

Queueing Theory: Empirical distributions.

EBB074A05

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2020:11:15

0.1 DONE Change the font size of emacs

0.2 DONE Change background to white

1 General info

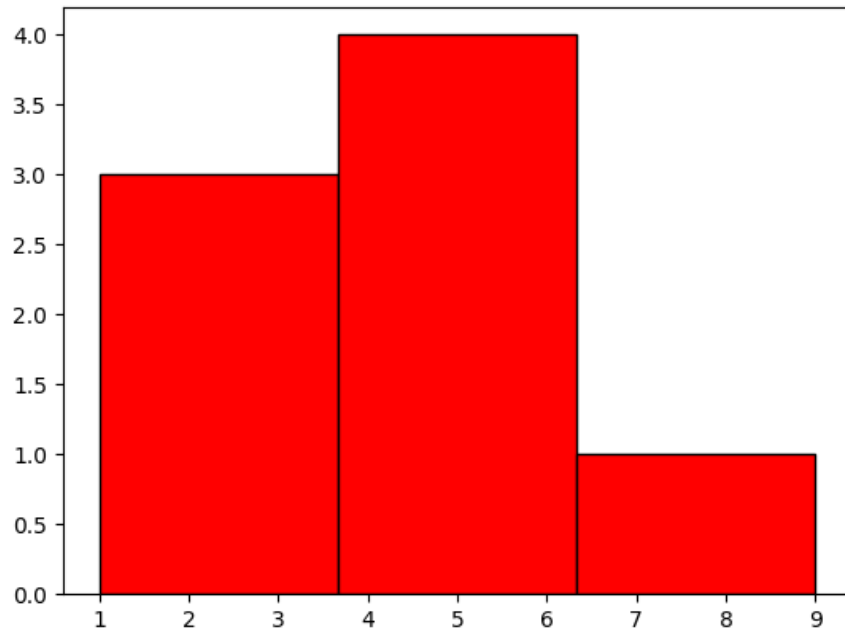
This file contains the code and the results that go with this youtube movie: <https://youtu.be/aKfv908uWqM>

2 Empirical distribution, how to make and plot

We want to know the fraction of periods the queue length is longer than some value q , say. For this we will make the empirical distribution of the queue lengths.

2.1 Plotting a PDF/histogram

```
1 import matplotlib.pyplot as plt
2
3 x = [2, 5, 2, 1, 9, 5, 5, 5]
4
5 plt.clf()
6 plt.hist(x, bins=3, facecolor='red', edgecolor='black', linewidth=1)
7 plt.savefig('emp0.png')
8 'emp0.png'
```



2.1.1 DONE Explain: the `plt.clf()` is necessary to clear earlier plots.

2.1.2 DONE Change the number of bins from 3 to 7.

you can remove the bins argument altogether.

2.2 First naive idea

Given a set of measurements x_1, \dots, x_n , the empirical CDF is defined as

$$F(x) = \sum_{i=1}^n I_{x_i \leq x} / n$$

This is a clean mathematical definition, but as is often the case with mathematical definitions, you should stay clear from using it to *compute* the CDF: the numerical performance is absolutely terrible.

```

1 x = [2, 5, 2, 1, 9, 5, 5, 5]
2
3 def F(y):
4     tot = 0
5     for xi in x:
6         tot += xi <= y
7
8     return tot/len(x)
9
10 F(5.5)

```

0.875

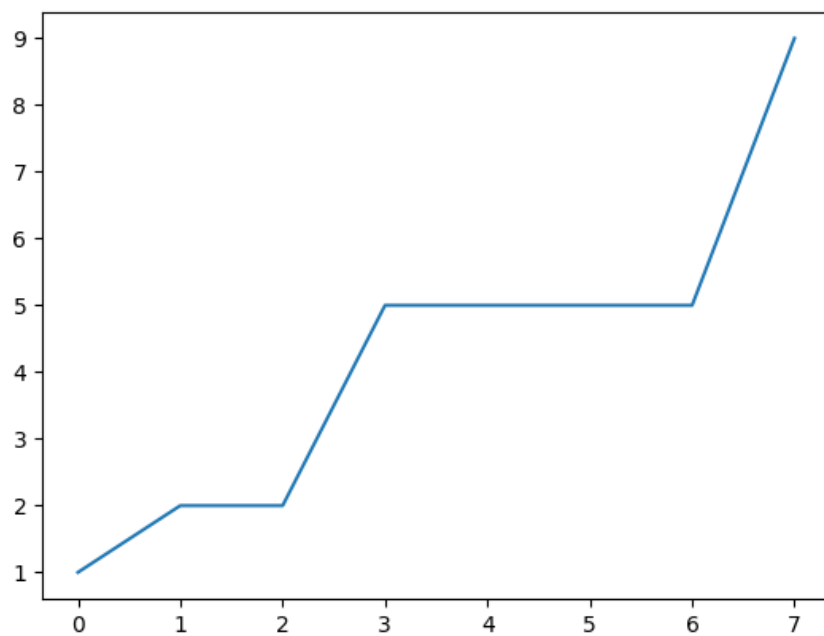
2.2.1 TODO Why is the numerical performance so bad?

2.3 A better idea

```
1 sorted(x)
```

1 2 2 5 5 5 5 9

```
1 plt.clf()
2 plt.plot(sorted(x))
3 plt.savefig("emp00.png")
4 "emp00.png"
```



2.4 Yet better idea

```
1 def cdf_better(x):
2     x = sorted(x)
3     n = len(x)
4     y = range(1, n + 1)
5     y = [z / n for z in y] # normalize
6     return x, y
7
8
9 x = [2, 5, 2, 1, 8, 5, 5]
10 x, F = cdf_better(x)
11 F
```

0.14285714285714285 0.2857142857142857 0.42857142857142855 0.5714285714285714 0.7142857142857143

2.4.1 DONE Explain

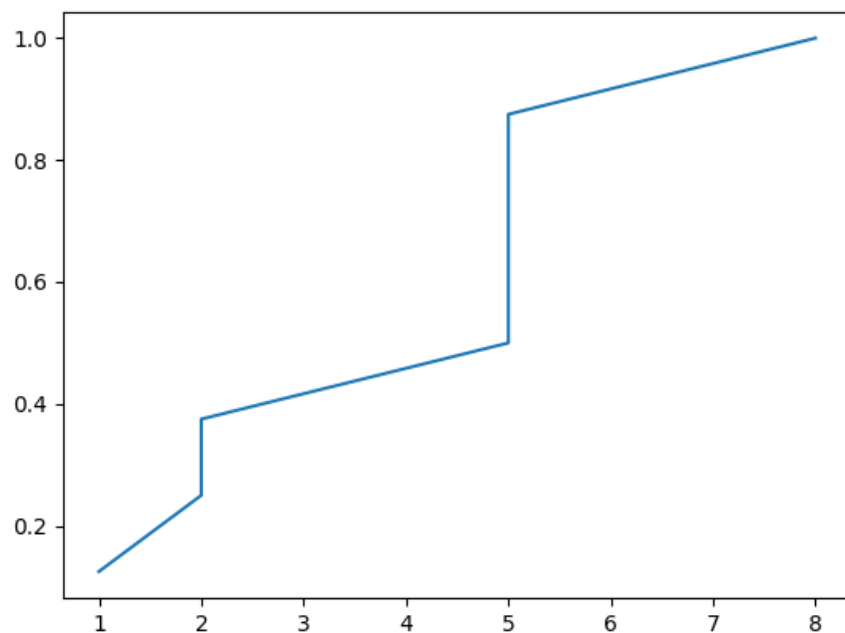
Why the $\sim n = \text{len}(x)$

2.4.2 TODO Explain

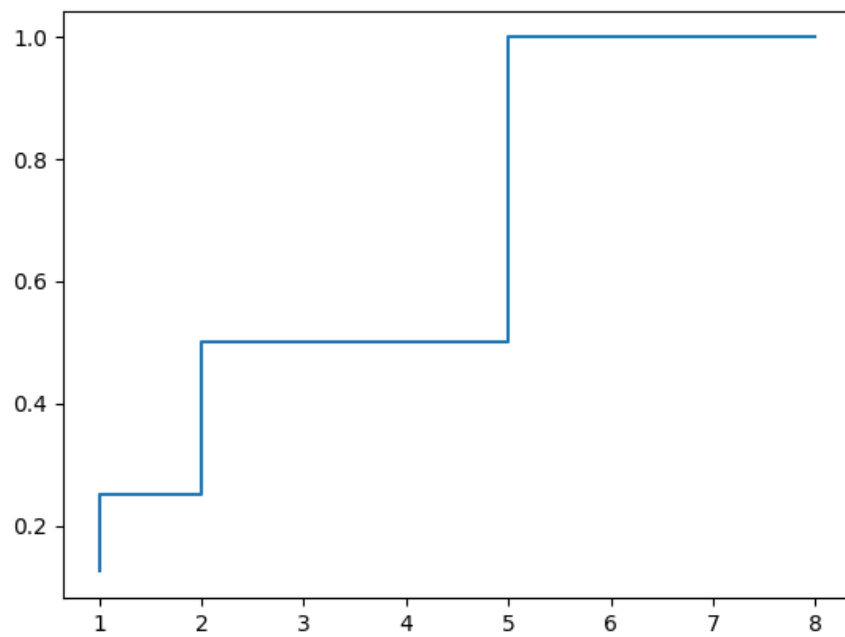
You should know that for loops in R and python are quite slow. We use this in the list comprehension in the line in which we `#normalize`. For larger amounts of data it is better to use numpy. This we do below.

2.5 Plot the cdf

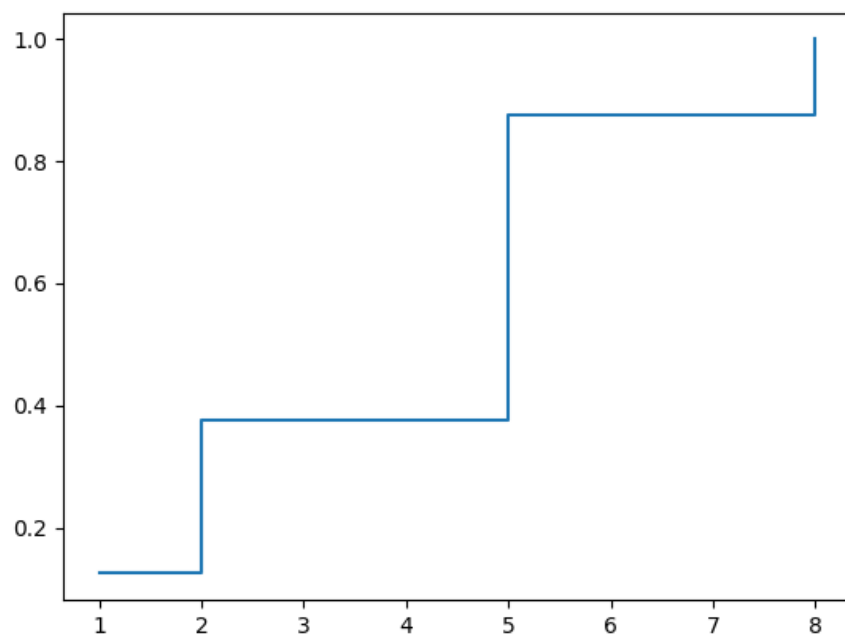
```
1 x = [2, 5, 2, 1, 8, 5, 5, 5]
2 x, F = cdf_better(x)
3
4 plt.clf()
5 plt.plot(x, F)
6 plt.savefig(fname)
7 fname
```



```
1 plt.clf()
2 plt.step(x, F)
3 plt.savefig(fname)
4 fname
```



```
1 plt.clf()
2 plt.plot(x, F, drawstyle="steps-post")
3 plt.savefig(fname)
4 fname
```



2.6 Faster with numpy

```
1 import numpy as np
2
3 def cdf(x):
4     y = np.arange(1, len(x) + 1) / len(x)
5     x = np.sort(x)
6     return x, y
7
8 x = [2, 5, 2, 1, 8, 5, 5]
9 x, F = cdf(x)
10 F
```

0.14285714 0.28571429 0.42857143 0.57142857 0.71428571 0.85714286 1

2.7 Remove duplicate values

Finally, we can make the computation of the cdf significantly faster with using the following numpy functions.

```
1 unique, count = np.unique(np.sort(x), return_counts=True)
2 unique, count
```

array ((1 2 5 8)) array ((1 2 3 1))

```
1 count.cumsum()/7
```

0.14285714 0.42857143 0.85714286 1

```
1 def cdf_fastest(x):
2     # remove multiple occurrences of the same value
3     unique, count = np.unique(np.sort(x), return_counts=True)
4     x = unique
5     y = count.cumsum() / count.sum()
6     return x, y
7
8 x = [2, 5, 2, 1, 8, 5, 5]
9 x, F = cdf_fastest(x)
10 F
```

0.14285714 0.42857143 0.85714286 1