

Methods of Computational Physics - I (PHY637MJ)

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Contents

Program 1 - Gaussian Integral Estimation

Aim: To estimate the integral $\int_{-\infty}^{+\infty} dx e^{-\alpha x^2}$ by the Trapezoidal method to a given precision (by calculating the truncation error from the analytical value) as well as through adaptive subintervals

We will numerically compute the integral $\int_0^{\infty} \exp(-\alpha x^2)$. As we necessarily have to set some finite upper limit in numerics, there are two sources of error in the obtained result: one from the truncation of the upper limit and the truncation error of the numerical method itself.

To eliminate the former, we choose the upper limit of $4/\sqrt{2\alpha}$ corresponding to a coverage of 8σ in the full integral. As for the latter, know that for a function $f(x)$ integrated over the interval $[a, b]$ with step-size h using the trapezoidal method, the truncation error is given by

$$E_T \leq \frac{(b-a)}{12} h^2 \times \max|f''(x)|$$

Analytically for the gaussian with as seen in the below figure, the absolute value of the second derivative attains its maximum at $x = 2\alpha$, hence

$$E_T \leq \frac{\sqrt{2\alpha}}{3} h^2$$

If δ is the maximum amount of error we are willing to tolerate, we may choose any h such that

$$h^2 \leq \frac{3\delta}{\sqrt{2\alpha}}$$

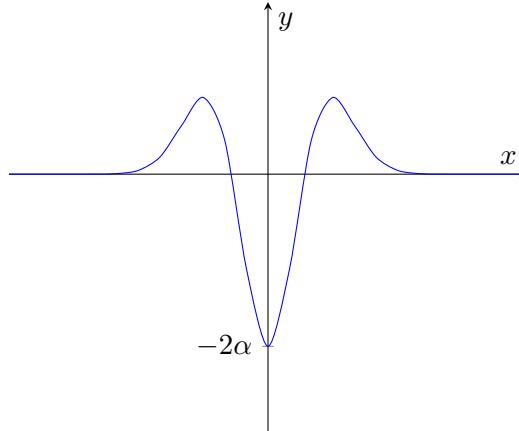


Figure 1: $f''(x) = 4\alpha^2 x^2 e^{-\alpha x^2} - 2\alpha e^{-\alpha x^2}$ for $\alpha = 1$

```

! AYUSH PRAVIN SHENOY
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!
! To estimate the gaussian integral by the Trapezoidal method to a given precision
! (by calculating the truncation error from the analytical value) as well as through
! adaptive subintervals

program GINTEG

    use INTEGRATE
    use FORMULA

    implicit none

    real :: ALPHA          ! Gaussian parameters
    real :: THR            ! Error tolerance

    real :: B              ! Integration Upper Limit
    real :: TRUVAL         ! Analytical value
    real :: INTEG          ! Numerical result
    real :: H              ! Subinterval size
    integer :: N            ! No of subintervals
    real :: DELTA

    integer::I

    read(*,*) ALPHA
    read(*,*) THR

    B = 4.0/sqrt(2*ALPHA)      ! 4 Sigma
    TRUVAL = 0.5*sqrt(PI/ALPHA) ! Half-Integral

    write(*,*) "Alpha           : ", ALPHA
    write(*,*) "Error Threshold (1E): ", -THR
    write(*,*)

    ! Trapezoidal with analytical error bound
    DELTA = 10**(-THR)
    H = sqrt(3.0/sqrt(2*ALPHA))*sqrt(DELTA)
    N = B/H

    write(*,*) "DELTA :", DELTA
    write(*,*) "      H :", H
    write(*,*) "      N :", N
    write(*,*)

    call TRAPEZOID(F,0.0,B,N,INTEG)

    write(*,*) "# ANALYTICAL ERROR BOUND"
    call PRINT_HEADER
    write(*,2) N, INTEG, abs(INTEG-TRUVAL), 100.0*(INTEG-TRUVAL)/TRUVAL
    call PRINT_RESULT(INTEG,TRUVAL)

```

```

! Trapezoidal with adaptive subintervals
write(*,*) "# TRAPEZOIDAL METHOD"
call PRINT_HEADER

N=1
do
    call TRAPEZOID(F,0.0,B,N,INTEG)
    write(*,2) N, INTEG, abs(INTEG-TRUVAL), 100.0*(INTEG-TRUVAL)/TRUVAL
    if (abs(2*(INTEG-TRUVAL)) <= 1E-3) exit
    N = N*2
end do
call PRINT_RESULT(INTEG,TRUVAL)

! Simpson with adaptive subintervals
write(*,*) "# SIMPSONS 1/3 RULE"
call PRINT_HEADER

N=1
do
    call SIMPSON(F,0.0,B,N,INTEG)
    write(*,2) N, INTEG, abs(INTEG-TRUVAL), 100.0*(INTEG-TRUVAL)/TRUVAL
    if (abs(2*(INTEG-TRUVAL)) <= 1E-3) exit
    N = N*2
end do
call PRINT_RESULT(INTEG,TRUVAL)
!write(*,*) H, 2*INTEG, TRUVAL, abs(2*INTEG - TRUVAL)

2 format (2X,I15,4X,3(F15.10,4X))      ! Values

contains

    real function F(X)

        real, intent(in) :: X

        F = exp(-ALPHA*(X**2))
    end function F

    subroutine PRINT_HEADER()
        write(*,1) "#", "N_INTERVALS", "INTEGRAL", "ABS ERROR", "% ERROR"
        write(*,1) "#", ("-----", I=1,4)

        1 format (A1,1X,4(A15,4X))      ! Table header
    end subroutine

    subroutine PRINT_RESULT(INTEG,TRUVAL)

        real, intent(in) :: INTEG
        real, intent(in) :: TRUVAL

        write(*,*)
        write(*,*) "Converged value : ", 2*INTEG
        write(*,*) "Analytical value : ", 2*TRUVAL

```

```
    write(*,*)
end subroutine

end program GINTEG
```

Flowchart:

Input:

1.0
3

Output:

```
Alpha : 1.00000000
Error Threshold (1E) : -3.00000000

DELTA : 1.00000005E-03
H : 4.60577980E-02
N : 61

# ANALYTICAL ERROR BOUND
# N_INTERVALS      INTEGRAL      ABS ERROR      % ERROR
# -----          -----          -----          -----
       61           0.8861705065   0.0000564456   -0.0063692038

Converged value : 1.77234101
Analytical value : 1.77245390

# TRAPEZOIDAL METHOD
# N_INTERVALS      INTEGRAL      ABS ERROR      % ERROR
# -----          -----          -----          -----
       1            1.4146879911   0.5284610391   59.6304397583
       2            0.8987370133   0.0125100613   1.4116091728
       4            0.8861057162   0.0001212358   -0.0136799999

Converged value : 1.77221143
Analytical value : 1.77245390

# SIMPSONS 1/3 RULE
# N_INTERVALS      INTEGRAL      ABS ERROR      % ERROR
# -----          -----          -----          -----
       1            0.9431253076   0.0568983555   6.4202919006
       2            0.7267533541   0.1594735980   -17.9946689606
       4            0.8818953037   0.0043316483   -0.4887741506
       8            0.8861675858   0.0000593662   -0.0066987611

Converged value : 1.77233517
Analytical value : 1.77245390
```

Program 2 - Pi Estimation by Acceptance Rejection Method

Aim: To estimate the value of π to within a given precision using the acceptance-rejection method along with ensemble-averaging

In this method, a variation of which was proposed by Laplace in 1812, we actually estimate the volume of a circle \mathcal{C} inscribed within the square region $\mathcal{S} = [-1, +1] \times [-1, +1] \subset \mathbb{R}^2$

We sample two uniformly distributed numbers x_i, y_i , with the distribution scaled such that $x_i, y_i \in [-1, +1] \subset \mathbb{R}$. As the two samples are independent, we may regard a series of N such pairs (x_i, y_i) as uniformly distributed points in \mathcal{S}

Given that the distribution of points is uniform, we have for a given sampled point p :

$$\begin{aligned}\text{Prob}(p \in \mathcal{C}) &= \frac{\text{Area}(\mathcal{C})}{\text{Area}(\mathcal{S})} \\ &= \frac{\pi}{4}\end{aligned}$$

The probability in the above expression is estimated by the fraction of points out of N that lie inside the circle. Thus we have

$$\pi = 4 \times \frac{\text{no. of points inside } \mathcal{C}}{N}$$

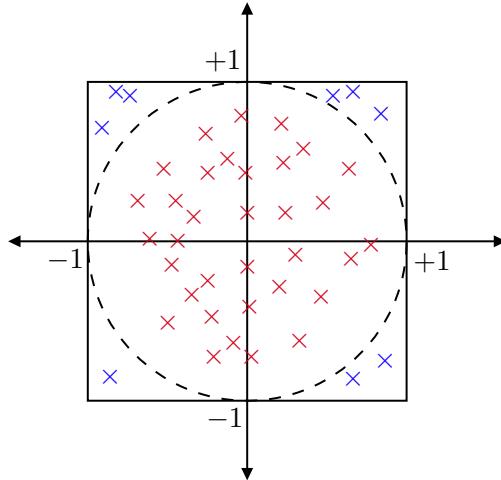


Figure 2: Diagrammatic representation of the acceptance-rejection method.

Flowchart:

Program:

```
! AYUSH PRAVIN SHENOY
! 24021014
!
! To estimate the value of pi within a given precision using the acceptance-rejection
! method along with ensemble-averaging

program PIMONTEC

    use iso_fortran_env, only: IK => int32, RK => real64

    use MONTECARLO

    implicit none

    integer :: N          ! ensemble size
    real   :: THR         ! ERR < 1.0E-(THR)
    integer :: M          ! Number of throws

    integer :: J
    real   :: GUESS
    real   :: ENSAV

    double precision,parameter::PI=4*atan(1.d0)

    read(*,*) THR
    read(*,*) N
    read(*,*) M

    write(*,*) "Error Threshold (1E): ", -THR
    write(*,*)
    write(*,1) "N_THROWS", "GUESS", "ABS ERROR", "PERCENT ERROR"
    write(*,1) ("-----", J=1,4)

    THR = 10**(-THR)

    do

        ! Populate ensemble
        ENSAV=0
        do J=1,N
            call ACCREJ_SPHERE(1.0,2,M,GUESS)
            ENSAV = ENSAV + GUESS
        end do
        ENSAV = ENSAV/N

        ! Output results
        write(*,2) M, ENSAV, ENSAV-PI, 100*(ENSAV-PI)/PI

        ! Check for convergence
        if (abs(ENSAV - PI) .lt. THR) exit

    end do
```

```
M = 10*M  
end do  
  
write(*,*)  
write(*,*) "Converged value of PI is: ", ENSAV  
  
1 format (A15,4X,3(A15,4X))  
2 format (I15,4X,3(F15.10,4X))  
  
end program PIMONTEC
```

Input:

```
4  
100  
10
```

Output:

```
Error Threshold (1E): -4.00000000
```

N_THROWS	GUESS	ABS ERROR	PERCENT ERROR
10	3.1279997826	-0.0135928710	-0.4326745230
100	3.1311995983	-0.0103930553	-0.3308212242
1000	3.1408393383	-0.0007533153	-0.0239787703
10000	3.1411597729	-0.0004328807	-0.0137790212
100000	3.1412582397	-0.0003344138	-0.0106447233
1000000	3.1415793896	-0.0000132640	-0.0004222068

```
Converged value of PI is: 3.14157939
```

Program 4 - Gaussian Integral by Monte Carlo Integration

Aim: To estimate the value of a gaussian integral to within a given precision using uniform-sample Monte Carlo integration along with ensemble-averaging

Consider an integral that is perhaps analytically intractable:

$$I = \int_a^b dx f(x)$$

The general Monte-Carlo integration problem is to write I in the form

$$I = \int_a^b dx g(x)h(x)dx$$

where $g(x)$ is a probability distribution we have an efficient method to sample from. We can then note that the integral we wish to solve is simply the expectation value of the function $h(X)$ where X is distributed as $g(x)$. This expectation value can now be estimated by drawing say N samples ξ_i from $g(x)$, thus yielding

$$I = \langle h \rangle \approx \frac{1}{N} \sum_{i=1}^N h(\xi_i)$$

In our case we wish to solve the integral

$$I_\alpha = \int_0^t dx e^{-\alpha x^2}$$

Applying the above procedure, we can insert a uniform distribution (scaled from 0 to t), $U(x) = 1/t$ and obtain

$$I = \frac{t}{N} \sum_{i=1}^N e^{-\alpha \xi_i^2}$$

To compute the error in estimation, we note that the well-tabulated error-function is given by

$$\begin{aligned} \text{erf}(t) &= \frac{2}{\sqrt{\pi}} \int_0^t dx e^{-\alpha x^2} \\ \implies I_\alpha &= \sqrt{\frac{\pi}{4\alpha}} \text{erf}(t) \end{aligned}$$

Flowchart:

```

! AYUSH PRAVIN SHENOY
! 24021014
!
! To estimate the value of a gaussian integral to within a given precision using
! uniform-sample Monte Carlo integration along with ensemble-averaging

program MCINTEG

use FORMULA

implicit none

real :: A ! Gaussian parameter
real :: T ! Interval Endpoint
integer :: N ! Number of samples
real :: THR ! Convergence threshold
integer :: M ! Ensemble size

real::INTEG
integer::I
real::TRUVAL
real,allocatable::R(:)

read(*,*) A
read(*,*) T
read(*,*) N
read(*,*) THR
read(*,*) M

write(*,*) "Error Threshold (1E): ", -THR
write(*,*) "ENSEMBLE SIZE", "INTEGRAL", "ABS ERROR", "PERCENT ERROR"
write(*,1) ("-----", I=1,4)

allocate(R(N))

THR = 10**(-THR)
TRUVAL = 0.5*sqrt(PI/A)*erf(T)

do
    INTEG = 0
    do I=1,M
        call RANDOM_NUMBER(R)
        R = R*T ! Scale dist to T
        R = GAUSSIAN(R,A) ! Evaluate function
        INTEG = INTEG + sum(R)*T/N ! Calculate integral
    end do
    INTEG = INTEG/M

    write(*,2) M, INTEG, INTEG-TRUVAL, 100.0*(INTEG - TRUVAL)/TRUVAL
end program MCINTEG

```

```
! Check convergence
if (abs(INTEG-TRUVAL) .lt. THR) exit

! Adjust ensemble size
M = 2*M

end do

write(*,*)
write(*,*) "The converged value is: ", INTEG
write(*,*) "The tabulated value is: ", TRUVAL

1 format (A15,4X,3(A15,4X))
2 format (I15,4X,3(F15.10,4X))

end program MCINTEG
```

Input:

```
2.0
5
10000
4
10
```

Output:

```
Alpha : 2.00000000
Error Threshold (1E) : -4.00000000
```

ENSEMBLE SIZE	INTEGRAL	ABS ERROR	PERCENT ERROR
10	0.6262360811	-0.0004209876	-0.0671799034
20	0.6258510351	-0.0008060336	-0.1286243498
40	0.6226068735	-0.0040501952	-0.6463176608
80	0.6238328218	-0.0028242469	-0.4506846070
160	0.6251795888	-0.0014774799	-0.2357716858
320	0.6275514364	0.0008943677	0.1427204311
640	0.6267522573	0.0000951886	0.0151899057

```
The converged value is: 0.626752257
```

```
The tabulated value is: 0.626657069
```

Program 5 - Box-Muller Transformation

Aim: To sample two uniformly distributed random variables and calculate their Box-Muller transform, verifying that the result is normally distributed.

The Box-Muller transform of two uniformly distributed random variables u and v is given by

$$\begin{aligned}x(u, v) &= \cos(2\pi v)\sqrt{-2 \ln u} \\y(u, v) &= \sin(2\pi v)\sqrt{-2 \ln u}\end{aligned}$$

Flowchart:

```

! AYUSH PRAVIN SHENOY
! 24021014
!
! To sample two uniformly distributed random variables and calculate their Box-Muller
! transform, verifying that the result is normally distributed

program BOXMULLER

use FORMULA

implicit none

integer :: N           ! Number of samples

integer      :: I
real         :: M(2,2)
real,allocatable :: O(:, :)
real,allocatable :: R(:, :)

read (*,*) N

allocate(O(N,2))
allocate(R(N,2))
call RANDOM_NUMBER(O)

R(:,1) = sqrt(-2*log(O(:,1)))*cos(2*PI*O(:,2))
R(:,2) = sqrt(-2*log(O(:,1)))*sin(2*PI*O(:,2))

do I=1,N
    write(*,1) O(I,:), R(I,:)
end do

M(1,1) = sum(R(:,1))/N
M(1,2) = sum(R(:,1)**2)/N - M(1,1)**2

M(2,1) = sum(R(:,2))/N
M(2,2) = sum(R(:,2)**2)/N - M(2,1)**2

write(*,2) "#"
write(*,2) "#", "X", "Y"
write(*,2) "#", ("-----", I =1,2)
write(*,3) "# Mean : ", M(1,1), M(2,1)
write(*,3) "# Sdev : ", M(2,1), M(2,2)

1 format(4(F15.10,4X))
2 format(A1,8X,A15,4X,A15)
3 format(A9,F15.10,4X,F15.10)

end program BOXMULLER

```

Output:

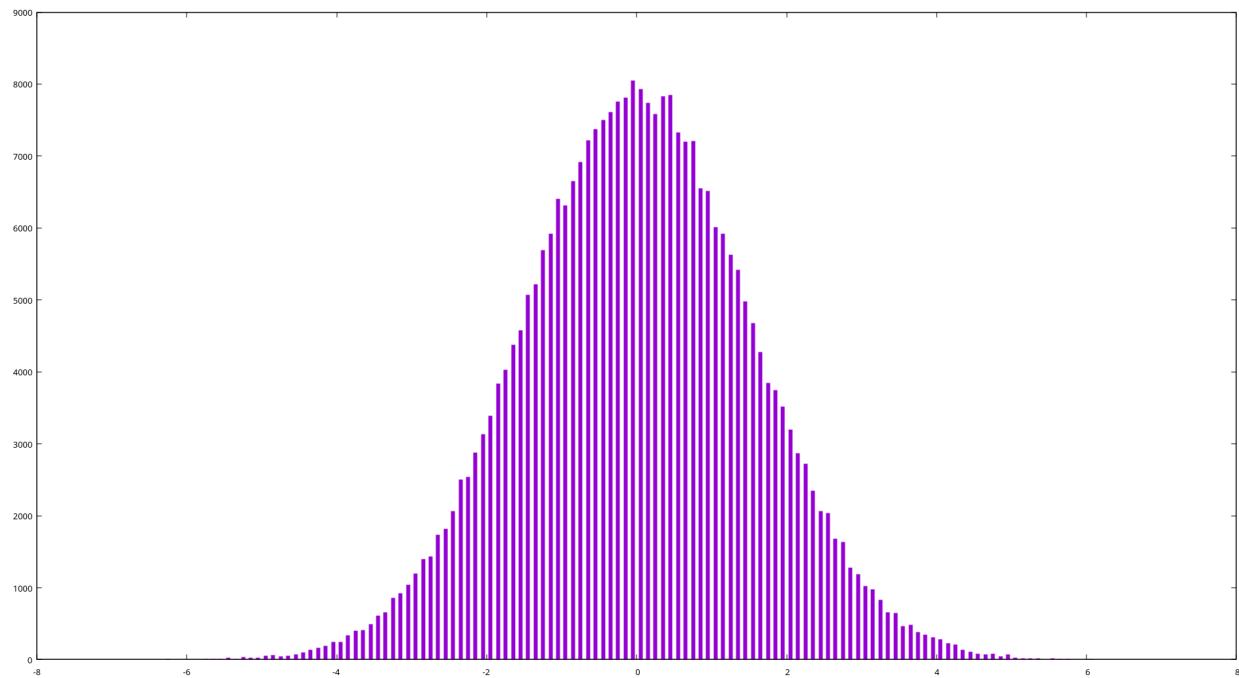


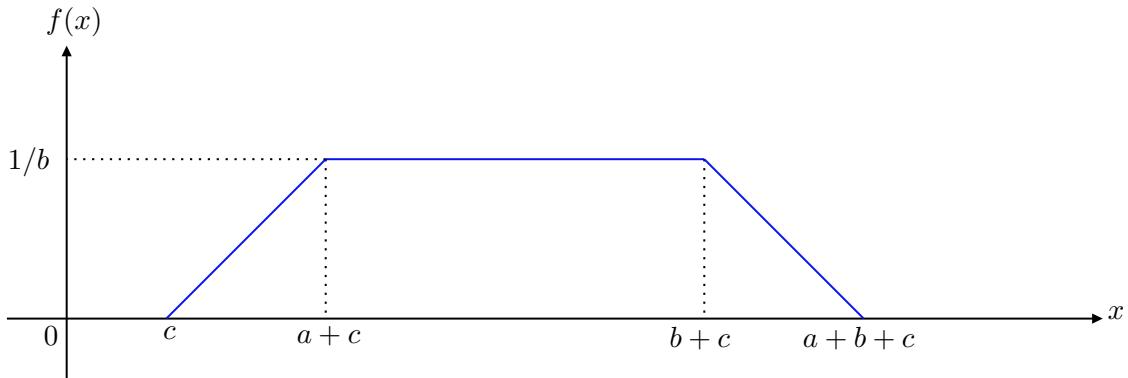
Figure 3: Histogram of $n = 10000$ Box-Muller transformed variables

Program 6 - Trapezoidal distribution

Aim: To sample the given trapezium-shaped distribution by implementing the appropriate transformation of a uniformly distributed random variable.

The given distribution is

$$t(x) = \begin{cases} (x - c)/ab & c \leq x \leq a + c \\ 1/b & a + c \leq x \leq b + c \\ (a + b + c - x)/ab & b + c \leq x \leq a + b + c \end{cases}$$



By integration, the CDF is found to be

$$F_t(x) = \begin{cases} \frac{x^2}{2ab} - \frac{c}{ab}x + \frac{c^2}{2ab} & c \leq x \leq a + c \\ \frac{a}{2b} + \frac{x}{b} - \frac{a+c}{b} & a + c \leq x \leq b + c \\ 1 - \frac{(a+b+c-x)^2}{2ab} & b + c \leq x \leq a + b + c \end{cases}$$

Given that we can efficiently sample a uniformly distributed random variable Y , our method is to find $F_t^{-1}(Y)$ which we know is distributed as $t(y)$. In our case we calculate analytically:

$$F_t^{-1}(Y) = \begin{cases} c + \sqrt{2abY} & 0 \leq x \leq \frac{a}{2b} \\ bY + \frac{a}{2} + c & \frac{a}{2b} \leq x \leq 1 - \frac{a}{2b} \\ a + b + c - \sqrt{2ab(1-Y)} & 1 - \frac{a}{2b} \leq x \leq 1 \end{cases}$$

The above piecewise function of a uniformly distributed random variable is straightforwardly implemented in Fortran. As shown in the below diagram, it is distributed as desired.

Flowchart:

```

program TRAP_DIST

implicit none

integer :: N
real :: A
real :: B
real :: C

real,allocatable :: U(:)
real,allocatable :: R(:)
real :: D
integer :: I

! Read parameters
read (*,*) N
read (*,*) A
read (*,*) B
read (*,*) C

allocate(U(N))
allocate(R(N))
call RANDOM_NUMBER(U)

D = A/(2*B)

! Calculate inverse CDF
do I = 1,N

    if ( U(I) >= 0 .and. U(I) <= D ) then
        R(I) = C + sqrt(2.0*A*B*U(I))
    elseif ( U(I) > D .and. U(I) < (1- D)) then
        R(I) = B*U(I) + 0.5*A + C
    elseif ( U(I) > (1-D) .and. U(I) <= 1.0 ) then
        R(I) = A+B+C - sqrt(2*A*B*(1-U(I)))
    endif

    write(*,1) U(I), R(I)

end do

1 format(2(F15.10,4X))

end program TRAP_DIST

```

Output:

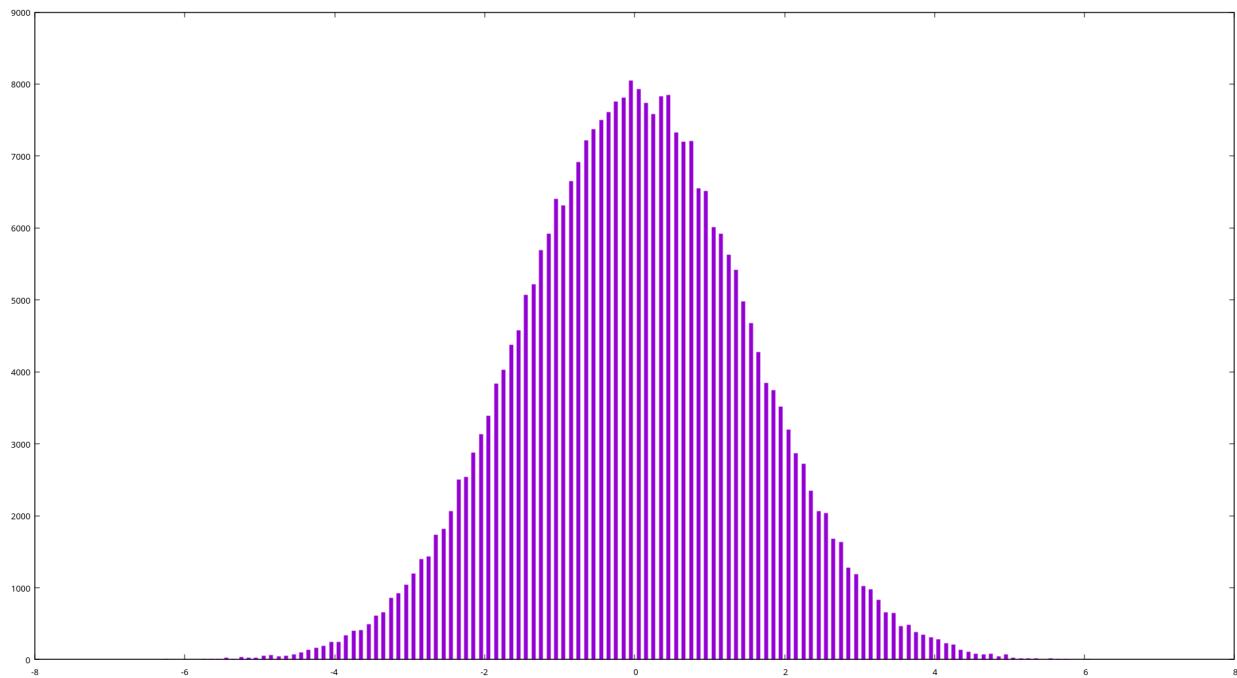


Figure 4: Histogram of $n = 10000$ Box-Muller transformed variables

Program 7 - Ising Model

Aim: To simulate the zero-field Ising model using the Metropolis-Hastings algorithm and calculate the energy, magnetization, magnetic susceptibility and specific heat as a function to temperature and lattice size.

The Ising model is a mathematical model for a ferromagnetic material, defined by placing spins that can either point up or down on a lattice of finite size, say L . To overcome the necessarily finite size in simulation, periodic boundary conditions are imposed. The system is considered to be placed in equilibrium with a bath of temperature T .

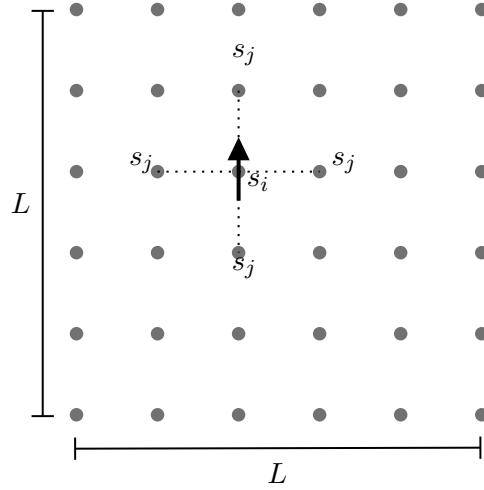


Figure 5: A spin s_i on the lattice and its nearest neighbours s_j

The Hamiltonian is defined to be

$$H = -J \sum_{\langle i,j \rangle} s_i s_j - G \sum_i s_i$$

where the sum $\langle ij \rangle$ is over nearest neighbours as shown in the above diagram. J represents the coupling strength between individual spins while G represents the interaction of the spins with a homogeneous external magnetic field. Observables of interest are the total energy E , total magnetization M , as well as the magnetic susceptibility χ and heat capacity C_V :

$$\begin{aligned}\chi &= \langle M^2 \rangle - \langle M \rangle^2 \\ C_V &= \langle E^2 \rangle - \langle E \rangle^2\end{aligned}$$

As the system is liable to exhibit thermal fluctuations, we must calculate ensemble averages of the above observables. However owing to the large (2^{L^2}) number of configurations, this is

not viable. Hence we resort to Monte Carlo techniques, sampling possible configurations in order to calculate the averages.

As the system is in equilibrium with a heat-bath we expect the configurations to follow Boltzman statistics. We use the Metropolis method to generate a finite of configurations that are representative of the entire phase space, thus yielding the correct ensemble averages. This is done using Markov chains.

The master equation for a Markov process is

$$\rho(X, t+1) - \rho(X, t) = - \sum_{X'} T(X \rightarrow X') \rho(X, t) + \sum_{X'} T(X' \rightarrow X) \rho(X', t)$$

We are interested in the stationary distribution $\rho(X)$ for which the left side vanishes. One possible solution is the detailed balance condition where in the long run the rates of both opposing processes become equal:

$$\frac{T(X \rightarrow X')}{T(X' \rightarrow X)} = \frac{\rho(X, t)}{\rho(X', t)}$$

We the write the transition probabilities in the form

$$T(X \rightarrow X') = A_{XX'} \omega_{XX'}$$

where $0 \leq \omega_{XX'} \leq 1$, $\sum_X \sum_{X'} \omega_{XX'} = 1$, and $\omega_{XX'} = \omega_{X'X}$

We then have

$$\frac{A_{XX'}}{A_{X'X}} = \frac{\rho(X')}{\rho(X)}$$

Our targetted stationary distribution is the Boltzmann distribution, so

$$\begin{aligned} \frac{\rho(X')}{\rho(X)} &= e^{-\beta(H[X'] - H[X])} = e^{-\beta\Delta H} \\ \implies \frac{A_{XX'}}{A_{X'X}} &= e^{-\beta\Delta H} \end{aligned}$$

Flowchart:

```

subroutine PRINT_LATTICE(LATT,U)

    integer, intent(in) :: U                  ! Output unit (6 for stdout)
    integer, intent(in) :: LATT(:,:)        ! Lattice to print
    integer             :: I
    integer             :: J
    integer             :: L

    L = size(LATT(1,:))

    do I = 1,L
        write(U,*) (LATT(I,J), J=1,L)
        write(U,*)
    end do
    !write(U,*) ("-----",I=1,2*L)
    write(U,*)

end subroutine

subroutine PRINT_OBSERVABLES(TIME,E,E_SQ,M,M_SQ,U)
    ! Write observables to output file

    integer, intent(in) :: TIME
    real,    intent(in) :: E
    real,    intent(in) :: E_SQ
    real,    intent(in) :: M
    real,    intent(in) :: M_SQ
    integer, intent(in) :: U

    105 format (2X,I8,4X,4(F20.10,4X))      ! Observables
    write(U,105) TIME, E ,E_SQ, M, M_SQ

end subroutine

integer function WRAP(X,L)
    ! Wrap co-ordinate around if needed

    integer, intent(in) :: X
    integer, intent(in) :: L

    if (X .eq. 0) then
        WRAP = X + L
    elseif (X .eq. L+1) then
        WRAP = X - L
    else
        WRAP = X
    end if

end function

program ISING_MODEL

```

```

implicit none

! Interfaces and other formalities
interface
  subroutine PRINT_LATTICE(LAT, U)
    integer, dimension(:, :, :), intent(in) :: LAT
    integer, intent(in)                   :: U
  end subroutine
end interface
integer, external :: WRAP

! Model Parameters
! -----
real    :: T_START   ! Starting Temperature          (units of kB)
real    :: T_STOP    ! Ending Temperature           (units of kB)
real    :: T_STEP    ! Temperature step            (units of kB)
real    :: J          ! Spin-Spin coupling strength (units of kB)
real    :: G          ! Spin-Field coupling strength (units of J)
integer :: L          ! Lattice size

! Numerical Parameters
! -----
real    :: CONV THR = 1E-3      ! Equilibration Threshold
integer :: EQ_SAMPLES = 100000   ! Number of post-equilibrium samples

! Observables
! -----
real :: M      = 0      ! Net magnetization
real :: M_SQ   = 0      ! Net magnetization squared
real :: E      = 0      ! Total energy
real :: E_SQ   = 0      ! Total energy sq

! Observable Averages
! -----
real :: MEAN_M   = 0
real :: MEAN_M_SQ = 0
real :: MEAN_E   = 0
real :: MEAN_E_SQ = 0

! Response functions
! -----
real :: CHI        ! Magnetic susceptibility
real :: C_V        ! Specific heat

! Other Variables
! -----
! IO and Dummy
character (len=4) :: DUMMY
integer           :: I,P,Q,S
logical           :: VERBOSE

```

```

! Simulation
integer, allocatable :: LATTICE(:,:,:) ! Current state
real :: T ! Current temperature
real :: E_DIFF ! Difference in energy after M-H step
real ,dimension(2) :: R_SF ! Spin-flip roll
real :: R_CA ! Configuration acceptance roll
integer :: X_F ! Spin-flip X-coord
integer :: Y_F ! Spin-flip Y-coord
integer :: TIME ! Timestep counter

! Model Initialization
! -----
! Read input file from standard input
read(*,*) T_START
read(*,*) T_STOP
read(*,*) T_STEP
read(*,*) J
read(*,*) G
read(*,*) L

! Open output files
open(unit=1,file=".isising.out")
open(unit=2,file=".observables.out")
open(unit=3,file=".state.out")

! Write initial inputs
write(1,*) "INPUT PARAMETERS"
write(1,*) ("-", I = 1,16)
write(1,103) "T_STRT = ", T_START
write(1,103) "T_STOP = ", T_STOP
write(1,103) "T_STOP = ", T_STEP
write(1,103) "J = ", J
write(1,103) "G = ", G
write(1,104) "L = ", L
write(1,*)

! Write output headers
write(2,101) "#", "TIME", "E", "E_SQ", "M", "M_SQ"
write(2,101) "#", "-----", ("-----", I=1,4)

! Initialize system
! -----
allocate(LATTICE(L,L))

LATTICE = +1

write(1,*) "INITIAL CONFIGURATION"
write(1,*) "-----"
write(*,*)
call PRINT_LATTICE(LATTICE,1)
write(*,*)

```

```

! Calculate Hamiltonian
E = - G*sum(LATTICE)      ! Spin-Field Interaction Energy
do P = 1,L                 ! Spin-Spin Interaction Energy
  do Q = 1,L
    E = E - J*LATTICE(P,Q)*(LATTICE(P           , WRAP(Q+1,L))& ! Right
                           +LATTICE(WRAP(P+1,L), Q           )) ! Down
  end do
end do

E_SQ = E**2
M = sum(LATTICE)
M_SQ = M**2

!call PRINT_OBSERVABLES(0,E,E_SQ,M,M_SQ,2)

write(1,103) "Initial E   =", E
write(1,103) "Initial M   =", M

T = T_START
write(*,106) "#", "TEMP", "CHI_S", "C_V_S", "E_S", "M_S"
write(*,106) "#", ("-----", I=1,5)

! Run T : [T_START,T_STOP] Experiment
! -----
do

  MEAN_E = 0
  MEAN_M = 0
  MEAN_E_SQ = 0
  MEAN_M_SQ = 0

  TIME = 0

  ! Timesteps
  ! -----

  do TIME = 1, EQ_SAMPLES
    ! Monte Carlo Cycle
    ! -----

    ! Metropolis steps
    do I = 1,SIZE(LATTICE)

      ! Propose configuration

      ! Flip a random spin
      call RANDOM_NUMBER(R_SF)
      R_SF = R_SF*L
      R_SF = int(ceiling(R_SF))
      X_F = R_SF(1)
      Y_F = R_SF(2)
      S = LATTICE(X_F,Y_F)
      LATTICE(X_F,Y_F) = -1*S

```

```

!write(3,*) "TEMP", T, "TIME", TIME
!write(3,*) I, "PROP", X_F, Y_F, S, -S

    ! Calculate spin-spin energy change (Up,Down Left Right)
    E_DIFF = +2*j*s*(LATTICE(WRAP(X_F-1,L),Y_F) + LATTICE(WRAP(X_F+1,L),Y_F)&
                      +LATTICE(X_F,WRAP(Y_F-1,L)) + LATTICE(X_F,WRAP(Y_F+1,L)))
    ! Calculate spin-field energy change in non-zero field
    if (abs(G) >= 1E-6) then
        E_DIFF = E_DIFF - SUM(LATTICE)
    end if

    ! Accept/Reject Config
    if (E_DIFF > 0) then
        ! Reject with 1 - P(Boltzmann)
        call RANDOM_NUMBER(R_CA)
        if (exp(-E_DIFF/T) <= R_CA) then
            !write(3,*) "POS REJECT", E_DIFF,R_CA, exp(-E_DIFF/T)

            LATTICE(X_F,Y_F) = S
            E_DIFF = 0.0
        else
            !write(3,*) "POS ACCEPT", E_DIFF,R_CA, exp(-E_DIFF/T)
        end if
    else
        !write(3,*) "NEG ACCEPT", E_DIFF,R_CA, exp(-E_DIFF/T)
    end if

    ! Update energy
    E = E+E_DIFF

    !call PRINT_LATTICE(LATTICE, 3)
    !write(3,*) E
    !call PRINT_OBSERVABLES(TIME,LATTICE,E,3)
    !write(3,*) "-----"
end do

    ! Calculate and accumulate observables

E_SQ = E**2
M = sum(LATTICE)
M_SQ = M**2

MEAN_E = MEAN_E + E
MEAN_M = MEAN_M + M
MEAN_E_SQ = MEAN_E_SQ + E_SQ
MEAN_M_SQ = MEAN_M_SQ + M_SQ

    ! Data and debug Outputs

!call PRINT_OBSERVABLES(TIME,E,E_SQ,M,M_SQ,2)

!write(3,*) "-----"
!write(3,*) "AFTER TIME", TIME

```

```

!write(3,*)
!call PRINT_LATTICE(LATTICE,3)
!call PRINT_OBSERVABLES(TIME,E,E_SQ,M,M_SQ,M_S,3)
!write(3,*) "-----"
end do

! Average out observables and normalize to lattice size

MEAN_E      = MEAN_E      / EQ_SAMPLES
MEAN_E_SQ   = MEAN_E_SQ  / EQ_SAMPLES
MEAN_M      = MEAN_M      / EQ_SAMPLES
MEAN_M_SQ   = MEAN_M_SQ  / EQ_SAMPLES

C_V = MEAN_E_SQ - MEAN_E**2
CHI = MEAN_M_SQ - MEAN_M**2

CHI = CHI/size(LATTICE)
C_V = C_V/size(LATTICE)

MEAN_E = MEAN_E/size(LATTICE)
MEAN_M = MEAN_M/size(LATTICE)

! Output to log and standard out

write(1,103) "T = ", T
write(1,103) "CHI_S =", CHI
write(1,103) "C_V_S =", C_V
write(1,103) "-----"

write(*,102) T, CHI, C_V, MEAN_E, MEAN_M
!write(*,*) T, CHI,C_V,E/EQ_SAMPLES,M/EQ_SAMPLES

T = T+T_STEP

if (T>T_STOP) stop

end do

! Format : Output
! -----
101 format (A1,1X,A8,4X,5(A20,4X))      ! Table header
106 format (A1,1X,5(A20,4X))            ! Table header
102 format (2X,5(F20.10,4X))          ! Average Observables
103 format (A15,F15.10)                 ! Single variables
104 format (A15,I8)

end program ISING_MODEL

```

Output:

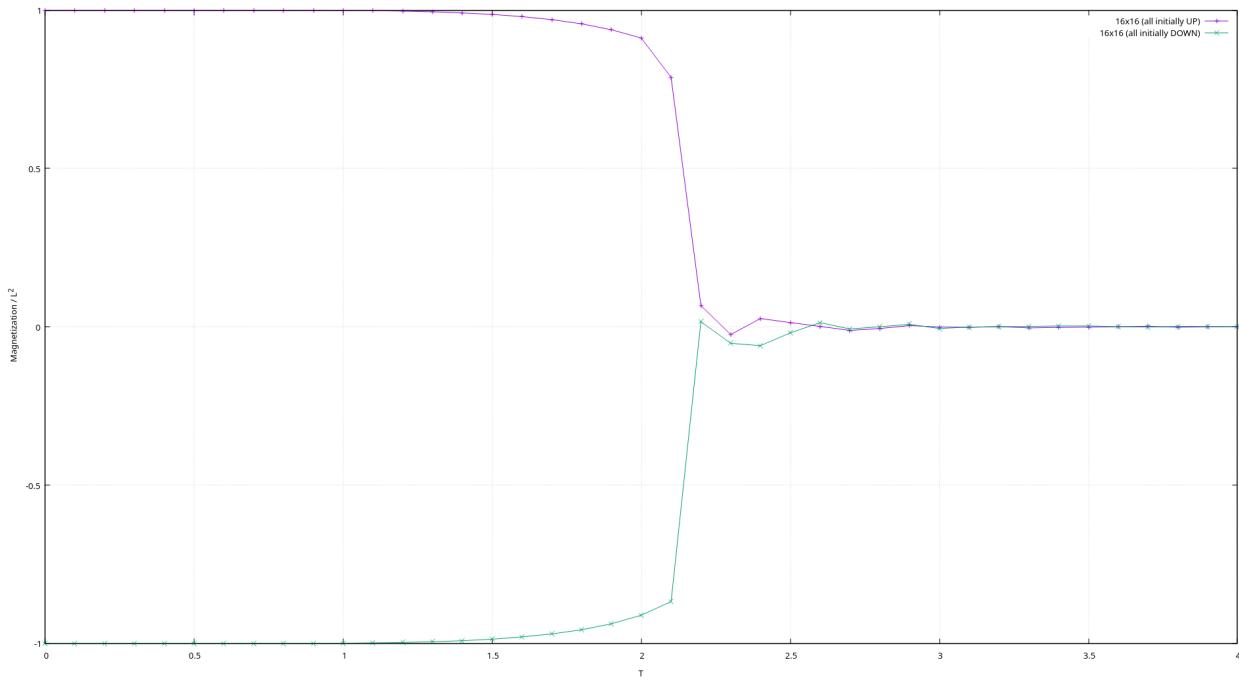


Figure 6: Magnetization per spin

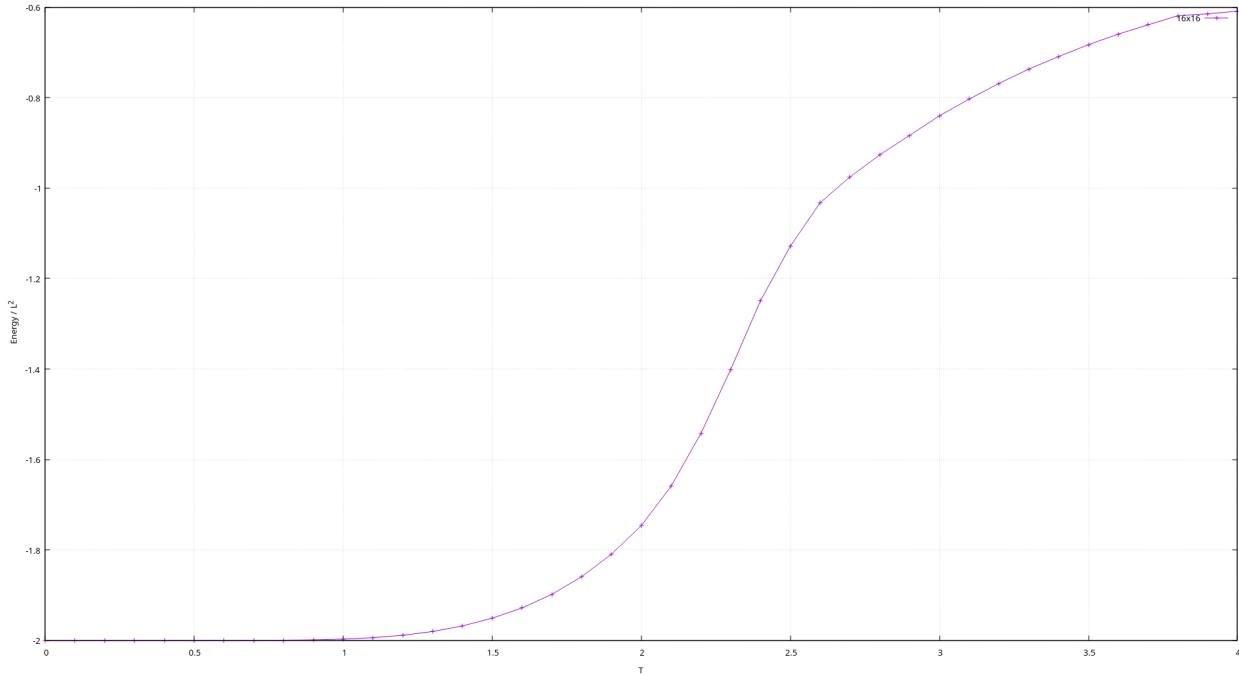


Figure 7: Energy per spin

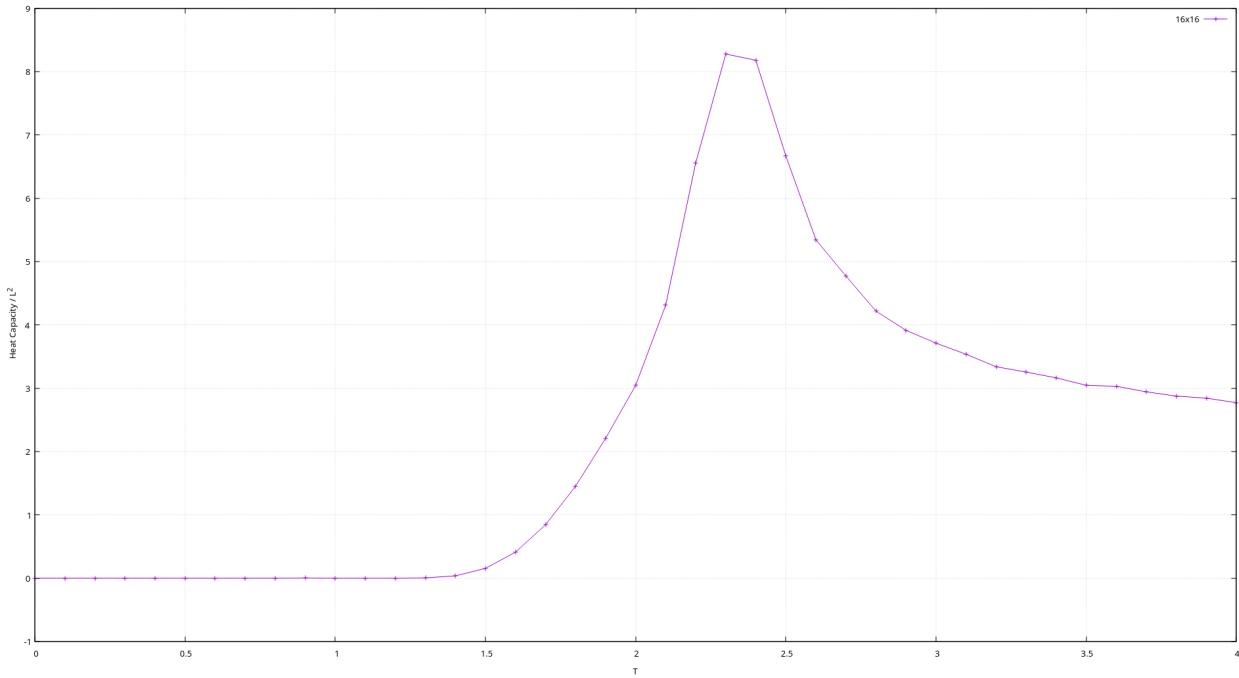


Figure 8: Heat Capacity per spin

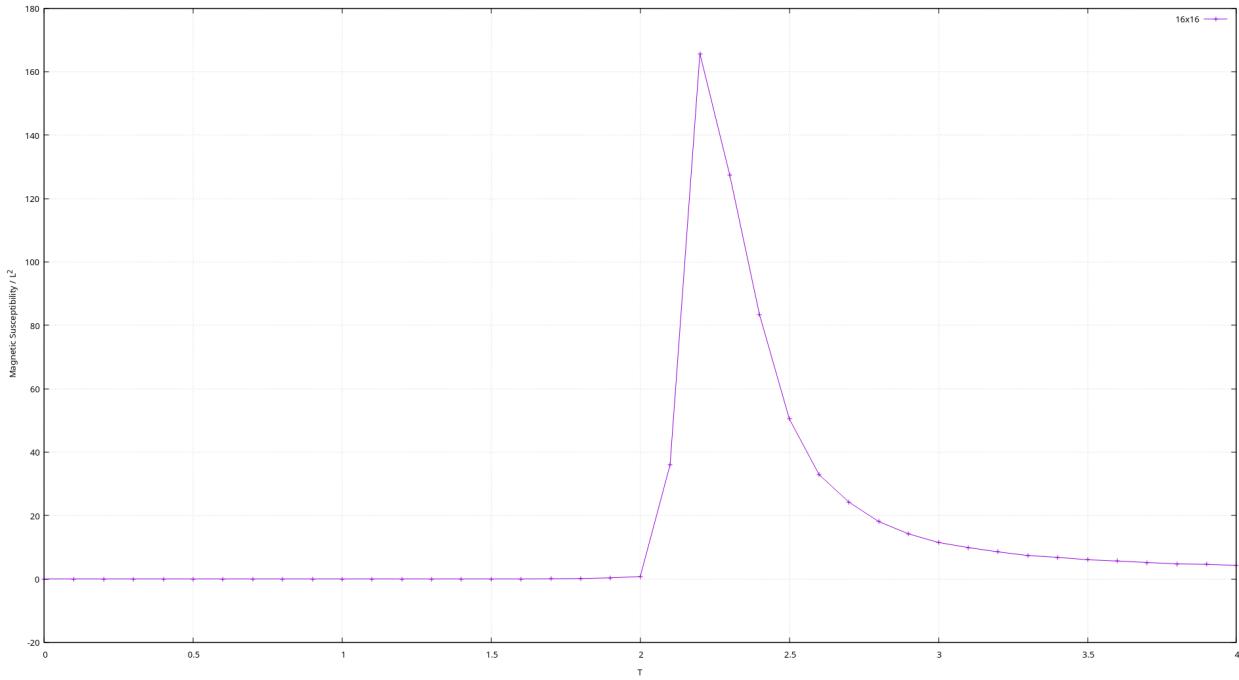


Figure 9: Magnetic Susceptibility per spin

APPENDIX - Common Subroutines & Functions

FORMULA

```

module FORMULA

use iso_fortran_env, only: IK => int32, RK => real64

implicit none
public::NBALL_VOLUME
double precision,parameter::PI = 4*atan(1.d0)

contains

real function NBALL_VOLUME(N,R)
    ! Compute the volume of a N-dimensional hypersphere of radius R

    integer,intent(in)::N
    real,intent(in)::R

    NBALL_VOLUME = (R**N)*(PI**(0.5*N))/gamma((0.5*N)+1)

end function NBALL_VOLUME

pure function GAUSSIAN(X,K) result(FUNC)

    real,dimension(:,),intent(in)::X
    real,intent(in)::K
    real,dimension(SIZE(X))::FUNC

    FUNC(:) = exp(-K*(X(:)**2))

end function GAUSSIAN

end module FORMULA

```

Flowcharts:

INTEGRATE

```
module INTEGRATE

use iso_fortran_env, only: IK => int32, RK => real64

implicit none
public::TRAPEZOID
public::SIMPSON

contains

subroutine TRAPEZOID(F,A,B,N,INTEGRAL)
    ! Composite Trapezoidal Rule
    !
    ! Integrates function F on [A,B] with N subintervals

    implicit none

    real, external      :: F          ! Integrand
    real, intent(in)   :: A          ! Interval start
    real, intent(in)   :: B          ! Interval end
    integer, intent(in) :: N          ! No. of subintervals
    real, intent(out)  :: INTEGRAL   ! Result

    integer :: I
    real    :: H

    H = (B-A)/N

    INTEGRAL = 0.5*(F(A) + F(B))

    do I = 1,N-1
        INTEGRAL = INTEGRAL + F(A+H*I)
    end do

    INTEGRAL = H*INTEGRAL

end subroutine TRAPEZOID

subroutine SIMPSON(F,A,B,N,INTEGRAL)
    ! Composite Simpson's 1/3rd Rule
    !
    ! Integrates function F on [A,B] with N subintervals

    implicit none

    real, external      :: F          ! Integrand
    real, intent(in)   :: A          ! Interval start
    real, intent(in)   :: B          ! Interval end
    integer, intent(in) :: N          ! No. of subintervals
    real, intent(out)  :: INTEGRAL   ! Result
```

```

integer          :: I
real            :: H
real, parameter :: R = 4.0/3.0

H = (B-A)/N
write(*,*) R

INTEGRAL = (F(A) + F(B))/3

do I = 1,N-1,2
    INTEGRAL = INTEGRAL + R*F(A+H*I)
end do

do I = 2,N-1,2
    INTEGRAL = INTEGRAL + 0.5*R*F(A+H*I)
end do

INTEGRAL = H*INTEGRAL

end subroutine SIMPSON

end module INTEGRATE

```

Flowcharts:

MONTECARLO

```
module MONTECARLO

use iso_fortran_env, only: IK => int32, RK => real64

implicit none
public::ACCREJ_SPHERE

contains

subroutine ACCREJ_SPHERE(RAD,DIMM,NRAND,ESTVOL)
    ! Compute volume of DIMM-dimensional hypersphere of radius RAD
    ! by acceptance-rejection method

    implicit none

    real,intent(in)::RAD          ! Scaling factor
    integer,intent(in)::DIMM        ! Dimension of RN vector
    integer,intent(in)::NRAND        ! Number of throws
    real,intent(out)::ESTVOL       ! Estimated Volume

    integer::I
    real::NIN
    double precision,allocatable::R(:,:)

    ! Generate and shift RN vector
    allocate(R(NRAND,DIMM))
    call RANDOM_NUMBER(R)
    R = (R - 0.5)*(2*RAD)

    ! Count internal points
    NIN = 0
    do I=1,NRAND
        if (sum(R(I,:)**2) <= RAD**DIMM) then
            NIN = NIN+1
        endif
    end do

    ! Compute volume
    ESTVOL = (NIN/NRAND)*(2**DIMM)

    deallocate(R)

end subroutine ACCREJ_SPHERE

end module MONTECARLO
```

Flowcharts: