

# Lesson 6

**6/6 points (100%)**

Quiz, 6 questions

 **Congratulations! You passed!**[Next Item](#)1 / 1  
points

1.

**For Questions 1-2, consider the following experiment:**

Suppose you are trying to calibrate a thermometer by testing the temperature it reads when water begins to boil. Because of natural variation, you take several measurements (experiments) to estimate  $\theta$ , the mean temperature reading for this thermometer at the boiling point.

You know that at sea level, water should boil at 100 degrees Celsius, so you use a precise prior with  $P(\theta = 100) = 1$ . You then observe the following five measurements: 94.6 95.4 96.2 94.9 95.9.

- What will the posterior for  $\theta$  look like?
  - ☐ Most posterior probability will be concentrated near the sample mean of 95.4 degrees Celsius.
  - ☐ Most posterior probability will be spread between the sample mean of 95.4 degrees Celsius and the prior mean of 100 degrees Celsius.
  - ☒ The posterior will be  $\theta = 100$  with probability 1, regardless of the data.

 **Correct**

Because all prior probability is on a single point (100 degrees Celsius), the prior completely dominates any data. If we are 100% certain of the outcome before the experiment, we learn nothing by performing it.

Clearly this was a poor choice of prior, especially in light of the data we collected.

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None of the above.

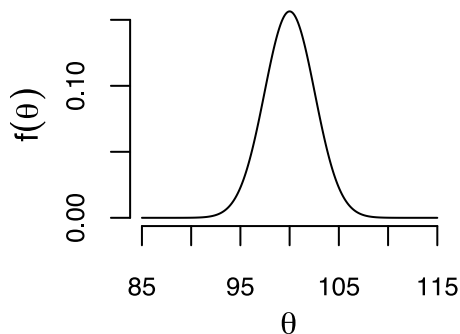
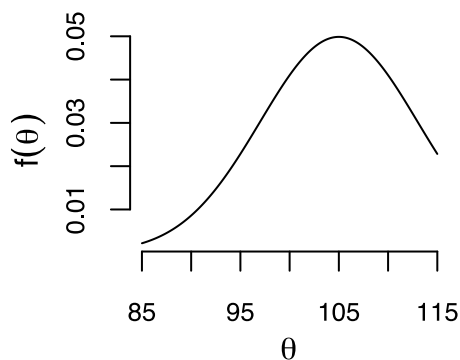
1 / 1  
points

2.

Thermometer:

Suppose you believe before the experiments that the thermometer is biased high, so that on average it would read 105 degrees Celsius, and you are 95% confident that the average would be between 100 and 110.

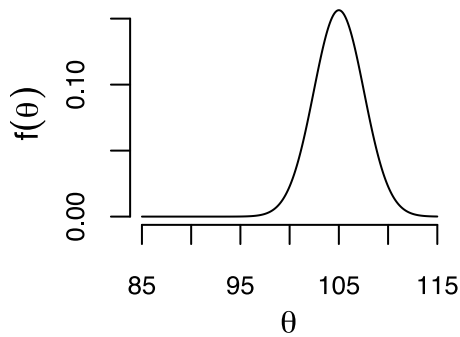
- Which of the following prior PDFs most accurately reflects this prior belief?



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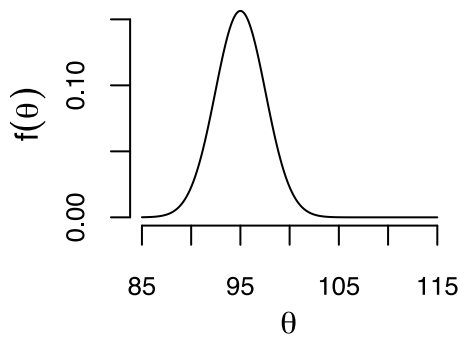
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**Correct**

The prior mean is 105 degrees Celsius and approximately 95% of the prior probability is assigned to the interval (100, 110).



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3.

Recall that for positive integer  $n$ , the gamma function has the following property:  $\Gamma(n) = (n - 1)!$ .

What is the value of  $\Gamma(6)$ ?

120



**Correct Response**

This is  $\Gamma(6) = 5! = 120$ .

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4.

Find the value of the normalizing constant,  $c$ , which will cause the following integral to evaluate to 1.

$$\int_0^1 c \cdot z^3(1-z)^1 dz.$$

Hint: Notice that this is proportional to a beta density. We only need to find the values of the parameters  $\alpha$  and  $\beta$  and plug those into the usual normalizing constant for a beta density.



$$\frac{\Gamma(1)}{\Gamma(z)\Gamma(1-z)} = \frac{0!}{(z-1)!1!}$$



$$\frac{\Gamma(4+2)}{\Gamma(4)\Gamma(2)} = \frac{5!}{3!1!} = 20$$



**Correct**

$\alpha = 3 + 1$  and  $\beta = 1 + 1$ .



$$\frac{\Gamma(3+1)}{\Gamma(3)\Gamma(1)} = \frac{3!}{2!0!} = 3$$



1 / 1  
points

5.

Consider the coin-flipping example from Lesson 5. The likelihood for each coin flip was Bernoulli with probability of heads  $\theta$ , or  $f(y | \theta) = \theta^y(1 - \theta)^{1-y}$  for  $y = 0$  or  $y = 1$ , and we used a uniform prior on  $\theta$ .

Recall that if we had observed  $Y_1 = 0$  instead of  $Y_1 = 1$ , the posterior distribution for  $\theta$  would have been  $f(\theta | Y_1 = 0) = 2(1 - \theta)I_{\{0 \leq \theta \leq 1\}}$ . Which of the following is the correct expression for the posterior predictive distribution for the next flip  $Y_2 | Y_1 = 0$ ?



$$f(y_2 | Y_1 = 0) = \int_0^1 2(1 - \theta)d\theta \text{ for } y_2 = 0 \text{ or } y_2 = 1.$$



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☒  $f(y_2 | Y_1 = 0) = \int_0^1 \theta^{y_2} (1 - \theta)^{1-y_2} 2(1 - \theta) d\theta$  for  $y_2 = 0$  or  $y_2 = 1$ .

**Correct**

This is just the integral over likelihood  $\times$  posterior. This expression simplifies to

$$\begin{aligned} \int_0^1 2\theta^{y_2} (1 - \theta)^{2-y_2} d\theta I_{\{y_2 \in \{0,1\}\}} &= \frac{2}{\Gamma(4)} \Gamma(y_2 + 1) \Gamma(3 - y_2) I_{\{y_2 \in \{0,1\}\}} \\ &= \frac{2}{3} I_{\{y_2=0\}} + \frac{1}{3} I_{\{y_2=1\}} \end{aligned}$$

☐  $f(y_2 | Y_1 = 0) = \int_0^1 2\theta^{y_2} (1 - \theta)^{1-y_2} d\theta$  for  $y_2 = 0$  or  $y_2 = 1$ .



1 / 1  
points

6.

The prior predictive distribution for  $X$  when  $\theta$  is continuous is given by  $\int f(x | \theta) \cdot f(\theta) d\theta$ . The analogous expression when  $\theta$  is discrete is  $\sum_{\theta} f(x | \theta) \cdot f(\theta)$ , adding over all possible values of  $\theta$ .

Let's return to the example of your brother's loaded coin from Lesson 5. Recall that he has a fair coin where heads comes up on average 50% of the time ( $p=0.5$ ) and a loaded coin ( $p=0.7$ ). If we flip the coin five times, the likelihood is binomial:  $f(x | p) = \binom{5}{x} p^x (1-p)^{5-x}$  where  $X$  counts the number of heads.

Suppose you are confident, but not sure that he has brought you the loaded coin, so that your prior is  $f(p) = 0.9 I_{\{p=0.7\}} + 0.1 I_{\{p=0.5\}}$ . Which of the following expressions gives the prior predictive distribution of  $X$ ?

☒  $f(x) = \binom{5}{x} .7^x (.3)^{5-x} (.9) + \binom{5}{x} .5^x (.5)^{5-x} (.1)$

**Correct**

This is a weighted average of binomials, with weights being your prior probabilities for each scenario (loaded or fair).

☐  $f(x) = \binom{5}{x} .7^x (.3)^{5-x} (.1) + \binom{5}{x} .5^x (.5)^{5-x} (.9)$

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$$f(x) = \{5 \text{ \choose } x\} .7^x (.3)^{5-x} (.5) + \{5 \text{ \choose } x\} .5^x (.5)^{5-x} (.5)$$

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$$f(x) = \{5 \text{ \choose } x\} .7^x (.3)^{5-x} + \{5 \text{ \choose } x\} .5^x (.5)^{5-x}$$

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