

Simulating Poisson processes

University of Toronto, ACT 350

Prof. S. Pesenti

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How to simulate a Poisson process

Recall that a Poisson process with rate $\lambda > 0$ is a counting process $\{N(t), t \geq 0\}$ that has the following properties

1. *starts at 0*: $N(0) = 0$ with probability 1
2. *is increasing*: $N(s) \leq N(t)$ for all $s \leq t$
3. *has Poisson distributed increments*: $P(N(s+t) - N(s) = n) \sim \text{Pois}(\lambda t)$ for all $s, t \geq 0$.

However, this definition does not help if we want to simulate a Poisson process. Thus, to simulate a Poisson process we use its alternative definition. That is, for independent random variables X_1, X_2, \dots that are Exponentially distributed with rate $\lambda > 0$, a Poisson process with rate $\lambda > 0$ can be written as

$$N(t) = \max \left\{ n \in \mathbb{N} \mid \sum_{i=1}^n X_i \leq t \right\}, \quad \text{for all } t \geq 0.$$

Thus, for simulating a Poisson process we first need to simulate exponential random variables, add them up and check whether they are smaller or equal to t . However, we will run into problems when we want to store the value of $N(t)$ for every $0 \leq t < \infty$. The way out is to implement $N(t)$ as a function of t in the following way. First generate the jump points of the Poisson process. Second, we know that a Poisson process can only jump one unit at a time. Thus, we can define the Poisson process as a step function that jumps exactly one unit at every jump point.

1. Generating jump times

This is an algorithm for generating the jump times of a Poisson process

1. Set $\tau_0 = 0$.
2. For $k \geq 1$ do
 - a) generate $X \sim \text{Exp}(\lambda)$
 - b) set $\tau_k = \tau_{k-1} + X$

The τ_k , $k = 0, 1, \dots$ are the time points, where the Poisson process jumps.

Numerical implementation: R code

```
# set the seed for generating random variables (for reproducibility)
# set.seed(2019)

# set the rate of the Poisson process/Exponential distribution
lambda <- 2
# number of jumps
n <- 50
# initialise the jump times
jump_times <- rep(0, length = n)

# note that the first value of the jump_times is already 0
for(i in 2:n){
  jump_times[i] <- jump_times[i - 1] + rexp(1, lambda)
}
jump_times
```

```
## [1] 0.00000000 0.08392964 0.40169155 0.69439969 0.86486407 0.95194932
## [7] 1.28498569 1.45903169 2.05518462 2.38224677 2.92635016 3.06556691
## [13] 4.61114863 5.18812490 6.01274892 6.79459443 8.80242283 10.00229765
## [19] 10.62725970 10.85193395 11.12956236 12.29444269 12.35255619 12.83112242
## [25] 13.52349274 14.05565723 14.83834796 15.12041067 15.20725859 15.61810617
## [31] 15.87047088 15.91764105 16.13303110 16.17630992 16.44527679 16.99007003
## [37] 17.50494681 18.24211092 18.88526868 18.97270443 19.05522498 19.72987822
## [43] 19.85945608 21.41060795 22.31358469 22.86029328 22.91795597 23.34722470
## [49] 23.59448295 24.00388081
```

A more efficient implementation, avoiding the for loop, is

```
jump_times_2 = cumsum(rexp(n-1, lambda))
# note that the two generated sample_paths are different,
# since the Exponential random variables are different.
jump_times_2
```

```
## [1] 0.1458827 0.4775990 0.6205391 3.5627986 3.9839791 4.2751437
## [7] 4.8299963 4.8538748 5.4361382 5.8027378 7.2884842 7.4042789
## [13] 7.6468467 7.7344969 8.0282881 8.8381585 9.0054665 9.0717844
## [19] 9.8012940 10.0508849 10.5103700 10.5135410 11.3223250 11.3289122
## [25] 11.3613650 11.4691510 11.5477209 12.6287205 12.9800112 13.1558413
## [31] 14.4728279 15.3039947 15.4729084 15.9731209 16.0664084 17.0613587
## [37] 17.2257266 17.5341449 18.4087557 19.2998072 19.6211126 20.9096770
## [43] 24.3958251 24.7768882 25.2472920 26.3649781 27.2638825 27.5002652
## [49] 27.6224412
```

2. Defining the Poisson process as a function

Next, we can define the Poisson process as a step function that jumps exactly one unit at every value in `jump_times`. Note that the implemented Poisson process is now a *function* of the time t .

```
# define the Poisson process as a function
poisson_process <- stepfun(jump_times[2 : 50], seq(0, n - 1, by = 1))
# the first argument of 'stepfun' is where the step function jumps
# the second argument is the height of the step function

# Let us check some values.
# The starting value is 0.
poisson_process(0)
```

```
## [1] 0
```

```
# What is the value of N(3)? (Note that we use jump_times and not jump_times_2)
poisson_process(3)
```

```
## [1] 10
```

```
poisson_process(10)
```

```
## [1] 16
```

```
poisson_process(15)
```

```
## [1] 26
```

The first jump time is 0.0839296. Let us check whether the Poisson process jumps at that time.

```
# The first jump of the Poisson process is
first_jump <- jump_times[2]
poisson_process(first_jump - 0.001)
```

```
## [1] 0
```

```
poisson_process(first_jump + 0.001)
```

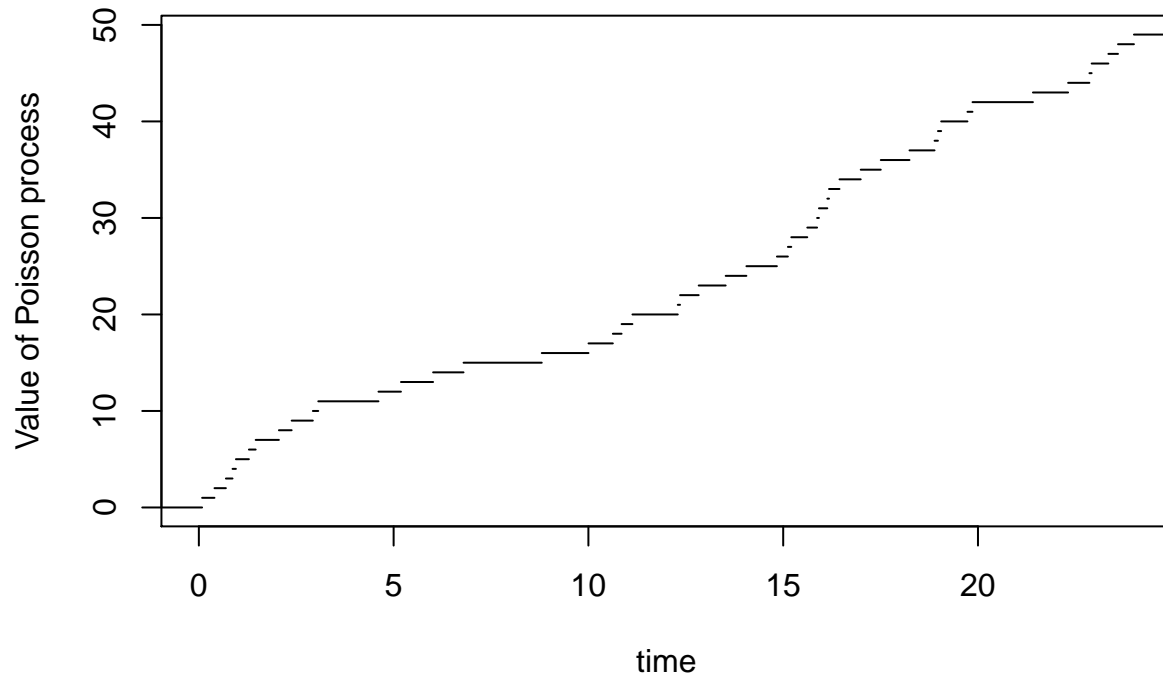
```
## [1] 1
```

Note you can evaluate the Poisson process for any time point t , however, keep in mind that we only simulated 50 jumps of the Poisson process. Thus, evaluating $N(t)$ for t large does not make sense.

3. Plotting a sample path

To plot a sample path we have to plot the implemented step function `poisson_process`.

```
plot(poisson_process, xlab = "time", ylab = "Value of Poisson process", main = NULL,
     verticals = FALSE, do.points = FALSE, xlim = c(0, jump_times[n]))
```



Question: Rerun the code, what do you observe?

Question: Rerun the code with a different λ or change the number of jumps. How does the sample path change?

Renewal Theory

A renewal process is defined as follows: Consider a sequence T_1, T_2, \dots of independent identically distributed (i.i.d.) non-negative random variables. Define the stochastic process

$$X_0 = 0, \quad X_n = T_1 + \dots + T_n, \quad n = 1, 2, \dots,$$

and set

$$N(t) = \max \left\{ n \in \mathbb{N} \mid \sum_{i=1}^n X_i \leq t \right\}, \quad \text{for all } t \geq 0.$$

Then the process $\{N(t) \mid t \geq 0\}$ is called a renewal process.

If the $T_i, i = 1, 2, \dots$ are i.i.d. Exponential with parameter $\lambda > 0$, then the renewal process $\{N(t) \mid t \geq 0\}$ is a Poisson process.

Task:

a) Generate a code to simulate from a renewal process, where the interarrival times $T_i, i = 1, 2, \dots$ are i.i.d. LogNormal distributed with mean 2 and standard deviation 0.5. *Hint: use the function `rlnorm`.*

- b)** Plot a sample path of the renewal process define in **a**).
- c)** What do you observe compared to the Poisson process?
- d)** Repeat tasks **a**) to **b**) with interarrival times being Gamma distributed with the same mean and standard deviation as in **a**). *Hint: use the function `rgamma`.*
- e)** Compare the Poisson process with the renewal process from **a**) and the renewal process from **d**) and give interpretations.