

## I. Poisson and Gamma distributions

1. Plot the PDF of the Poisson distribution for  $\lambda = 10.5$ . What is its mode?
2. Explore the Gamma( $s, r$ ) distribution using the Desmos applet on our course page. Note that they call the parameters  $s = \alpha$  and  $r = \beta_{\text{rate}}$ . For a fixed rate  $r = 1$ , what happens as the shape passes from  $s > 1$  through  $s = 1$  to  $s < 1$ ?
3. The Gamma distribution has mean  $\frac{s}{r}$ , mode  $\frac{s-1}{r}$ , and variance  $\frac{s}{r^2}$ .
  - (a) Find Gamma( $s, r$ ) with mean 5 and sd 1/2. Check by simulating with `rgamma`.
  - (b) Find Gamma( $s, r$ ) with mode 6 and mean 8.

## II. Posterior simulation

Consider a Poisson variable  $Y \sim \text{Pois}(\lambda)$  with an unknown rate  $\lambda$  of success. Model our prior understanding with  $\lambda \sim \text{Gamma}(1, 1)$ . We make two observations and find  $Y_1 = 5, Y_2 = 3$ .

1. What is the theoretical posterior distribution of  $\lambda$ ?
2. Simulate 1,000,000 values of  $\lambda$ .
3. For each value of  $\lambda$ , use `rpois` to choose a  $Y_1$  and a  $Y_2$  value.
4. Select just the  $\lambda$  values corresponding to  $Y_1 = 5, Y_2 = 3$ . This is a sample from the posterior distribution of  $f(\lambda | Y_1 = 5, Y_2 = 3)$ . How many of the 1,000,000 values did you get in your sample?
5. Make a density plot of these values of  $\lambda$  and then add the theoretical posterior distribution to the plot.

## III. Soccer goals

Pick your favorite soccer team, and think about their rate  $\lambda$  of goals scored per game.

1. Tune a Gamma prior to your group's knowledge of  $\lambda$ . What are the mean, mode, and sd for your prior?
2. Make a list of how many goals your team scored in each of their last ten games.
3. Find the posterior Gamma distribution for  $\lambda$ . What are the mean, mode, and sd?
4. Plot the prior and posterior distributions on the same graph. Use `bayesrules::plot_gamma_poisson()` if you have it.

## IV. Maximum likelihood

Consider the Gamma-Poisson model with data  $y_1, \dots, y_n$ . The likelihood function is given (up to a constant) by

$$L(\lambda) \propto \lambda^{\sum y_i} e^{-n\lambda}$$

Show that the maximum of the likelihood function occurs when  $\lambda = \bar{y} = \frac{1}{n} \sum_{i=1}^n y_i$