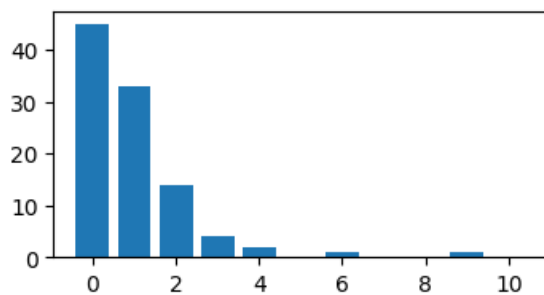
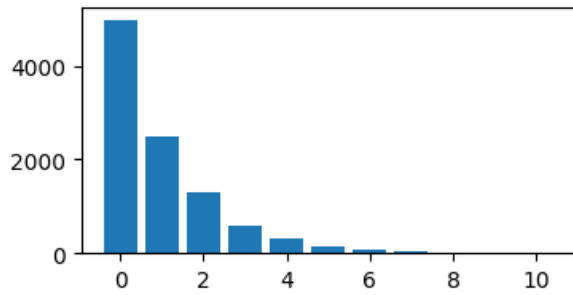
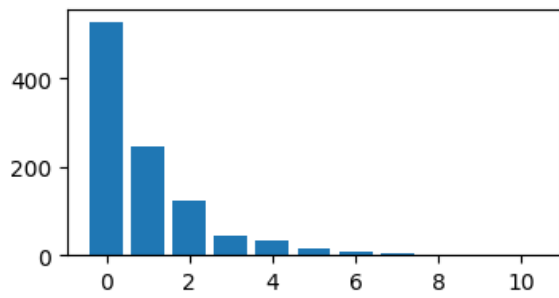


1.圣彼得堡问题

```
1 import numpy as np
2 import math
3 import random as rd
4 import matplotlib.pyplot as plt
5
6 # 硬币
7 def isFront():
8     return rd.choice([True, False])
9
10 # St. Petersburg question
11 def saint():
12     k = 0
13     while isFront() and k < 10:
14         k += 1
15     return k
16
17 # k次实验
18 def kSaint(k):
19     result = [0 for i in range(11)]
20     for i in range(k):
21         positive = saint()
22         result[positive] += 1
23     return result
24
25 # 作图
26 def kSaintPlot(k):
27     result = kSaint(k)
28     plt.figure(figsize=(4, 2))
29     plt.bar([i for i in range(len(result))], result)
30     plt.show()
31
32 kSaintPlot(100)
33 kSaintPlot(1000)
34 kSaintPlot(10000)
```

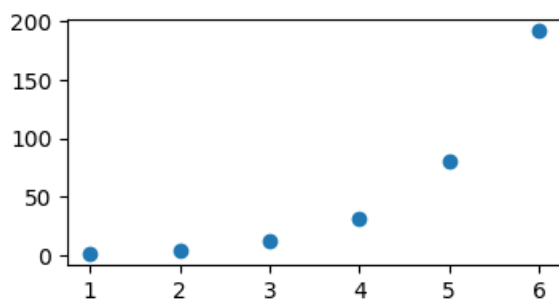




```

1 def earn():
2     earn = 0
3     k = 1
4     earn_values = []
5     while k < 7:
6         earn = k * 2 ** (k - 1)
7         earn_values.append(earn)
8         k += 1
9
10    plt.figure(figsize=(4, 2))
11    x_values = list(range(1, len(earn_values) + 1))
12    plt.scatter(x_values, earn_values)
13    plt.show()
14
15    earn()

```



3.

$$\int_{-\infty}^{\infty} g(x)dx = 0.1c + 0.6 * 2c + 0.3 * c/3 = 1 \Rightarrow c = \frac{5}{7}$$

当y取值为 $c = \frac{5}{7}$ 时，在[0,0.1)之间随机取一个数即得到对应的样本，同理，当y为 $2c, c/3$ 时，在区间[0.1,0.7),[0,7,1)随机取一个数即为样本

5

$$\int_{-\infty}^{\infty} cg(t)dt = c(\int_{-2}^2 \frac{8}{7} + \frac{118}{63}x^2 - \frac{74}{63}x^4 + \frac{10}{63}x^6)dx = c\frac{2}{63}(36x + \frac{59}{3}x^3 - \frac{37}{5}x^5 + \frac{5}{7}x^7)|_{-2}^2 = c\frac{35264}{6615} = 1$$

$$\Rightarrow c = \frac{6615}{35264}$$

显然偶函数 $f(x)$ 在 $[-2,0]$ 递减， $[0,2]$ 递增，其在 $[0, \infty]$ 的反函数显然存在，记为 $f^{-1}(x) = \begin{cases} 0 & x < -2 \\ g_1^{-1}(x) & x \in [-2, 0] \\ g_2^{-1}(x) & x \in (0, 2] \\ 0 & x > 2 \end{cases}$,

那么取 $h(x) = \begin{cases} 0 & x < -2 \\ -g_1^{-1}(x) & x \in [-2, 0] \\ g_2^{-1}(x) & x \in (0, 2] \\ 0 & x > 2 \end{cases}$ ，再取 $(-\infty, +\infty)$ 的均匀分布 X ，那么 $h(X)$ 即为所需抽样

7

不妨记 $A = (a_{ij})_{2 \times 2}$, $\mu = (\mu_1, \mu_2)^T$ 那么 $Y_1 \sim N(\mu_1, a_{11})$, $Y_2 \sim N(\mu_2, a_{22})$

其概率密度函数为 $p(y_1, y_2) = \frac{1}{2\pi\sqrt{(a_{11}^2 + a_{12}^2)(a_{21}^2 + a_{22}^2)}} \exp\{-\frac{1}{2}[\frac{(y_1 - \mu_1)^2}{a_{11}^2 + a_{12}^2} + \frac{(y_2 - \mu_2)^2}{a_{21}^2 + a_{22}^2}]\}$

Gibbs采样模拟圆周率随机投点

```

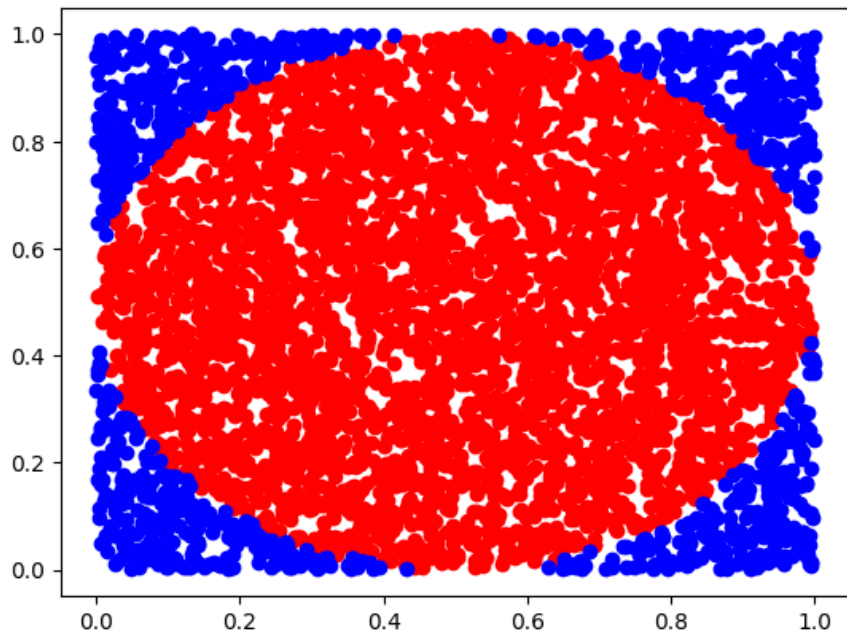
1 def Gibbs_pi(num):
2     def In_cir(x,y):
3         result = math.sqrt((x-0.5)**2 + (y-0.5)**2)
4         return result < 0.5
5
6     n = 0
7     fig, ax = plt.subplots()
8     inLstX, inLstY, outLstX, outLstY = [], [], [], []
9     for _ in range(num):
10        x, y = rd.uniform(0,1), rd.uniform(0,1)
11        if (In_cir(x,y)):
12            n += 1
13            inLstX.append(x)
14            inLstY.append(y)
15        else:
16            outLstX.append(x)
17            outLstY.append(y)
18    plt.scatter(inLstX,inLstY,color = 'red')
19    plt.scatter(outLstX,outLstY,color = 'blue')

```

```

20     plt.show()
21     return 4*n/num
22
23 print(Gibbs_pi(5000))
24
25
26

```



```

1 3.1744

```

蓄水池算法

- 1.先准备一个大小为m的返回结果ans，即蓄水池。
 - 2.对于先来到到的前m个数，先来先得，直接丢入蓄水池中，占好位置。
 - 3.对于后面来到的第cnt（cnt>m）个数，有m/cnt的概率保留，如果保留，则用其替换出蓄水池中的任意一个数。
- 对于样本中的每个数来说，保留在蓄水池中的概率都相同

```

1 sample = [rd.randint(1,100) for _ in range(100)]
2 def reserSamp(lst, m):
3     if m >= len(lst):
4         return sample
5     else:
6         cnt = len(lst) - m
7         mindex = list(range(m))
8         def isSave(m):
9             return rd.uniform(0,1) < m/cnt
10        def swap(lst, n):
11            del lst[rd.choice(mindex)]
12            lst.append(n)
13        return 0

```

```
14
15     ans = [lst[i] for i in range(m)]
16     for i in range(m, len(lst)):
17         if isSave(cnt):
18             swap(ans, lst[i])
19     return ans
20
21 print(reserSamp(sample, 5))
22
```

```
1 | [29, 82, 68, 22, 73]
```