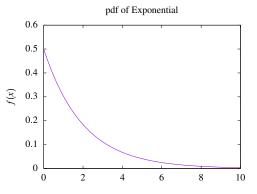
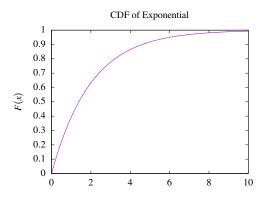
$Exponential(\mu)$ and Geometric(p)

October 2, 2025

The $Exponential(\mu)$ Distribution and Variate

What f(x) models random arrivals?





- If there are *n* arrivals in the time interval $[0, a_n]$, then the average interarrival time is $\bar{r} = a_n/n$
- ▶ The mean of the f(x) is \bar{r} , by **convention** the variable μ is used in the exponential pdf. In the case of SSQs: $\mu = \bar{r}$.
- The variance of f(x) is $\sigma^2 = \mu^2$.
- ightharpoonup f(x) crosses the y-axis at $1/\mu$

Here we are talking about a probability distribution again — what do you think is next?

Variate for $Exponential(\mu)$

Warning: The textbook uses average interarrival time $\bar{r} = \mu = \frac{a_n}{n} = \frac{1}{\lambda}$, where λ is the arrival rate (jobs/sec). About half the statistics and simulation textbooks use λ , the other half use $\mu = \frac{a_n}{n}$. Neither is right or wrong — just be careful when blindly following recipes from other sources.

The exponential pdf is

$$f(x) = \frac{1}{\mu} e^{-\frac{x}{\mu}}$$

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The exponential pdf is

$$f(x) = \frac{1}{\mu} e^{-\frac{x}{\mu}}$$

and the first step is integration to find the cdf

$$F(x) = \int_0^x \frac{1}{\mu} e^{-\frac{t}{\mu}} dt = -e^{-\frac{t}{\mu}} \Big|_0^x = -e^{-\frac{x}{\mu}} - (-1) = 1 - e^{-\frac{x}{\mu}}$$

Variate for $Exponential(\mu)$

The second step is setting F(x) = u (which we determine programmatically with Random()) and solving for x

$$u = 1 - e^{-\frac{x}{\mu}}$$

$$e^{-\frac{x}{\mu}} = 1 - u$$

$$-\frac{x}{\mu} = \ln(1 - u)$$

$$x = -\mu \ln(1 - u)$$

Since u is random in (0,1), 1-u is random in (0,1) as well and without loss of generality, we can write:

$$x = -\mu \ln(1 - u) = -\mu \ln(u) \qquad \mu = \frac{a_n}{n} = \frac{1}{\lambda}$$

See why we prefer the range of Random() to be (0,1)?

 λ is the **arrival rate** (jobs/sec)

Geometric(p) the discrete analogue to $Exponential(\mu)$

We have seen that $Uniform(a,b) \in \Re$ has a discrete analogue of Equilikely(a,b); it is fair to ask the same about $Exponential(\mu)$.

Let

$$y = \lfloor Exponential(\mu) \rfloor$$

then what is the discrete probability distribution of y?

$$Pr(y \ge 1) = Pr(Exponential(\mu) \ge 1) = Pr(-\mu \ln(u) \ge 1)$$
 (1)

Well this isn't that hard, since u is from Random() it has a simple distribution and we have hope...

$$-\mu \ln(u) \geq 1$$

$$\ln(u) \leq -\frac{1}{\mu}$$

$$u \leq e^{-\frac{1}{\mu}}$$

Geometric(p) the discrete analogue to $Exponential(\mu)$

Conclusion:

$$\lfloor Exponential(\mu) \rfloor \geq 1$$
 when u (from Random()) is $\leq e^{-\frac{1}{\mu}}$

But we would like a simpler way to think of $e^{-\frac{1}{\mu}}$

When $\mu > 0$, we have

$$0 < e^{-\frac{1}{\mu}} < 1$$

and we think of it as a probability

$$p = e^{-\frac{1}{\mu}} \tag{2}$$

You may be familiar with the geometric distribution reflecting the number of "heads" flipped before the first tail with a *p*-bias coin.

A Variate for Geometric(p)

Now consolidating the algebra from (eqn 1)...

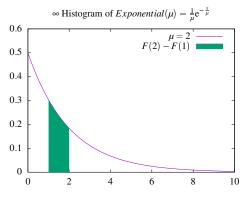
$$y = \lfloor Exponential(\mu) \rfloor = \lfloor -\mu \ln(u) \rfloor = \left\lfloor \frac{\ln(u)}{-\frac{1}{\mu}} \right\rfloor$$
$$y = \left\lfloor \frac{\ln(u)}{\ln\left(e^{-\frac{1}{\mu}}\right)} \right\rfloor$$

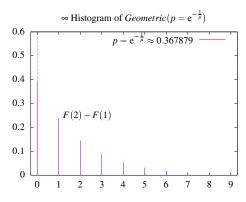
... and plugging in our defintion of p in terms of μ (eqn 2):

$$y = \left| \frac{\ln(u)}{\ln(p)} \right| = \left| \frac{\ln(1-u)}{\ln(p)} \right|$$

With u from Random(), y will follow the geometric distribution parameterized with probability p.

The $Exponential(\mu)$ and $Geometric(p = e^{-\frac{1}{\mu}})$ Variate distributions





- Exponential distribution: pdf $f(\mu, x) = \frac{1}{\mu} e^{-\frac{x}{\mu}}$, $F(\mu, x) = 1 e^{-\frac{x}{\mu}}$
- ▶ Geometric distribution: **point mass function** $g(\mu, x) = F(\mu, x + 1) F(\mu, x)$
- Exponential variate: $x = -\mu \ln(u) \equiv -\mu \ln(1 u)$
- Geometric variate: $y = \left\lfloor \frac{\ln u}{\ln p} \right\rfloor \equiv \left\lfloor \frac{\ln(1-u)}{\ln p} \right\rfloor$