```
In [1]: | import Pkg
        Pkg.add("DataStructures")
        Pkg.add("Shuffle")
        Pkg.add("Plots")
        Pkg.add("Distributions")
        Pkg.add("Random")
        using DataStructures
        using Shuffle
        using Plots
        using Random
        using Distributions
        #takes in an integer n
        #returns array of length n of integers +/- 1 (randomly chosen) that represent sp.
        function initial_config(n::Int)
          config = zeros(n)
          if n%2 != 0
            println("Size n must be even.")
            return
          else
            for i=1:n
              config[i] = rand([-1, 1])
            end
          end
          #print(config)
          return config
          \# essentially \ from \ N, \ we \ get \ a \ randomized \ array \ of \ spin \ ups \ and \ downs
          \#e.g., N=2 may equal [1,-1]; N=4 may equal [1,-1,-1,1] as our initial configura
        function gaussian_rf(N)
          nd = Normal(0, 1)
          return rand(nd, N)
        function unit_rf(N)
          field = zeros(N)
          for i=1:N
            field[i] = rand([-.5, .5])
          end
          return field
        end
        #standard function for getting the hamilitonian of 1/r^2
        #Ising Model
        function get_energy(s::AbstractArray, h::AbstractArray, J)
          E0 = 0.0
          E1 = 0.0
          E2 = 0.0
          for i=1:length(s)
            #if i != length(s)
              \#E0 += J*s[i]*s[i+1]
            #else
              \#E0 += J*s[i]*s[1]
            #end
            for j=i:length(s)
              if j != i
                E1 += J*(s[i]-s[j])^2/(i-j)^2
              end
            end
            E2 += h[i]*s[i]
          end
          E = -E1/2 - E2
          return E
        end
        #magnization equals the summation of spins in
        #a configuration.
        #magnetization per spin = <M>/N, where
        #N is basically length of the configuration
        function get_magnetization(config::AbstractArray)
          M = sum(config)
          return M
        end
        function get_susceptibility(M_list, Msq_list, kT, N)
          avg M = 0
          #avg_M = sum(M_list)/mcsteps
          avg_Msq = sum(Msq_list)/mcsteps
          X = (1/kT)*(avg_Msq - avg_M^2)/N
          return X
        end
        function do_MC_Step(config, kT, J, h, M_list, Msq_list)
          N = length(config)
          E = get_energy(config, h, J)
          for i=1:N
            site = rand(1:N)
            config[site] = -1*config[site]
            #attempt to update one site of the configuration
            E_new = get_energy(config, h, J)
            #look at the hamiltonian of the configuration
            #given this updated site
            dE = E_new - E
            #look at the difference between old and new hamiltonian
            prob = exp(-dE/(kT))
            #Metropolis acceptance ratio
            r = rand(Float64)
            if min(1, prob) > r
             #if true, configuration is "updated"
              #accepted_states += 1
              E += dE
            else
               config[site] = -1*config[site]
              #if false, then configuration stays the same
          end
          M = get_magnetization(config)
          M sq = M^2
          #println("M squared: ", M_sq)
          #push!(M_list, abs(M))
push!(Msq_list, M_sq)
          #we get the magnization per spin of the updated system
        function metropolis(config_initial::AbstractArray, kT, J, h, mcsteps)
          config = copy(config_initial)
          #starts out with a configuration of randomized N spin array
          mags = Vector{Float64}()
          square_mags = Vector{Float64}()
          M = get_magnetization(config)
          push!(mags, M)
          push!(square_mags, M^2)
          accepted_states = 0
          #initial state of the configuration before a MCMC update
          E = get_energy(config, h, J)
          # hamiltonian of a particular configuration
          for i=1:mcsteps
            #For MCMC an arbitrary number steps are executed for precision
            do MC_Step(config, kT, J, h, mags, square_mags)
          end
          return mags, square_mags
        end
        #CONSTANTS:
        #Number of spins in initialized configuration
        N = 1000
        #Interaction constant
        J = 1.0
        #Initialize number of montecarlo steps
        mcsteps = 10000
        config0 = initial config(N)
        #Initialize random field(s)
        h1 = unit_rf(N)
        h2 = gaussian_rf(N)
        h3= zeros(N) #PURE STATE
        initkT = 0.2
        iter = 0.2
        finalkT = 3.0
            Updating registry at `~/.julia/registries/General.toml`
```

```
Updating registry at `~/.julia/registries/General.toml`
Resolving package versions...
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Project.toml`
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Manifest.toml`
Resolving package versions...
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Project.toml`
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Manifest.toml`
Resolving package versions...
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Project.toml`
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Project.toml`
Resolving package versions...
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Project.toml`
No Changes to `/gpfs/home/agassaml/.julia/environments/v1.8/Manifest.toml`
Resolving package versions...
```

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```
1RDFM N=1000 Pure vs Gauss - Jupyter Notebook
                In [2]: #Pure State:
                          @time begin X_list1 = Vector{Float64}()
                          for kT = initkT:iter:finalkT
                            println("At ", kT, " kT.")
                            data = metropolis(config0, kT, J, h3, mcsteps)
X = get_susceptibility(data[1], data[2], kT, length(config0))
                            push!(X_list1, X)
                          end
                          plot(initkT:iter:finalkT, X_list1, xlabel = "Temperature", ylabel = "Susceptibil")
                          end
                          At 0.2 kT.
                          At 0.4 kT.
                          At 0.6 kT.
                          At 0.8 kT.
                          At 1.0 kT.
                          At 1.2 kT.
                          At 1.4 kT.
                          At 1.6 kT.
                          At 1.8 kT.
                          At 2.0 kT.
                          At 2.2 kT.
                          At 2.4 kT.
                          At 2.6 kT.
                          At 2.8 kT.
                          At 3.0 kT.
                          193510.523665 seconds (4.43 M allocations: 245.146 MiB, 0.00% gc time, 0.00% co
                          mpilation time: 22% of which was recompilation)
                Out[2]:
                             0.150
                             0.125
                          Susceptibility
                             0.100
                             0.075
                              0.050
                                                                                                 Pure State
                                            0.5
                                                        1.0
                                                                                             2.5
                                                                                                         3.0
                                                                 Temperature
                In [3]: #Gaussian Random Field:
                          @time begin X_list2 = Vector{Float64}()
                         for kT = initkT:iter:finalkT
  println("At ", kT, " kT.")
                            data = metropolis(config0, kT, J, h2, mcsteps)
X = get_susceptibility(data[1], data[2], kT, length(config0))
                            push!(X_list2, X)
                          end
                          plot!(initkT:iter:finalkT, X_list2, xlabel = "Temperature", ylabel = "Susceptibi")
                          At 0.2 kT.
                          At 0.4 kT.
                          At 0.6 kT.
                          At 0.8 kT.
                          At 1.0 kT.
                          At 1.2 kT.
                          At 1.4 kT.
                          At 1.6 kT.
                          At 1.8 kT.
                          At 2.0 kT.
                          At 2.2 kT.
                          At 2.4 kT.
                          At 2.6 kT.
                          At 2.8 kT.
                          At 3.0 kT.
                          170957.952665 seconds (335.45 k allocations: 22.476 MiB, 0.00% compilation tim
                          e)
                Out[3]:
                              1.2
                                                                                      Pure State
Gaussian Random Field
                              1.0
                          Susceptibility
                             8.0
                             0.6
                             0.4
                              0.2
                              0.0
                                          0.5
                                                      1.0
                                                                   1.5
                                                                                2.0
                                                                                             2.5
                                                                                                         3.0
```

```
In [ ]:
In [ ]:
```

Temperature

 $[0.250079999999997, \ 0.28896, \ 0.2437200000000005, \ 0.23897, \ 0.223208, \ 0.218373]$ 3333333334, 0.2132457142857143, 0.201125, 0.1976266666666667, 0.1914239999999 9998, 0.177927272727274, 0.180396666666668, 0.1668, 0.16540285714285716, 0.

In [17]: print(X_list2)

15995199999999998]

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