CO28

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[]: import numpy as np

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import matplotlib.pyplot as plt
     from matplotlib import animation
     from matplotlib.colors import ListedColormap
     from scipy.signal import convolve2d
     import numba
[]:N=30
     spins = np.random.choice([-1, 1], (N, N))
     # Visualise a random arrangement of spins
     cmp = ListedColormap(['red', 'blue'])
     plt.imshow(spins, cmap=cmp, vmin=-1, vmax=1)
     plt.tick_params(bottom=False, top=False, left=False,
                     right=False, labelbottom=False, labelleft=False)
     plt.show()
[]: # Performs the Metropolis algorithm, sweep through the lattice and calculate
     # acceptance ratio r from energies, and flip spin with probability r
     # Use numba (compiles Python to machine code) for almost 100x speedup
     # Input: S_init (initial matrix of spins), JkT, BkT (values of J/kT and B/kT)
     # Output S_next (matrix of spins after sweep iteration)
     @numba.jit(nopython=True)
     def sweep(S_init, JkT, BkT):
         S_next = np.copy(S_init)
         for iy, ix in np.ndindex(S_next.shape):
             spinCoupling = S_next[iy, ix] * S_next[(iy + 1) % N, ix]
             spinCoupling += S_next[iy, ix] * S_next[iy - 1, ix]
             spinCoupling += S_next[iy, ix] * S_next[iy, (ix + 1) % N]
             spinCoupling += S_next[iy, ix] * S_next[iy, ix - 1]
             deltaE = 2 * JkT * spinCoupling - 2 * BkT * S_next[iy, ix]
             r = np.exp(-deltaE)
            rand = np.random.uniform()
             if r > rand:
                 S_next[iy, ix] *= -1
         return S_next
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[]: neighbours = np.array([[0,1,0]],
                             [1,0,1],
                            [0,1,0])
     # Calculate energy by adding up spin-spin interaction and applied magnetic field
     # Remember to divide first term by 2 to account for double counting of pairs
     # Use convolution for efficient neighbour sum
     def energy(spins, JkT, BkT):
         convolve = convolve2d(spins, neighbours, mode='same', boundary='wrap')
         return -JkT * (spins * convolve).sum() / 2 + BkT * spins.sum()
[]: # Simulate the system for particular values of JkT and BkT
     # and store magnetic moments and energies
     spinsList = []
     magneticMoments = []
     energies = []
     JkT = 0.5
     BkT = 0.0
     n_{iterations} = 2000
     spins = np.random.choice([-1, 1], (N, N))
     for i in range(n_iterations):
         spinsList.append(spins)
         magneticMoments.append((spins.sum()))
         energies.append(energy(spins, JkT, BkT))
         spins = sweep(spins, JkT, BkT)
     energies = np.array(energies)
     magneticMoments = np.array(magneticMoments)
[]: # Plot the magnetic moment and energy
     fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(8,4))
     fig.suptitle(f"\{N\}x\{N\} grid, \{J/k_BT\} = \{JkT\}, \{B/k_BT\} = \{BkT\}")
     ax1.set_title("Energy")
     ax1.plot(energies)
     ax1.set_xlabel("Sweep iteration")
     ax1.set_ylabel("$E/k_BT$")
     ax2.set_title("Magnetic Moment")
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ax2.plot(magneticMoments)

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ax2.set_xlabel("Sweep iteration")
ax2.set_ylabel("$M$")
fig.tight_layout()
plt.show()
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[]: fig, [ax1, ax2] = plt.subplots(2, 1, figsize=(8,10))
     n_{iterations} = 5000
     BList = [0, 0.01, 0.1]
     JList = np.linspace(0, 1, 40)
     colors = ["red", "blue", "green"]
     \# Simulate system for various values of JkT and BkT
     for i, BkT in enumerate(BList):
         avgMagMoments = []
         avgHeatCapacities = []
         avgSusceptibilities = []
         for JkT in JList:
             magneticMoments = []
             energies = []
             spins = np.random.choice([-1, 1], (N, N))
             for n in range(n_iterations):
                 magneticMoments.append(np.abs(spins.sum()))
                 energies.append(energy(spins, JkT, BkT))
                 spins = sweep(spins, JkT, BkT)
             # Calculate averages of heat capacity and magnetic moment at equilibrium
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magneticMoments = np.array(magneticMoments)
             energies = np.array(energies)
             avgMagMoments.append(np.mean(magneticMoments[2000:]))
             avgHeatCapacities.append(np.var(energies[2000:]))
         # Plot heat capacity/magnetic moment vs JkT for different values of BkT
         ax1.plot(JList, avgMagMoments, c=colors[i],
                  marker="d", label=f"$B/k_BT$ = {BkT}")
         ax2.plot(JList, avgHeatCapacities, c=colors[i],
                  marker="d", label=f"$B/k_BT$ = {BkT}")
     criticalPoint = 0.44
     ax1.axvline(criticalPoint, color="orange", alpha=0.6,
                 linestyle='--', label=f"$J/k_BT$={criticalPoint}")
     ax1.set_xlabel("$J/k_BT$")
     ax1.set_ylabel("$|M|$")
     ax1.set_title("Equilibrium magnetic moment vs $J/k_BT$")
     ax1.legend()
     ax2.axvline(criticalPoint, color="orange", alpha=0.6,
                 linestyle='--', label=f"$J/k BT$={criticalPoint}")
     ax2.set_xlabel("$J/k_BT$")
     ax2.set ylabel("$C/k B$")
     ax2.set_title("Equilibrium heat capacity vs $J/k_BT$")
     ax2.legend()
     fig.tight_layout()
    plt.show()
[]: # Extra code for creating animation of the lattice of spins
     fig, ax = plt.subplots()
     ax.tick_params(bottom=False, top=False, left=False,
                    right=False, labelbottom=False, labelleft=False)
     grid = ax.imshow(spins, cmap=cmp, vmin=-1, vmax=1)
     def animate(i):
         grid.set_data(spinsList[i])
         return grid
     anim = animation.FuncAnimation(fig, animate,
                                    frames=n_iterations, interval=50)
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Wait roughly 2000 iterations for burn in period

anim.save("ising_model.mp4")