

Exercises Unit VIII: Metropolis - Monte Carlo algorithm for importance sampling integration

1. Sampling physical quantities with gaussian distribution: direct sampling and Metropolis sampling

Consider the quantum harmonic oscillator and its ground state. The exact solution and the expectation values of kinetic, potential and total energy are known analytically, and can be used to compare the numerical results.

- (a) *Direct sampling.* Estimate kinetic energy, potential energy, first moments $\langle x^i \rangle$ of the wavefunction $\psi(x) = Ae^{-x^2/(4\sigma^2)}$ with a sample-mean Monte Carlo calculation of the integral of the expectation values using a sequence of random points *directly* obtained for instance from the `gasdev` subroutine (see previous Lectures). See for instance the code `direct_sampling.f90`. Study the numerical accuracy and the convergence of the previous quantities as a function of the number of sampling points.
- (b) Is the normalization constant A of the wavefunction important for our purposes?
- (c) *Metropolis sampling.* Repeat the sampling using the Metropolis algorithm. See for instance the code `metropolis_sampling.f90`. Evaluate the numerical accuracy and convergence of the more relevant quantities as a function of the number of sampling points.

2. Correlations

- (a) Calculate the autocorrelation function $C(j) = \frac{\langle x_i x_{i+j} \rangle - \langle x_i \rangle^2}{\langle x_i^2 \rangle - \langle x_i \rangle^2}$ for a sequence of random numbers with a gaussian distribution using the Metropolis method, with different values of δ/σ : 1, 5, 10, 25, 50. Comment the results.
- (b) For a fixed value of σ compare the autocorrelation function for two sequences of random numbers with a gaussian distribution (i) using the Metropolis method and (ii) using some ad-hoc routine, like for instance `gasdev` based on the Box-Muller algorithm. Discuss the results.

3. Verification of the Boltzmann distribution

We can verify directly that the Metropolis algorithm yields the Boltzmann distribution. We consider a **single classical particle** in one dimension in equilibrium with a heat bath (*canonical ensemble*). We fix therefore the temperature T , which labels a *macrostate*. The energy E can vary according to the particular *microstate* (in this particular case, it is enough to label a microstate, a part from the sign of the velocity).

- (a) Write a code (see e.g. `boltzmann_metropolis.f90`) to determine the form of the probability distribution $P(E)$ that is generated by the Metropolis algorithm. Let for instance $T=1$, the initial velocity $v_{initial}=0$, the number of Monte Carlo steps $nmcs=1000$, and the maximum variation of the velocity $dvmax=2$. Calculate the mean energy, the mean velocity, and the probability density $P(E)$.
- (b) Consider $\ln P(E)$ as a function of E . Can you recognize the expected behavior? (see slides for the analytic derivation of $P(E)$) You should recognise that the asymptotic behavior is a straight line whose slope is $-1/T$.
- (c) How many $nmcs$ do you need to have a reasonable estimate of the mean energy and mean velocity?
- (d) Verify that your results do not depend from the initial conditions by changing $v_{initial}$. What does it change? What does it change by changing instead $dvmax$?
- (e) Modify the program to simulate an ideal gas of **N particles** in one dimension. [Hint: modify the subroutine *Metropolis* inserting a loop over the particles] Consider for instance $N=20$, $T=100$, $nmcs=200$. Assume all particles to have the same initial velocity $v_{initial}=10$. Determine the value of $dvmax$ so that the acceptance ratio is about 50%? What are the mean energy $\langle E \rangle$ (i.e., total energy of the system $\langle E_{tot} \rangle$ divided by the number of particles) and the mean velocity? [the symbol $\langle \rangle$ indicates temporal(statistical) averages]
- (f) Calculate $P(E)$ (E now indicates the mean energy per particle), make a plot and describe its behaviour. Is it similar to the case $N=1$? Comment on that.
- (g) Calculate the total energy E_{tot} for $T=10, 20, 30, 90, 100$, and 110 , and estimate the heat capacity as the numerical derivative of the energy with respect to the temperature, $C = \partial \langle E_{tot} \rangle / \partial T$. [C is the heat capacity, i.e. referred to the whole system; you may consider, alternatively, the specific heat, referred to a single particle...]
- (h) Calculate the mean square energy fluctuation $\langle \Delta E_{tot}^2 \rangle = \langle E_{tot}^2 \rangle - \langle E_{tot} \rangle^2$ for $T=10$ and $T=40$. Compare the magnitude of the ratio $C = \langle \Delta E_{tot}^2 \rangle / T^2$ numerically estimated from the mean square energy fluctuation with that obtained in (f).

4. MC simulation of a simple N-particles model

Consider an ideal gas of N non interacting, distinguishable particles, **confined** in a box (fixed \mathbf{V}) and **isolated** (fixed \mathbf{E}), divided into left/right with the possibility for one particle at a time to pass through the separation wall, with equal probability from the left to the right or viceversa.

A **macrostate** is specified for instance by the number of particles on the left side, say n , that can correspond to different **microstates** depending on the list of the specific particles there. A Monte Carlo approach consists in generating a certain number of movements, randomly, and consider them as representative of all the possible movements. The program `box.f90` is a possible implementation of the algorithm describing the time evolution of the system in terms of macrostates, i.e. –given an initial number of particles on the left, n – the approach to equilibrium and which is the equilibrium macrostate.

- (a) Choose $N=4, 10, 20, 40, 80$, and $n=N$ initially. Make a plot of n (or, better, of n/N) with respect to time. What is the equilibration time τ_{eq} (=how many MC steps)?
- (b) Modify the program so that at each time step t it calculates the number of particles $\langle n(t) \rangle$ averaged over different runs (e.g. 5 runs). Make a plot to compare $n(t)$ over the individual runs and averaged $\langle n(t) \rangle$.
- (c) (*Optional; do it at home!*) Compare the numerical value of $\langle n(t) \rangle$ with the exact analytic results for a simple case, for instance $N=4$.
- (d) (*Optional*) Consider only one run. Modify the program to calculate numerically the probability P_n of having *at equilibrium* a macrostate with n particles on the left, by simply counting the number of occurring microstates that correspond to the macrostate n and dividing for the total number of microstates generated in the time evolution. Plot the histogram P_n for $N=20, 40, 80$ and a “sufficiently” long run. Comment.
- (e) Modify the program to measure the statistical fluctuations at the equilibrium, by calculating the variance $\sigma^2 = \langle n^2 \rangle - \langle n \rangle^2$, where the average is done over a time interval *after* reaching the equilibrium.
- (f) Determine $\langle n \rangle$ and $\sigma / \langle n \rangle$ at equilibrium for $N=20, 40, 80$. Which is the dependence of these quantities on N ?

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!cccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccc
! metropolis.f90
!
! campionamento METROPOLIS di varie quantita' (tra cui l'energia,
! hamiltoniana:      h = -1/2 \nabla^2 + 1/2 x^2),
! confronto stime numeriche con valori analitici attesi,
! su psi^2(x), con  psi(x) = exp(-\beta x^2)
! (beta = 1/4*sigma^2; con beta=.25 => psi^2(x) = costante * gaussiana normale
! P(x) = exp(-x**2/(2*sigma**2))/sqrt(2*pi*sigma**2)
!cccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccc

```

```

program metropolis
  implicit none
  integer, parameter :: dp=selected_real_kind(13)
  integer :: i,n
  integer, dimension(1) :: seed
  real(kind=dp):: sigma,beta,etot,ekin,epot,rnd
  real(kind=dp):: x,x1,x2,x3,x4,xi,delta,lnpsi,lnpsip,p,acc
  character(len=13), save :: format1 = "(a7,2x,2f9.5)"

  acc = 0.0_dp
  x1 = 0.0_dp
  x2 = 0.0_dp
  x3 = 0.0_dp
  x4 = 0.0_dp
  ekin = 0.0_dp
  epot = 0.0_dp
  print*, "seed, n, beta, x0, delta"
  read*,   seed(1),n,beta,x,delta
  call random_seed(put=seed)
  sigma=1.0_dp/sqrt(4.0_dp*beta)

  do i=1,n
    ekin = ekin - 0.5_dp * ((2*beta*x)**2 - 2*beta)
    epot = epot + 0.5_dp * x**2
    etot = ekin + epot
    x1 = x1 + x
    x2 = x2 + x**2
    x3 = x3 + x**3
    x4 = x4 + x**4
    !cccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccc
    lnpsi = -beta * x**2      !
    call random_number(rnd)   !
    xi = x + delta * (rnd-0.5_dp) !
    lnpsip = -beta * xi**2    !   metropolis
    p = exp ( 2 * (lnpsip-lnpsi) ) !   algorithm
  end do

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        call random_number(rnd)          !
        if (p > rnd) then                !
            x = xp                        !
            !cccccccccccccccccccccccccccccccccccccccc
            acc=acc+1.0_dp
        endif
    enddo

write(unit=*,fmt=*)"acceptance ratio = ",acc/n
write(unit=*,fmt=*)"Risultati (simulazione vs.risultato esatto):"
write(unit=*,fmt=format1)"etot = ",etot/n,1.0_dp/(8.0_dp*sigma**2)&
    +0.5_dp*sigma**2
write(unit=*,fmt=format1)"ekin = ",ekin/n,1.0_dp/(8.0_dp*sigma**2)
write(unit=*,fmt=format1)"epot = ",epot/n,0.5_dp*sigma**2
write(unit=*,fmt=format1)"<x> = ",x1/n,0.0_dp
write(unit=*,fmt=format1)"<x^2>= ",x2/n,sigma**2
write(unit=*,fmt=format1)"<x^3>= ",x3/n,0.0_dp
write(unit=*,fmt=format1)"<x^4>= ",x4/n,3.0_dp*sigma**4

end program metropolis

```

```

!cccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccc
! diretto.f90
!
! campionamento DIRETTO di varie quantita' (tra cui l'energia,
! hamiltoniana:      h = -1/2 \nabla^2 + 1/2 x^2),
! confronto stime numeriche con valori analitici attesi,
! su psi^2(x), con psi(x) = exp(-\beta x^2)
! (beta = 1/4*sigma^2; con beta=.25 => psi^2(x) = costante * gaussiana normale
! P(x) = exp(-x**2/(2*sigma**2))/sqrt(2*pi*sigma**2)
!cccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccc

```

```

Module gaussian
  implicit none
  public :: gasdev

```

```
contains
```

```

  SUBROUTINE gasdev(x)

    REAL, INTENT(OUT) :: x
    REAL :: rsq, v1, v2
    REAL, SAVE :: g
    LOGICAL, SAVE :: gaus_stored=.false.

    if (gaus_stored) then
      x=g
      gaus_stored=.false.
    else
      do
        call random_number(v1)
        call random_number(v2)
        v1 = 2.0*v1 - 1.0
        v2 = 2.0*v2 - 1.0
        rsq = v1**2 + v2**2
        if (rsq > 0.0 .and. rsq < 1.0) exit
      end do
      rsq = sqrt(- 2.0*log(rsq)/rsq)
      x = v1*rsq
      g = v2*rsq
      gaus_stored = .true.
    end if

```

```
  END SUBROUTINE gasdev
```

```
end module gaussian
```

```

program diretto
  use gaussian
  implicit none
  integer, parameter :: dp=selected_real_kind(13)
  integer :: i,n
  integer, dimension(1) :: seed
  real :: rnd
  real(kind=dp):: sigma,beta,etot,ekin,epot
  real(kind=dp):: x,x1,x2,x3,x4
  character(len=13), save :: format1 = "(a7,2x,2f9.5)"

  x1 = 0.0_dp
  x2 = 0.0_dp
  x3 = 0.0_dp
  x4 = 0.0_dp
  ekin = 0.0_dp
  epot = 0.0_dp
  print*, "seed, n, beta ="
  read*, seed(1),n,beta
  call random_seed(put=seed)
  sigma=1.0_dp/sqrt(4.0_dp*beta)
  do i=1,n
    !cccccccccccccccccccccccccccccccc
    call gasdev(rnd) !
    x=rnd*sigma ! campionamento diretto
    !cccccccccccccccccccccccccccccccc!
    ekin = ekin - 0.5_dp * ((2*beta*x)**2 - 2*beta)
    epot = epot + 0.5_dp * x**2
    etot = ekin + epot
    x1 = x1 + x
    x2 = x2 + x**2
    x3 = x3 + x**3
    x4 = x4 + x**4
  end do

  write(unit=*,fmt=*)"Risultati (simulazione verso risultato esatto):"
  write(unit=*,fmt=format1)"etot = ",etot/n,1.0_dp/(8.0_dp*sigma**2)&
    +0.5_dp*sigma**2
  write(unit=*,fmt=format1)"ekin = ",ekin/n,1.0_dp/(8.0_dp*sigma**2)
  write(unit=*,fmt=format1)"epot = ",epot/n,0.5_dp*sigma**2
  write(unit=*,fmt=format1)"<x> = ",x1/n,0.0_dp
  write(unit=*,fmt=format1)"<x^2>= ",x2/n,sigma**2
  write(unit=*,fmt=format1)"<x^3>= ",x3/n,0.0_dp
  write(unit=*,fmt=format1)"<x^4>= ",x4/n,3.0_dp*sigma**4

end program diretto

```

```

!cccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccc
! boltzmann_metropolis.f90
!
! Metropolis algorithm used as importance-sampling:
! generation of microstates with Boltzmann distribution,
! here for a classical particle in 1D.
! The interesting quantity is the probability P(E)dE for a particle
! to have energy between E and E+dE (here E can label a microstate,
! a part from the sign +/- of the velocity)
!cccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccccc

module common
  implicit none
  public :: initial, Metropolis, data, probability, averages
  real, public :: E,T,del_E,beta,dvmax,vel,accept
  integer, public, dimension(:), allocatable :: seed
  integer, public :: nbin,nmcs,sizer
  real, public, dimension(:), allocatable :: P
contains

  subroutine initial(nequil,vcum,ecum,e2cum)
    real, intent(out) :: vcum,ecum,e2cum
    integer, intent(out) :: nequil
    print*," number of MC steps >"
    read *, nmcs
    print*," absolute temperature >"
    read *, T
    print*," initial velocity >"
    read *, vel
    print*," maximum variation of the velocity (hint: 4*sqrt(T)=",4*sqrt(T),") >"
    read *, dvmax
    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)
    beta = 1/T
    nequil = 0.1 * nmcs ! WARNING : VERIFY this choice !
    E = 0.5 * vel * vel
    del_E = T/20 ! a reasonable width of the bin for the histogram of P(E)
    nbin = int(4*T / del_E) ! max. number of bins
    print *,"# T : ",T
    print *,"# <E0> : ",E
    print *,"# <v0> : ",vel
    print *,"# dvmax : ",dvmax
    print *,"# nMCsteps:",nmcs

```



```

print *,"# deltaE  :",del_E
print *,"# nbin   :",nbin
open(unit=9,file="boltzmann.dat",status="replace",action="write")
write(unit=9,fmt=*)"# T      :",T
write(unit=9,fmt=*)"# <E0>  :",E
write(unit=9,fmt=*)"# <v0>  :",vel
write(unit=9,fmt=*)"# dvmax  :",dvmax
write(unit=9,fmt=*)"# nMCsteps:",nmcs
write(unit=9,fmt=*)"# deltaE  :",del_E
write(unit=9,fmt=*)"# nbin   :",nbin
allocate (P(0:nbin))
ecum = 0.0
e2cum = 0.0
vcum = 0.0
P = 0.0
accept= 0.0
end subroutine initial

subroutine Metropolis()
real :: dv,vtrial,de,rnd
call random_number(rnd)
dv = (2*rnd - 1) * dvmax           ! trial variation for v
vtrial = vel + dv                 ! trial velocity v
de = 0.5 * (vtrial*vtrial - vel*vel) ! corresponding variation of E
call random_number(rnd)
if (de >= 0.0) then
    if ( exp(-beta*de) < rnd ) return ! trial step not accepted
end if
vel = vtrial
accept = accept + 1
E = E + de
end subroutine Metropolis

subroutine data(vcum,ecum,e2cum)
real, intent(inout) :: vcum,ecum,e2cum
Ecum = Ecum + E
E2cum = E2cum + E*E
vcum = vcum + vel
call probability()
end subroutine data

subroutine probability()
integer :: ibin
ibin = int(E/del_E)
if ( ibin <= nbin ) P(ibin) = P(ibin) + 1
end subroutine probability

```

```

subroutine averages(nequil,vcum,Ecum,E2cum)
  integer, intent(in) :: nequil
  real, intent(in) :: vcum,Ecum,E2cum
  real :: znorm, Eave, E2ave, vave, sigma2
  integer :: ibin
  znorm = 1.0/nmcs
  accept = accept / (nmcs+nequil) ! acceptance ratio
  Eave = Ecum * znorm ! average energy
  E2ave = E2cum * znorm !
  vave = vcum * znorm ! average velocity
  sigma2 = E2ave - Eave*Eave
  print *, "# <E2>num.:", E2ave
  print *, "# <E> num.:", Eave
  print *, "# <E> th. :", T/2
  print *, "# <v> :", vave
  print *, "# accept. :", accept
  print *, "# sigma :", sqrt(sigma2)
  write(unit=9,fmt=*) "# <E2>num:", E2ave
  write(unit=9,fmt=*) "# <E> num.:", Eave
  write(unit=9,fmt=*) "# <E> th. :", T/2
  write(unit=9,fmt=*) "# <v> :", vave
  write(unit=9,fmt=*) "# accept. :", accept
  write(unit=9,fmt=*) "# sigmaE :", sqrt(sigma2)
  write(unit=9,fmt=*) "# ibin*del_E, P(E)"
  do ibin = 0, nbin
    write(unit=9,fmt=*) ibin*del_E, P(ibin) * znorm
  end do
  close(unit=9)
end subroutine averages
end module common

program Boltzmann
  use common
  real :: vcum, ecum, e2cum
  integer :: imcs, nequil
  ! parameters and variable initialization
  call initial(nequil,vcum,ecum,e2cum)
  do imcs = 1, nmcs + nequil
    call Metropolis()
    ! data accumulation after each Metropolis step
    if ( imcs > nequil ) call data(vcum,ecum,e2cum)
  end do
  call averages(nequil,vcum,Ecum,E2cum)
  deallocate(P)
end program Boltzmann

```



```
! elsewhere from the right to the left  
call initial()  
call move()  
end program box
```