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1  Course: Statistical_Inference
2  Lesson: ConditionalProbability
3
4
5  - Class: text
6  Output: "Conditional Probability. (Slides for this and other Data Science courses may
be found at github https://github.com/DataScienceSpecialization/courses/. If you care
to use them, they must be downloaded as a zip file and viewed locally. This lesson
corresponds to 06_Statistical_Inference/03_Conditional_Probability.)"
7
8  - Class: text
9  Output: In this lesson, as the name suggests, we'll discuss conditional probability.
10
11 - Class: mult_question
12 Output: If you were given a fair die and asked what the probability of rolling a 3
is, what would you reply?
13 AnswerChoices: 1/6; 1/2; 1/3; 1/4; 1
14 CorrectAnswer: 1/6
15 AnswerTests: omnitest(correctVal='1/6')
16 Hint: There are 6 possible outcomes and you want to know the probability of 1 of them.
17
18 - Class: mult_question
19 Output: Suppose the person who gave you the dice rolled it behind your back and told
you the roll was odd. Now what is the probability that the roll was a 3?
20 AnswerChoices: 1/6; 1/2; 1/3; 1/4; 1
21 CorrectAnswer: 1/3
22 AnswerTests: omnitest(correctVal='1/3')
23 Hint: Given that there are 3 odd numbers on the die your possibilities have been
reduced to 3 and you want to know the probability of 1 of them.
24
25 - Class: text
26 Output: The probability of this second event is conditional on this new information,
so the probability of rolling a 3 is now one third.
27
28 - Class: text
29 Output: We represent the conditional probability of an event A given that B has
occurred with the notation  $P(A|B)$ . More specifically, we define the conditional
probability of event A, given that B has occurred with the following.
30
31 - Class: text
32 Output:  $P(A|B) = P(A \& B) / P(B)$ .  $P(A|B)$  is the probability that BOTH A and B occur
divided by the probability that B occurs.
33
34 - Class: mult_question
35 Output: Back to our dice example. Which of the following expressions represents
 $P(A\&B)$ , where A is the event of rolling a 3 and B is the event of the roll being odd?
36 AnswerChoices: 1/6; 1/2; 1/3; 1/4; 1
37 CorrectAnswer: 1/6
38 AnswerTests: omnitest(correctVal='1/6')
39 Hint: Here A is a subset of B so the probability of both A AND B happening is the
probability of A happening.
40
41 - Class: mult_question
42 Output: Continuing with the same dice example. Which of the following expressions
represents  $P(A\&B)/P(B)$ , where A is the event of rolling a 3 and B is the event of the
roll being odd?
43 AnswerChoices: (1/6)/(1/2); (1/2)/(1/6); (1/3)/(1/2); 1/6
44 CorrectAnswer: (1/6)/(1/2)
45 AnswerTests: omnitest(correctVal='(1/6)/(1/2)')
46 Hint: Here A is a subset of B so the probability of both A AND B happening is the
probability of A happening. The probability of B is the reciprocal of the number of
odd numbers between 1 and 6 (inclusive).
47
48 - Class: text
49 Output: From the definition of  $P(A|B)$ , we can write  $P(A\&B) = P(A|B) * P(B)$ , right?
Let's use this to express  $P(B|A)$ .
50
51 - Class: text
52 Output:  $P(B|A) = P(B\&A)/P(A) = P(A|B) * P(B)/P(A)$ . This is a simple form of Bayes'

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Rule which relates the two conditional probabilities.

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53
54 - Class: text
55 Output: Suppose we don't know  $P(A)$  itself, but only know its conditional
probabilities, that is, the probability that it occurs if B occurs and the
probability that it occurs if B doesn't occur. These are  $P(A|B)$  and  $P(A|\sim B)$ ,
respectively. We use  $\sim B$  to represent 'not B' or 'B complement'.
56
57 - Class: text
58 Output: We can then express  $P(A) = P(A|B) * P(B) + P(A|\sim B) * P(\sim B)$  and substitute
this is into the denominator of Bayes' Formula.
59
60 - Class: text
61 Output:  $P(B|A) = P(A|B) * P(B) / ( P(A|B) * P(B) + P(A|\sim B) * P(\sim B) )$ 
62
63 - Class: text
64 Output: Bayes' Rule has applicability to medical diagnostic tests. We'll now discuss
the example of the HIV test from the slides.
65
66 - Class: text
67 Output: Suppose we know the accuracy rates of the test for both the positive case
(positive result when the patient has HIV) and negative (negative test result when
the patient doesn't have HIV). These are referred to as test sensitivity and
specificity, respectively.
68
69 - Class: mult_question
70 Output: Let 'D' be the event that the patient has HIV, and let '+' indicate a
positive test result and '-' a negative. What information do we know? Recall that we
know the accuracy rates of the HIV test.
71 AnswerChoices:  $P(+|D)$  and  $P(-|\sim D)$ ;  $P(+|\sim D)$  and  $P(-|D)$ ;  $P(+|D)$  and  $P(-|D)$ 
and  $P(-|\sim D)$ 
72 CorrectAnswer:  $P(+|D)$  and  $P(-|\sim D)$ 
73 AnswerTests: omnitest(correctVal='P(+|D) and P(-|\sim D)')
74 Hint: The clue here is accuracy. The test is positive when the patient has the
disease and negative when he doesn't.
75
76 - Class: mult_question
77 Output: Suppose a person gets a positive test result and comes from a population with
a HIV prevalence rate of .001. We'd like to know the probability that he really has
HIV. Which of the following represents this?
78 AnswerChoices:  $P(+|D)$ ;  $P(D|+)$ ;  $P(\sim D|+)$ ;  $P(D|-)$ 
79 CorrectAnswer:  $P(D|+)$ 
80 AnswerTests: omnitest(correctVal='P(D|+)')
81 Hint: We've already been given the information that the test was positive '+'. We
want to know whether D is present given the positive test result.
82
83 - Class: text
84 Output: By Bayes' Formula,  $P(D|+) = P(+|D) * P(D) / ( P(+|D) * P(D) + P(+|\sim D) * P(\sim D) )$ 
85
86 - Class: text
87 Output: We can use the prevalence of HIV in the patient's population as the value for
 $P(D)$ . Note that since  $P(\sim D)=1-P(D)$  and  $P(+|\sim D) = 1-P(-|\sim D)$  we can calculate  $P(D|+)$ .
In other words, we know values for all the terms on the right side of the equation.
Let's do it!
88
89 - Class: cmd_question
90 Output: Disease prevalence is .001. Test sensitivity (+ result with disease) is 99.7%
and specificity (- result without disease) is 98.5%. First compute the numerator,
 $P(+|D)*P(D)$ . (This is also part of the denominator.)
91 CorrectAnswer: .997*.001
92 AnswerTests: equiv_val(0.000997)
93 Hint: Multiply the test sensitivity by the prevalence.
94
95 - Class: cmd_question
96 Output: Now solve for the remainder of the denominator,  $P(+|\sim D)*P(\sim D)$ .
97 CorrectAnswer:  $(1-.985)*(1-.001)$ 
98 AnswerTests: equiv_val(.014985)
99 Hint: Multiply the complement of test specificity by the complement of prevalence.
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100
101 - Class: cmd_question
102 Output: Now put the pieces together to compute the probability that the patient has
the disease given his positive test result,  $P(D|+)$ . Plug your last two answers into
the formula  $P(+|D) * P(D) / (P(+|D) * P(D) + P(+|\sim D) * P(\sim D))$  to compute  $P(D|+)$ .
103 CorrectAnswer: .000997/(.000997+.014985)
104 AnswerTests: equiv_val(.06238268)
105 Hint: Divide (.997*.001) by (.997*.001 + .015*.999)
106
107 - Class: text
108 Output: So the patient has a 6% chance of having HIV given this positive test result.
The expression  $P(D|+)$  is called the positive predictive value. Similarly,  $P(\sim D|-)$ , is
called the negative predictive value, the probability that a patient does not have
the disease given a negative test result.
109
110 - Class: mult_question
111 Output: The diagnostic likelihood ratio of a positive test,  $DLR_+$ , is the ratio of
the two + conditional probabilities, one given the presence of disease and the other
given the absence. Specifically,  $DLR_+ = P(+|D) / P(+|\sim D)$ . Similarly, the  $DLR_-$  is
defined as a ratio. Which of the following do you think represents the  $DLR_-$ ?
112 AnswerChoices:  $P(-|D) / P(-|\sim D)$ ;  $P(+|\sim D) / P(-|D)$ ;  $P(-|D) / P(+|\sim D)$ ; I haven't a clue.
113 CorrectAnswer:  $P(-|D) / P(-|\sim D)$ 
114 AnswerTests: omnitest(correctVal='P(-|D) / P(-|\sim D)')
115 Hint: The signs of the test in both the numerator and denominator have to agree as
they did for the  $DLR_+$ .
116
117 - Class: text
118 Output: Recall that  $P(+|D)$  and  $P(-|\sim D)$ , (test sensitivity and specificity
respectively) are accuracy rates of a diagnostic test for the two possible results.
They should be close to 1 because no one would take an inaccurate test, right? Since
 $DLR_+ = P(+|D) / P(+|\sim D)$  we recognize the numerator as test sensitivity and the
denominator as the complement of test specificity.
119
120 - Class: mult_question
121 Output: Since the numerator is close to 1 and the denominator is close to 0 do you
expect  $DLR_+$  to be large or small?
122 AnswerChoices: Large; Small; I haven't a clue.
123 CorrectAnswer: Large
124 AnswerTests: omnitest(correctVal='Large')
125 Hint: What happens when you divide a large number by a much smaller one?
126
127 - Class: mult_question
128 Output: Now recall that  $DLR_- = P(-|D) / P(-|\sim D)$ . Here the numerator is the
complement of sensitivity and the denominator is specificity. From the arithmetic and
what you know about accuracy tests, do you expect  $DLR_-$  to be large or small?
129 AnswerChoices: Large; Small; I haven't a clue.
130 CorrectAnswer: Small
131 AnswerTests: omnitest(correctVal='Small')
132 Hint: What happens when you divide by small number by a larger one?
133
134
135 - Class: text
136 Output: Now a little more about likelihood ratios. Recall Bayes Formula.  $P(D|+) =$ 
 $P(+|D) * P(D) / (P(+|D) * P(D) + P(+|\sim D) * P(\sim D))$  and notice that if we replace all
occurrences of 'D' with ' $\sim D$ ', the denominator doesn't change. This means that if we
formed a ratio of  $P(D|+)$  to  $P(\sim D|+)$  we'd get a much simpler expression (since the
complicated denominators would cancel each other out). Like this....
137
138 - Class: text
139 Output:  $P(D|+) / P(\sim D|+) = P(+|D) * P(D) / (P(+|\sim D) * P(\sim D)) = P(+|D)/P(+|\sim D) * P(D)/P(\sim D)$ .
140
141 - Class: mult_question
142 Output: The left side of the equation represents the post-test odds of disease given
a positive test result. The equation says that the post-test odds of disease equals
the pre-test odds of disease (that is,  $P(D)/P(\sim D)$ ) times
143 AnswerChoices: the  $DLR_+$ ; the  $DLR_-$ ; I haven't a clue.
144 CorrectAnswer: the  $DLR_+$ 
145 AnswerTests: omnitest(correctVal='the  $DLR_+$ ')

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146 **Hint:** Do you recognize the expression $P(+|D) / P(+|\sim D)$? The '+' signs are a big clue.

147

148 - **Class:** text

149 **Output:** In other words, a DLR_+ value equal to N indicates that the hypothesis of disease is N times more supported by the data than the hypothesis of no disease.

150

151 - **Class:** text

152 **Output:** Taking the formula above and replacing the '+' signs with '-' yields a formula with the DLR_- . Specifically, $P(D|-) / P(\sim D|-) = P(-|D) / P(-|\sim D) * P(D)/P(\sim D)$. As with the positive case, this relates the odds of disease post-test, $P(D|-) / P(\sim D|-)$, to those of disease pre-test, $P(D)/P(\sim D)$.

153

154 - **Class:** mult_question

155 **Output:** The equation $P(D|-) / P(\sim D|-) = P(-|D) / P(-|\sim D) * P(D)/P(\sim D)$ says what about the post-test odds of disease relative to the pre-test odds of disease given negative test results?

156 **AnswerChoices:** post-test odds are greater than pre-test odds; post-test odds are less than pre-test odds; I haven't a clue.

157 **CorrectAnswer:** the DLR_+

158 **AnswerTests:** omnitest(correctVal='post-test odds are less than pre-test odds')

159 **Hint:** Remember that we argued (hopefully convincingly) that DLR_- is small (less than 1). Post-test odds = Pre-test odds * DLR_- so post-test odds are a fraction of the pre-test odds.

160

161 - **Class:** text

162 **Output:** Let's cover some basics now.

163

164 - **Class:** text

165 **Output:** Two events, A and B, are independent if they have no effect on each other. Formally, $P(A\&B) = P(A)*P(B)$. It's easy to see that if A and B are independent, then $P(A|B)=P(A)$. The definition is similar for random variables X and Y.

166

167 - **Class:** mult_question

168 **Output:** We've seen examples of independence in our previous probability lessons. Let's review a little. What's the probability of rolling a '6' twice in a row using a fair die?

169 **AnswerChoices:** 1/6; 2/6; 1/36; 1/2

170 **CorrectAnswer:** 1/36

171 **AnswerTests:** omnitest(correctVal='1/36')

172 **Hint:** Square the probability of rolling a single '6' since the two rolls are independent of one another.

173

174 - **Class:** mult_question

175 **Output:** You're given a fair die and asked to roll it twice. What's the probability that the second roll of the die matches the first?

176 **AnswerChoices:** 1/6; 2/6; 1/36; 1/2

177 **CorrectAnswer:** 1/6

178 **AnswerTests:** omnitest(correctVal='1/6')

179 **Hint:** Now the events aren't independent. You don't care what the first roll is so that's a probability 1 event. The second roll just has to match the first, so that's a 1/6 event.

180

181 - **Class:** mult_question

182 **Output:** If the chance of developing a disease with a genetic or environmental component is p, is the chance of both you and your sibling developing that disease $p*p$?

183 **AnswerChoices:** Yes; No

184 **CorrectAnswer:** No

185 **AnswerTests:** omnitest(correctVal='No')

186 **Hint:** The events aren't independent since genetic or environmental factors likely will affect the outcome.

187

188 - **Class:** text

189 **Output:** We'll conclude with iid. Random variables are said to be iid if they are independent and identically distributed. By independent we mean "statistically unrelated from one another". Identically distributed means that "all have been drawn from the same population distribution".

190

191 - **Class:** text

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192     Output: Random variables which are iid are the default model for random samples and
193 many of the important theories of statistics assume that variables are iid. We'll
194 usually assume our samples are random and variables are iid.
195
196
197 - Class: text
198 Output: Congrats! You've concluded this lesson on conditional probability. We hope
199 you liked it unconditionally.
200
201 - Class: mult_question
202 Output: "Would you like to receive credit for completing this course on
203 Coursera.org?"
204 CorrectAnswer: NULL
205 AnswerChoices: Yes;No
206 AnswerTests: coursera_on_demand()
207 Hint: ""
```