Chapter 2

Hard question 1

Suppose there are two species of Species Bear. Both are equally common in the wild and live in the same places. They look exactly alike and eat the same food, and there is yet no genetic assay capable of telling them apart. They differ however in their family sizes. Species A gives birth to twins 10% of the time, otherwise birthing a single infant. Species B births twins 20% of the time, otherwise birthing singleton infants. Assume these numbers are known with certainty, from many years of field research.

Now suppose you are managing a captive Species Breeding program. You have a new female panda of unknown species, and she has just given birth to twins. What is the probability that her next birth will also be twins?

The important information about the two species.

Species A:

- 10% Twins, P(Twin | Species A)
- 90% Single infant

Species B:

- 20% Twins, P(Twin | Species B)
- 80 % Single infant

The problem is divided up using total probaility:

 $P(Twin \mid Previous \; twin) = P(Twin \mid Species \; A) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Twin \mid Species \; A \mid Previous \; twin) + P(Twin \mid Species \; B) \times P(Twin \mid Species \; B) + P(Twin \mid Species \; B)$

Two of the probabilites are given directly in the excersise. Inserting these into the expression gives

$$P(Twin \mid Previous twin) = 0.1 \times P(Species A \mid Previous twin) + 0.2 \times P(Species A \mid Previous twin)$$

Calculating the missing probabilities is done using bayes rule. First for Species A

$$P(Species\ A\ |\ Previous\ twin) = \frac{P(Previous\ twin\ |\ Species\ A) \times P(Species\ A)}{P(Previous\ twin)} = \frac{0.1 \times 0.5}{P(Previous\ twin)}$$

To complete the calculation the probability of a twin is needed. Total probability can be used

 $P(\text{Previous twin}) = P(\text{Previous twin} \mid \text{Species A}) \times P(\text{Species A}) + P(\text{Previous twin} \mid \text{Species B}) \times P(\text{Species B}) = 0.1 \times 0.5 + 0.2 \times 10^{-3} \text{ Mpc}$

Inserting this back into the bayes rule applied on the probability the panda being from Species A completes that calculation

$$P(\text{Species A} \mid \text{Previous twin}) = \frac{0.1 \times 0.5}{0.15} = 0.33$$

The same calculation is repeated for Species B

$$P(Species \; B \; | \; Previous \; twin) = \frac{P(Previous \; twin \; | \; Species \; B) \times P(Species \; B)}{P(Previous \; twin)} = \frac{0.2 \times 0.5}{0.15} = 0.66$$

Finally the two probabilites calcutated using Bayes rule can go back into the original expression

$$P(Twin \mid Previous twin) = 0.1 \times P(Species A \mid Previous twin) + 0.2 \times P(Species A \mid Previous twin) = 0.1 \times 0.33 + 0.2 \times 0.66 = 0.1$$

16.5% makes sence. The probability is more than just taking the average probability of having twins from the two species. Since the previous litter was twins, it increased the probability of the panda being from species B and therefore the probability of the panda giving birth to twins.

Hard Question 2

Recall all the facts from the problem above. Now compute the probability that the panda we have is from species A, assuming we have observed only the first birth and that it was twins.

This was calculated in question 1

$$P(Species\ A\ |\ Previous\ twin) = \frac{P(Previous\ twin\ |\ Species\ A) \times P(Species\ A)}{P(Previous\ twin)} = \frac{0.1 \times 0.5}{0.15} = 0.33 = 33\%$$

Hard Question 3

Continuing on from the previous problem, suppose the same panda mother has a second birth and that it is not twins, but a singleton infant. Compute the posterior probability that this panda is species A.

The question expressed in symbols becomes

$$P(Species A \mid (Second singulation \cap First twin))$$

Aproach 1

This can be solved using bayes

$$P(Species \; A \; | \; (Second \; singelton \cap First \; twin)) = \frac{P((Second \; singelton \cap First \; twin) \; | \; Species \; A) \times P(Species \; A)}{P(Second \; singelton \cap First \; twin)}$$

$$\frac{P((Second\ singelton \cap First\ twin)\ |\ Species\ A) \times P(Species\ A)}{P(Second\ singelton \cap First\ twin)} = \frac{P(Second\ singelton\ |\ Species\ A) \times P(First\ twin\ |\ Species\ A)}{P(Second\ singelton \cap First\ twin)}$$

Using that the two events: the second being a singelton given the species is A i and the first being a twin given the species is A, are independent. We can calulate the probability useing multiplication.

$$\frac{P(Second \ singelton \ | \ Species \ A) \times P(First \ twin \ | \ Species \ A)) \times P(Species \ A)}{P(Second \ singelton \ \cap \ First \ twin)} = \frac{0.9 \times 0.1 \times 0.5}{P(Second \ singelton \ \cap \ First \ twin)}$$

To address the denominator we use total pobability.

 $P(Second\ singelton \cap First\ twin) = P((Second\ singelton \cap First\ twin) \mid Species\ A) \times P(Species\ A) + P((Second\ singelton \cap First\ twin) \mid Species\ A) \times P(Species\ A) + P((Second\ singelton \cap First\ twin) \mid Species\ A) \times P(Species\ A) + P((Second\ singelton \cap First\ twin) \mid Species\ A) \times P(Species\ A) + P((Second\ singelton \cap First\ twin) \mid Species\ A) \times P(Species\ A) \times P(S$

Using that the conditional probabilities are independent the probabilities can be calculated using multiplication.

P(Second singelton
$$\cap$$
 First twin) = $0.9 \times 0.1 \times 0.5 + 0.8 \times 0.2 \times 0.5 = 0.125$

Inserting this

$$P(Species~A~|~(Second~singelton \cap First~twin)) = \frac{0.9 \times 0.1 \times 0.5}{0.125} = 0.36 = 36\%$$

Aproach 2 Appraoch 1 was allot of work. Stating with what we know from question 2 it can be done faseter.

The prior is now updated and with it so are the probabilites. Keeping the notation simple $P(Species\ A\ now\ takes\ a\ new\ value\ without\ an\ updated\ notation.$

$$P(Species\ A\ |\ Second\ singelton) = \frac{P(Second\ singelton\ |\ Species\ A)P(Species\ A)}{P(Second\ singeltion)} = \frac{0.9\times0.33}{P(Second\ singeltion)}$$

$$P(Species~A~|~Second~singelton) = \frac{0.9 \times 0.33}{P(Second~singeltion)} = \frac{0.9 \times 0.33}{0.825} = 0.36 = 36\%$$

Hard Question 4

A common boast of Bayesian statisticians is that Bayesian inference makes it easy to use all of the data, even if the data are of different types.

So suppose now that a veterinarian comes along who has a new genetic test that she claims can identify the species of our mother panda. But the test, like all tests, is imperfect. This is the information you have about the test: - The probability it correctly identifies a species A panda is 0.8. - The probability it correctly identifies a species B panda is 0.65. The vet administers the test to your Species And tells you that the test is positive for species A. First ignore your previous information from the births and compute the posterior probability that your panda is species A. Then redo your calculation, now using the birth data as well.

Ignoring the adjustemnts to the priors we use bayes

$$P(Species\ A\mid Tested\ as\ A) = \frac{P(Tested\ as\ A\mid Species\ A)\times P(Species\ A)}{P(Tested\ as\ A)} = \frac{0.8\times 0.5}{P(Tested\ as\ A)}$$

 $P(\text{Tested as A}) = P(\text{Tested as A} \mid \text{Species A}) \times P(\text{Species A}) + P(\text{Tested as A} \mid \text{Species B}) \times P(\text{Species B}) = 0.8 \times 0.5 + 0.35 \times 0.5 \times 0.5$

P(Species A | Tested as A) =
$$\frac{0.8 \times 0.5}{P(Tested \text{ as A})} = \frac{0.8 \times 0.5}{0.575} = 0.696 = 69.6\%$$

Including the updated priors due to the testing gives anoter result. As in question 3 approach 2 this means the probabilities are updated without updating the notation.

$$P(Species\ A\mid Tested\ as\ A) = \frac{P(Tested\ as\ A\mid Species\ A)\times P(Species\ A)}{P(Tested\ as\ A)} = \frac{0.8\times 0.36}{P(Tested\ as\ A)}$$

 $P(Tested as A) = P(Tested as A \mid Species A) \times P(Species A) + P(Tested as A \mid Species B) \times P(Species B) = 0.8 \times 0.36 + 0.35 \times 0.36 \times 0.36 + 0.35 \times 0.36 \times 0$

P(Species A | Tested as A) =
$$\frac{0.8 \times 0.36}{P(Tested \text{ as A})} = \frac{0.8 \times 0.36}{0.484} = 0.595 = 59.5\%$$