# Exercises Lecture III:

# Random numbers with non uniform distributions;

## 1. Exponential distribution with Inverse Transformation Method

- (a) With the Inverse Transformation Method we can generate random numbers according to the exponential distribution  $f(z) = \lambda e^{-\lambda z}$ , starting from random numbers with uniform distribution: if x is the random variable with uniform distribution in [0,1], then z = -ln(x) is distributed according to  $e^{-z}$ . Write a code implementing the algorithm. An example is given in expdev.f90.
- (b) Check—doing a histogram—that the random variate z generated with that algorithm is actually exponentially distributed.
  (What is convenient to plot in order to check this behavior? Hint: with gnuplot you can print the log of your data (e.g., suppose you saved the values of z in column 1 and its frequency in column 2, plot with u 1:(log(\$2)) or u 1:(log10(\$2)) )).
- (c) With gnuplot you can also do the fit of the histogram with an exponential function using the least-square method, with  $\lambda$  as fitting parameter. Check whether you get the expected value of  $\lambda$ . (It is convenient to make a semilog plot as suggested above and then make a least-square linear fit; the slope is  $\lambda$ )

## 2. An example of non uniform distribution: comparison between different algorithms

Suppose you want to generate a random variate x in (-1,1) with distribution

$$p(x) = \frac{1}{\pi} (1 - x^2)^{-1/2}$$

Consider both methods suggested below, do the histograms and check that both methods give correct results.

- (a) From the Inverse Transformation Method: generate a random number y with uniform distribution in [0,1] and consider  $x = \cos(\pi y)$  (or  $x = \sin(2\pi y)$ , or  $x = \sin \pi(2y-1)...$  Why?).
- (b) Generate two random numbers U and V with uniform distribution in [0,1]. Disregard them if  $U^2 + V^2 > 1$ . Otherwise consider

$$x = \frac{U^2 - V^2}{U^2 + V^2}$$

Note 1: the last method has the advantage of using only elementary operations.

Note 2: since x is also negative, pay attention to the algorithm used to make the histogram; you should notice the difference between the intrinsic functions int and nint; see also floor. From Chapman's book:

```
AINT(A,KIND):
                Real elemental function
- Returns A truncated to a whole number.
AINT(A) is the largest integer which is smaller than |A|, with the sign of A.
For example, AINT(3.7) is 3.0, and AINT(-3.7) is -3.0.
- Argument A is Real; optional argument KIND is Integer
ANINT(A,KIND):
                Real elemental function
- Returns the nearest whole number to A.
For example, ANINT(3.7) is 4.0, and AINT(-3.7) is -4.0.
- Argument A is Real; optional argument KIND is Integer
FLOOR(A,KIND):
                Integer elemental function
- Returns the largest integer < or = A.
For example, FLOOR(3.7) is 3, and FLOOR(-3.7) is -4.
  Argument A is Real of any kind; optional argument KIND is Integer
  Argument KIND is only available in Fortran 95
NINT(A[,KIND])
- Integer elemental function
- Returns the nearest integer to the real value A.
- A is Real
```

#### 3. Gaussian distribution: Box-Muller algorithm

Consider the Box-Muller algorithm to generate a random number gaussian distribution (see for instance **boxmuller.f90**; the **gasdev** subroutine used inside is similar to what you can find in "Numerical Recipes": it gives a gaussian distribution with  $\sigma = 1$  and average  $\mu = 0$ ). Do a histogram of the data generated, calculate *numerically* from the sequence the average value and the variance, check with the expected results.

#### 4. Gaussian distribution: the central limit theorem

Use the central limit theorem in order to produce random numbers with gaussian distribution. Remind that given a sequence of independent random numbers  $r_i$ , their average

$$x_N = \frac{1}{N} \sum_{i=1}^N r_i$$

is distributed according to

$$P_N(x) = \frac{1}{\sqrt{2\pi\sigma^2}} exp\left(-\frac{(x_N - \mu)^2}{2\sigma^2}\right)$$

with  $\mu = < r >$ ,  $\sigma^2 = (< r^2 > - < r >^2)/N$ . Furthermore, defining

$$z_N = \frac{x_N - \mu}{\sigma}$$

we have  $\langle z_N^4 \rangle \approx 3 \langle z_N^2 \rangle^2$ .

- (a) Use random numbers  $r_i$  uniformly distributed in [-1,1], choosing N $\approx 50$  and generating at least  $\approx 1000$  points  $x_N$ . Verify (doing an histogram) that  $x_N$  have a Gaussian distribution.
- (b) Calculate numerically  $\langle x_N \rangle$  and  $\sigma_x^2$  and compare them with  $\mu$  and  $\sigma^2$  analytically calculated.
- (c) Consider  $z_N$  and verify numerically the relationship  $< z_N^4 > \approx 3 < z_N^2 >^2.$
- (d) Do again the exercise using random numbers  $r_i$  with exponential distribution  $(p(r) = e^{-r}$  if  $r \ge 0$ , 0 elsewhere). Calculate  $\langle x_N \rangle$ ,  $\sigma_{x_N}^2$  and  $\mu$ ,  $\sigma^2$  numerically and analytically, respectively.
- (e) (optional) Consider now random numbers  $r_i$  with distribution

$$p(r) = \frac{a}{\pi} \frac{1}{r^2 + a^2}$$

(Lorentz's), with a = 1, for instance. Do  $\langle x \rangle$ ,  $\langle x^2 \rangle$  and  $\sigma_x^2$  exist? Try to determine the characteristic of the distribution of the variable sum  $x_N = \frac{1}{N} \sum_{i=1}^N r_i.$ 

5. Various random distributions (test of a ready-to-use module) (Optional)

You can try t\_random.f90 which uses the module random.f90 to generate random deviates with other distributions. Remember to compile first the module: gfortran random.f90 t\_random.f90

```
! expdev.f90
program test_expdev
 implicit none
 real :: lambda,delta,x
 integer :: i,n,nbin,ibin, sizer
 integer, dimension(:), allocatable :: histo, seed
print*, " Generates random numbers x distributed as exp(-lambda*x)"
 call random_seed(sizer)
 allocate(seed(sizer))
 print *,'Here the seed has ',sizer,' components; insert them (or print "/") >\'
 read(*,*)seed
 call random_seed(put=seed)
 print *," length of the sequence >"
 read *, n
 print *," exponential decay factor (lambda)>"
 read *, lambda
 print *," Collecting numbers generated up to 2/lambda (disregard the others)"
 print *," and normalizing the distribution in [0,+infinity[ "
 print *," Insert number of bins in the histogram>"
 read *, nbin
 delta = 2./lambda/nbin
   allocate (histo(nbin))
 histo = 0
 do i = 1,n
    call expdev(x)
    ibin = int (x/lambda/delta) + 1
    if (ibin <= nbin)histo(ibin) = histo(ibin) + 1</pre>
 end do
 open (unit=7,file="expdev.dat",status="replace",action="write")
 do ibin= 1 ,nbin
    write(unit=7,fmt=*)(ibin-0.5)*delta,histo(ibin)/float(n)/delta
 end do
contains
 subroutine expdev(x)
                       °/_----
   REAL, intent (out) :: x
   REAL :: r
   do
      call random_number(r)
      if(r > 0) exit
   end do
   x = -\log(r)
                       %-----
 END subroutine expdev
end program test_expdev
```

```
! boxmuller.90
! uses the Box-Muller algorithm to generate
! a random variate with a gaussian distribution (sigma = 1)
ļ
program boxmuller
  implicit none
 real :: rnd,delta
 real, dimension(:), allocatable :: histog
 integer :: npts,i,ibin,maxbin,m
 print*,' input npts, maxbin >'
 read*, npts,maxbin
 allocate(histog(-maxbin/2:maxbin/2))
 histog = 0
 delta = 10./maxbin
 do i = 1, npts
    call gasdev(rnd)
    ibin = nint(rnd/delta)
    if (abs(ibin) < maxbin/2) histog(ibin) = histog(ibin) + 1</pre>
 end do
 open(1,file='gasdev.dat',status='replace')
 do ibin = -maxbin/2, maxbin/2
    write(1,*)ibin*delta, histog(ibin)/real(npts)/delta
 end do
 close(1)
 deallocate(histog)
 stop
contains
 SUBROUTINE gasdev(rnd)
   IMPLICIT NONE
   REAL, INTENT(OUT) :: rnd
   REAL :: r2,x,y
   REAL, SAVE :: g
   LOGICAL, SAVE :: gaus_stored=.false.
   if (gaus_stored) then
      rnd=g
      gaus_stored=.false.
   else
      do
         call random_number(x)
         call random_number(y)
         x=2.*x-1.
         y=2.*y-1.
```

```
r2=x**2+y**2
if (r2 > 0. .and. r2 < 1.) exit
end do
r2=sqrt(-2.*log(r2)/r2)
rnd=x*r2
g=y*r2
gaus_stored=.true.
end if
END SUBROUTINE gasdev
end program boxmuller
```