## Step 1: let's define our prior as a Gamma distribution

Please note that we have two way to define a Gamma distribution

$$f(\lambda|\alpha,\beta) = \frac{\lambda^{\alpha-1}e^{-\beta\lambda}\beta^{\alpha}}{\Gamma(\alpha)} \text{ for } x > 0 \text{ and } \alpha > 0, \beta > 0$$
$$f(\lambda|k,\theta) = \frac{\lambda^{k-1}e^{-\lambda/\theta}}{\theta^{k}\Gamma(k)}, \text{ for } x > 0 \text{ and } k, \theta > 0$$

Note that

In[275]:=

$$k = \alpha$$
 and  $\theta = \beta$ 

Mathematica automatically assumes the  $(k, \theta)$  methods.

However, in Bayesian statistics, we prefer the  $(\alpha, \beta)$  way.

Hence, the parameters will be

$$k = \alpha = 7$$
  
$$\theta = 1/\beta = 1/5$$

based on the parameter values, we define the prior as a function of  $\boldsymbol{\lambda}$ 

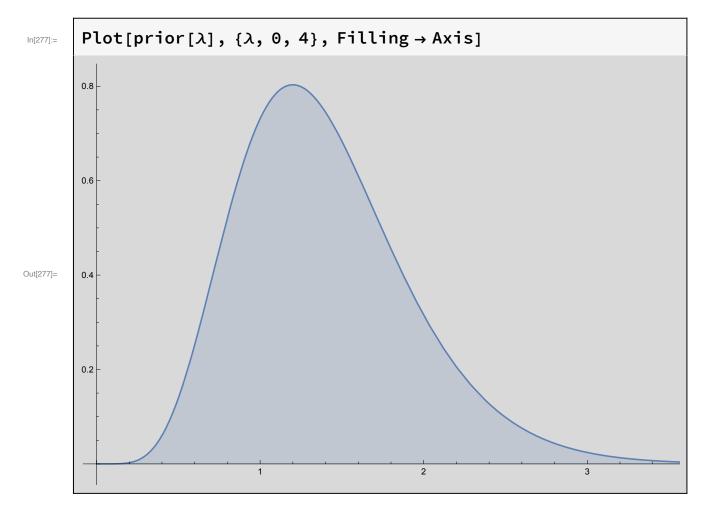
prior[ $\lambda$ ] := PDF[GammaDistribution[7, 1/5],  $\lambda$ ]

let's print its analytic form to see whether it works

In[276]:= **prior**[λ]

$$\begin{cases} \frac{15625}{144} e^{-5\lambda} \lambda^{6} & \lambda > 0 \\ 0 & \text{True} \end{cases}$$

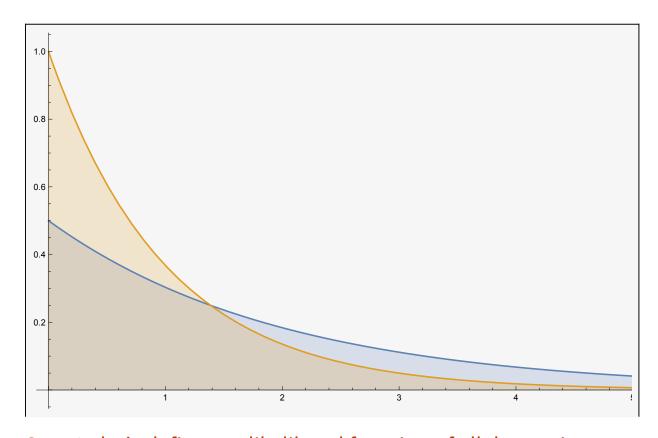
let's plot it as well



Step 2: let's define our likelihood function of each data point as an Exponential Distribution

```
lf[x] := PDF[ExponentialDistribution[\lambda], x]
        lf[x]
          e^{-x \lambda} \lambda x \ge 0
Out[279]=
                    True
```

```
Plot[Table[lf[x], \{\lambda, \{0.5, 1\}\}] // Evaluate,
 \{x, 0, 5\}, Filling \rightarrow Axis
```



Step 3: let's define our likelihood function of all data points as the product of Exponential PDFs

```
likelihood[f_, data_] := Product[PDF[f, x], {x, data}]
 In[281]:=
                     likelihood[ExponentialDistribution[\lambda], {a<sub>1</sub>, a<sub>2</sub>, a<sub>3</sub>, a<sub>4</sub>, a<sub>5</sub>}]
 In[282]:=
Out[282]=
                               \begin{array}{lll} e^{-\lambda \; a_3} \; \lambda & a_3 \, \geq \, 0 \\ 0 & \quad True \end{array} \right) \; \left( \left\{ \begin{array}{lll} e^{-\lambda \; a_4} \; \lambda & a_4 \, \geq \, 0 \\ 0 & \quad True \end{array} \right) \; \left( \left\{ \begin{array}{lll} e^{-\lambda \; a_5} \; \lambda & a_5 \, \geq \, 0 \\ 0 & \quad True \end{array} \right) \right. \right.
```

```
l = Simplify[likelihood[
        In[283]:=
                                                                                                                                                                                                          ExponentialDistribution[\lambda], {a<sub>1</sub>, a<sub>2</sub>, a<sub>3</sub>, a<sub>4</sub>, a<sub>5</sub>}]]
                                                                                                                                                                      e^{-\lambda \ (a_1 + a_2 + a_3 + a_4 + a_5)} \ \lambda^5 \quad a_1 \, \geq \, 0 \, \&\& \, a_2 \, \geq \, 0 \, \&\& \, a_3 \, \geq \, 0 \, \&\& \, a_4 \, \geq \, 0 \, \&\& \, a_5 \, \geq \, 0 \, \&\& \, a_5 \, \geq \, 0 \, \&\& \, a_6 \, \geq \, 0 \, \&\& \, a_8 \, a_8 \, \geq \, 0 \, \&\& \, a_8 \, a_8 \, \geq \, 0 \, \&\& \, a_8 \, a_8 \, \geq \, 0 \, \&\& \, a_8 \,
Out[283]=
                                                                                                                                                                     0
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                  True
```

In[285]:=

likelihood[ExponentialDistribution[ $\lambda$ ], {1, 2, 3, 4, 5}]

Out[285]=

 $e^{-15\lambda}\lambda^5$ 

## Step 4: let's calculate our posterior using Bayes Rule

The Bayes Rule suggests that

$$\pi(\lambda|x) = \frac{L(x|\lambda)p(\lambda)}{\int L(x|\lambda)p(\lambda) d\lambda}$$

Since the denominator is just a constant, we just need to focus on the numerator calculation as follows

num = Simplify[l \* prior[λ]] In[286]:=  $\frac{15\,625}{144} \ \mathbb{e}^{-\lambda \ (5+a_1+a_2+a_3+a_4+a_5)} \ \lambda^{11} \quad a_1 \geq 0 \ \&\& \ a_2 \geq 0 \ \&\& \ a_3 \geq 0 \ \&\& \$  $a_4 \ge 0 \&\& a_5 \ge 0 \&\& \lambda > 0$ True

It looks like a Gamma distribution, which is

$$f(\lambda|\alpha,\beta) = \frac{\lambda^{\alpha-1}e^{-\beta\lambda}\beta^{\alpha}}{\Gamma(\alpha)} \text{ for } x > 0 \text{ and } \alpha > 0, \beta > 0$$

Let's vary our sample points (add one extra sample points) to find the pattern of Gamma paratmeter.

l = Simplify[likelihood[ In[287]:= ExponentialDistribution[ $\lambda$ ], {a<sub>1</sub>, a<sub>2</sub>, a<sub>3</sub>, a<sub>4</sub>, a<sub>5</sub>, a<sub>6</sub>}]]; num = Simplify[l \* prior[λ]]  $\frac{15\,625}{144} \ \mathbb{e}^{-\lambda \ (5+a_1+a_2+a_3+a_4+a_5+a_6)} \ \lambda^{12} \quad \lambda > 0 \ \& \ a_1 \geq 0 \ \& \ a_2 \geq 0 \ \& \& \ a_2 \geq 0 \ \& \& \ a_3 \geq 0 \ \& \ a_4 \geq 0 \ \& \ a_3 \geq 0 \ \& \& \ a_4 \geq 0 \ \& \& \ a_5 \geq 0 \ \& \& \ a_$ Out[288]=  $a_3 \ge 0 \&\& a_4 \ge 0 \&\& a_5 \ge 0 \&\& a_6 \ge 0$ True

Still a Gamma distribution

Again, let's vary our sample points (add one extra sample points) to find the pattern of Gamma paratmeter.

In[289]:=

Out[290]=

Now we can find the pattern of Gamma paratmeter:

$$\beta = 5 + \sum_{i} x_i$$

 $\alpha = 7 + n$  where n is sample size.