

# Homework\_5

January 3, 2020

## 1 Homework 5 - STAT 7730

### 1.1 Random Sampling

#### 1.1.1 Problem 1

Consider the pmf  $p(x)$  defined by  $p(x) = c\sqrt{(x)}\mathbb{I}(x \in \{1, 2, \dots, 50\})$

(a) What is the numerical value of  $c$ ?

```
[2]: import math
import numpy as np
import matplotlib.pyplot as plt
```

```
[31]: x = np.array(range(0,51))**(.5)
c = x.sum()**(-1)

print("The numerical value of c is %f" % c)
```

The numerical value of  $c$  is 0.004183

(b) For  $n = 10^3$ , generate  $X_1, \dots, X_n$  iid with common pmf  $p$ . Let

$$\hat{p}(x) = \#\{k : X_k = x\}/n$$

be the *empirical pmf* of  $X_1, \dots, X_n$ . Make a graph comparing  $p(x)$  and  $\hat{p}(x)$ . Repeat with  $n = 10^6$ . Interpret your results.

```
[32]: prob = x*c;

n = 1000
sample = np.zeros(n)

for i in range(0,n):
    rnd = np.random.uniform()
    sample[i] = sum(x <= rnd for x in np.cumsum(prob))

p_hat = np.zeros(50)

for i in range(0,50):
    p_hat[i] = sum(sample == i+1)
```

```
p_hat = p_hat/n
```

```
[36]: fig, ax = plt.subplots()
ax.plot(list(range(1,51)), prob[1:51], 'r.', label='Real pmf')
ax.plot(list(range(1,51)), p_hat, 'b.', label = 'Empirical pmf')
plt.ylabel('probability')
plt.xlabel('x')
ax.legend()
plt.show()
```

The real pmf and the empirical pmf differ significantly in the first case when  $n = 1000$ . Nevertheless, in the next case ( $n = 1000000$ ) we can observe that both pmfs coincide.

```
[7]: n = 1000000
sample = np.zeros(n)
cmsm = np.cumsum(prob)
for i in range(0,n):
    rnd = np.random.uniform()
    sample[i] = sum(cmsm <= rnd)

p_hat2 = np.zeros(50)

for i in range(0,50):
    p_hat2[i] = sum(sample == i+1)
```

```
p_hat2 = p_hat2/n
```

```
[26]: fig, ax = plt.subplots()
ax.plot(list(range(1,51)), prob[1:51], 'yo', label='Real pmf')
ax.plot(list(range(1,51)), p_hat2, 'b.', label = 'Empirical pmf')
plt.ylabel('probability')
plt.xlabel('x')
ax.legend()
plt.show()
```

## 2. Consider the pdf

$$f(x) = \frac{1}{2\sqrt{x}} \mathbb{I}(0 < x < 1)$$

(a) How would you convert a uniform (0,1) random variable  $U$  into a random variable  $X$  that had pdf  $f$ ?

Take  $X = U^2$ , then if  $0 \leq x \leq 1$  we have

$$F_X(x) = P[X \geq x] = P[U^2 \geq x] = P[U \geq \sqrt{x}] = \sqrt{x}$$

and

$$f_X(x) = \frac{1}{2\sqrt{x}} \mathbb{I}(0 < x < 1)$$

(b) Generate a sample of size  $10^3$  from  $f$  and compare a plot of  $f$  to a histogram of your samples. Repeat with a sample of size  $10^6$ .

As in the previous problem, the empirical distribution seems to converge only in the second case, with  $n = 10^6$ . The plot of  $f$  and this last histogram have similar shapes.

```
[55]: n = 1000

u = np.random.uniform(size = n)

sample_f = u**2
weights = np.ones_like(sample_f)/len(sample_f)
plt.hist(sample_f, 20, weights=weights, alpha=0.8)
plt.show()
plt.plot([x/20 for x in range(1,21)], \
         [0.5*math.sqrt(x/20)**(-1) for x in range(1,21)], 'r')
plt.show()
```

```

[56]: n = 1000000

u = np.random.uniform(size = n)

sample_f = u**2
weights = np.ones_like(sample_f)/len(sample_f)
plt.hist(sample_f, 20,weights=weights, alpha=0.8)
plt.show()

plt.plot([x/20 for x in range(1,21)],\
         [0.5*math.sqrt(x/20)**(-1) for x in range(1,21)], 'r')
plt.show()sum = lambda arg1, arg2: arg1 + arg2

```

## 1.2 Problem 3

Consider the function

$$g(x) = (\sin(10x))^2 |x^3 + 2x - 3| \not\equiv (x \in (-1, 0) \cup (1/2, 1))$$

and the pdf  $f(x) = \frac{1}{6}g(x)$  that is the same shape as  $g$ . Use rejection sampling to generate a sample of size  $10^3$  from  $f$  and compare a plot of  $g$  to a histogram of your samples. Repeat with a sample of size  $10^6$ .

Note that  $x^3 + 2x - 3 = (x - 1)(x^2 + x + 3) = ([x + 0.5]^2 + 2.75) < 0$  when  $x < 1$ .

Hence,  $|x^3 + 2x - 3| = 3 - 2x - x^3$ , and

$$\frac{d}{dx}(3 - 2x - x^3) = -3x^2 - 2 < 0,$$

so that  $3 - 2x - x^3$  is decreasing and attains its maximum when  $x = -1$ . Because  $(\sin(10x))^2 \leq 1$ , we have that

$$g(x) \leq (3 - 2x - x^3)\mathbb{1}(x \in (-1, 0) \cup (1/2, 1)) \leq 3 - 2(-1) - (-1)^3 = 6.$$

Therefore, we can use rejection sampling on the rectangle  $[0, 6] \times [-1, 1]$ . [Figure] shows how with  $n = 6$ , the empirical distribution shows a very similar shape to the one of  $g$ .

```
[21]: g = lambda x: (math.sin(10*x)**2)*abs(x**3 + 2*x- 3)*(((x > -1)and (x < 0)) +
    ↪((x > 1/2) and (x < 1)));

n = 1000
i = 0
sample = np.zeros(n)

while (i < n):
    x = np.random.uniform()*2 - 1
    y = np.random.uniform()*6
    if (y <= g(x)):
        sample[i] = x
        i = i + 1

weights = np.ones_like(sample)/len(sample)
plt.hist(sample, 30,weights=weights, alpha=0.8)
plt.show()

plt.plot([float(x)/40-1 for x in range(0,81)],\
         [g(float(x)/40-1) for x in range(0,81)], 'r')
plt.show()
```

```
[37]: n = 1000000  
  
i = 0  
sample = np.zeros(n)
```



```

while (i < n):
    x = np.random.uniform()*2 - 1
    y = np.random.uniform()*6
    if (y <= g(x)):
        sample[i] = x
        i = i + 1

weights = np.ones_like(sample)/len(sample)
plt.hist(sample, 30, weights=weights, alpha=0.8)
plt.show()

plt.plot([float(x)/40-1 for x in range(0,81)], \
         [g(float(x)/40-1) for x in range(0,81)], 'r')
plt.show()

```

