

Poisson distribution

Poisson approximation to Binomial

Binomial trials “n” is at least 30  
Probability of success “p” is at most 0.05

Binomial  $E[X] = np$        $\lambda = np$



How many times can we expect to get a 6 if we throw 600 times?

1 throw

600 throws

$\frac{1}{6}$   
?

$(600)\frac{1}{6} = 100$

How many times can are we expected to throw to get the first 6?

1 throw

?

$\frac{1}{6}$   
1

$\frac{1}{1/6} = 6$



```
binom.pmf(k, n, p)
binom.expect(args=(n, p))
binom.cdf(k, n, p)
```

Suppose we toss a coin once every 10 minutes. The probability of heads is 0.2778

What is the probability of getting one heads in 30 minutes?  $n = 3$   $p = 0.2778$

$$P[X = 1] = \text{binom.pmf}(k=1, n=3, p=0.2778) = 0.4346$$

What is the expected number of heads in 30 minutes?

$$E[X] = n * p = 3 * 0.2778 = 0.8334 \quad E[X] = \text{binom.expect}(args=(3, 0.2778)) = 0.8334$$

What is the probability of getting one heads in 90 minutes?  $n = 9$   $p = 0.2778$

$$P[X = 1] = \text{binom.pmf}(k=1, n=9, p=0.2778) = 0.185$$

What is the expected number of heads in 90 minutes?

$$E[X] = n * p = 9 * 0.2778 = 2.5 \quad E[X] = \text{binom.expect}(args=(9, 0.2778)) = 2.5$$



```
binom.pmf(k, n, p)
binom.expect(args=(n, p))
binom.cdf(k, n, p)
```

	Once in 10 mins $p = 0.2778$
30 minutes	$P[X = 1] = 0.4346$ $E[X] = 0.8334$
90 minutes	$P[X = 1] = 0.185$ $E[X] = 2.5$





```
binom.pmf(k, n, p)
binom.expect(args=(n, p))
binom.cdf(k, n, p)
```

Suppose we toss a coin once every minute.

The probability of heads is 0.02778

What is the probability of getting one heads in 30 minutes?  $n = 30$   $p = 0.02778$

$$P[X = 1] = \text{binom.pmf}(k=1, n=30, p=0.02778) = 0.368$$

What is the expected number of heads in 30 minutes?

$$E[X] = n * p = 30 * 0.02778 = 0.8334 \quad E[X] = \text{binom.expect}(args=(30, 0.02778)) = 0.8334$$

What is the probability of getting one heads in 90 minutes?  $n = 90$   $p = 0.02778$

$$P[X = 1] = \text{binom.pmf}(k=1, n=90, p=0.02778) = 0.203$$

What is the expected number of heads in 90 minutes?

$$E[X] = n * p = 90 * 0.02778 = 2.5 \quad E[X] = \text{binom.expect}(args=(90, 0.02778)) = 2.5$$



`binom.pmf(k, n, p)`  
`binom.expect(args=(n, p))`  
`binom.cdf(k, n, p)`

	Once in 10 mins $p = 0.2778$	Once per min $p = 0.02778$
30 minutes	$P[X = 1] = 0.4346$ $E[X] = 0.8334$	$P[X = 1] = 0.368$ $E[X] = 0.8334$
90 minutes	$P[X = 1] = 0.185$ $E[X] = 2.5$	$P[X = 1] = 0.203$ $E[X] = 2.5$





`binom.pmf(k, n, p)`

`binom.expect(args=(n, p))`

`binom.cdf(k, n, p)`

Suppose we toss a coin 10 times every minute. The probability of heads is 0.002778

What is the probability of getting one heads in 30 minutes?  $n = 300$   $p = 0.002778$

$$P[X = 1] = \text{binom.pmf}(k=1, n=300, p=0.002778) = 0.362$$

What is the expected number of heads in 30 minutes?

$$E[X] = n * p = 300 * 0.002778 = 0.8334 \quad E[X] = \text{binom.expect}(args=(300, 0.002778)) = 0.8334$$

What is the probability of getting one heads in 90 minutes?  $n = 900$   $p = 0.002778$

$$P[X = 1] = \text{binom.pmf}(k=1, n=900, p=0.002778) = 0.205$$

What is the expected number of heads in 90 minutes?

$$E[X] = n * p = 900 * 0.002778 = 2.5 \quad E[X] = \text{binom.expect}(args=(90, 0.02778)) = 2.5$$



```
binom.pmf(k, n, p)
binom.expect(args=(n, p))
binom.cdf(k, n, p)
```

	Once in 10 mins $p = 0.2778$	Once per min $p = 0.02778$	Ten times per min $p = 0.002778$
30 minutes	$P[X = 1] = 0.4346$ $E[X] = 0.8334$	$P[X = 1] = 0.368$ $E[X] = 0.8334$	$P[X = 1] = 0.362$ $E[X] = 0.8334$
90 minutes	$P[X = 1] = 0.185$ $E[X] = 2.5$	$P[X = 1] = 0.203$ $E[X] = 2.5$	$P[X = 1] = 0.205$ $E[X] = 2.5$





`binom.pmf(k, n, p)`

`binom.expect(args=(n, p))`

`binom.cdf(k, n, p)`

Suppose the rate at which “heads” comes is 2.5 in 90 mins

What is the probability of getting one heads in 30 minutes?

$$\lambda = \frac{2.5}{3} = 0.8334$$

$$P[X = 1] = \text{poisson.pmf}(k=1, mu=0.8334) = 0.362$$

What is the probability of getting one heads in 90 minutes?

$$\lambda = 2.5$$

$$P[X = 1] = \text{poisson.pmf}(k=1, mu=2.5) = 0.205$$

What is the expected number of heads in 90 minutes?

$$E[X] = n * p = 900 * 0.002778 = 2.5$$

$$E[X] = \text{binom.expect}(args=(90, 0.02778)) = 2.5$$





```
binom.pmf(k, n, p)
binom.expect(args=(n, p))
binom.cdf(k, n, p)
```

	Once in 10 mins $p = 0.2778$	Once per min $p = 0.02778$	Ten times per min $p = 0.002778$	Poisson $\lambda = 2.5$ per 90 min
30 minutes	$P[X = 1] = 0.4346$ $E[X] = 0.8334$	$P[X = 1] = 0.368$ $E[X] = 0.8334$	$P[X = 1] = 0.362$ $E[X] = 0.8334$	$P[X = 1] = 0.362$ $E[X] = 0.8334$
90 minutes	$P[X = 1] = 0.185$ $E[X] = 2.5$	$P[X = 1] = 0.203$ $E[X] = 2.5$	$P[X = 1] = 0.205$ $E[X] = 2.5$	$P[X = 1] = 0.205$ $E[X] = 2.5$

## Poisson distribution

`poisson.pmf(k, mu)`

$$P[X = k] = \frac{\lambda^k e^{-\lambda}}{k!}$$

**Football matches have an average of 2.5 goals in 90 mins**

Q1) What is the probability of having one goal in 30 mins?

90 mins      2.5  
30 mins      ?



$$\lambda = \frac{30 * 2.5}{90} = 0.833$$

$$P[X = 1] = \text{poisson.pmf}(k=1, mu=0.833) = 0.362$$

Q2) What is the probability of having one goal in 90 mins?  $\lambda = 2.5$

$$P[X = 1] = \text{poisson.pmf}(k=1, mu=2.5) = 0.205$$



## Poisson distribution

`poisson.pmf(k, mu)`

$$P[X = k] = \frac{\lambda^k e^{-\lambda}}{k!}$$

**You receive 240 messages per hour on average - assume Poisson distributed.**

Q1) What is the average or expected number of messages in 30 seconds?

$$\begin{array}{ccc} 1 \text{ hour (3600 seconds)} & \times & 240 \text{ messages} \\ 30 \text{ seconds} & & ? \end{array} \quad \frac{30 * 240}{3600} = 2$$

Q2) What is the probability of one message arriving over a 30 second time interval?

If we consider 30 seconds as one unit interval, then  $\lambda = 2$

$$P[X = 1] = \text{poisson.pmf}(k=1, mu=2) = 0.27$$

$$P[X = 1] = \frac{(2)^1 e^{(-2)}}{1!} = 0.27$$

Q3) What is the probability that there are no messages in 15 seconds?

$$P[X = 0] = \text{poisson.pmf}(k=0, mu=1) = 0.367$$

$$P[X = 0] = \frac{(1)^1 e^{(-1)}}{1!} = 0.367$$

$$\lambda = \frac{15 * 240}{3600} = 1$$

Q4) What is the probability that there are 3 messages in 20 seconds?

$$P[X = 3] = \text{poisson.pmf}(k=3, mu=1.33) = 0.104$$

$$P[X = 3] = \frac{(1.33)^3 e^{(-1.33)}}{3!} = 0.104$$

$$\lambda = \frac{20 * 240}{3600} = 1.33$$



## Poisson distribution

`poisson.pmf(k, mu)`

$$P[X = k] = \frac{\lambda^k e^{-\lambda}}{k!}$$

You receive 240 messages per hour on average - assume Poisson distributed.

Q1) What is the average time to wait between two messages?

1 hour (3600 seconds)  $\times$  240 messages = 15 seconds

?

Q2) What is the average number of messages per second?  $\lambda_1 = \frac{1}{15} = 0.067$  per second

Q3) What is the probability of having no messages in 10 seconds?  $\lambda_{10} = \frac{10}{15} = 0.67$  per 10-seconds

$$P[X = 0] = \frac{\lambda_{10}^0 e^{-\lambda_{10}}}{0!} = e^{-\lambda_{10}} = e^{-10\lambda_1} = 0.5134$$

$$\lambda_{10} = 10 * \lambda_1$$

$$P[X = 0] = \text{poisson.pmf}(k=0, mu=10/15)$$

$$\lambda_t = t * \lambda_1$$

Q4) What is the probability of waiting for more than 10 seconds for the next message?

Let  $T$  denote the time to wait for the next message

$$P[T > 10] = e^{-10\lambda_1} = 0.5134$$

Q5) What is the probability of waiting less than or equal to 10 seconds?

$$P[T \leq 10] = 1 - e^{-10\lambda_1} = 0.4865$$

`from scipy.stats import expon`

$$P[T \leq x] = 1 - e^{-x\lambda}$$

$$P[T \leq 10] = \text{expon.cdf}(x=10, scale=15) = 0.4865$$

You are working as a data engineer who has to resolve any bugs/  
failures of machine learning models in production

$$P[T \leq x] = 1 - e^{-x\lambda}$$

expon.cdf(x, scale)

The time taken to debug is exponentially distributed with mean of 5 minutes

Q1) Find the probability of debugging in 4 to 5 minutes

$$P[4 < T < 5] = \text{expon.cdf}(x=5, \text{scale}=5) - \text{expon.cdf}(x=4, \text{scale}=5) = 0.0814$$

Q2) Find the probability of needing more than 6 minutes to debug

$$P[T > 6] = 1 - \text{expon.cdf}(x=6, \text{scale}=5) = 0.3012$$

Q3) Given that you have already spent 3 minutes, what is the probability of needing more than 9 minutes

$$P[T > 9 | T > 3] = \frac{P[(T > 9) \cap (T > 3)]}{P[T > 3]} = \frac{P[T > 9]}{P[T > 3]} = \frac{1 - \text{expon.cdf}(x=9, \text{scale}=5)}{1 - \text{expon.cdf}(x=3, \text{scale}=5)} = 0.3012$$

$$\frac{P[T > 9]}{P[T > 3]} = \frac{e^{-9\lambda}}{e^{-3\lambda}} = e^{-6\lambda} = P[T > 6]$$

Memoryless:

The fact that you took three minutes so far does not affect how much more you might take to debug



**Waiting for first goal**

**Rate = 2.5 / 90 mins**

**Rate = 1.25 / 45 mins**

**Rate = 0.0278 / min**

Let  $T_1$  denote the amount of time to wait for first goal

Is  $T_1$  discrete or continuous?



$T_1$  Time taken for first goal from the start of the match

$T_2$  Time taken for second goal from point the first goal is scored

Under the assumptions of the poisson process,  $T_1$  and  $T_2$  are independent

These are called “Inter arrival times”, and are exponentially distributed



A call centre gets 3.5 calls per hour

$$P[T \leq x] = 1 - e^{-x\lambda}$$

expon.cdf(x, scale)

Q) Calculate the probability that the next call will arrive at least 30 minutes after the previous call.

Approach 1: Interval = 1 minute

$$\begin{array}{ccc} 1 \text{ hour (60 min)} & \times & 3.5 \\ 1 \text{ minute} & & ? \end{array} \quad \frac{3.5}{60} = 0.0583$$

$$P[T > 30] = 1 - \text{expon.cdf}(x=30, \text{scale}=1/0.0583) = 0.1739$$

Approach 2: Interval = 30 minute

$$\begin{array}{ccc} 1 \text{ hour (60 min)} & \times & 3.5 \\ 30 \text{ minute} & & ? \end{array} \quad \frac{3.5}{2} = 1.75$$

$$P[T > 1] = 1 - \text{expon.cdf}(x=1, \text{scale}=1/1.75) = 0.1739$$

# Log-Normal distribution

Consider an example of number of days of hospitalisation

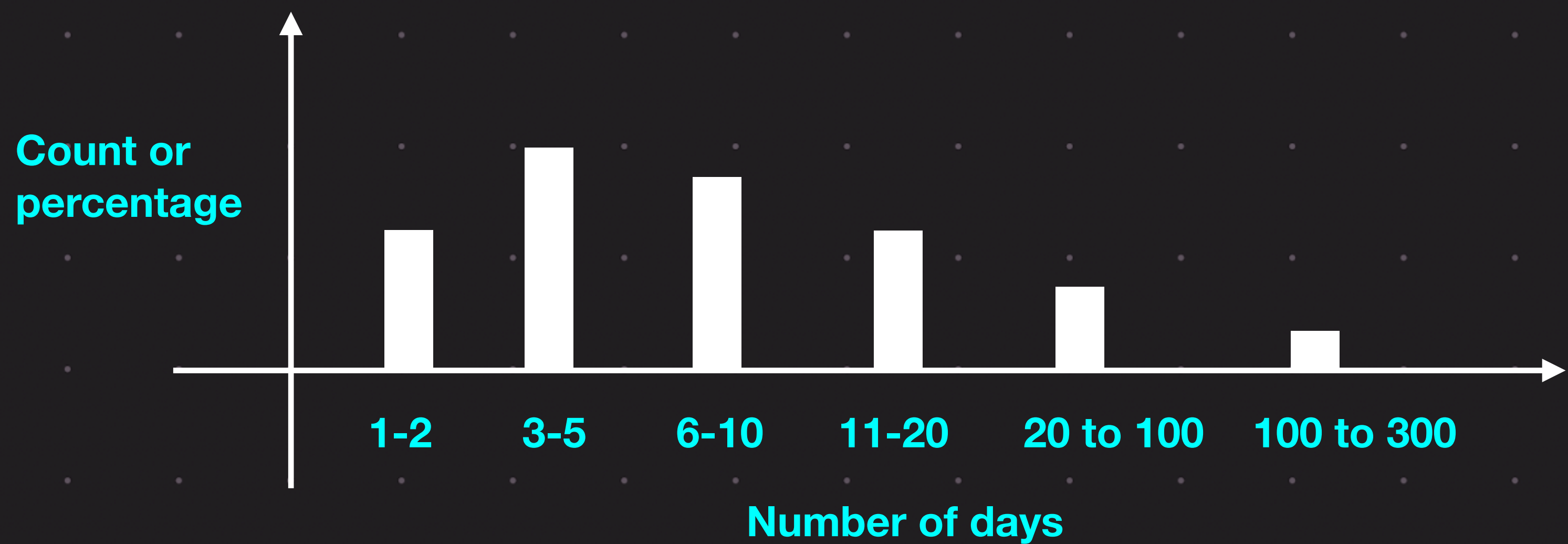
Most people may be hospitalised for 3 to 5 days

Quite often, just 1 or 2 days

But, there can be a few extreme cases in 20 to 30 days

Extremely rare cases of 300 days (coma etc.)

# Log-Normal distribution



Number of days is discrete

What if we had exact number of hours?

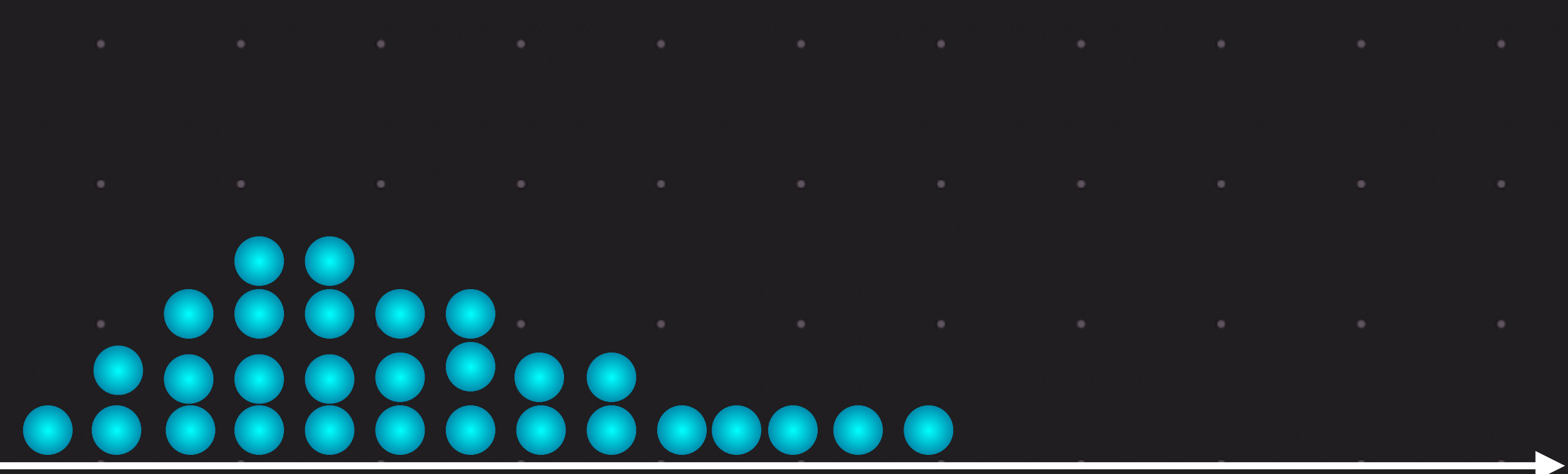
This would be continuous



# Log-Normal distribution

Count or  
percentage

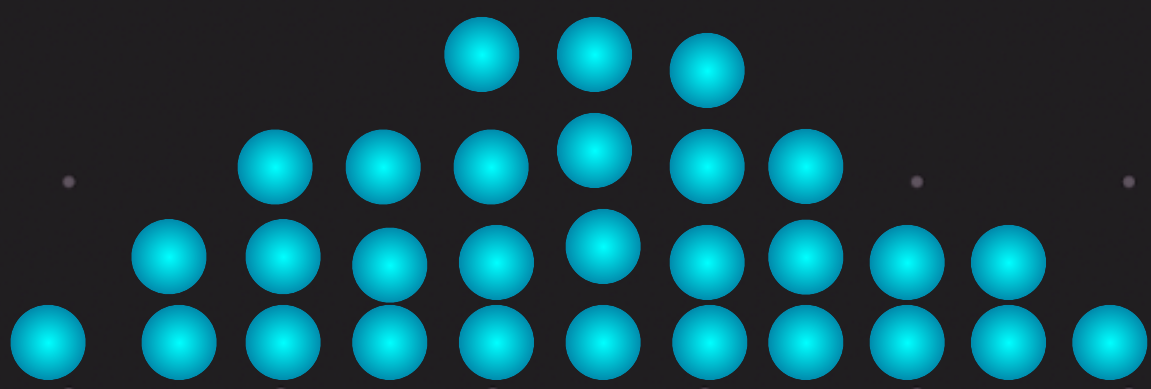
Hours



# Log-Normal distribution

Count or  
percentage

Logarithm of hours



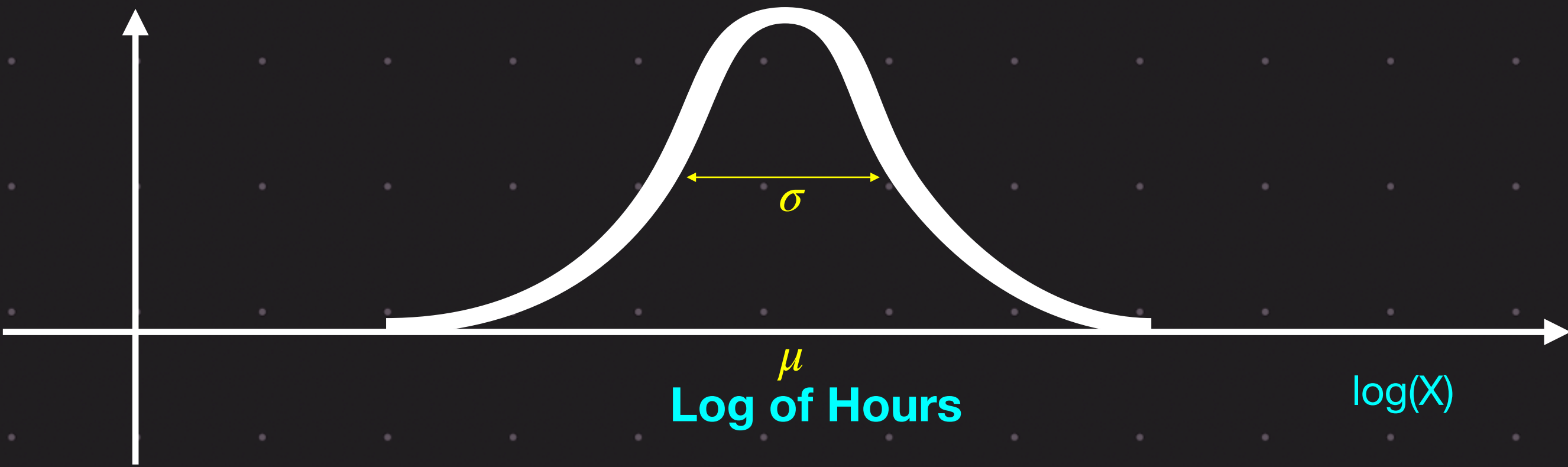
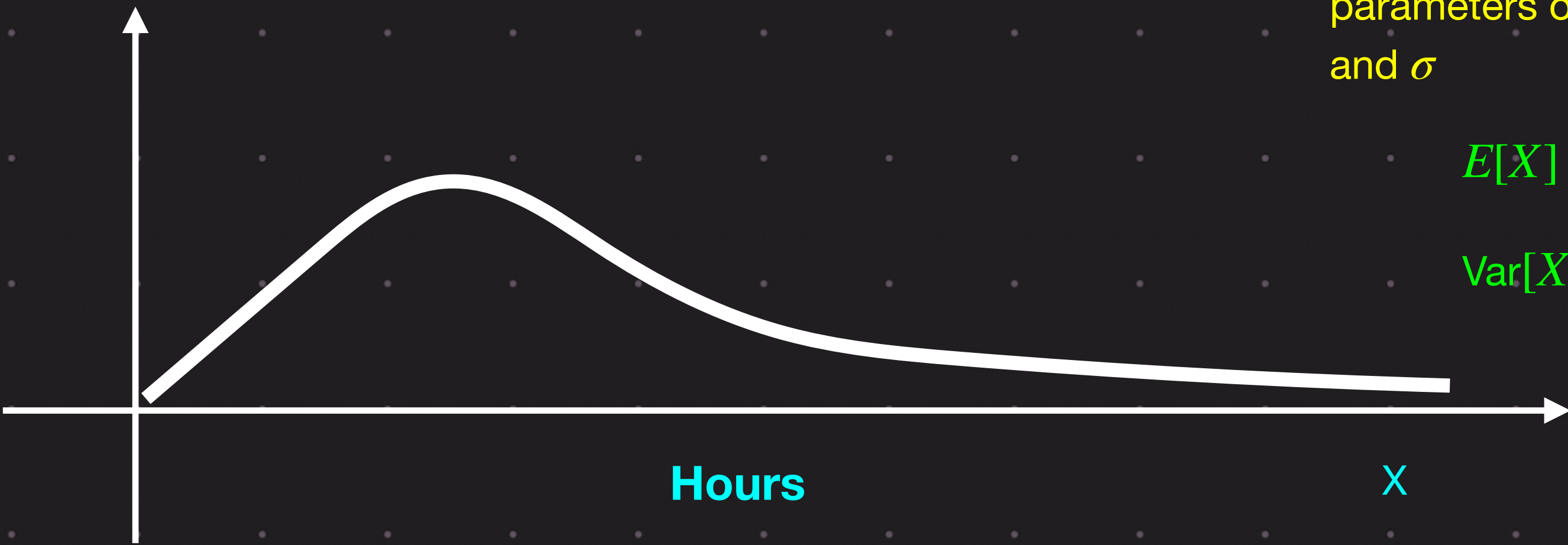
# Log-Normal distribution

For the log normal, we use the parameters of the Gaussian:  $\mu$  and  $\sigma$

$$E[X] = e^{\mu + \sigma^2/2}$$

$$\text{Var}[X] = (e^{\sigma^2} - 1) \left( e^{2\mu + \sigma^2} \right)$$

Always confirm these equations from some source (eg: wiki)



log(X)