

## Exercises Lecture III:

### Random numbers with non uniform distributions; simulations of simple random processes

homework: n. 1,3,4

1. **Random numbers with non uniform distributions:  
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  $\lambda$* )

*Remember that with the method of the least-square fit we get for a linear regression:  $y = ax + b$ :*

$$a = \frac{\overline{xy} - \bar{y}\bar{x}}{(\Delta x)^2}; \quad b = \bar{y} - a\bar{x}$$

*where  $(\Delta x)^2 = \overline{x^2} - \bar{x}^2$  (other definitions are trivial ...).*

## 2. Random numbers with non uniform distributions: 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  $U$  with uniform distribution in  $[0,1]$  and consider  $x = \sin \pi(2U - 1)$ .
- (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:*

`AINTE(A,KIND):` Real elemental function  
- Returns A truncated to a whole number.  
`AINTE(A)` is the largest integer which is smaller than  $|A|$ , with the sign of A.  
For example, `AINTE(3.7)` is 3.0, and `AINTE(-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 `ANINT(-3.7)` is -4.0.  
- Argument A is Real; optional argument KIND is Integer

`FLOOR(A,KIND):` Integer elemental function  
- Returns the largest integer  $< \text{or} = A$ .  
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. Random numbers with 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. Simulation of radioactive decay

- Write a program for a numerical simulation of the radioactive decay, with a decay parameter  $\lambda$  in input. (See for instance `decay.f90`).
- Use the code with “reasonable” values of the parameters (e.g.,  $N(0)$  about 1000) and save  $N(t)$  in a data file. Check whether  $N(t) = N(0)e^{-\lambda t}$  as expected. (*Hint: As for the exercise 1, you could make use of a least-square fit by considering  $\ln N(t)$  vs.  $t$ , i.e. the relationship in a semilog form in order to manage a linear fit.*)
- Change  $N(0)$  (100 or less; 10000 or more). What do you see?

*Notice that in `decay.f90` the upper bound of the inner loop (`nleft`) is changed within the execution of the loop; but in the execution the loop goes on up to the `nleft` set at the beginning of the loop; this ensures that the implementation of the algorithm is correct. See the programs `checkloop.f90` and `decay_checkloop.f90` in the same directory.*

### 5. Random deviates with other distributions (*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: `g95 (or gfortran) random.f90 t_random.f90`

```
!CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC
! 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  $\exp(-\lambda \cdot x)$ "
    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
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

!CCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCCC
! boxmuller.90
! uses the Box-Muller algorithm to generate
! a random variate with a gaussian distribution (sigma = 1)
!
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(histogram(-maxbin/2:maxbin/2))
histogram = 0
delta = 10./maxbin
do i = 1, npts
    call gasdev(rnd)
    ibin = nint(rnd/delta)
    if (abs(ibin) < maxbin/2) histogram(ibin) = histogram(ibin) + 1
end do

open(1, file='gasdev.dat', status='replace')
do ibin = -maxbin/2, maxbin/2
    write(1, *) ibin*delta, histogram(ibin)/real(npts)/delta
end do
close(1)
deallocate(histogram)
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

```

