ZJU Computational Physics: Homework #4Due on Monday, December 13, 2024

Github: https://github.com/NAKONAKO4/ZJU-computational-physics-NAKO

管一沣

Problem 1

Random Number Generator

选取前十组数据进行随机数生成,采用线性同余生成器,数学公式表示为 $X_{n+1} = (a \cdot X_n + c) \mod m$,生成 2500 个随机数。测试方法采用四种,分别为:分布均匀性测试、自相关性测试、Kolmogorov-Smirnov测试、频谱分析(FFT 后观察各个频率分布是否均匀)

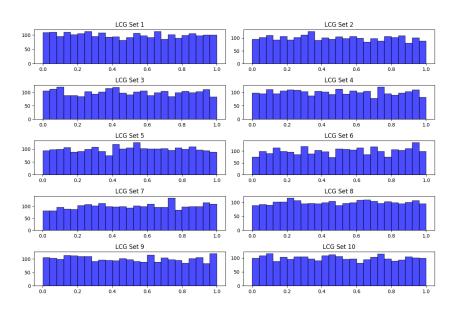


图 1: 在 10 组参数下生成的随机数分布的均匀性

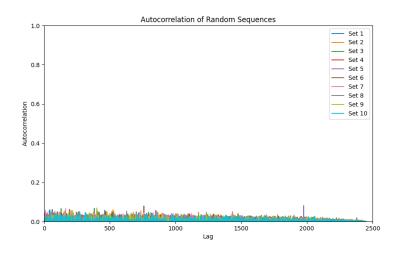


图 2: 在 10 组参数下生成的随机数分布的自相关性

```
Kolmogorov-Smirnov Test Results:

Set 1: KS Statistic = 0.02053, p-value = 0.23940

Set 2: KS Statistic = 0.02132, p-value = 0.20305

Set 3: KS Statistic = 0.01532, p-value = 0.59502

Set 4: KS Statistic = 0.01214, p-value = 0.85017

Set 5: KS Statistic = 0.02075, p-value = 0.22859

Set 6: KS Statistic = 0.02566, p-value = 0.07304

Set 7: KS Statistic = 0.02717, p-value = 0.04902

Set 8: KS Statistic = 0.01238, p-value = 0.83373

Set 9: KS Statistic = 0.02206, p-value = 0.17299

Set 10: KS Statistic = 0.01439, p-value = 0.67335
```

图 3: 在 10 组参数下生成的随机数分布的 Kolmogorov-Smirnov 测试

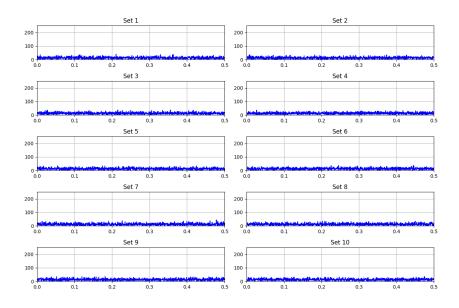
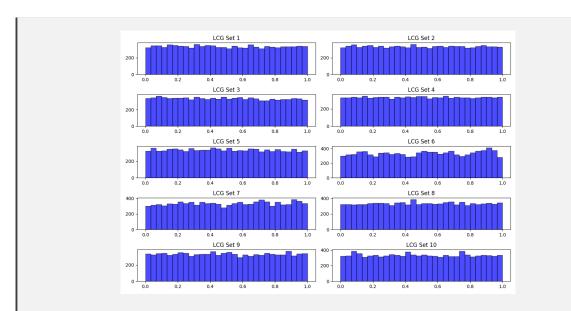


图 4: 在 10 组参数下生成的随机数分布的频谱分析

可以看到,分布均匀性较好,自相关性基本为 0,Kolmogorov-Smirnov 测试的 stat 较小且 p-value 较大,说明随机数序列接近均匀分布,频谱分析可以看到十组数据在各个频率上分布接近均匀。事实上 2500 个随机数相对并不多,但由于在较多随机数生成时存在绘图问题,只进行了生成 10000 个随机数时的分布均匀性测试,得到结果为:



可以看到随着生成随机数越多,随机数生成器的均匀性能体现的越好,这说明随机数生成器的性能是好的,能够生成较随机的伪随机数。

```
1 import numpy as np
 2 import matplotlib.pyplot as plt
 4 def lcg(m, a, c, seed, n):
      x = seed
      numbers = []
      for _ in range(n):
          x = (a * x + c) % m
          numbers.append(x / m)
      return numbers
10
11
12 params = [
      (214326, 1807, 45289),
13
      (244944, 1597, 51749),
14
      (233280, 1861, 49297),
15
      (175000, 2661, 26979),
      (121500, 4081, 25673),
17
      (145800, 3661, 30809),
      (139968, 3877, 29573),
      (214326, 3613, 45289),
20
      (714025, 1366, 150889),
21
      (134456, 8121, 28411),
22
23 ]
24
25 seed = 42
26 \text{ num\_samples} = 2500
28 random_sequences = []
```

```
29
30 for m, a, c in params:
      random_sequences.append(lcg(m, a, c, seed, num_samples))
32
33 def plot_histograms(sequences, n_cols=2):
      n_rows = len(sequences) // n_cols + (len(sequences) % n_cols > 0)
35
      fig, axs = plt.subplots(n_rows, n_cols, figsize=(12, 8))
      axs = axs.flatten()
36
      for i, seq in enumerate(sequences):
          axs[i].hist(seq, bins=25, alpha=0.7, color='blue', edgecolor='black')
39
          axs[i].set_title(f"LCG_Set_{1}1+1}")
      plt.tight_layout()
40
      plt.show()
41
42
43 plot_histograms(random_sequences)
45 def autocorrelation(sequence):
      n = len(sequence)
46
      mean = np.mean(sequence)
47
      autocorr = np.correlate(sequence - mean, sequence - mean, mode='full')
48
      autocorr = autocorr[n - 1:] / autocorr[n - 1]
      return autocorr
52 plt.figure(figsize=(10, 6))
53 for i, seq in enumerate(random_sequences):
      acorr = autocorrelation(seq)
      plt.plot(acorr[50:], label=f"Set_{\( \) {i+1}}")
55
56 plt.ylim([0, 1])
57 plt.xlim([0, 2500])
58 plt.title("Autocorrelation_of_Random_Sequences")
59 plt.xlabel("Lag")
60 plt.ylabel("Autocorrelation")
61 plt.legend()
62 plt.show()
64 from scipy.stats import kstest
66 print("Kolmogorov-Smirnov, Test, Results:")
67 for i, seq in enumerate(random_sequences):
      stat, p_value = kstest(seq, 'uniform')
68
      print(f"Set_{\sqcup}\{i+1\}:_{\sqcup}KS_{\sqcup}Statistic_{\sqcup}=_{\sqcup}\{stat:.5f\},_{\sqcup}p-value_{\sqcup}=_{\sqcup}\{p\_value:.5f\}")
69
70
71 def spectrum_analysis_all(sequences):
      n_{cols} = 2
72
      n_rows = len(sequences) // n_cols + (len(sequences) % n_cols > 0)
73
      fig, axs = plt.subplots(n_rows, n_cols, figsize=(12, 8))
74
      axs = axs.flatten()
75
```

```
76
      freq_range = (0, 0.5)
77
      magnitude_max = 250
78
79
       for i, sequence in enumerate(sequences):
80
          sequence = sequence - np.mean(sequence)
81
82
          fft_result = np.fft.fft(sequence)
          magnitude = np.abs(fft_result)
83
          freq = np.fft.fftfreq(len(sequence))
86
          axs[i].plot(freq[:len(freq)//2], magnitude[:len(magnitude)//2], color='blue')
          axs[i].set_title(f"Set_|{i+1}")
87
          axs[i].set_xlim(freq_range)
88
          axs[i].set_ylim(0, magnitude_max)
89
          axs[i].grid()
90
91
      for ax in axs[len(sequences):]:
92
          ax.axis('off')
93
94
      plt.tight_layout()
95
      plt.show()
98 spectrum_analysis_all(random_sequences)
100 def find_period(sequence):
      seen = {}
101
      for i, value in enumerate(sequence):
102
          if value in seen:
103
              return i - seen[value]
104
          seen[value] = i
105
      return None
106
108 print("Period_Lengths:")
109 for i, seq in enumerate(random_sequences):
      period = find_period(seq)
      print(f"Set_{i+1}:_Period_Length_=_{period_if_period_else_'No_Period_Detected'}")
```

Problem 2

Random walks in two dimensions

代码运行结果为:

```
Simple random walks:  \langle x(N) \rangle = 0.000, \ \langle y(N) \rangle = 0.000, \ \langle \Delta R^2(N) \rangle = 4.000  non-reversal random walks:  \langle x(N) \rangle = 0.000, \ \langle y(N) \rangle = -0.000, \ \langle \Delta R^2(N) \rangle = 4.971  self-avoiding random walks:  \langle x(N) \rangle = 0.000, \ \langle y(N) \rangle = 0.000, \ \langle \Delta R^2(N) \rangle = 6.350  Total number of self-avoiding walks: 618
```

值得注意的是,在实验过程中我发现由于浮点误差,代码中的"is_self_avoiding()" 函数会因为设置的随机行走方向存在浮点数,而导致判断错误得到有 634 步 self-avoiding random walks,但这是错误的,所以我修改随机行走方向全部为整数用来输入到"is_non_reversal()"和"is_self_avoiding()" 函数中用来判断随机行走类型,然后在输入到"calculate_statistics()" 函数中进行 x 和 y 相关的计算时,对 y 数据乘以 $\sqrt{3}$ 来将随机行走方向回到真实数值。

```
1 import itertools
2 import numpy as np
4 directions = [
     (1, 0),
    (-1, 0),
     (0.5, 1),
    (-0.5, 1),
     (0.5, -1),
     (-0.5, -1)
10
11 ]
12
13
14 def enumerate_walks(N):
      return list(itertools.product(range(6), repeat=N))
17 def calculate_trajectory(steps):
     x, y = 0, 0
    trajectory = [(x, y)]
     for step in steps:
20
        dx, dy = directions[step]
21
         x += dx
22
         y += dy
23
         trajectory.append((x, y))
24
    return trajectory
25
26
28 def is_non_reversal(steps):
```

```
29
      for i in range(1, len(steps)):
          if steps[i] == (steps[i - 1] + 3) % 6:
30
             return False
31
      return True
32
33
35 def is_self_avoiding(trajectory):
      visited = set()
      for point in trajectory:
37
          if point in visited:
             return False
          visited.add(point)
40
      return True
41
42
43 def is_self_avoiding_1(trajectory):
      visited = set()
44
      a=0
45
      for point in trajectory:
46
          if point in visited:
47
             return False
48
49
          visited.add(point)
          a += 1
      return True, a
51
53
54
55 def calculate_statistics(trajectories):
      x_vals = [traj[-1][0] for traj in trajectories]
56
      y_vals = [traj[-1][1]*np.sqrt(3)/2 for traj in trajectories]
57
      # TIMES \SQRT{3}/2 TO RETURN THE DIRECTION TO REAL VALUE.
58
59
60
      mean_x = np.mean(x_vals)
      mean_y = np.mean(y_vals)
61
      mean_x2 = np.mean(np.array(x_vals) ** 2)
62
      mean_y2 = np.mean(np.array(y_vals) ** 2)
      mean_delta_r2 = mean_x2 + mean_y2 - mean_x ** 2 - mean_y ** 2
      return mean_x, mean_y, mean_delta_r2
65
66
68 def main():
      N = 4
69
      all_walks = enumerate_walks(N)
70
71
      simple_trajectories = [calculate_trajectory(steps) for steps in all_walks]
72
      #PRINT(SIMPLE_TRAJECTORIES)
73
      simple_mean_x, simple_mean_y, simple_mean_delta_r2 = calculate_statistics(
74
           simple_trajectories)
```

```
75
                                    print("Simple_random_walks:")
                                     \textbf{print}(\mathbf{f}'' < \mathbf{x}(\mathbb{N}) >_{\sqcup} =_{\sqcup} \{ \texttt{simple\_mean\_x} : .3\mathbf{f} \}, _{\sqcup} < \mathbf{y}(\mathbb{N}) >_{\sqcup} =_{\sqcup} \{ \texttt{simple\_mean\_y} : .3\mathbf{f} \}, _{\sqcup} < \Delta \mathbb{R}^2(\mathbb{N}) >_{\sqcup} =_{\sqcup} \{ \mathbf{f}'' < \mathbf{x}(\mathbb{N}) >_{\sqcup} =_{\sqcup} \{ \mathbf{f}
76
                                                               simple_mean_delta_r2:.3f}")
77
                                    non_reversal_walks = [steps for steps in all_walks if is_non_reversal(steps)]
78
                                    non_reversal_trajectories = [calculate_trajectory(steps) for steps in non_reversal_walks
79
80
                                    non_reversal_mean_x, non_reversal_mean_y, non_reversal_mean_delta_r2 =
                                                               calculate_statistics(
                                                        non_reversal_trajectories)
                                    print("\nnon-reversal_random_walks:")
82
83
                                    print(
                                                         f''(x(N)) = \{non_reversal_mean_x: .3f\}, (y(N)) = \{non_reversal_mean_y: .3f\}, (AR^2(N)) = \{non_reversal_mean_y: .3f\}, (AR^2(N
84
                                                                                   □{non_reversal_mean_delta_r2:.3f}")
85
                                    self_avoiding_trajectories = [traj for traj in simple_trajectories if is_self_avoiding(
86
                                                             traj)]
                                    self_avoiding_mean_x, self_avoiding_mean_y, self_avoiding_mean_delta_r2 =
87
                                                             calculate_statistics(
                                                          self_avoiding_trajectories)
                                    print("\nself-avoiding_random_walks:")
                                    print(
                                                         f''(x(N)) = \{self_avoiding_mean_x: .3f\}, (y(N)) = \{self_avoiding_mean_y: .3f\}, (\Delta R^2(N)) \}
                                                                                   □=□{self_avoiding_mean_delta_r2:.3f}")
92
                                    for traj in simple_trajectories:
93
                                                         if is_self_avoiding(traj):
94
                                                                              a+=1
95
                                    print(f"Total\_number\_of\_self-avoiding\_walks: \_\{a\}")
96
97
98 if __name__ == "__main__":
QΩ
                                    main()
```

Problem 3

Numerical solution of the potential within a rectangular region.

```
a. 对 n_x=n_y=9, n_x=n_y=45, n_x=n_y=72 进行分析,结果如下,其中 n_x=n_y=45 是为了与 c 题进行对比。

Iterations for 9x9 grid: 57
Iterations for 45x45 grid: 393
Iterations for 72x72 grid: 257
```

```
1 import numpy as np
3 def initialize_grid(nx, ny, boundary_top, boundary_bottom, boundary_left, boundary_right):
      grid = np.zeros((ny, nx))
      grid[0, :] = boundary_top
5
      grid[-1, :] = boundary_bottom
      grid[:, 0] = boundary_left
      grid[:, -1] = boundary_right
      return grid
10
11 def jacobi_relaxation(grid, tol=1e-2, max_iter=10000):
      ny, nx = grid.shape
      new_grid = grid.copy()
13
      for iteration in range(max_iter):
          for i in range(1, ny-1):
15
             for j in range(1, nx-1):
16
                 new_grid[i, j] = 0.25 * (grid[i+1, j] + grid[i-1, j] + grid[i, j+1] + grid[i,
17
          diff = np.abs(new_grid - grid).max()
18
          if diff < tol:</pre>
19
20
             return new_grid, iteration
          grid[:, :] = new_grid
21
      return new_grid, max_iter
22
24 # FOR NX=NY=9
25 \text{ nx}_9, \text{ ny}_9 = 9, 9
26 boundary_top, boundary_bottom = 9, 9
27 boundary_left, boundary_right = 5, 5
28 grid_9x9 = initialize_grid(nx_9, ny_9, boundary_top, boundary_bottom, boundary_left,
       boundary_right)
29 final_grid_9x9, iterations_9x9 = jacobi_relaxation(grid_9x9, tol=0.01)
30
31 # FOR NX=NY=45
32 \text{ nx}_45, \text{ ny}_45 = 45, 45
33 grid_45x45 = initialize_grid(nx_45, ny_45, boundary_top, boundary_bottom, boundary_left,
       boundary_right)
34 final_grid_45x45, iterations_45x45 = jacobi_relaxation(grid_45x45, tol=0.01)
35
36 # FOR NX=NY=72
37 \text{ nx}_{72}, \text{ ny}_{72} = 72, 72
38 grid_72x72 = initialize_grid(nx_72, ny_72, boundary_top, boundary_bottom, boundary_left,
       boundary_right)
39 final_grid_72x72, iterations_72x72 = jacobi_relaxation(grid_72x72, tol=0.01)
41 print(f"Iterations_for_9x9_grid:__{iterations_9x9}")
42 print(f"Iterations_1for_45x45_grid:_{(iterations_45x45}")
```

```
43 print(f"Iterations_for_72x72_grid:_{(iterations_72x72}")
```

b. 结果如下,可以看到进行边界值平均作为初始猜测值的方法是更好的,需要的迭代次数远少于随机猜测初始值的方法。对两种方法得到的结果进行差值分析,可以看到两种方法的结果相差很小。

```
Iterations with average guess: 9
Iterations with random guess: 83
Mean Absolute Difference: 0.005835184233804118
Max Absolute Difference: 0.027777134676482795
```

```
1 import numpy as np
  3 def initialize_grid_with_guess(nx, ny, boundary_top, boundary_bottom, boundary_left,
                    boundary_right, guess_type, vmax=20):
                  grid = np.zeros((ny, nx))
                  grid[0, :] = boundary_top
                  grid[-1, :] = boundary_bottom
                  grid[:, 0] = boundary_left
                  grid[:, -1] = boundary_right
                  if guess_type == "average":
10
11
                            avg_value = (boundary_top + boundary_bottom + boundary_left + boundary_right) / 4
                            grid[1:-1, 1:-1] = avg_value
12
                  elif guess_type == "random":
13
                            grid[1:-1, 1:-1] = np.random.uniform(-vmax, vmax, size=(ny-2, nx-2))
14
15
16
                  return grid
18 def jacobi_relaxation(grid, tol=1e-2, max_iter=10000):
                  ny, nx = grid.shape
19
                 new_grid = grid.copy()
                  for iteration in range(max_iter):
21
                            for i in range(1, ny-1):
22
                                      for j in range(1, nx-1):
23
                                                 new_grid[i, j] = 0.25 * (grid[i+1, j] + grid[i-1, j] + grid[i, j+1] + grid[i, j
24
                                                              j-1])
                            diff = np.abs(new_grid - grid).max()
25
                            if diff < tol:</pre>
26
                                      return new_grid, iteration
27
                            grid[:, :] = new_grid
28
                  return new_grid, max_iter
29
30
31 nx, ny = 9, 9
32 boundary_top, boundary_bottom = 9, 9
```

```
33 boundary_left, boundary_right = 5, 5
34 \text{ vmax} = 20
36 # CASE 1: AVERAGE BOUNDARY VALUES GUESS
37 grid_average = initialize_grid_with_guess(nx, ny, boundary_top, boundary_bottom,
       boundary_left, boundary_right, "average")
38 final_grid_avg, iterations_avg = jacobi_relaxation(grid_average, tol=0.01)
40 # CASE 2: RANDOM GUESS
41 grid_random = initialize_grid_with_guess(nx, ny, boundary_top, boundary_bottom,
       boundary_left, boundary_right, "random", vmax=vmax)
42 final_grid_random, iterations_random = jacobi_relaxation(grid_random, tol=0.01)
44 print(f"Iterations_uwith_uaverage_uguess:_u{iterations_avg}")
45 print(f"Iterations_with_random_guess:_{(iterations_random)")
47 difference = final_grid_avg - final_grid_random
48 abs_difference = np.abs(difference)
50 mean_diff = np.mean(abs_difference)
51 max_diff = np.max(abs_difference)
53 print(f"Mean_Absolute_Difference_of_two_methods:_{mean_diff}")
54 print(f"Max_Absolute_Difference_of_two_methods:_{{max_diff}}")
```

c. 结果如下,可以看到比 a 中的 $n_x = n_y = 45$ 情况运算更快,迭代次数更少。

Checkerboard methods' iterations for grid 45x45: 344

```
import numpy as np

def initialize_grid(nx, ny, boundary_top, boundary_bottom, boundary_left, boundary_right):
    grid = np.zeros((ny, nx))
    grid[0, :] = boundary_top
    grid[-1, :] = boundary_bottom
    grid[:, 0] = boundary_left
    grid[:, -1] = boundary_right
    return grid

def checkerboard_jacobi_relaxation(grid, tol=1e-2, max_iter=10000):
    ny, nx = grid.shape
    new_grid = grid.copy()

for iteration in range(max_iter):
```

```
for i in range(1, ny-1):
16
                                            for j in range(1, nx-1):
 17
                                                        if (i + j) % 2 == 0: # RED POINT
 18
                                                                   new\_grid[i, j] = 0.25 * (grid[i+1, j] + grid[i-1, j] + grid[i, j+1] + grid[i+1, j] + grid[i+1,
 19
                                                                                  [i, j-1])
20
21
                                 for i in range(1, ny-1):
                                           for j in range(1, nx-1):
                                                        if (i + j) % 2 == 1: # BLACK POINT
                                                                   new_grid[i, j] = 0.25 * (new_grid[i+1, j] + new_grid[i-1, j] + new_grid[i,
24
                                                                                      j+1] + new_grid[i, j-1])
25
                                diff = np.abs(new_grid - grid).max()
26
                                 if diff < tol:</pre>
27
                                            return new_grid, iteration
28
29
                                 grid[:, :] = new_grid
30
31
                     return new_grid, max_iter
32
34 \text{ nx}, \text{ ny} = 45, 45
35 boundary_top, boundary_bottom = 9, 9
36 boundary_left, boundary_right = 5, 5
38 grid = initialize_grid(nx, ny, boundary_top, boundary_bottom, boundary_left, boundary_right)
40 final_grid, iterations = checkerboard_jacobi_relaxation(grid, tol=0.01)
42 print(f"Checkerboard\_methods'\_iterations\_for\_grid\_45x45: {iterations}")
```