Probability mass functions (PMFs)

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1 Introduction

The purpose of this lab is to teach you how to generate different discrete probability distributions in different programming languages. You may work in C++, Java, Matlab or Python.

1.1 Linear congruential generator

In C++, you need to create an instance of a random number generator. Most widely studied and used random number generator is "Linear congruential generator". Another popular random number generator is "Mersenne Twister generator". I have explained only "Linear congruential generator" here. The pseudo-random sequence of number using the linear congruential generator is given by the following equation

$$x_{n+1} = (ax_n + b) \bmod m \tag{1}$$

where m > 0, 0 < a < m, $0 \le b < m$ and $0 \le x_0 < m$ is the seed. In C++, we can create an instance of a LCG by creating an object of type "minstd_rand". The parameter of the constructor is the seed value.

1.2 Uniform distribution

We can create a uniform distribution (discrete) by creating an object like the following

$$uniform_int_distribution < int > uniform_int_number(m, n);$$
 (2)

where m,n are the lower and upper limit of the discrete uniform distribution.

1.3 Geometric distribution

We can create a realization of geometric distribution by creating an object of type geometric_distribution

$$qeometric_distribution < int > qeometric_number(p);$$
 (3)

where p is the probability of success.

1.4 Binomial distribution

We can create a realization of binomial distribution by creating an object of type binomial_distribution

$$binomial_distribution < int > binomial_number(n, p);$$
 (4)

where n is the total number of trials and p is the probability of success.

1.5 Poisson distribution

We can create a realization of Poisson distribution by creating an object of type poisson_distribution

$$poisson_distribution < int > poisson_number(\mu);$$
 (5)

where μ is the average number of occurrences.

2 Generating any discrete random variable

In order to generate a discrete random realization according to any pmf, we first need to generate a continuous standard uniform realization. F(x) is the CDF(Cumulative distribution function).

2.1 Continuous uniform random variable

In C++, we generate a standard continuous uniform realization using the following line of code

$$uniform_real_distribution < double > uniform_real_number(a, b);$$
 (6)

where a and b are the lower and upper bounds of continuous uniform random variable. In this case, they should be a=0, b=1.

In order to generate any discrete random variable, we follow the following algorithm

- 1. For each x_i , find $F(x_i)$ i = 1 to N.
- 2. Generate a random continuous uniform $u \sim U(0,1)$.
- 3. Locate the smallest x_i where $u < F(x_i)$.
- 4. Set $x = x_i$.
- 5. Return x.

2.1.1 Weights of the packages and cost of shipping a package

Let the pmf of weights of the packages X and the shipping cost Y = g(X) defined as below

$$p_X[x] = \begin{cases} 0.15 & x = 1, 2, 3, 4\\ 0.1 & x = 5, 6, 7, 8\\ 0 & otherwise \end{cases}$$
 (7)

$$Y = \begin{cases} 105X - 5X^2 & 1 \le X \le 5\\ 500 & 6 \le X \le 10 \end{cases}$$
 (8)

- 1. Write a function that outputs n samples of the shipping cost.
- 2. Write a function which computes the expectation and variance of Y random variable.

2.1.2 Histogram of the Poisson distribution

Generate n independent samples of the Poisson (5) random variable Y. For each $y \in S_Y$, let $\mathbf{n}(\mathbf{y})$ denote the number of times that y was observed. Thus $\sum_{y \in S_Y} n(y) = \mathbf{n}$, and the relative frequency of y is $\mathbf{R}(\mathbf{y}) = \mathbf{n}(\mathbf{y}) / \mathbf{n}$. Compare the relative frequency of y against $p_y[y]$ by plotting $\mathbf{R}(\mathbf{y})$ and $p_y[y]$ on the same graph as functions of y for $\mathbf{n} = 100$, $\mathbf{n} = 1000$ and $\mathbf{n} = 10,000$. How large should \mathbf{n} , be to have reasonable agreement?