

# CS146 #3

[Code ▼](#)

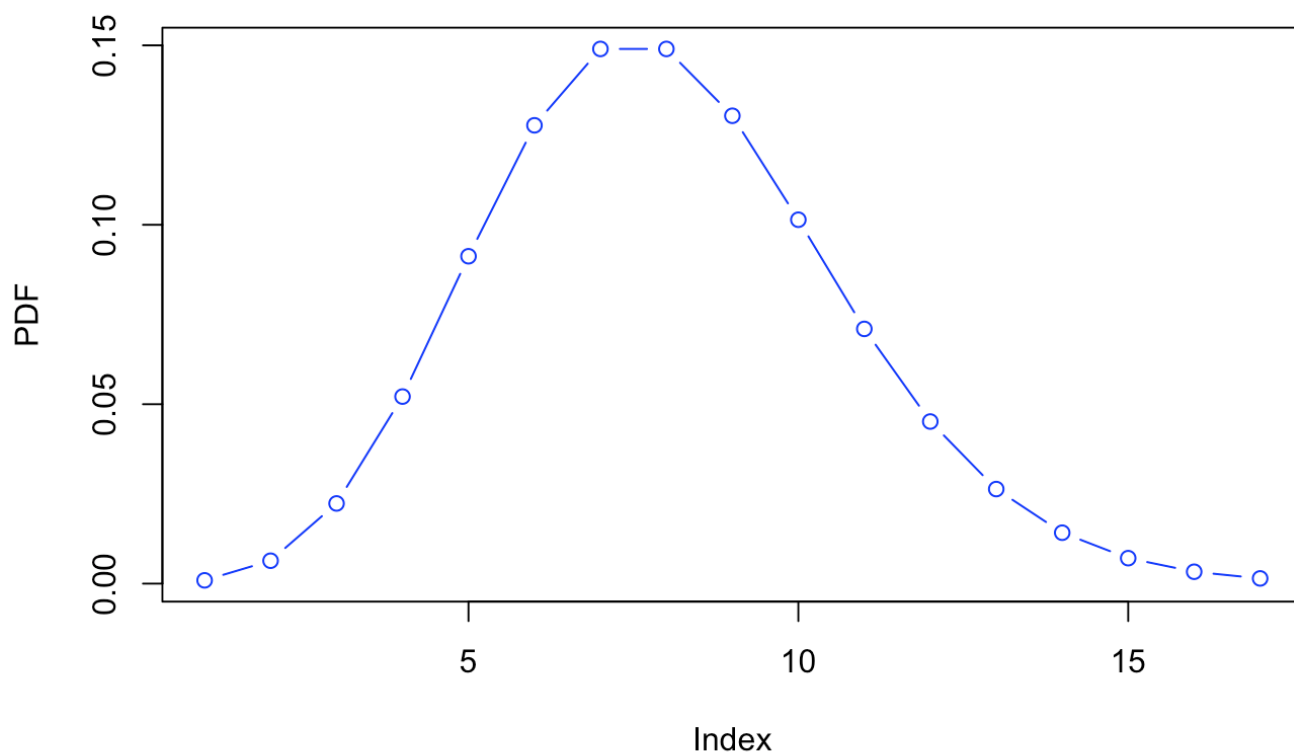
## Create your own probability distribution.

I chose to go with the poisson distribution. the support is all natural numbers  $\mathbb{N}$ , and 0 for  $K$ . i.e.

$$k \in \mathbb{N} \cup 0$$

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```
Lambda <- 7
K = seq(0,16,1)
PDF<-dpois(K,lambda = Lambda)
plot(PDF, col="blue", type="b")
```



**Will your pdf have one mode or multiple modes?**

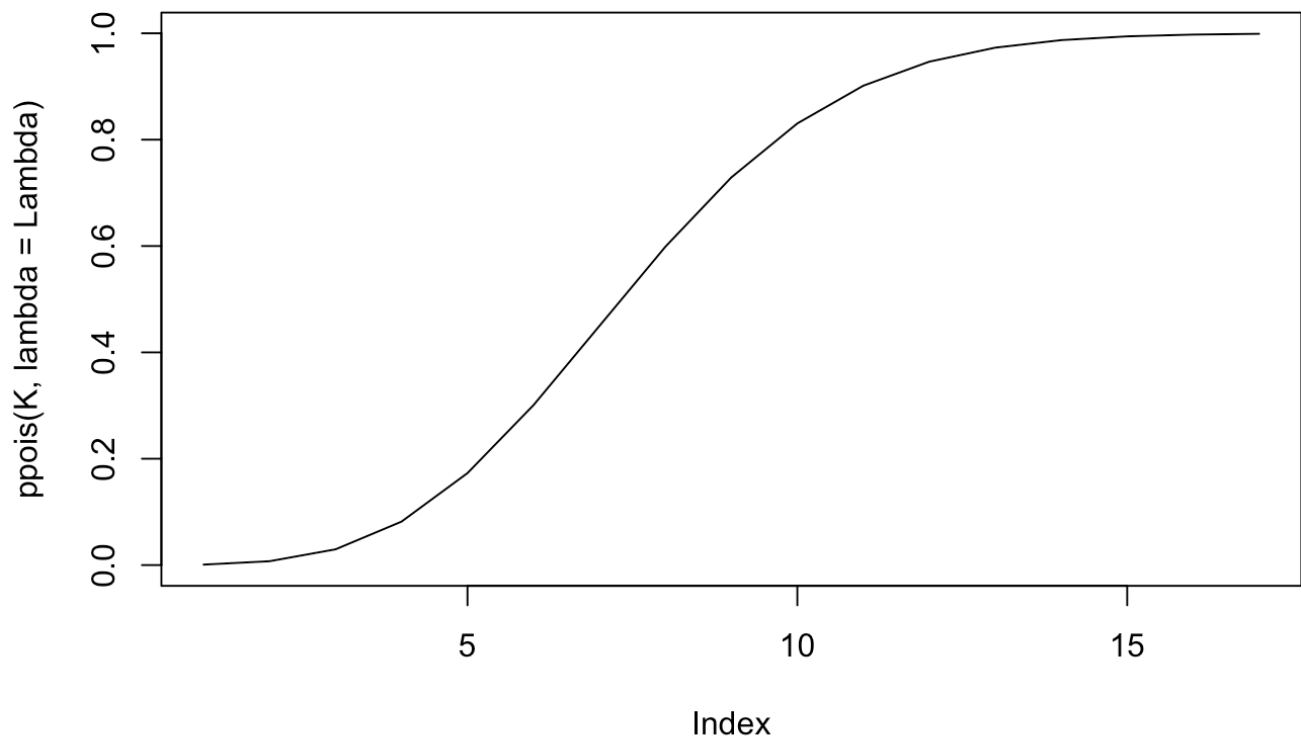
One, over Lambda.

**Will it be skew or symmetric?** It is skewed because of the vector I decided to draw (1:20) which is not symmetric in relation to the mean (or Lambda).

**Where will most of the probability mass be?**

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```
# Perhaps the interval with most of the mass is between ~6 and ~10.
# We can also compute sort of "threshold" for most of the mass starting from 0 (i.e.
  CDF):
plot(ppois(K, lambda = Lambda), type = "l")
```



What are the parameters of your distribution?

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```
# The parameter is Lambda, which describes some known mean, or average rate in a give
n volume/time.
Lambda
```

(Optional) Stretch goal: write R functions for the cdf, pdf, qf and random samples from your distribution.

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```
LLambda = 10
KK=10
#the CDF function, instead of using ppois:
sum(dpois(0:KK,LLambda))
```

```
[1] 0.5830398
```

**3.2 Derive the posterior distribution, including its parameter values, from an exponential likelihood function with a gamma prior over the exponential parameter  $\lambda$ , and with data  $y_i$  for  $i = 1, 2, \dots, n$ .**

Given: posterior is proportional to exponential\_pdf\_likelihood \* gamma\_pdf\_prior.

exponential\_pdf\_likelihood =

$$\lambda e^{-\lambda y(i \dots n)}$$

gamma\_pdf\_prior. =

$$\frac{\beta^\alpha}{\Gamma(\alpha)} y^{\alpha-1} e^{-\beta y}$$

Therefore the posterior for  $\lambda$  with data  $y_i$  for  $i = 1, 2, \dots, n$ . is:

$$\lambda e^{-\lambda(y_1 + \dots + y_n)} \times \frac{\beta^\alpha}{\Gamma(\alpha)} \lambda^{\alpha-1} e^{-\beta\lambda}$$

Which is proportional to:

$$\lambda^n e^{-\lambda y(i \dots n)} \times (\lambda^{\alpha-1} e^{-\beta\lambda})$$

Further simplification:

$$\lambda^{n+\alpha-1} e^{-\lambda y(i \dots n) - \beta\lambda}$$

$$\lambda^{n+\alpha-1} e^{-\lambda(y(i \dots n) + \beta)}$$

which is the Gamma, with parameters:

$$\alpha + n$$

$$\sum_{i=1}^n y_i + \beta$$