

SAMPLING THE GAUSSIAN DISTRIBUTION WITH THE BOX–MULLER ALGORITHM

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1 INTRODUCTION

In this note I discuss how to sample the Gaussian distribution by means of the Box–Muller algorithm. The source codes reported in this notes are used to illustrate the basic usage of vectors and arrays. Moreover, I also present a first example of *dynamical memory allocation*, which may be possibly used with such structures. A simple routine to make the histogram of a 1D random variable is quickly introduced. Such routine, as I discussed in previous notes, can be either written in the same file containing the source code for the main program or, alternatively, in a different file. In this note, I also discuss how to recall a subroutine written in an external file, namely how to *link two fortran codes*

2 THE BOX–MULLER ALGORITHM

Gaussian Random Variables (RVs) are highly important in Science, because the natural emergence of the Central Limit Theorem [[Bouchaud and Georges, 1990](#); [Forte, 2014](#); [Forte et al., 2014](#)].

A (continuous) RV x is distributed according a Gaussian with mean μ and variance σ^2 if the probability density to get the variable within a “volume–element” dx around x is given by

$$P(x) = \left(\frac{1}{2\pi\sigma^2} \right)^{1/2} \exp \left[-\frac{(x-\mu)^2}{2\sigma^2} \right] \quad (1)$$

When x is distributed according the above density, we write $x \sim \mathcal{N}[\mu, \sigma^2]$. Equivalently, the variable $z = \sigma^{-1}(x - \mu)$ is distributed according a so called standard Gaussian, i. e. $z \sim \mathcal{N}[0, 1]$ and

$$P(z) = \left(\frac{1}{2\pi}\right)^{1/2} \exp\left(-\frac{z^2}{2}\right) \quad (2)$$

A simple algorithm to sample from the Gaussian in Eq. (2) has been developed by [Box and Muller \[1958\]](#). The algorithm can be easily implemented on a personal laptop. Given two random variables s, t , both uniformly distributed in the unitary interval¹, the random variables

$$Z_c = \sqrt{-2 \ln s} \cos(2\pi t) \quad (3)$$

$$Z_s = \sqrt{-2 \ln s} \sin(2\pi t) \quad (4)$$

are both distributed according a standard Gaussian, i. e. $Z_{c,s} \sim \mathcal{N}[0, 1]$. As an immediate consequence related to the transformation between RVs [[Gardiner, 2012](#)], the RV x defined as

$$x = \mu + \sigma Z_{c,s} \quad (5)$$

will be distributed according a Gaussian $\mathcal{N}[\mu, \sigma^2]$. The derivation of the Box-Muller algorithm follows from the following observation. Two independent standard Gaussian variables Z_c and Z_s are characterized by the joint PDF (Probability Density Function)

$$P(Z_c, Z_s) = \frac{1}{2\pi} \exp\left(-\frac{Z_c^2 + Z_s^2}{2}\right)$$

or, if we define polar coordinates,

$$\begin{cases} Z_c = R \cos \theta \\ Z_s = R \sin \theta \end{cases}$$

the original joint distribution becomes

$$P(R, \theta) dR d\theta = \left(\frac{1}{2\pi}\right) e^{-R^2/2} R dR d\theta$$

We can immediately integrate out the θ contribution from the above expression by integrating over all the possible allowed angles ($\theta \in [0, 2\pi]$). In particular,

$$\frac{1}{2\pi} \int_0^{2\pi} d\theta = 1$$

i. e., $\theta \sim \mathcal{U}[0, 2\pi]$. What remains is the distribution

$$P(R) dR = e^{-R^2/2} R dR$$

or, by introducing the variable transformation $\xi = R^2/2$ ($d\xi = R dR$)

$$P(\xi) = e^{-\xi} d\xi$$

The above PDF is associate to a distribution function

$$F(Q) = \int_0^Q d\xi e^{-\xi}$$

¹ $s, t \sim \mathcal{U}[0, 1]$

We remember that $F(Q)$ is the probability that $0 \leq \xi \leq Q$ (i. e. $F(Q) \in [0, 1]$). Performing the above integral we get

$$-\ln[1 - F(Q)] = Q$$

with $1 - F(Q)$ a number between 0 and 1. Thus, if we draw from the uniform distribution between 0 and 1 a RV s , the variable

$$\sqrt{-2 \ln s} = R$$

will be distributed according the density $P(R)dR = e^{-R^2/2} R dR$. Drawing $s \sim \mathcal{U}[0, 1]$ and $t \sim \mathcal{U}[0, 1]$, the variables

$$\begin{cases} Z_c = R \cos \theta \\ Z_s = R \sin \theta \end{cases}$$

with $R = \sqrt{-2 \ln s}$ and $\theta = 2\pi t$, by construction, are distributed according two independent Gaussians of 0 mean and unitary variance. At this point, using Eq. (5), we can sample from any kind of Gaussian $\mathcal{N}[\mu, \sigma^2]$

3 FORTRAN CODE

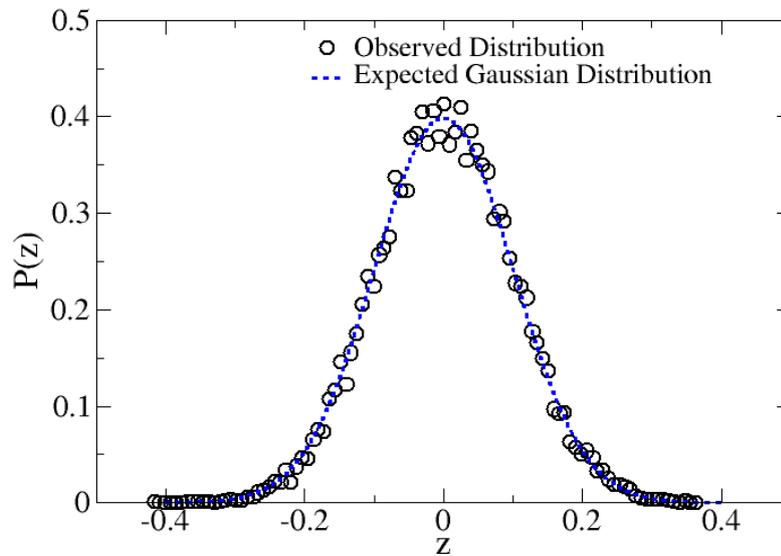


Figure 1: The Listing 1 simulates N_s Independent Gaussian variables with mean $\mu = 1.2$ and standard deviation $\sigma = 10^{-1}$ using the Box–Muller algorithm introduced in Sec. 2. The code in Listing 1 returns a data file “Histo.dat” with three columns. the first one is the abscissa “ x ”, the second one gives the *integer* counts in x and $x + dx$; finally, the last column gives the probability density (normalized to one). Here we show the density of the standardized variable z , obtained by replacing the first column of the file “Histo.dat” with the variable $z = \sigma^{-1}(x - \mu)$. At the same time, the third column, say y , is replaced by $\sigma y \equiv P(z)$.

Listing 1: Fortran code for the implementation of the Box–Muller algorithm introduced in Sec. 2

```

1  | *****
2  | ! This program Shows how to generate Gaussian Random Variables using
3  | ! the Box - Muller Algorithm. To make the histogram of the output
4  | ! data, I also introduce an easy subroutine for 1D histograms.
5  | ! The program shows the basic usage of vectors in Fortran.
6  | *****
7  |
8  | *****
9  | ! Author:      Giuseppe Forte
10 | ! First release: 23/09/2017
11 | ! e-mail:      giuseppe.forte@giuseppeforte.me
12 | ! Website:     https://www.giuseppeforte.me
13 | ! License:     Creative Commons Attribution-ShareAlike 4.0
14 | !              International (CC BY-SA 4.0)
15 | !              (https://creativecommons.org/licenses/by-sa/4.0/)
16 | *****
17 |
18 |
19 | *****
20 | ! The following module contains the subroutine to make the Histogram
21 | ! of a 1D Random variable.
22 | ! Remember to use the current module in the main program by
23 | ! "uploading" it with the instruction
24 | !
25 | !       use Histogram
26 | !
27 | ! written immediately after the instruction program, i.e.
28 | !
29 | !       program Prog_Name
30 | !---> use Histogram
31 | !     implicit none
32 | !     [main program body]
33 | !     end program Prog_Name
34 | *****
35 |
36 |     module Histogram
37 |     contains
38 |
39 |     subroutine MakeHist(Vec,Nbin,Mids,Counts,Density)
40 |     implicit none
41 |     real*8, dimension(:),intent(in):: Vec ! Vec is a vector
42 |     ! variable, its dimension is not specified (otherwise
43 |     ! I would have used the instruction dimension(N), with N the
44 |     ! predetermined dimension of the vector).
45 |     ! That means that a certain memory space will be pre-allocated
46 |     ! by fortran, thus the memory is left free and
47 |     ! definitively allocated only elsewhere. This procedure is
48 |     ! called dynamical memory allocation, in contrast with a
49 |     ! fixed allocation, like the vector Counts(Nbin) below, for example,
50 |     ! whose dimension is fixed and given by Nbin. In particular,
51 |     ! Counts is a set of numbers, Counts(1), Counts(2), ...,
52 |     ! Counts(Nbin). Such dimension (Nbin) cannot be changed during
53 |     ! the execution of the program. On the other hand,
54 |     ! Vec is an array whose dimension is not specified in the
55 |     ! declaration section and can be allocated and REALLOCATED many
56 |     ! times, as the program develops. Here we shall be concerned
57 |     ! with a single allocation of the memory. An example with
58 |     ! multiple allocations will be discussed in another note
59 |     integer, intent(in):: Nbin ! Number of bins
60 |     integer, intent(out):: Counts(Nbin) ! counting the number
61 |     ! of events within a given bin
62 |     real*8, intent(out):: Mids(Nbin), Density(Nbin)
63 |     ! Density = the renormalized counts to make the integral of
64 |     ! the density equal to one

```

```

65 | !      Minds = mid points within the range of the random variable
66 |      real*8   xmax,xmin,dx,Pos,Norm,inflim,suplim
67 |      integer  k
68 |      logical,allocatable:: Cond(:) ! this line is completely
69 | !      equivalent to the statement
70 | !      logical, dimension(:):: Cond (Here we are defining a logical
71 | !      vector, whose dimension is not declared, again, dynamical
72 | !      memory allocation)
73 |
74 |      ! In order to allocate a vector dynamically allocated, like
75 | !      the vector "Vec" (or Cond) we should write an instruction like
76 | !      allocate(Vec(SpecifyDimension)), however, in this case the
77 | !      vector Vec is AUTOMATICALLY allocated. Indeed, Vec is an input
78 | !      argument of the subroutine, meaning that, once we pass the
79 | !      argument to the subroutine, Vec will be allocated in
80 | !      agreement with the dimension of the passed vector
81 |      xmax = maxval(Vec)
82 |      xmin = minval(Vec)
83 |      dx   = (xmax-xmin)/dble(Nbin)
84 |      Pos  = xmin
85 |      k    = 1
86 |      open(unit=15, file="Histo.dat",status="unknown")
87 |
88 |      do 108 while(Pos.le.xmax)
89 |          suplim = Pos + dx
90 |          inflim = Pos
91 |          Cond   = Vec.lt.suplim ! here, the logical vector "Cond"
92 | !          is AUTOMATICALLY allocated by comparing it with the vector
93 | !          Vec. In order to make the equivalence syntactically correct
94 | !          Cond must be characterized by the same dimension of the expression
95 | !          on the right, thus it has to have the same dimension of Vec
96 |          Cond   = Cond.and.(Vec.ge.inflim)
97 |          Counts(k) = count(Cond) !counts the number of true conditions
98 | !          in cond (i.e. the number of random variables in the current
99 | !          interval)
100 |          Mids(k)  = inflim + 0.5d0*dx
101 |          Pos      = Pos + dx
102 |          k        = k + 1
103 | 108      end do
104 |      Norm   = sum(Counts)*dx
105 |      Density = dble(Counts)/Norm ! sum(Density)/Norm = 1!!!!
106 |
107 |      do k=1, Nbin
108 |          write(15,*)real(Mids(k)),Counts(k),real(Density(k))
109 |      end do
110 |      close(15)
111 |      return
112 |      end subroutine MakeHist
113 |
114 |      end module Histogram
115 |
116 |
117 | !*****
118 | !      MAIN PROGRAM
119 | !*****
120 |      program BoxMULLER
121 |      use Histogram
122 |      implicit none
123 |      integer i !dummy index
124 |      integer Ns, Nbin ! Ns =Numbers of samples
125 |      parameter(Ns = 10000, Nbin = 100)
126 | !      Nbin = number of bins in the histogram
127 |      real*8 x(Ns)! x is a vector of dimension Ns, i.e.
128 | !      it has got Ns components. We can access the generic
129 | !      component "i" (i=1,2,3,...,Ns) in the vector x
130 | !      simply by writing x(i)
131 |      real*8 Mids(Nbin), Density(Nbin)

```

```

132     integer Counts(Nbin)
133     ! The abscissa of the histogram of the RV x runs from
134     ! the minimum of x to the maximum of x, divided in
135     ! Nbin intervals. The mid point of such intervals is stored
136     ! in the vector Mids. The number of RVs falling in a given
137     ! interval is stored in the vector CoutS. Finally the
138     ! normalized histogram is stored in the vector Density
139     real*8 u1,u2 !convenience variables
140     real*8 Mu, Sigma ! Mu = mean of the target Gaussian
141                     ! Sigma = Standard Deviation of the target
142                     ! Gaussian
143     parameter(Mu = 1.2d0, Sigma = 0.1d0)
144     real*8 TwoPi ! TwoPi = 2 * Greek Pi = 2 * 3.141593... =
145     ! = 6.283185...
146     parameter(TwoPi = 8.0d0*atan(1.0d0))
147
148     ! Initialization of the vector "x"
149     x = 0.0d0 ! this instruction puts ALL the components
150     ! of the vector "x" equal to 0
151     !the above instruction is "equivalent" to the set
152     ! of instructions
153     ! do i=1, Ns
154     !   x(i) = 0.0d0
155     ! end do
156     ! Actually, the instruction is not exactly the same. indeed, if we
157     ! use the cycle do i=1...end do, Fortran will execute the instructions
158     ! sequentially on the same processor. If we use the instruction x=0.0d0
159     ! Fortran will run the instruction AUTOMATICALLY on ALL the AVAILABLE
160     ! processors: in other words x=0.0d0 will be parallelized (not bad,
161     ! considering that we are not making explicit use of MPI. This is a
162     ! feature which has been introduced only in the modern versions
163     ! of Fortran.)
164
165     ! Generation of the Gaussian random variables with mean Mu
166     ! and Standard Deviation Sigma (Variance = Sigma^2)
167     do 100 i=1, Ns
168         call random_number(u1)
169         call random_number(u2)
170         ! Implementation of the Box-Muller Algorithm
171         x(i) = Mu + Sigma*sqrt(-2.0d0*log(u1))*cos(TwoPi*u2)!
172         ! Box-Muller
173     100 end do
174
175     ! Construction of the histogram (frequencies) and
176     ! associated probability density
177     call MakeHist(x,Nbin,Mids,Counts,Density)
178     ! the subroutine MakeHist is contained in the module Histogram
179     ! Such subroutine takes in input the vector "x", i.e. the
180     ! "observed" random values (here we are simulating such values)
181     ! and the number of bins (Nbin). The output are the vectors Mids,
182     ! Counts and Density, TOGETHER with a data file "Histo.dat"
183     ! containing three columns, specifically
184     !
185     ! Mids(k), Counts(k), Density(k) (k=1,2,...,Nbin)
186     !
187
188     end program BoxMULLER

```

A FURTHER COMMENTS AND EASY MODIFICATIONS

Listing 2: external module "Histogram" used by the code in Listing 3

```

1 | | /*****

```

```

2  ! This module contains the subroutine to make the Histogram of a 1D
3  ! Random variable. This is the code for the module ONLY.
4  ! You might thus use it in any other Fortran code, provided that
5  ! you properly recall it and compile all the files in the correct
6  ! way (please, check Listing3.f in order to learn how to compile
7  ! a fortran code which calls a routine written in another
8  ! file)
9  !*****
10
11 !*****
12 ! Remember to use the current module in the main program by
13 ! "uploading" it with the instruction
14 !
15 !         use Histogram
16 !
17 ! written immediately after the instruction program ....i.e.
18 !
19 !         program Prog_Name
20 !---> use Histogram
21 !         implicit none
22 !         [main program body]
23 !         end program Prog_Name
24 !         YOU ALSO NEED TO COMPILE YOUR MAIN PROGRAM CALLING AT
25 !         THIS MODULE PROPERLY (please, check Listing3.f, which is
26 !         code calling at this module)
27 !*****
28
29     module Histogram
30     contains
31
32     subroutine MakeHist(Vec,Nbin,Mids,Counts,Density)
33     implicit none
34     real*8, dimension(:),intent(in):: Vec
35     integer, intent(in):: Nbin
36     integer, intent(out):: Counts(Nbin)
37     real*8, intent(out):: Mids(Nbin),Density(Nbin)
38     real*8  xmax,xmin,dx,Pos,Norm,inflim,suplim
39     integer k
40     logical,allocatable:: Cond(:)
41
42
43     xmax = maxval(Vec)
44     xmin = minval(Vec)
45     dx   = (xmax-xmin)/dble(Nbin)
46     Pos  = xmin
47     k    = 1
48     open(unit=15, file="Histo.dat",status="unknown")
49
50     do 108 while(Pos.le.xmax)
51         suplim = Pos + dx
52         inflim = Pos
53         Cond   = Vec.lt.suplim
54         Cond   = Cond.and.(Vec.ge.inflim)
55         Counts(k) = count(Cond)
56         Mids(k)   = inflim + 0.5d0*dx
57         Pos      = Pos + dx
58         k        = k + 1
59 108     end do
60     Norm   = sum(Counts)*dx
61     Density = dble(Counts)/Norm
62
63     do k=1, Nbin
64         write(15,*)real(Mids(k)),Counts(k),real(Density(k))
65     end do
66     close(15)
67     return
68     end subroutine MakeHist

```

```

69 |
70 |         end module Histogram

```

So far, we have used a code to generate Gaussian RVs and make the histogram of such variables. Now, let us assume that we have an experimental sample $x = (x_1, x_2, \dots, x_{N_s})$ of unspecified size N_s , collected in the file “Experimental_Observations.dat”. Our goal is to write a program to determine the size of the sample, i. e. N_s and then make the distribution related to such data. Following the coding technique suggested by [Markus and Metcalf \[2012\]](#) (see also the associated web-site [here](#)), we propose the code in Listing 3. Such code uses the external module “Histogram” coded in Listing 2. This is the same module used in Listing 1, however, now such module is written in a file which is external to the main program shown in Listing 3. We thus need to learn how to compile the code in Listing 3 in the correct way, i. e. we need to know how to *pass* to the compiler the module in Listing 2. Let us assume that we call the code in Listing 2 with the name “Listing2.f”. Now, let us assume that the *source code containing the main program is written in “Listing3.f”*. In order to put everything together we will compile the main program in Listing 3 using the instruction

- `gfortran -O3 Listing2.f Listing3.f -o out.x`

Listing 3: Fortran Code to make the histogram of an external file of unspecified size

```

1 | !*****
2 | ! This program Shows how to make the histogram of a RV contained
3 | ! in a given external file, say Experimental_Observations.dat. This
4 | ! file might be a file passed to you by anyone of your collaborators.
5 | ! The program makes use of an external module, i.e. the Fortran code
6 | ! for such an external module is written elsewhere in a
7 | ! file DIFFERENT from the current one
8 | !*****
9 |
10 | !*****
11 | ! Author:      Giuseppe Forte
12 | ! First release: 23/09/2017
13 | ! e-mail:      giuseppe.forte@giuseppeforte.me
14 | ! Website:     https://www.giuseppeforte.me
15 | ! License:     Creative Commons Attribution-ShareAlike 4.0
16 | !              International (CC BY-SA 4.0)
17 | !              (https://creativecommons.org/licenses/by-sa/4.0/)
18 | !*****
19 |
20 | ! MAIN PROGRAM !!!!!
21 |     program hist
22 |     use HistoGram ! Look: here we are using a module (HistoGram).
23 |     ! Such module is not coded in the present file. It is, however,
24 |     ! contained in another file (Listing2.f). We use it here.
25 |     ! however, in order to properly recall such module, we need to be
26 |     ! carefull when we compile the present file (say Listing3.f: the
27 |     ! present file, THIS FILE)
28 |     ! IF WE COMPILE THE CURRENT FILE IN THE USUAL WAY, i.e.
29 |     ! gfortran -O3 Listing3.f -o out.x
30 |     ! we WILL GET AN ERROR MESSAGE from the compiler. No problem
31 |     ! In order to compile our codes correctly we just need to perform
32 |     ! two operations:
33 |     ! 1) Move the file Listing2.f IN THE SAME FOLDER OF THE FILE
34 |     !    Listing3.f
35 |     ! 2) Compile Listing3.f with the following instruction
36 |     !    gfortran -O3 Listing2.f Listing3.f -o out.x
37 |     !    that's it :-)
38 |     ! typically, if you write

```

```

39 | !      gfortran -O3 Listing3.f Listing2.f -o out.x
40 | !      should be the same, however, some Operative Systems
41 | !      require you to put Listing2.f first
42 | !      some others require you to put Listing3.f first
43 | !      (....it's a matter of machines...who knows
44 | !      what inside their "mind"? :-))
45 | !      implicit none
46 | !      integer Nbin,Ns
47 | !      parameter(Nbin = 80)
48 | !      real*8, allocatable:: x(:)
49 | !      real*8 Density(Nbin),Mids(Nbin)
50 | !      integer Counts(Nbin)
51 | !      real*8 u
52 | !      integer i, nodata, ierr, k
53 |
54 |
55 | !      open(unit=10,file="Experimental_Observations.dat",
56 | &      status="old",iostat=ierr) ! when we write iostat = ierr
57 | ! the INTEGER number ierr receives the error status from
58 | ! an open instruction. This means that the program
59 | ! executes the open instruction and, if no errors
60 | ! happen during such instruction the INTEGER variable
61 | ! ierr is set equal to 0. Otherwise it will acquire
62 | ! some positive number. For exemple, if in line 55, we
63 | ! write file= "Eperimental_Opservation.dat", i.e. the
64 | ! name of the file does not match the name written in the
65 | ! instruction, ierr will be assigned a positive number.
66 | ! If everything is correct, ierr will be assigned 0.
67 |
68 | !      if(ierr.ne.0)then
69 | !      write(*,*)"file Experimental_Observations.dat could not"
70 | !      write(*,*)"be opened. Program ended!"
71 | !      goto 200 ! because a problem occurred, the programe
72 | !      ! is ended
73 | !      end if
74 |
75 | !      i = 1
76 | !      ! if everything went well, at this point we still have
77 | !      ! ierr = 0.
78 | !      do 100 while(ierr == 0)
79 | !      read( 10, *, iostat = ierr )u !here we are using
80 | !      ! the read instruction together with the iostat
81 | !      ! assignment. What does that mean? After executing
82 | !      ! the read instruction in line 79, the fortran
83 | !      ! compiler will put an INTEGER number in the variable
84 | !      ! ierr. Three different situation might occur
85 | !      ! 1) if ierr = 0, the read instruction works in the
86 | !      ! usual way (and the do while statement keeps on
87 | !      ! going)
88 | !      ! 2) if ierr > 0, it means that an error has occurred
89 | !      ! during the file reading. For example, if in the file
90 | !      ! "Experimental_Observations.dat", at a certain point,
91 | !      ! there is an alphanumeric variable, such as NaN,
92 | !      ! that would cause the compiler to report an
93 | !      ! error. That is because in THIS PROGRAM
94 | !      ! , i.e. the CURRENT EXAMPLE, the variable
95 | !      ! we would like to read from the file
96 | !      ! Experimental_Observations.dat is defined in line 51
97 | !      ! as a double precision real variable
98 | !      ! and NOT an alphanumeric variable. In this
99 | !      ! case the fortran compiler will assign a
100 | !      ! positive number to ierr
101 | !      ! 3) if ierr < 0, it means that the end of the file
102 | !      ! has been reached (the last line has been read)
103 | !      if(ierr>0)then
104 | !      write(*,*) "Error reading the data!"
105 | !      goto 200

```

```

106         elseif(ierr<0)then
107             exit ! Reached the end of the file
108         endif
109         i = i + 1
110     100     end do
111         Ns = i - 1 ! Ns = size of the sample
112         close(10)
113         ! Now we know the size of the sample in
114         ! Experimental_Observations.dat, we call such size Ns
115         ! and we are ready to make the histogram
116         open(unit=13,file="Experimental_Observations.dat",
117             & status="old")
118
119         allocate(x(Ns)) !allocation of the variable x using the
120         ! size Ns
121         do 101 i =1, Ns
122             read(13,*)x(i) !store in x the content of the file
123             ! Experimental_Observations.dat
124     101     end do
125         close(13)
126
127         call MakeHist(x,Nbin,Mids,Counts,Density) ! call the
128         ! histogram routine. Such routine
129         ! IS IN THE MODULE HistoGram contained in an
130         ! external file, for example called Listing2.f
131         deallocate(x) !deallocate the memory reserved for x
132     200     end program hist
133     !! REMEMBER REMEMBER REMEMBER REMEMBER REMEMBER REMEMBER
134     !! REMEMBER REMEMBER REMEMBER REMEMBER REMEMBER REMEMBER
135
136     ! In this example we have
137     ! 1) Listing3.f (the current file)
138     ! 2) Listing3.f DEPENDS ON a routine CONTAINED in
139     !    Listing2.f
140     ! 3) MOVE Listing2.f in the same folder of Listing3.f
141     ! 4) COMPILE WITH THE INSTRUCTION
142     ! 5) gfortran -O3 Listing2.f Listing3.f -o out.x
143     ! HAVE FUN :-) !!!!

```

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