

STAT 753: Stochastic Models and Simulations

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Homework 1

1. If a fair coin is flipped, find the probability that the first Head appears on the third toss.

Solution

The sample space for a coin flipped 3 times is

$$\{HHH, TTT, HHT, THH, HTH, THT, TTH, TTH\}. \quad (1)$$

The probability of an event occurring is:

$$P(E) = \frac{\# \text{ favorable outcomes}}{\# \text{ total outcomes}}. \quad (2)$$

In this case, there is one favorable outcome out of eight possible outcomes, so the probability is $\frac{1}{8}$ or 12.5%. Alternatively, we can see that each coin toss is an independent event, with $P(H) = P(T) = \frac{1}{2}$. Therefore, the probability of flipping heads on the third toss is equal to

$$P(T) \cdot P(T) \cdot P(H) = \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} = \frac{1}{8}. \quad (3)$$

Code

```
1 np.random.seed(10)
2 toss_trials = []
3
4 p=0.5 #probability for a fair coin
5 N=1000 #desired simulations
6
7 #define Heads=1, Tails=0
8 for i in range(N):
9     toss_trials.append(np.random.binomial(1,p, size=3))
10
11 desired_outcome=[0,0,1]
12 total_TTH_events = [toss_trials[i] for i in range(N) if all(toss_trials[i]
13     ↪ == desired_outcome)== True]
13 prob1 = (len(total_TTH_events)/N)
14 print('Probability of event {T,T,H} based on 1000 simulations: %s' % prob1)
15 print('Theoretical probability of event {T,T,H}: 0.125')
16 print('Percent error: %s%%' % np.round(((prob1 - 0.125)/0.125)*100,3))
17
18 >> Probability of event {T,T,H} based on 1000 simulations: 0.129
19 >> Theoretical probability of event {T,T,H}: 0.125
20 >> Percent error: 3.2%
```

2. Assume 100 houses are independently on fire, each with probability 2.5%. Find the probability that there are exactly 2 fires, using: (a) the binomial distribution; (b) the Poisson approximation.

Solution

(a) The probability that exactly 2 houses are on fire can be found by simply plugging in the known values to the binomial pdf:

$$P(k) = \binom{n}{k} p^k (1-p)^{n-k} = \binom{100}{2} 0.025^2 (1-0.025)^{100-2} \quad (4)$$

The resulting probability is approximately 0.25878.

(b) The Poisson pdf is defined as:

$$P(k) = \frac{\lambda^k e^{-\lambda}}{k!}. \quad (5)$$

We can compute the probability of exactly 2 houses on fire by noting that $k=2$ and $E[\lambda] = n \cdot p = 100 \cdot 0.025 = 2.5$. Therefore, the probability is calculated as:

$$\frac{\lambda^2 e^{-\lambda}}{2!} = \frac{2.5^2 e^{-2.5}}{2} = 0.2565. \quad (6)$$

Code

Part (a)

```

1  np.random.seed(100)
2  N=1000 #desired simulations
3  n_houses=100
4  p=0.025
5
6  binomial_sample = np.random.binomial(n_houses, p, size=N)
7  prob2a_sim = len([binomial_sample[k] for k in range(N) if
   ↪ binomial_sample[k]==2])/N
8  prob2a_theory = binom.pmf(k=2, n=100, p=0.025)
9
10 print('Probability of exactly 2 fires based on 1000 simulations with
   ↪ binomial dist: %s' % prob2a_sim)
11 print('Theoretical probability of exactly 2 fires with binomial dist: %s' %
   ↪ np.round(prob2a_theory,4))
12 print('Percent error: %s%%' % np.abs(np.round(((prob2a_sim -
   ↪ prob2a_theory)/prob2a_theory)*100,3)))
13
14 >> Probability of exactly 2 fires based on 1000 simulations with binomial
   ↪ dist: 0.263
15 >> Theoretical probability of exactly 2 fires with binomial dist: 0.2588
16 >> Percent error: 1.629%
17

```

Part (b)

```
1 np.random.seed(1000)
2 N=1000 #desired simulations
3 n_houses=100
4 p=0.025
5 poisson_sample = np.random.poisson(lam=n_houses*p,size=N)
6
7 prob2b_sim = len([poisson_sample[k] for k in range(N) if
  ↳ poisson_sample[k]==2])/N
8 prob2b_theory = poisson.pmf(k=2, mu=100*0.025)
9
10 print('Probability of exactly 2 fires based on 1000 simulations with
  ↳ poisson dist: %s' % prob2b_sim)
11 print('Theoretical probability of exactly 2 fires with poisson dist: %s' %
  ↳ np.round(prob2b_theory,4))
12 print('Percent error: %s%%' % np.round(((prob2b_sim -
  ↳ prob2b_theory)/prob2b_theory)*100,3))
13
14 >> Probability of exactly 2 fires based on 1000 simulations with poisson
  ↳ dist: 0.273
15 >> Theoretical probability of exactly 2 fires with poisson dist: 0.2565
16 >> Percent error: 6.426%
```

3. Historical annual returns of S&P 500 have mean 0.067 and standard deviation 0.165. Assuming they are independent, find the probability that the return is positive if the distribution is: (a) Gaussian; (b) Laplace.

Solution

(a) For a Gaussian distribution $X \sim \mathcal{N}(0.067, (0.165)^2)$, the probability that an annual return is positive can be written as $P(0 < x < \infty) = 1 - F_x(0)$, where

$$F_x(0) = \Phi\left(\frac{x - \mu}{\sigma}\right) = \Phi\left(\frac{0 - 0.067}{0.165}\right) = \Phi(-0.4067). \quad (7)$$

By symmetry, $\Phi(-z) = 1 - \Phi(z)$. The probability can then be written as

$$P(0 < x < \infty) = 1 - (1 - \Phi(0.4067)) = \Phi(0.4067). \quad (8)$$

Using a standard normal distribution chart, the approximate probability is 0.6576.

(b) The Laplace pdf is defined as

$$f(x) = \frac{1}{2\lambda} e^{-\frac{|x-\mu|}{\lambda}} \quad (9)$$

where $\mu = 0.067$ and $\lambda = 0.165$. To find $P(0 < x < \infty)$, we need to integrate $f(x)$ over the desired bounds. Doing so yields

$$F(x > 0) = 1 - \frac{1}{2} e^{-\frac{|x-\mu|}{\lambda}} = 1 - \frac{1}{2} e^{-\frac{|0-0.067|}{0.165}} \quad (10)$$

The resulting probability is approximately 0.6669.

Code

Part (a)

```

1 np.random.seed(1000)
2 mu = 0.067
3 sigma= 0.165
4 N =1000
5 gaussian_sample = np.random.normal(mu, sigma, N)
6
7 prob3a_sim = len([gaussian_sample[k] for k in range(N) if
  ↳ gaussian_sample[k]>0])/N
8 prob3a_theory = 1 - norm.cdf(0, mu, sigma)
9
10 print('Probability of S&P 500 positive return with gaussian dist: %s' %
  ↳ prob3a_sim)
11 print('Theoretical probability of S&P 500 positive return with gaussian
  ↳ dist: %s' % np.round(prob3a_theory,4))
12 print('Percent error: %s%%' % np.round(((prob3a_sim -
  ↳ prob3a_theory)/prob3a_theory)*100,3))

```

```
13
14 >> Probability of S&P 500 positive return with gaussian dist: 0.662
15 >> Theoretical probability of S&P 500 positive return with gaussian dist:
    ↪ 0.6577
16 >> Percent error: 0.661%
```

Part (b)

```
1 np.random.seed(1000)
2 mu = 0.067
3 sigma= 0.165
4 N =1000
5 laplace_sample = np.random.laplace(mu, sigma, N)
6
7 prob3b_sim = len([laplace_sample[k] for k in range(N) if
    ↪ laplace_sample[k]>0])/N
8 prob3b_theory = 1 - laplace.cdf(0, mu, sigma)
9
10 print('Probability of S&P 500 positive return with laplace dist: %s' %
    ↪ prob3b_sim)
11 print('Theoretical probability of S&P 500 positive return with laplace
    ↪ dist: %s' % np.round(prob3b_theory,4))
12 print('Percent error: %s%%' % np.abs(np.round(((prob3b_sim -
    ↪ prob3b_theory)/prob3b_theory)*100,3)))
13
14 >> Probability of S&P 500 positive return with laplace dist: 0.652
15 >> Theoretical probability of S&P 500 positive return with laplace dist:
    ↪ 0.6669
16 >> Percent error: 2.229%
```