First course in probability and statistics (MS-A0501, MS-A0502, MS-A0503, MS-0504) Department of Mathematics and Systems Analysis, Aalto University
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## Exam 21.2.2024 suggested solutions

## SOLUTION 1

a) A student passes the exam if 1) the teacher does not ask any question, 2) the teacher asks a single question and the student answers it correctly or 3) the teacher poses two questions and the student answers both of them correctly

$$
\begin{aligned}
\mathrm{P}(\text { pass }) & =p+p \pi+p \pi^{2} \\
& =\frac{1}{3}+\frac{1}{3} \pi+\frac{1}{3} \pi^{2}
\end{aligned}
$$

The three events ( 0,1 , or 2 questions) are mutually exclusive and collectively exhaustive. The probability of passing the exam is a weighted average of the conditional probability of passing under each of these events, weighted by the probability of each event occurring.
b) The probability to fail - given the adopted policy that all questions must be answered correctly - equals the probability to incorrectly answer a question, $1-\pi$.

Alternatively, one can arrive at this result by calculating the conditional probability $\mathrm{P}(A \mid B)$, where A denotes the event that the student incorrectly answers the second question and $B$ the event that the teacher asks 2 questions and the student answers correctly the first of them.

$$
\mathrm{P}(A \mid B)=\frac{p \pi(1-\pi)}{p \pi}=1-\pi
$$

## SOLUTION 2

In statistical terms, 20 Bernoulli experiments are carried out. In each experiment a green cube is chosen with probability 0.5 . Let $Y$ be the number of green cubes chosen so that $Y \sim \operatorname{Bin}(20,0.5)$.
a) Six-months-olds pick 17 green cubes with probability $\mathbf{0 . 0 0 1 1}$ :

$$
\begin{aligned}
\mathrm{P}(Y=17) & =\binom{20}{17} 0.5^{17} \times 0.5^{(20-17)} \\
& =\frac{20!}{(20-3)!\times 3!} \times 0.5^{17} \times 0.5^{3} \\
& =\frac{20!}{17!\times 3!} \times 0.5^{20} \\
& =\frac{20 \times 19 \times 18}{3!} \times 0.5^{20} \\
& =\frac{20 \times 19 \times 18}{3 \times 2 \times 1} \times 0.5^{20} \\
& =20 \times 19 \times 3 \times 0.5^{20} \\
& =1140 \times 0.5^{20} \\
& \approx 0.001087189
\end{aligned}
$$

$$
\approx 0.0011
$$

R command dbinom(17,20, 0.5 ) returns the same probability.
b) Six-months-olds pick at least 17 green cubes with probability $\mathbf{0 . 0 0 1 3}$ :

$$
\begin{aligned}
\mathrm{P}(Y \geqslant 17) & =\sum_{i=17}^{20} \mathrm{P}(Y=i) \\
& =\sum_{i=17}^{20}\binom{20}{i} 0.5^{i} \times 0.5^{(20-i)} \\
& =\binom{20}{17} 0.5^{17} \times 0.5^{3}+\binom{20}{18} 0.5^{18} \times 0.5^{2} \\
& +\binom{20}{19} 0.5^{19} \times 0.5^{1}+\binom{20}{20} 0.5^{20} \times 0.5^{0} \\
& \approx 0.001087+0.00018+0.000019+0.0000095 \\
& \approx \mathbf{0 . 0 0 1 3}
\end{aligned}
$$

The probability can be obtained conveniently by means of the probability of the complementary event

$$
\mathrm{P}(Y \geqslant 17)=1-\mathrm{P}(Y<17)=1-\mathrm{P}(Y \leqslant 16)
$$

with the R command 1 -pbinom $(16,20,0.5)$. The calculations reveal that it would be very improbable for six-months-olds to choose 17 or more green cubes if they picked a green and red cube with equal probability. This sort of probability calculations and inference underlie the reasoning of the researchers.

## SOLUTION 3

a) From the problem statement we have that

$$
\mathrm{P}(T \mid D)=\mathbf{0 . 9 5}
$$

By the chain rule:

$$
\mathrm{P}(T \cap D)=\mathrm{P}(D) \mathrm{P}(T \mid D)=\frac{1}{100} \cdot 0.95=\mathbf{0 . 0 0 9 5}
$$

By the total probability law:

$$
\begin{aligned}
\mathrm{P}(T) & =\mathrm{P}(T \cap D)+\mathrm{P}\left(T \cap D^{c}\right) \\
& =\mathrm{P}(D) \mathrm{P}(T \mid D)+\mathrm{P}\left(D^{c}\right) \mathrm{P}\left(T \mid D^{c}\right) \\
& =0.01 \cdot 0.95+0.99 \cdot 0.02 \\
& \approx 0.0293
\end{aligned}
$$

By the Bayes' rule:

$$
\mathrm{P}(D \mid T)=\frac{\mathrm{P}(T \cap D)}{\mathrm{P}(T)}=\frac{0.0095}{0.0293} \approx \mathbf{0 . 3 2 4 2}
$$

By complementing the probability to 1 :

$$
\mathrm{P}\left(D^{c} \mid T\right)=1-\mathrm{P}(D \mid T) \approx \mathbf{0 . 6 7 5 8}
$$

By the Bayes' rule:

$$
\begin{aligned}
\mathrm{P}\left(D \mid T^{c}\right) & =\frac{\mathrm{P}(D) \mathrm{P}\left(T^{c} \mid D\right)}{\mathrm{P}\left(T^{c}\right)} \\
& =\frac{\mathrm{P}(D) \mathrm{P}\left(T^{c} \mid D\right)}{1-\mathrm{P}(T)}= \\
& =\frac{(1 / 100) \cdot 0.05}{1-0.0293} \\
& \approx \mathbf{0 . 0 0 0 5}
\end{aligned}
$$

(Grade: 1 point question)
b) not graded
c) $\mathrm{P}(D \mid T)$ (Grade: 0.5 point) d) $0++-+$ (Grade: 0.5 point)

## SOLUTION 4

a) The test statistic is

$$
z=\frac{\bar{X}-\mu}{\sigma / \sqrt{n}}=\frac{4.98-4}{1.30 / \sqrt{289}} \approx \mathbf{1 2 . 8 1 5}
$$

b) The central limit theorem states that the sample average $\bar{X}$ of independent identically distributed random variables $X_{i}$ with mean $\mu$ and variance $\sigma^{2}$ approximately follows the normal distribution $\mathrm{N}\left(\mu, \sigma^{2} / n\right)$ for large $n$. It follows that

$$
z=\frac{\bar{X}-\mu}{\sigma / \sqrt{n}} \stackrel{\text { large } n}{\sim} \mathcal{N}(0,1) .
$$

According to the question, the assumptions for the central limit theorem apply. The figure 4.98 is a sample average of independent evaluations on the Likert-scale. The standard deviation 1.30 is known. The numerical value of the test statistic, 12.815, can be compared to the standard normal. Its 0.9995 -fractile is 3.290527 (qnorm(0.9995)). The probability for the test statistic to take a value outside the range ( $-3.290527,3.290527$ ) is 0.001 under the null hypothesis. Null hypothesis of mean 4 can be rejected at significance level 0.001 , because $12.815>3.290527$. Our conclusion is that the people (of the intended target population) like Became Rich people more than "moderately" as described by integer 4.

## SOLUTION 5

a) The likelihood function when observing 2 heads and 6 tails is given by

$$
\mathcal{L}(\theta)=\theta^{2}(1-\theta)^{6} .
$$

The function is continuously differentiable in $[0,1]$ and zero at the extremes of the interval. We follow the steps we delineated in the lecture namely we take the logarithm of the likelihood

$$
l(\theta)=2 \log (\theta)+6 \log (1-\theta) .
$$

Taking its derivative, we get

$$
l^{\prime}(\theta)=\frac{2}{\theta}-\frac{6}{1-\theta},
$$

which is 0 at $\theta=\mathbf{1} / \mathbf{4}$. The likelihood is positive here, so the maximum is at $\theta=\mathbf{1} / \mathbf{4}$ (not at the endpoints, where it evaluates to 0 ).
b) The unnormalised posterior is the product of the prior (given in the question) and the likelihood (described in the question and derived in a)):

$$
30 \theta^{2}(1-\theta)^{2} \cdot \theta^{2}(1-\theta)^{6}=30 \cdot \theta^{4}(1-\theta)^{8} .
$$

c) To find the MAP we note that the unnormalised posterior is continuously differentiable in $[0,1]$, and is zero at the end points. To maximise the function, we can again take the logarithm to ease our job

$$
4 \log (\theta)+8 \log (1-\theta) .
$$

Its derivative is

$$
\frac{4}{\theta}-\frac{8}{1-\theta}
$$

and is zero at $\theta=\mathbf{1} / \mathbf{3}$.
Alternatively, we can take the derivative of the unnormalised posterior to get

$$
4 \theta^{3}(1-\theta)^{8}-8 \theta^{4}(1-\theta)^{7}
$$

Setting it to 0 , we obtain $1-\theta=2 \theta$, which gives $\theta=\mathbf{1} / \mathbf{3}$.
The unnormalized posterior is positive here, so the maximum is at $\theta=\mathbf{1 / 3}$ (not at the endpoints, where the value is zero).

