ML - Probability Practice

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Part A. Visitors to your website are asked to answer a single survey question before they get access to the content on the page. Among all of the users, there are two categories: Random Clicker (RC), and Truthful Clicker (TC). There are two possible answers to the survey: yes and no. Random clickers would click either one with equal probability. You are also giving the information that the expected fraction of random clickers is 0.3. After a trial period, you get the following survey results: 65% said Yes and 35% said No. What fraction of people who are truthful clickers answered yes? Hint: use the rule of total probability.

Solution:

Step 1: Define Initial Probabilities

Random Clicker (RC): Clicks "Yes" or "No" with equal probability:

- P(Yes|RC) = 0.5
- P(No|RC) = 0.5

Truthful Clickers (TC): Someone who always clicks the correct answer:

• Probability of a user being a Random Clicker (RC):

$$P(RC) = 0.3$$

Probability of a user being a Truthful Clicker (TC):

$$P(TC) = 1 - P(RC) = 0.7$$

Step 2: Survey Results After the trial period, the survey results indicate:

P(Yes)=0.65 - P(No)=0.35

Step 3: Calculate the Fraction of Truthful Clickers Who Answered "Yes" Using the Rule of Total Probability:

• $P(Yes) = P(Yes|TC) \times P(TC) + P(Yes|RC) \times P(RC)$

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# Calculate the probability of a truthful clicker answering "Yes"
prob_yes_truth <- (0.65 - (0.5 * 0.3)) / 0.7

# Convert to percentage
fraction <- prob_yes_truth * 100

# Print the result
print(paste("The fraction of people who are truthful clickers answered 'Yes' is", round(fraction, 2), "%"))</pre>
```

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## [1] "The fraction of people who are truthful clickers answered 'Yes' is 71.43 %"
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Part B

Imagine a medical test for a disease with the following two attributes:

- The sensitivity is about 0.993. That is, if someone has the disease, there is a probability of 0.993 that they will test positive.
- The specificity is about 0.9999. This means that if someone doesn't have the disease, there is probability of 0.9999 that they will test negative.
- In the general population, incidence of the disease is reasonably rare: about 0.0025% of all people have it (or 0.000025 as a decimal probability).

Suppose someone tests positive. What is the probability that they have the disease?

Solution:

Step 1: Define Initial Probabilities:

- P(D) = 0.000025 (Probability of having the disease)
- P(NoD) = 1 0.000025 = 0.999975 (Probability of not having the disease)

Sensitivity of the test is 0.993, leading to:

- P(+Pos|D) = 0.993 (Probability of testing positive given that the individual has the disease)
- P(-Neg|D) = 1 0.993 = 0.007 (Probability of testing negative given that the individual has the disease)

Specificity of the test is 0.9999, leading to:

- P(-Neg|NoD) = 0.9999 (Probability of testing negative given that the individual does not have the disease)
- P(+Pos|NoD) = 1−0.9999 = 0.0001 (Probability of testing positive given that the individual does not have the disease)

Question: "Suppose someone tests positive. What is the probability that they actually have the disease?"

We calculate the following joint probabilities using Bayes' Theorem:

- P(+Pos 'and' D) = P(+Pos|D) × P(D)
- P(+Pos 'and' NoD) = P(+Pos|NoD) × P(NoD)

```
# Calculating joint probabilities
joint_prob_pos_disease <- 0.993 * 0.000025
joint_prob_pos_no_disease <- 0.0001 * 0.9999

# Calculate the total probability of testing positive
total_prob_positive <- joint_prob_pos_disease + joint_prob_pos_no_disease
prob_disease_given_pos <- joint_prob_pos_disease / total_prob_positive
prob_no_disease_given_pos <- joint_prob_pos_no_disease / total_prob_positive

# Display the result
prob_disease_given_pos</pre>
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## [1] 0.1988944
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print(paste("The probability of someone having the disease given that they tested positive is",
round(prob_disease_given_pos * 100, 2), "%"))

[1] "The probability of someone having the disease given that they tested positive is 19.89 %"