

Week 6: Random number generators, exponential distribution

Armenak Petrosyan

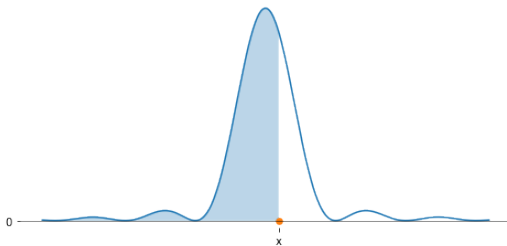
Last time

Definition (Continuous random variable, pdf)

X is called a **continuous random variable** if there exists a function $f(x)$ such that

$$F(x) = \int_{-\infty}^x f(y) dy.$$

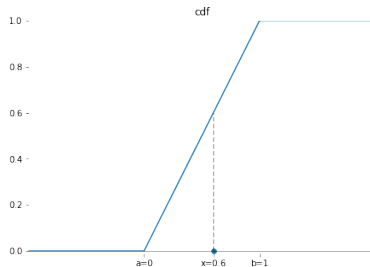
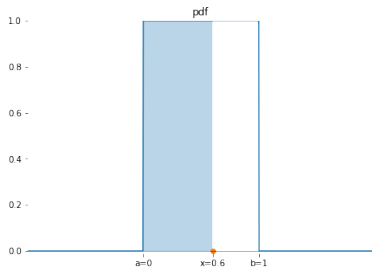
$f(x)$ is called the **probability density function** or **pdf** of continuous random variable X .



Uniform distribution (reminder)

$$f(x) = F'(x) = \begin{cases} 0 & x < a \\ \frac{1}{b-a} & a \leq x \leq b \\ 0 & x > b \end{cases}$$

$$F(x) = \begin{cases} 0 & x < a \\ \frac{x-a}{b-a} & a \leq x \leq b \\ 1 & x \geq b \end{cases}$$



Exercise 1

Problem (3.1-1 in the textbook)

Compute the variance and mgf of the uniform distribution on $[a, b]$

Solution

$$\blacktriangleright E[X^2] = \int_a^b x^2 \frac{1}{b-a} dx = \frac{1}{b-a} \frac{x^3}{3} \Big|_a^b = \frac{a^2+ab+b^2}{3}.$$

$$\blacktriangleright \text{Var}(X) = E[X^2] - E[X]^2 = \frac{a^2+ab+b^2}{3} - \left(\frac{a+b}{2}\right)^2 = \frac{(b-a)^2}{12}.$$

$$\blacktriangleright M(t) = \int_a^b e^{tx} \frac{1}{b-a} dx = \frac{e^{tx}}{t(b-a)} \Big|_a^b = \frac{e^{tb}-e^{ta}}{t(b-a)} \text{ when } t \neq 0.$$

$$\blacktriangleright M(0) = \int_a^b \frac{1}{b-a} dx = 1.$$

Definition (Median)

Median of the continuous random variable X is a number $m \in \mathbb{R}$ such that

$$P(X \leq m) = 0.5.$$

Definition (Percentiles)

Let $0 \leq p \leq 1$. The $100 \cdot p$ -**th percentile** of the X is called a number π_p such that

$$P(X \leq \pi_p) = p.$$

$\pi_{0.25}, \pi_{0.5}, \pi_{0.75}$ are called **first, second and third quartiles**.

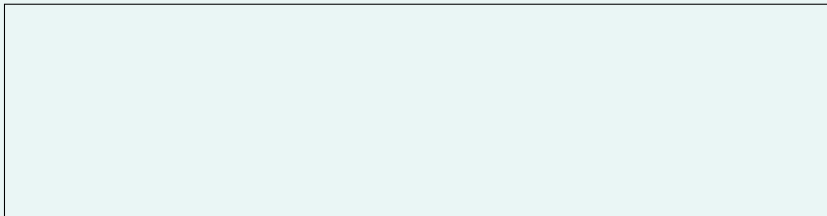
Exercise 2

Problem (3.1-14 from the textbook)

Let $f(x) = 1/2$, $0 < x < 1$ or $2 < x < 3$, zero elsewhere, be the pdf of X

- a Sketch the graph of this pdf.
- b Define the cdf of X and sketch its graph.
- c Find $q_1 = \pi_{0.25}$.
- d Find $m = \pi_{0.5}$. Is it unique?
- e Find $q_3 = \pi_{0.75}$.

Solution



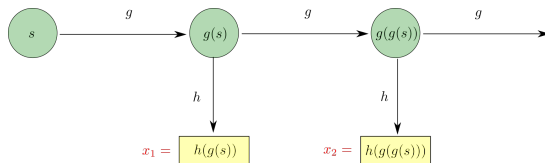
Thursday class

Generating random numbers

1. **True random number generator (TRNG)**: uses physical process (e.g. atmospheric noise).
2. **Pseudo-random number generator (PRNG)**: it starts with a value s (called seed) and applies operations to generate a sequence of numbers that has the following form

$$s, h(g(s)), h(g(g(s))), \dots$$

The seed s is chosen "randomly" (usually uses system time).



- **Linear Congruential Generator** PRNG generates random numbers in $[0, 1)$

$$s_0 = s, \quad s_{n+1} = as_n + b$$

$$x_n = s_n \bmod 1$$

- The Python standard library uses **Mersenne's Twister** PRNG algorithm.

Random sampling from a discrete distribution

- ▶ We want to sample from a discrete distribution that has cdf $F(x)$.
- ▶ Sample a value y from uniform distribution on $[0, 1)$.
- ▶ Let x_i be the largest value for which $F(x_i) \leq y$.

Theorem

Let $F(x)$ be a continuous cdf and let Y be a uniformly distributed random variable on $[0, 1]$. Let $F^{-1}(y)$ be the inverse of F . Then the random variable $X = F^{-1}(Y)$ is distributed as F :

$$P(X \leq x) = F(x).$$

Inverse transform method: (based on above theorem)

- ▶ We want to sample from a continuous random variable with cdf $F(x)$.
- ▶ Sample a value y from uniform distribution on $[0, 1]$.
- ▶ Let x be the largest value for which $F(x) \leq y$ or, equivalently, $x = F^{-1}(y)$.



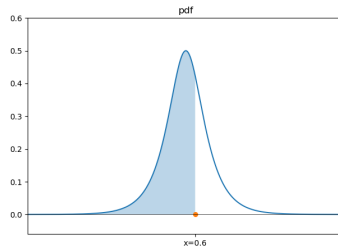
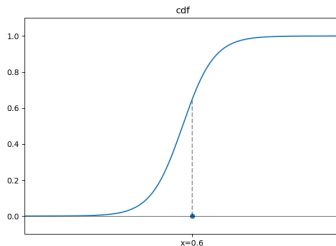
Rejection method: (uses area argument)

- ▶ We want to sample from a random variable with pdf $f(x)$.
- ▶ Let $[a, b]$ be large enough to contain most of histogram of $f(x)$.
- ▶ Sample a value (x, y) from uniform distribution on $[0, 1) \times [a, b]$.
- ▶ If $y \leq f(x)$, let X be a sample of X . Otherwise, reject it and try another point (x, y) .



Logistic distribution

$$F(x) = \frac{e^x}{1 + e^x} \Rightarrow f(x) = \frac{e^x}{(1 + e^x)^2}$$



- $y = \frac{e^x}{1+e^x} \Rightarrow e^x = \frac{1-y}{y}.$
- $F^{-1}(y) = \log(1-y) - \log(y).$

```
import numpy as np

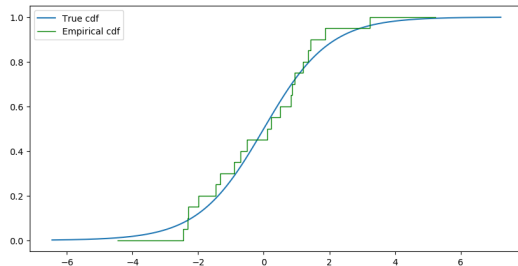
def logistic_inv(y):
    return np.log(1-y)-np.log(y)

# random sample from uniform distribution on [0,1)
y = np.random.sample((20,))

# random sample from new distribution
x = logistic_inv(y)
```

$$F_{\text{emp}}(x) = \frac{\text{number of values} \leq x}{\text{size of data}}$$

(Empirical cdf)



Definition (Poisson process)

Suppose we have an action that happens regularly over time (bus arrival, visits of a website by users, etc) that satisfies the following properties:

1. The occurrence of the action at any time interval in the future is independent of past occurrences.
2. The average number of actions happening in a given time interval is proportional to the interval length.

Denote by λ the average number of times the action happens in the unit time interval $[0, 1]$. Then we call this type of action a **Poisson process with rate λ** .

- Poisson process is the continuous version of Bernoulli trials.
- If we denote by Y the number of actions in a Poisson process that happened in a certain fixed interval, $[a, b]$, then Y will have a Poisson distribution with parameter $\lambda \cdot (b - a)$.

Exponential distribution: motivation

Let X be the time when the action happens for the first time in a Poisson process with rate λ .

- ▶ X is called **waiting time** (it is the time we need to wait for the action to happen).
- ▶ X is a continuous random variable as we see below.
- ▶ $\text{Range}(X) = [0, \infty)$.
- ▶ $F(X \leq x) = 0$ when $x < 0$.
- ▶ For $x \geq 0$,

$$F(x) = P(X \leq x) = 1 - P(X > x).$$

- ▶ $P(X > x)$ is the probability that there are no actions in the interval $[0, x]$. Since the number of actions in the interval $[0, x]$ is a Poisson random variable with parameter $\lambda \cdot x$,

$$P(X > x) = e^{-\lambda x} \frac{(\lambda x)^0}{0!} = e^{-\lambda x}.$$

- ▶ Hence,

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

- ▶ For the pdf of X , we have

$$f(x) = F'(x) = \begin{cases} 0 & x < 0 \\ \lambda e^{-\lambda x} & x \geq 0 \end{cases}.$$

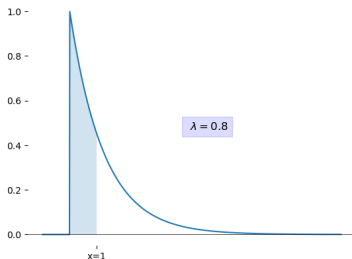
Exponential distribution: definition

Definition (Exponential distribution)

We say that a continuous random variable X has **exponential distribution** with parameter θ if its pdf is given by

$$f(x) = \begin{cases} 0 & x < 0 \\ \frac{1}{\theta} e^{-\frac{x}{\theta}} & x \geq 0 \end{cases}.$$

- ▶ We prefer $\theta = \frac{1}{\lambda}$ parametrization because it represents the average waiting time as will be apparent soon.
- ▶ Exponential distribution is the continuous version of the geometric distribution.



Mean and variance of exponential distribution

- From definition

$$M(t) = \int_0^{\infty} e^{tx} \frac{1}{\theta} e^{-\frac{x}{\theta}} dx = \frac{1}{\theta} \int_0^{\infty} e^{x(t-\frac{1}{\theta})} dx = \frac{1}{\theta} \cdot \frac{1}{t-\frac{1}{\theta}} \cdot e^{x(t-\frac{1}{\theta})} \Big|_{x=0}^{\infty} = \frac{1}{1-\theta t} \text{ when } t < \frac{1}{\theta}.$$

- Hence

$$M'(t) = \frac{\theta}{(1-\theta t)^2}, \quad M''(t) = \frac{2\theta^2}{(1-\theta t)^3}.$$

- Therefore

$$E[X] = M'(0) = \theta$$
$$\text{Var}(X) = E[X^2] - E[X]^2 = M''(0) - E[X]^2 = \theta^2.$$