

Machine Learning Worksheet 2

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Problem 1

- $T \in \{0, 1\}$: the scanned person is a civilian or terrorist respectively
- $S \in \{0, 1\}$: scanner identifies the person as a civilian or terrorist respectively

$$P(S = 1|T = 1) = 0.95$$

$$P(S = 0|T = 0) = 0.95$$

$$P(T = 1) = 0.01$$

$$\Rightarrow P(T = 0) = 1 - P(T = 1) = 0.99$$

$$\Rightarrow P(S = 1|T = 0) = 1 - P(S = 0|T = 0) = 0.05$$

$$\begin{aligned} P(T = 1|S = 1) &= \frac{P(S = 1|T = 1) \cdot P(T = 1)}{P(S = 1)} \\ &= \frac{P(S = 1|T = 1) \cdot P(T = 1)}{\sum_{t \in \{0,1\}} P(S = 1|T = t) \cdot P(T = t)} \\ &= \frac{0.95 \cdot 0.01}{0.05 \cdot 0.99 + 0.95 \cdot 0.01} \\ &= \frac{1}{6} \end{aligned}$$

Problem 2

- $C = \{0, 1\}$: the tossed coin shows head or tail respectively

$$C \sim \text{Ber}\left(\frac{1}{2}\right)$$

- $T = \{0, 1, 2\}$: the number of two coin tosses that resulted in *tail*

$$\Rightarrow T \sim \text{Bin}\left(2, \frac{1}{2}\right)$$

- $R = \{0, 1, 2, 3\}$: number of red balls drawn from the box

$$R \sim \text{Bin}\left(3, \frac{t}{2}\right)$$

$$\begin{aligned} P(T = t) &= \binom{2}{t} \cdot \left(\frac{1}{2}\right)^t \cdot \left(\frac{1}{2}\right)^{2-t} = \frac{1}{4} \cdot \binom{2}{t} \\ P(R = r|T = t) &= \binom{3}{r} \cdot \left(\frac{t}{2}\right)^r \cdot \left(1 - \frac{t}{2}\right)^{3-r} \\ P(T|R) &= \frac{P(R|T) \cdot P(T)}{P(R)} \end{aligned}$$

$$\begin{aligned} \Rightarrow P(T = 2|R = 3) &= \frac{P(R = 3|T = 2) \cdot P(T = 2)}{P(R = 3)} \\ &= P(R = 3|T = 2) \cdot \frac{P(T = 2)}{P(R = 3)} \\ &= P(R = 3|T = 2) \cdot \frac{P(T = 2)}{\sum_{t \in T} P(R = 3|T = t) \cdot P(T = t)} \\ &= 1 \cdot \frac{\frac{1}{4}}{0 \cdot \frac{1}{4} + \frac{1}{8} \cdot \frac{1}{2} + 1 \cdot \frac{1}{4}} \\ &= \frac{\frac{1}{4}}{\frac{5}{16}} = \frac{4}{5} \end{aligned}$$

Problem 3

random variables:

- U : the selected urn

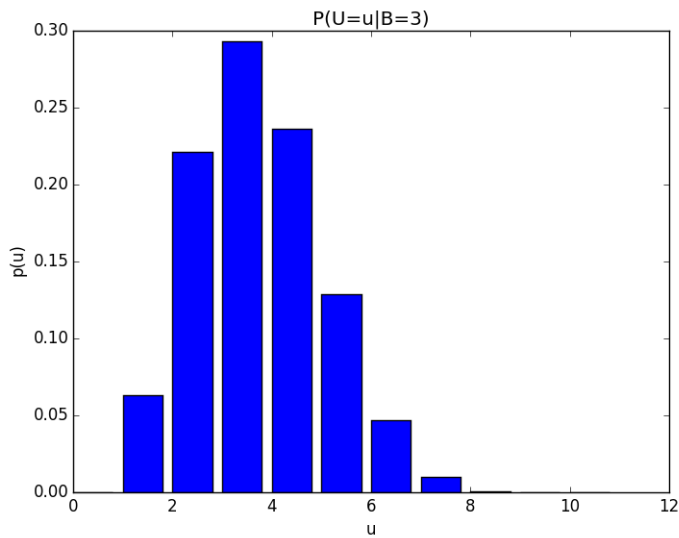
- B : the number of black balls drawn from an urn

$$\begin{aligned}
U &\sim \text{U}(-1, 10) \\
B &\sim \text{Bin}\left(10, 10 \cdot \frac{u}{10}\right) = \text{Bin}(10, u) \\
\Rightarrow P(U) &= \frac{1}{11} \\
\Rightarrow P(B|U) &= \binom{10}{b} \cdot \left(\frac{u}{10}\right)^b \cdot \left(1 - \frac{u}{10}\right)^{10-b}
\end{aligned}$$

$$\begin{aligned}
P(U|B) &= \frac{P(B|U) \cdot P(U)}{P(B)} \\
P(B) &= \sum_{u' \in U} P(B|U = u') \cdot P(U = u') \\
&= \sum_{u' \in U} \binom{10}{b} \cdot \left(\frac{u'}{10}\right)^b \cdot \left(1 - \frac{u'}{10}\right)^{10-b} \cdot \frac{1}{11} \\
&= \frac{1}{11} \cdot \binom{10}{b} \cdot \sum_{u' \in U} \left(\frac{u'}{10}\right)^b \cdot \left(1 - \frac{u'}{10}\right)^{10-b} \\
&= \frac{1}{11} \cdot \binom{10}{b} \cdot \sum_{u'=0}^{10} \left(\frac{u'}{10}\right)^b \cdot \left(1 - \frac{u'}{10}\right)^{10-b}
\end{aligned}$$

$$\begin{aligned}
P(U = u|B = b) &= \frac{P(B = b|U = u) \cdot P(U = u)}{P(B = b)} \\
&= \frac{\binom{10}{b} \cdot \left(\frac{u}{10}\right)^b \cdot \left(1 - \frac{u}{10}\right)^{10-b} \cdot \frac{1}{11}}{\frac{1}{11} \cdot \binom{10}{b} \cdot \sum_{u'=0}^{10} \left(\frac{u'}{10}\right)^b \cdot \left(1 - \frac{u'}{10}\right)^{10-b}} \\
&= \frac{\left(\frac{u}{10}\right)^b \cdot \left(1 - \frac{u}{10}\right)^{10-b}}{\sum_{u'=0}^{10} \left(\frac{u'}{10}\right)^b \cdot \left(1 - \frac{u'}{10}\right)^{10-b}} \\
P(U = u|B = 3) &= \frac{\left(\frac{u}{10}\right)^3 \cdot \left(1 - \frac{u}{10}\right)^7}{\sum_{u'=0}^{10} \left(\frac{u'}{10}\right)^3 \cdot \left(1 - \frac{u'}{10}\right)^7}
\end{aligned}$$

$$\begin{aligned}
&\Rightarrow P(U = 0|B = 3) = 0 \\
&\Rightarrow P(U = 1|B = 3) = \frac{1594323}{25277575} \\
&\Rightarrow P(U = 2|B = 3) = \frac{16777216}{75832725} \\
&\Rightarrow P(U = 3|B = 3) = \frac{7411887}{25277575} \\
&\Rightarrow P(U = 4|B = 3) = \frac{5971968}{25277575} \\
&\Rightarrow P(U = 5|B = 3) = \frac{390625}{3033309} \\
&\Rightarrow P(U = 6|B = 3) = \frac{1179648}{25277575} \\
&\Rightarrow P(U = 7|B = 3) = \frac{250047}{25277575} \\
&\Rightarrow P(U = 8|B = 3) = \frac{65536}{75832725} \\
&\Rightarrow P(U = 9|B = 3) = \frac{243}{25277575} \\
&\Rightarrow P(U = 10|B = 3) = 0
\end{aligned}$$



(result calculated using *ipython* interactive console – log attached to mail)

Problem 4

- $X = 1$: next ball drawn from one of the urns is black

- $X = 0$: next ball drawn from one of the urns is white

$$X \sim \text{Ber}\left(\frac{u}{10}\right)$$

$$\begin{aligned} P(X = 1) &= \sum_{u=0}^{10} P(X = 1|U = u) \cdot P(U = u) \\ &= \frac{673471}{2022206} \\ &\approx 0.33304 \end{aligned}$$

(result calculated using *ipython* interactive console – log attached to mail)

Problem 5

$$p(x) = \begin{cases} 1, & x \in [0, 1] \\ 0, & x \notin [0, 1] \end{cases}$$

$$\begin{aligned} \mathbb{E}[X] &= \int_{-\infty}^{\infty} p(x) \cdot x \, dx \\ &= \int_{-\infty}^0 p(x) \cdot x \, dx + \int_0^1 p(x) \cdot x \, dx + \int_1^{\infty} p(x) \cdot x \, dx \\ &= \int_{-\infty}^0 0 \cdot x \, dx + \int_0^1 1 \cdot x \, dx + \int_1^{\infty} 0 \cdot x \, dx \\ &= \int_{-\infty}^0 0 \, dx + \int_0^1 x \, dx + \int_1^{\infty} 0 \, dx \\ &= 0 + \left[\frac{1}{2} \cdot x^2 \right]_{x=0}^1 + 0 \\ &= \frac{1}{2} \end{aligned}$$

$$\begin{aligned}
\mathbb{E}[X^2] &= \int_{-\infty}^{\infty} p(x) \cdot x^2 \, dx = \int_0^1 x^2 \, dx \\
&= \left[\frac{1}{3} \cdot x^3 \right]_{x=0}^1 \\
&= \frac{1}{3}
\end{aligned}$$

$$\begin{aligned}
\text{Var}[X] &= \mathbb{E}[X^2] - \mathbb{E}[X]^2 \\
&= \frac{1}{3} - \left(\frac{1}{2}\right)^2 = \frac{1}{3} - \frac{1}{4} \\
&= \frac{1}{12}
\end{aligned}$$

Problem 6

- $\mathbb{E}[X] = \mathbb{E}_Y[\mathbb{E}_{X|Y}[X]]$:

$$\begin{aligned}
\mathbb{E}_Y[\mathbb{E}_{X|Y}[X]] &= \mathbb{E}_Y \left[\int \frac{p(x, y)}{p(y)} \cdot x \, dx \right] \\
&= \int p(y) \cdot \int \frac{p(x, y)}{p(y)} \cdot x \, dx \, dy \\
&= \int p(y) \cdot \frac{1}{p(y)} \int p(x, y) \cdot x \, dx \, dy \\
&= \int \int p(x, y) \cdot x \, dx \, dy \\
&= \mathbb{E}[X]
\end{aligned}$$

- $\text{Var}[x] = \mathbb{E}_Y[\text{Var}_{X|Y}[X]] - \text{Var}_Y[\mathbb{E}_{X|Y}[X]]$:

$$\begin{aligned}
\mathbb{E}_Y[\text{Var}_{X|Y}[X]] - \text{Var}_Y[\mathbb{E}_{X|Y}[X]] &= \mathbb{E}_Y[\mathbb{E}_{X|Y}[X^2] - \mathbb{E}_{X|Y}[X]^2] - \text{Var}_Y[\mathbb{E}_{X|Y}[X]] \\
&= \mathbb{E}_Y[\mathbb{E}_{X|Y}[X^2] - \mathbb{E}_{X|Y}[X]^2] - \mathbb{E}_Y[\mathbb{E}_{X|Y}[X]^2] + \mathbb{E}_Y[\mathbb{E}_{X|Y}[X]]^2 \\
&= \mathbb{E}_Y[\mathbb{E}_{X|Y}[X^2]] - \mathbb{E}_Y[\mathbb{E}_{X|Y}[X]]^2 \\
&= \mathbb{E}[X^2] - \mathbb{E}[X]^2 \\
&= \text{Var}[X]
\end{aligned}$$

Problem 7

- $P[X > c] \leq \frac{E[X]}{c}$:

$$\begin{aligned}
 \frac{E[X]}{c} &= \frac{\sum (p(x) \cdot x)}{c} \\
 &= \frac{\sum_{x=-\inf}^{c-1} (p(x) \cdot x) + \sum_{x=c}^{\inf} (p(x) \cdot x)}{c} \\
 &\geq \frac{\sum_{x=c}^{\inf} (p(x) \cdot x)}{c} \\
 &\geq \frac{\sum_{x=c}^{\inf} (p(x) \cdot c)}{c} \\
 &= \sum_{x=c}^{\inf} (p(x)) = P(X \geq c) \\
 &\geq P(X > c)
 \end{aligned}$$

- $P[X > \frac{3}{4}]$:

$$P[X > \frac{3}{4}] \leq \frac{\frac{n}{2}}{n * \frac{3}{4}}$$

Problem 8

- $P[|X - E[x]| > \alpha] \leq \frac{\text{Var}[X]}{\alpha^2}$:

$$\begin{aligned}
 \frac{\text{Var}[X]}{\alpha} &= \frac{E[(X - E[x])^2]}{\alpha^2} \\
 &\geq P((X - E[x])^2 > \alpha^2) \\
 &= P(|X - E[x]| > \alpha)
 \end{aligned}$$

- $P[|X| > \frac{3}{4}]$:

$$\begin{aligned}
 P[|X| > \frac{3}{4}] &\leq \frac{n * \frac{1}{4}}{n * n * \frac{9}{12}} \\
 P[X > \frac{3}{4}] &\leq \frac{\frac{n * \frac{1}{4}}{n * n * \frac{9}{12}}}{2}
 \end{aligned}$$

Problem 9

Basics

λ Trick :

$$\sum_{i=0}^{n+1} \lambda_i = 1 \wedge \lambda'_i = \frac{\lambda_i}{(1-\lambda_{n+1})} \rightarrow \sum_{i=0}^n \lambda'_i = 1$$

Definition Convex:

$$\forall x_1, x_2 \forall t \in [0, 1] \ f(tx_1 + (1-t)x_2) \leq t * f(x_1) + (1-t) * f(x_2)$$

Induction

Hypothese

$$f(\sum_{i=0}^n \lambda_i x_i) \leq \sum_{i=0}^n \lambda_i f(x_i) \mid \sum_{i=0}^n \lambda_i = 1$$

Induction Start

$$n = 1 \wedge \lambda_1 = 1$$

$$f(\lambda_1 x_1) \leq \lambda_1 * f(x_1)$$

$$f(1 * x_1) \leq 1 * f(x_1)$$

$$f(x_1) \leq f(x_1)$$

Induction Step

$$f(\sum_{i=0}^{n+1} \lambda_i x_i) \leq \sum_{i=0}^{n+1} \lambda_i f(x_i)$$

Steps:

$$f(\sum_{i=0}^{n+1} \lambda_i x_i) = f(\sum_{i=0}^n \lambda_i x_i + \lambda_{n+1} x_{n+1})$$

$$f(\sum_{i=0}^n \lambda_i x_i + \lambda_{n+1} x_{n+1}) = f((1 - \lambda_{n+1}) \sum_{i=0}^n \lambda'_i x_i + \lambda_{n+1} x_{n+1})$$

$$f((1 - \lambda_{n+1}) \sum_{i=0}^n \lambda'_i x_i + \lambda_{n+1} x_{n+1}) \leq (1 - \lambda_{n+1}) f(\sum_{i=0}^n \lambda'_i x_i) + \lambda_{n+1} f(x_{n+1})$$

$$(1 - \lambda_{n+1}) f(\sum_{i=0}^n \lambda'_i x_i) + \lambda_{n+1} f(x_{n+1}) \leq (1 - \lambda_{n+1}) \sum_{i=0}^n \lambda'_i f(x_i) + \lambda_{n+1} f(x_{n+1})$$

$$(1 - \lambda_{n+1}) \sum_{i=0}^n \lambda'_i f(x_i) + \lambda_{n+1} f(x_{n+1}) = \sum_{i=0}^n \lambda_i f(x_i) + \lambda_{n+1} f(x_{n+1})$$

$$\sum_{i=0}^n \lambda_i f(x_i) + \lambda_{n+1} f(x_{n+1}) = \sum_{i=0}^{n+1} \lambda_i f(x_i)$$