

Specific methods for non-uniform #s:

1. The use of central limit thm:

Note that if $u_1, u_2, \dots, u_n \stackrel{i.i.d}{\sim} \text{unif}(0,1)$, then

$$\frac{\sum_{i=1}^n u_i - E\left[\sum_{i=1}^n u_i\right]}{\sqrt{\text{var}\left(\sum_{i=1}^n u_i\right)}} \xrightarrow{d} N(0,1) \text{ by CLT.}$$

As $n \rightarrow \infty$, the distr. tends to normal distr. But, of course, in practice, we settle for some finite # of n . So, how large should n be? $n=2$ is not suitable, bc. the distr. has a triangular shape for $n=2$. For $n=3$, the distr. is already nicely "bell-shaped". The answer depends on how close an approximation is required.

Consider $n=12$. Then,

$$E(u_i) = \frac{1}{2} \Rightarrow E\left(\sum_{i=1}^{12} u_i\right) = \sum_{i=1}^{12} E(u_i) = 12 \cdot \frac{1}{2} = 6$$

$$V(u_i) = \frac{1}{12} \Rightarrow V\left(\sum_{i=1}^{12} u_i\right) = \sum_{i=1}^{12} V(u_i) = 12 \cdot \frac{1}{12} = 1$$

since indep.

$$\Rightarrow \sum_{i=1}^{12} u_i - 6 \xrightarrow{d} N(0,1)$$

(Note that >7 values do not occur, which could be a problem).

2. The Box-Muller method:

If U_1, U_2 are two indep. $U(0,1)$ r.v.s, then Box & Muller (1958) showed that,

$$X = \sqrt{(-2 \ln U_1)} \cos(2\pi U_2) \text{ and}$$

$$Y = \sqrt{(-2 \ln U_1)} \sin(2\pi U_2) \text{ are indep. } N(0,1) \text{ r.v.s.}$$

Since this method requires the computation of sine & cosine functions, it is computationally time-consuming.

3. The Polar method:

A rejection method that gets around the time-consuming calculations in Box-Muller.

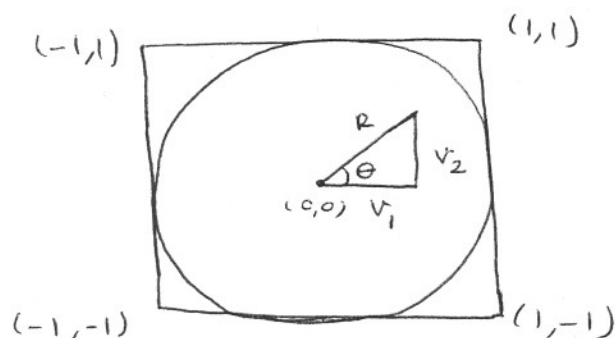
Note that: if $U \sim \text{Unif}(0,1)$ then
 $2U \sim \text{Unif}(0,2)$
 $2U-1 \sim \text{Unif}(-1,1)$

\$ we generate #'s U_1, U_2 , set

$V_1 = 2U_1 - 1$ & $V_2 = 2U_2 - 1$, and accept #'s only if

$$V_1^2 + V_2^2 \leq 1.$$

Note that this will lead to (V_1, V_2) uniformly distributed in the circle.



$$\sin(\theta) = \frac{V_2}{R} = \frac{V_2}{\sqrt{V_1^2 + V_2^2}}$$

$$\cos(\theta) = \frac{V_1}{R} = \frac{V_1}{\sqrt{V_1^2 + V_2^2}}$$

⇒ Box-muller transformation can be written as

$$X = \sqrt{-2 \log u} \cdot \frac{V_1}{\sqrt{V_1^2 + V_2^2}}$$

$$Y = \sqrt{-2 \log u} \cdot \frac{V_2}{\sqrt{V_1^2 + V_2^2}}$$

Since $e^2 = V_1^2 + V_2^2$ is $\text{unif}(0,1)$ & indep. of Θ , we can use it for u above. Let $S = e^2$.

$$\Rightarrow X = \sqrt{-2 \ln S} \cdot \frac{V_1}{\sqrt{S}} = V_1 \left(\frac{-2 \ln S}{S} \right)^{1/2} \text{ and}$$

$$Y = V_2 \left(\frac{-2 \ln S}{S} \right)^{1/2}$$

are indep. $N(0,1)$ s.

The Algorithm is:

1. Generate U_1 & U_2 .
2. Set $V_1 = 2U_1 - 1$, $V_2 = 2U_2 - 1$, $S = V_1^2 + V_2^2$
3. If $S > 1$, go to step 1.
4. Set $X = \left(\frac{-2 \ln S}{S} \right)^{1/2} \cdot V_1$, $Y = \left(\frac{-2 \ln S}{S} \right)^{1/2} \cdot V_2$

4. Generating from χ_m^2 :

Recall: $\sum_{i=1}^m Z_i^2 \sim \chi_m^2$ where $Z_i \sim N(0,1)$.

So, to generate from χ_m^2 , simply generate m std. normal variates, square them, and sum them up. But, since simulating from $N(0,1)$ might be time-consuming, the approach below might be easier:

If m is even, χ_m^2 is simply $\text{Gamma}\left(\frac{m}{2}, \frac{1}{2}\right)$.

(We already talked about how to generate from Gamma by using Exp.)

If m is odd, first generate from χ^2_{m-1} by using $\text{Gamma}(\frac{m-1}{2}, \frac{1}{2})$ and then adding Z_i^2 .

To generate from Gamma when shape parameter is not integer, refs:

Cheng (1977) Applied Statistics, 26(1), 71-74.

Kinderman & Monahan (1980) New methods for generating Student's t & gamma vrs. Computing, 25, 369-377.