

01Simulating risk data

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1 Simulating risk data

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1.1 Data generating process

The data generating process of real-life transportation risk is very complex. As an illustrating example, here we assume that the risk was generated from a Poisson distribution with the following data generating process:

$$Y_i \sim \text{Poisson}(d_i \cdot \lambda_i) \quad (1)$$

$$\log(\lambda_i) = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \epsilon_i \quad (2)$$

$$\epsilon_i \sim \text{Normal}(0, 2^2), \quad (3)$$

where d_i is the distance traveled in the i -th trip, x_1 is precipitation, and x_2 is road traffic.

We assume the sample size $N = 10,000$ and the parameters and data has following values or distributions:

- $\beta_0 = -3$,
- $\beta_1 = 0.5$,
- $\beta_2 = 0.9$,
- $d \sim \text{Poisson}(1000)$
- $x_1 \sim \text{Bernoulli}(0.15)$,
- $x_2 \sim \text{Beta}(2, 2)$

1.2 Simulating data

1.2.1 Import packages and print package version

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import beta
import sys

print("Python version: " + sys.version)
print("pandas version: " + pd.__version__)
```

```
print("numpy version: " + np.__version__)
```

Python version: 3.7.5 (default, Oct 31 2019, 15:18:51) [MSC v.1916 64 bit (AMD64)]

pandas version: 0.25.3

numpy version: 1.17.4

1.2.2 set parameters and generate data

```
[2]: def simulate_distance(N_size):  
      return(np.random.poisson(lam = 1000, size = N_size))  
def simulate_precipitation(N_size):  
      return(np.random.binomial(n = 1, p = 0.15, size = N_size))  
def simulate_traffic(N_size):  
      return(np.random.beta(a = 2, b = 2, size = N_size))
```

```
[3]: np.random.seed(123) # set random seed  
N = 10**4  
b0, b1, b2 = -10, 0.5, 0.9  
  
d = simulate_distance(N)  
x1 = simulate_precipitation(N)  
x2 = simulate_traffic(N)  
epsilon = np.random.normal(loc = 0, scale = 1, size = N)  
  
lambda_i = np.exp(b0 + b1*x1 + b2*x2 + epsilon)  
y = np.random.poisson(d*lambda_i)
```

1.2.3 Plot distributions of simulated data

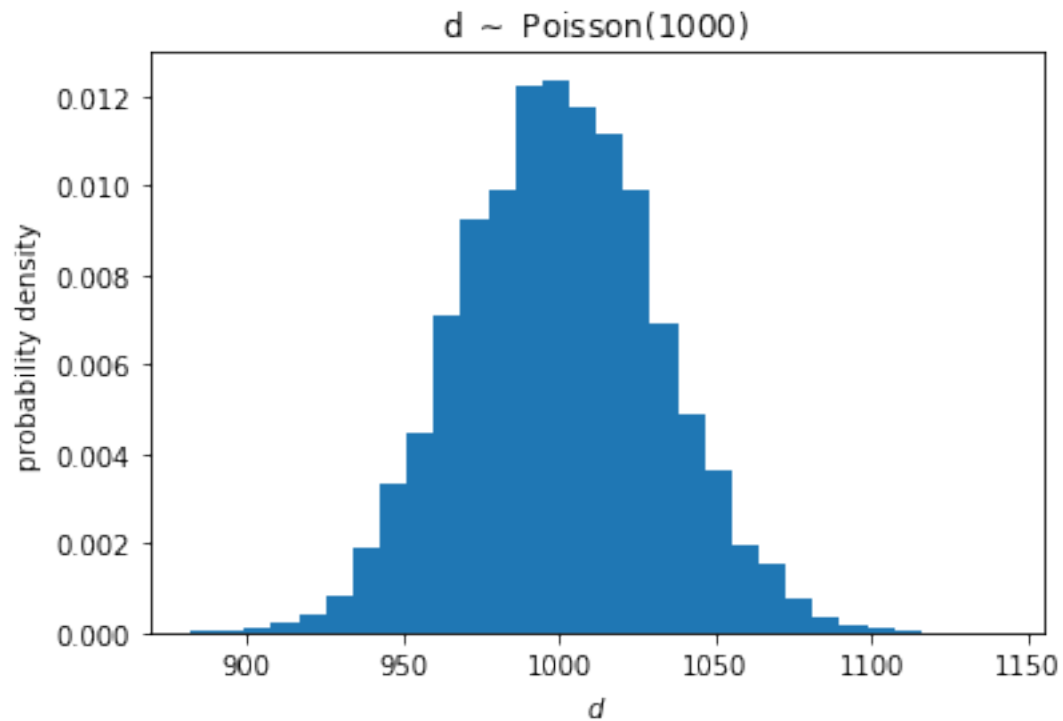
```
[4]: # The distribution of y  
from collections import Counter  
print("The maximum potential crash in a trip is " + str(max(y)))  
print("The distribution of y is: \n" + str(Counter(y)))
```

The maximum potential crash in a trip is 5

The distribution of y is:

Counter({0: 8910, 1: 937, 2: 117, 3: 24, 4: 6, 5: 6})

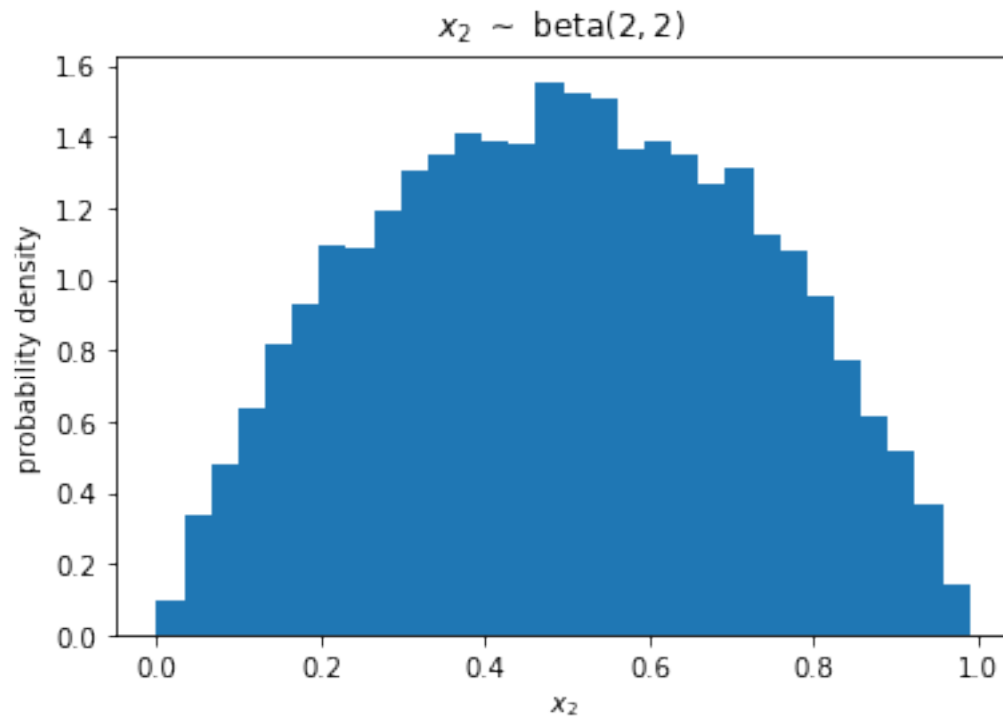
```
[5]: # The distribution of d  
from scipy.special import factorial  
count, bins, ignored = plt.hist(d, 30, density=True)  
plt.title("d $\sim$ Poisson(1000)$")  
plt.xlabel('$d$')  
plt.ylabel('probability density')  
plt.show()
```



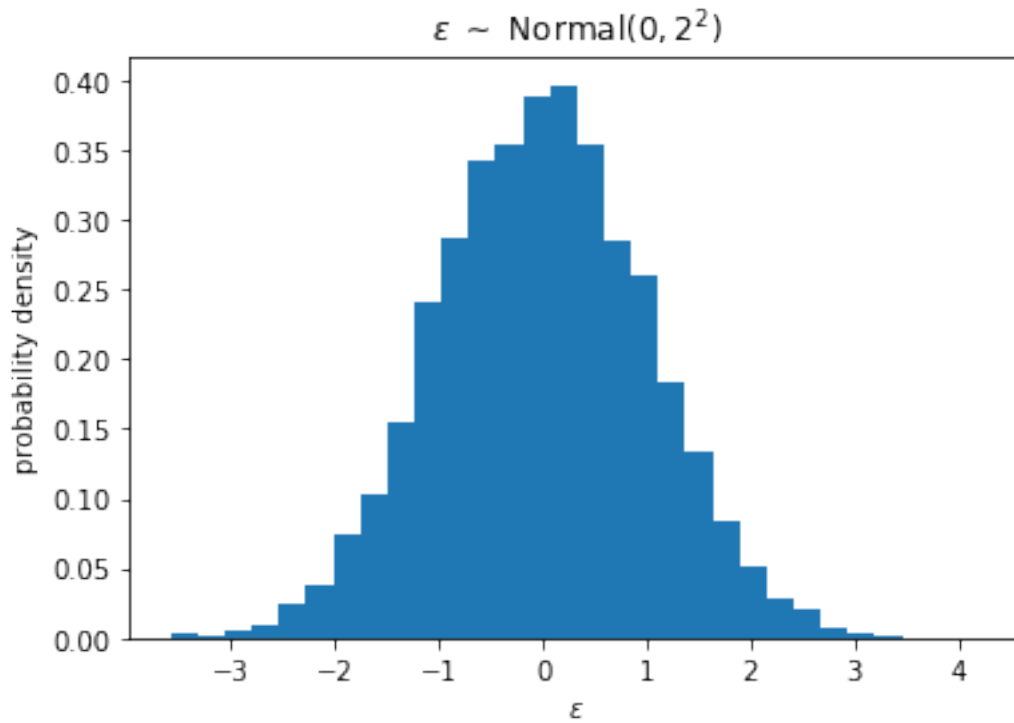
```
[6]: # The distribution of x1
Counter(x1)
```

```
[6]: Counter({0: 8508, 1: 1492})
```

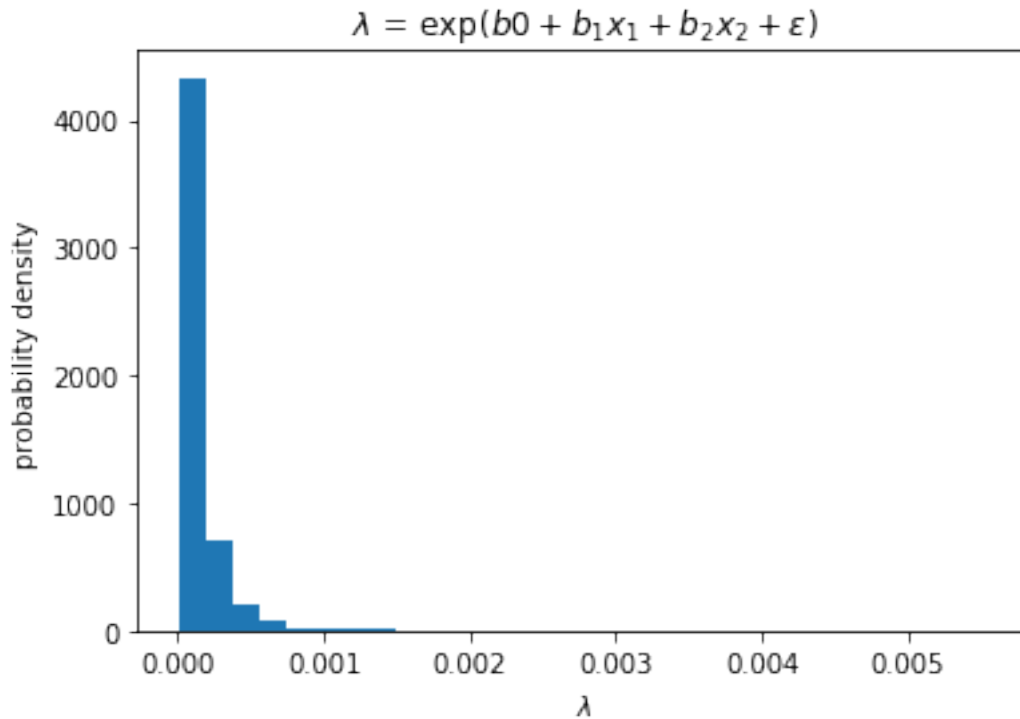
```
[7]: # The distribution of x2
count, bins, ignored = plt.hist(x2, 30, density=True)
plt.title("$x_2$ $\sim$ beta$(2,2)$")
plt.xlabel('$x_2$')
plt.ylabel('probability density')
plt.show()
```



```
[8]: # The distribution of  $\epsilon$ 
count, bins, ignored = plt.hist(epsilon, 30, density=True)
plt.title(" $\epsilon \sim \text{Normal}(0, 2^2)$ ")
plt.xlabel(' $\epsilon$ ')
plt.ylabel('probability density')
plt.show()
```



```
[9]: # The distribution of  $\lambda$ 
count, bins, ignored = plt.hist(lambda_i, 30, density=True)
plt.title(" $\lambda = \exp(b_0 + b_1x_1 + b_2x_2 + \epsilon)$ ")
plt.xlabel(' $\lambda$ ')
plt.ylabel('probability density')
plt.show()
```



1.3 Create and save dataframe

```
[10]: df = pd.DataFrame({
    'y': y,
    'Distance': d,
    'Precipitation': x1,
    'Traffic': x2
})
df.head(10)
```

```
[10]:   y  Distance  Precipitation  Traffic
0  0     1018             0  0.299886
1  0      973             0  0.565617
2  0     1021             0  0.414564
3  0      998             0  0.559767
4  0      985             0  0.777217
5  0      994             0  0.716722
6  0      991             0  0.782984
7  1      967             0  0.587694
8  0     1003             0  0.595906
9  0     1012             0  0.742293
```

```
[11]: df.to_csv("data/simulated_data.csv", sep=',', encoding='utf-8')
```

1.4 Simulate precipitation and traffic variables for links data

```
[12]: links = pd.read_csv('data/links.csv')
links.head(20)
```

```
[12]:
```

	Node A	Node Z	Distance
0	node 1	node 7	800
1	node 1	node 11	800
2	node 1	node 12	2400
3	node 2	node 6	1200
4	node 2	node 10	900
5	node 3	node 6	1100
6	node 3	node 8	800
7	node 3	node 12	600
8	node 4	node 8	700
9	node 4	node 10	700
10	node 4	node 14	2800
11	node 5	node 6	2000
12	node 5	node 7	500
13	node 5	node 11	300
14	node 6	node 13	2000
15	node 7	node 10	500
16	node 9	node 12	1000
17	node 9	node 13	600
18	node 9	node 14	1100
19	node 10	node 11	500

```
[13]: np.random.seed(0)
links['Precipitation'] = simulate_precipitation(links.shape[0])
links['Traffic'] = simulate_traffic(links.shape[0])
links.head(20)
```

```
[13]:
```

	Node A	Node Z	Distance	Precipitation	Traffic
0	node 1	node 7	800	0	0.254345
1	node 1	node 11	800	0	0.243435
2	node 1	node 12	2400	0	0.254188
3	node 2	node 6	1200	0	0.424037
4	node 2	node 10	900	0	0.573477
5	node 3	node 6	1100	0	0.834188
6	node 3	node 8	800	0	0.190679
7	node 3	node 12	600	1	0.689211
8	node 4	node 8	700	1	0.507221
9	node 4	node 10	700	0	0.129693
10	node 4	node 14	2800	0	0.192684
11	node 5	node 6	2000	0	0.334005
12	node 5	node 7	500	0	0.568231
13	node 5	node 11	300	1	0.499119

14	node 6	node 13	2000	0	0.441698
15	node 7	node 10	500	0	0.167268
16	node 9	node 12	1000	0	0.776060
17	node 9	node 13	600	0	0.514337
18	node 9	node 14	1100	0	0.122787
19	node 10	node 11	500	1	0.741051

```
[14]: links.to_csv("data/links_traffic_precipitation.csv", sep=',', encoding='utf-8')
```