## General probability questions

- 1. Exercises 2E1 in the book.
- 2. Exercises 2E2 in the book.
- 3. Exercises 2E3 in the book.

### Bayes theorem

1. In this exercise we will use Bayes theorem for spell correction. Let  $\theta$  be the word one wanted to spell, and let Y be the word that is actually spelled, which in this case is 'radom'. Suppose we know that it should be one of the three words  $\{'radom', 'radom', radon'\}$ . The three words are not equally like a priori (unnormalized prior taken from frequency of the word occurring):

$\theta$	$P(\theta)$
random'	$7 \times 10^{-5}$
'radon'	$7 \times 10^{-6}$
'radom'	$3 \times 10^{-7}$

The conditional probability of  $\theta$  given one has spelled 'radom' (the likelihood) is, by Google spell checker, given by (again not normalized):

$\theta$	$P(Y = 'radom' \theta)$
random'	0.00193
'radon'	0.000143
'radom'	0.975

What is the posterior distribution of  $\theta$  given that Y = radom'?

# Proportionality

Recall that two functions  $f_1(x)$  and  $f_2(x)$  are proportional,  $\infty$ , to each other if there exists a positive constant c > 0 such that  $f_1(x) = cf_2(x)$ .

- 1. Suppose  $f(x) = e^{-x^2}$  which of the following functions are f(x) proportional to
  - a)  $f_1(x) = e^{-\frac{x^2}{2}}$ .
  - b)  $f_2(x) = e^{-(x-\mu)^2}$
  - c)  $f_3(x) = e^{-x^2 \mu^2}$ .
- 2. Suppose  $f(p) = \frac{1}{B(\alpha,\beta)}p^{\alpha-1}(1-p)^{\beta-1}$  which of the following functions are f(p) proportional to
  - a)  $f_1(p) = p^{\alpha 1} (1 p)^{\beta 1}$ .
  - b)  $f_2(p) = p^{\alpha+2-1}(1-p)^{\beta-1}$ .
  - c)  $f_3(p) = p^{\alpha 1} (1 p)^{\beta + 3 1}$ .

# Independence

Independence recall that the definition of independence for probabilities is given by

$$P(A \cap B) = P(A)P(B).$$

The same holds for densities, i.e.

$$p_{X,Y}(x,y) = p_X(x)p_Y(y).$$

1. Suppose that the joint density of  $X_1$  and  $X_2$  given p is proportional to:

$$p(x_1, x_2|p) \propto p^{x_1+x_2}(1-p)^{2-x_1-x_2}$$

Show that given p that  $X_1$  and  $X_2$  two are independent.

2. Suppose we have a uniform prior on p (i.e. p(p) = 1), and suppose we observe n independent observations of a Bernoulli random variable (i.e.  $p(X|p) = p^X(1-p)^{1-X}$ ). Show that (using Bayes theorem and independence)

$$p(p|X_1, X_2, \dots, X_n) \propto p^{\sum_{i=1}^n X_i} (1-p)^{n-\sum_{i=1}^n X_i}$$

#### **Distributions**

A key concept in Bayesian modeling is distribution.

1. The most famous distribution is the Normal distribution:

$$p(x|\mu,\sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{\frac{(x-\mu)^2}{2\sigma^2}}.$$

- a) Is the distribution of x continuous or discrete?
- b) Suppose we want to put a prior on  $\mu$ , if we have no special information should the prior be continuous or discrete?
- c) Suppose we want to put a prior on  $\sigma$ , if we have no special information should the prior be continuous or discrete? Should there be any other restriction on the domain of  $\sigma$  (which values  $\sigma$  can take)?
- d) Write the integral form of  $P(X > 2 | \mu = 2, \sigma = 4)$ .
- 2. An other classical distribution is the Poisson distribution:

$$p(x|\lambda) = \frac{\lambda^x e^{-\lambda}}{r!} x = 0, 1, 2 \dots$$

- a) Is the distribution of x continuous or discrete?
- b) Suppose we want to put a prior on  $\lambda$ , if we have no special information should the prior be continuous or discrete?
- c) Write the sum form of  $P(X > 2|\lambda = 2)$

d) Suppose we observe two independent observation  $(X_1, X_2)$  observations of  $p(x|\lambda)$ . Further assume a an exponential prior on  $\lambda$   $(p(\lambda) \propto e^{-\beta\lambda})$ . Use Bayes theorem and independence to derive the posterior distribution of  $\lambda$   $(p(\lambda|X_1, X_2))$ 

Hint: For the normalizing constant note that the density of the Gamma distribution is given by

$$f(x|a,b) = \frac{b^a}{\Gamma(a)} x^{a-1} e^{-bx}$$

### Bayesian models

- 4E1
- 4E2
- 4E3
- 4E4
- 4E5
- In the model definition below, which line is the prior:

$$y \sim Poisson(\lambda)$$
  
 $\lambda \sim exp(\alpha)$ 

- 4M2
- 4M3
- Translate the mao model below into a mathematical definition.

```
flist <- alist (
    height ~ dpois(mu),
    mu = exp(a + b * x),
    a ~ dnorm(0,20),
    b ~ dnorm(0,20)
)
```