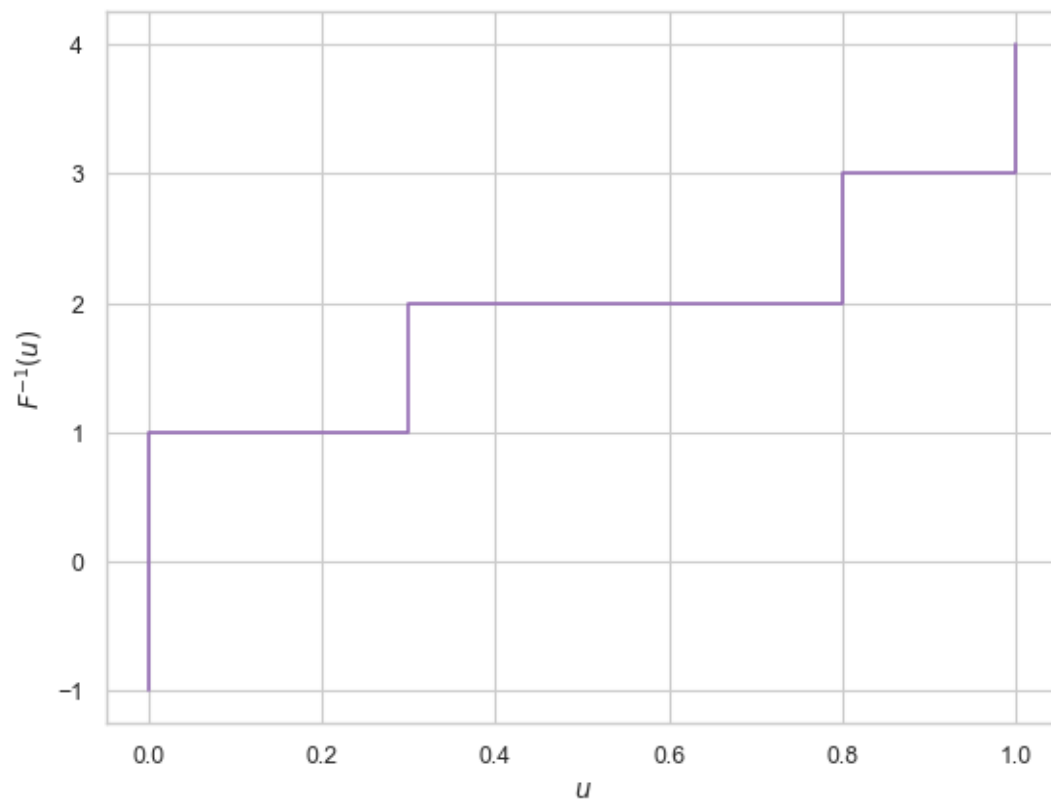
$$\begin{aligned}
 F_X(x) = u &\iff u \in (0, 0.3] \Rightarrow x = 1, \\
 &\iff u \in (0.3, 0.8] \Rightarrow x = 2, \\
 &\iff u \in (0.8, 1] \Rightarrow x = 3.
 \end{aligned}$$

Entonces:

$$F_X^{-1}(u) = \begin{cases} 1, & 0 < u \leq 0.3, \\ 2, & 0.3 < u \leq 0.8, \\ 3, & 0.8 < u \leq 1. \end{cases}$$

```
[14]: u_vals = [0, 0, 0.3, 0.8, 1.0]
      F_inv_values = [-1, 1, 2, 3, 4]

      plt.step(u_vals, F_inv_values, where='post')
      plt.xlabel('$u$')
      plt.ylabel('$F^{-1}(u)$')
      plt.show()
```



d) Valor esperado

$$E[X] = \sum_{x=1}^3 x \cdot f_X(x) = 1 \cdot 0.3 + 2 \cdot 0.5 + 3 \cdot 0.2 = 1.9.$$

e) Varianza

$$\mathbb{V}[X] = \sum_{x=1}^3 (x - \mathbb{E}[X])^2 \cdot f_X(x) = (1 - 1.9)^2 \cdot 0.3 + (2 - 1.9)^2 \cdot 0.5 + (3 - 1.9)^2 \cdot 0.2 = 0.49.$$

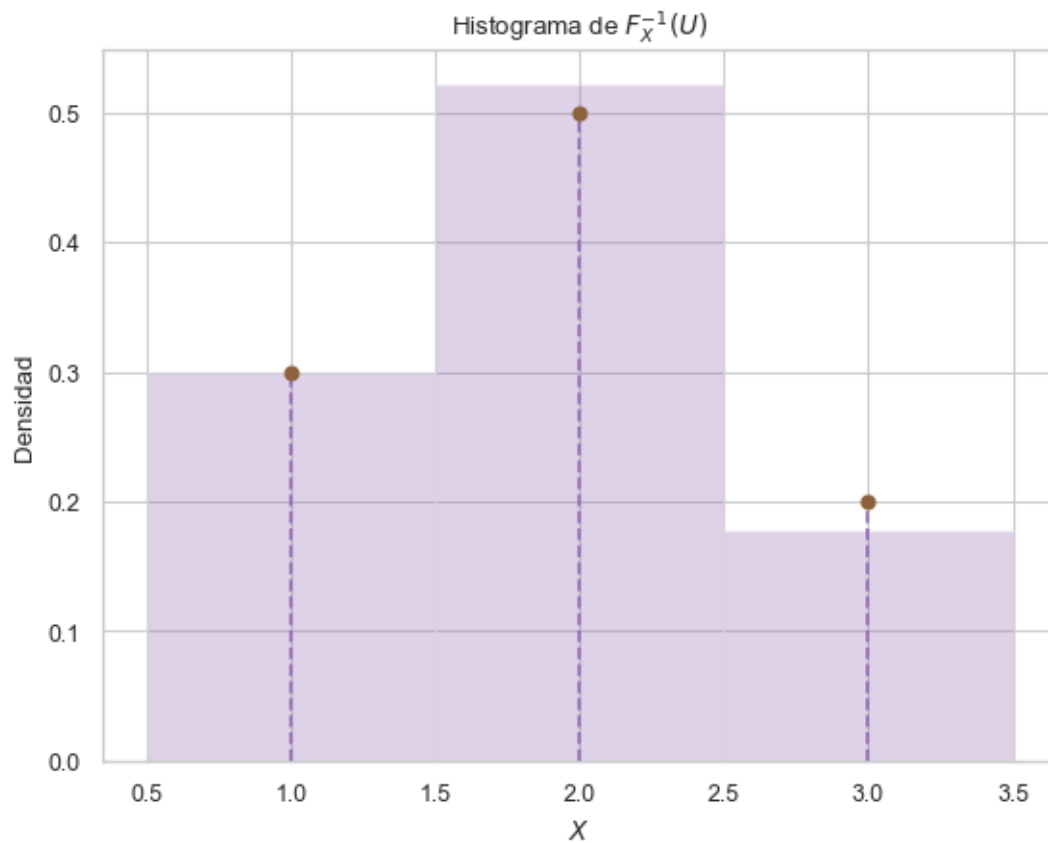
$$\mathbb{V}[X] = \mathbb{E}[X^2] - (\mathbb{E}[X])^2 = 0.3 \cdot 1^2 + 0.5 \cdot 2^2 + 0.2 \cdot 3^2 - (1.9)^2 = 2.79 - (1.9)^2 = 0.49.$$

f)

```
[15]: def F_inv(u):  
    if 0.3 < u <= 0.8:  
        return 2  
    elif u <= 0.3:  
        return 1  
    else:  
        return 3  
  
[16]: u_samples = np.random.uniform(0, 1, 500)  
    x_samples = [F_inv(u) for u in u_samples]
```

g)

```
[17]: plt.hist(x_samples, bins=np.arange(0.5, 4.5, 1), density=True, alpha=0.3)
plt.vlines(x_vals, 0, pmf, linestyle='--')
plt.plot(x_vals, pmf, 'o')
plt.xlabel("$X$")
plt.ylabel("Densidad")
plt.title("Histograma de  $F_X^{-1}(U)$ ")
plt.show()
```



h)

```
[18]: x_mean = np.mean(x_samples)
      print(f"Media muestral de X: {x_mean}")
      print(f"Media teórica de X: 1.9")
      print(f"Diferencia: {x_mean - 1.9:.4f}")
```

Media muestral de X: 1.878

Media teórica de X: 1.9

Diferencia: -0.0220

i)

```
[19]: x_var = np.var(x_samples)
      print(f"Varianza muestral de X: {x_var}")
      print(f"Varianza teórica de X: 0.49")
      print(f"Diferencia: {x_var - 0.49:.4f}")
```

Varianza muestral de X: 0.463116

Varianza teórica de X: 0.49

Diferencia: -0.0269

Binomial, Geométrica y Poisson

Implementa generadores por inversión para: - $X \sim \text{Bin}(m, p)$ con $m = 10$ y $p = 1/3$ - $X \sim \text{Geo}(p)$ con $\mathbb{P}(X = k) = p(1 - p)^{k-1}$, $k \geq 1$ con $p = 3/4$ - $X \sim \text{Poisson}(\lambda)$ con $\lambda = 2$

a) Binomial

```
[20]: def rbinom_inv(ntrials: int, p: float, size: int):
    """Binomial(ntrials, p) por transformada inversa."""
    if not (0 <= p <= 1):
        raise ValueError("p debe estar en [0,1].")
    if p == 0:
        return [0]*size
    if p == 1:
        return [ntrials]*size

    c = p/(1.0 - p)                # paso 2: c
    out = []
    for _ in range(size):
        U = random.random()        # paso 1
        i = 0                      # paso 2
        pr = (1.0 - p)**ntrials     # pr = P(X=0)
        F = pr
        if U < F:                  # paso 3-5
            out.append(i); continue
        while True:                # paso 6
            pr *= c * (ntrials - i) / (i + 1) # paso 7
            F += pr
            i += 1
            if U < F or i == ntrials:      # paso 8-11 (con tope)
                out.append(i)
                break
    return out

m = rbinom_inv(ntrials=10, p=0.3, size=100)
print(m[:20])
```

[4, 3, 2, 3, 3, 2, 3, 2, 3, 3, 4, 0, 5, 6, 2, 3, 3, 3, 4, 1]

b) Geometrica

```
[21]: def rgeom_inv_trials(p: float, n: int):  
    """Geom(p) en {1,2,...} por inversión: floor(log(U)/log(1-p))+1."""  
    out = []  
    for _ in range(n):  
        u = random.random()  
        x = math.ceil(math.log(u) / math.log(1.0 - p))  
        out.append(x)  
    return out  
  
# Ejemplo: p=0.3, n=100  
muestras = rgeom_inv_trials(0.3, 100)  
print(muestras[:20])
```

[3, 8, 3, 7, 4, 1, 1, 1, 8, 1, 9, 3, 3, 4, 1, 4, 3, 2, 1, 3]

c) Poisson

```
[22]: def rpois_inv(lam: float, n: int):
    """Poisson(lam) por transformada inversa, siguiendo el algoritmo dado."""
    if lam <= 0:
        raise ValueError("lam debe ser > 0")

    out = []
    for _ in range(n):
        # 1)  $U \sim U(0,1)$ 
        U = random.random()
        # 2)  $i=0, p=e^{-\text{lam}}, F=p$ 
        i = 0
        p = math.exp(-lam)
        F = p
        # 3) if  $U < F$ :  $X=i$ 
        if U < F:
            out.append(i)
            continue
        # 6-11) loop:  $p = (\text{lam}/(i+1))*p$ ;  $F = F + p$ ;  $i = i + 1$ ; if  $U < F$ :  $X=i$ 
        while True:
            i += 1
            p = p * lam / i
            F = F + p
            if U < F:
                out.append(i)
                break
    return out

muestras = rpois_inv(lam=2.0, n=100)
print(muestras[:20])
```

[3, 2, 2, 4, 0, 4, 1, 3, 2, 4, 1, 2, 3, 1, 3, 5, 1, 1, 2, 0]