

# 1A

Machine Learning II  
ID: 5684926 Tristan Scheidemann

## 1A-1.

Let  $X$  represent the probability of  $X$  phone calls and  $Y$  be the waiting time between different calls. In this case  $X$  would be Poisson and  $Y$  exponentially distributed.

## 1A-2.

We have the following constraints:

Given the above restrictions  $R$ , we have to prove / disprove:

$$R \Rightarrow P(B, C|X) = P(B|X) \cdot P(C|X)$$

Solution: Transform the joint probability  $P(B, C|X)$  into

$$P(B, C|X) = P(C|B, X) \cdot P(B|X).$$

Excluding the degenerate case  $P(B|X) = 0$ , this leads to:

$$\begin{aligned} P(B, C|X) &= P(B|X) \cdot P(C|X) \\ \Leftrightarrow P(C|B, X) \cdot P(B|X) &= P(B|X) \cdot P(C|X) \\ \Leftrightarrow P(C|B, X) &= P(C|X) \\ \Leftrightarrow \frac{P(C, B, X)}{P(B, X)} &= P(C|X). \end{aligned}$$

The above only holds iff  $P(C|B, X) = P(C|X)$ , which is generally not true, so conditional independence of A,B and A,C is not sufficient for transitivity.

## 1A-4.

Definition of the events:

$G$  = Person is guilty,

$T$  = Person passes the test.

- (i) The negations  $\bar{T}$ ,  $\bar{G}$  can be read as *not*.

$$P(G|\bar{T}) = \frac{P(\bar{T}|G) \cdot P(G)}{P(\bar{T})} = \frac{\frac{5}{6} \cdot \frac{1}{3}}{\frac{7}{18}} = \frac{5}{7},$$

with

$$P(\bar{T}|G) = \frac{5}{6},$$

$$P(G) = \frac{1}{3},$$

$$P(\bar{T}) = P(G) \cdot P(\bar{T}|G) + P(\bar{G}) \cdot P(\bar{T}|\bar{G}) = \frac{1}{3} \cdot \frac{5}{6} + \frac{2}{3} \cdot \frac{1}{6} = \frac{7}{18}$$

(ii)

$$P(G|\bar{T}, \bar{T}) = \frac{P(\bar{T}, \bar{T}|G)}{P(\bar{T}, \bar{T})} = \frac{P(\bar{T}|G) \cdot P(\bar{T}|G) \cdot P(G)}{\underbrace{P(\bar{T}, \bar{T})}_{\text{conditional independence}}} = \frac{\left(\frac{5}{6}\right)^2 \cdot \frac{1}{3}}{\frac{1}{3} \cdot \left(\frac{5}{6}\right)^2 + \frac{2}{3} \cdot \left(\frac{1}{6}\right)^2} = 0.\overline{925}.$$

Using the conditional independence of  $P(\bar{T}, \bar{T}|G)$  and independence of testing  $P(\bar{T}, \bar{T}|G)$ .

1A-5.

$$E[X] = \sum_{i=1}^6 i \cdot P(X=i) = (1+2+3) \cdot \frac{1}{12} + (4+5) \cdot \frac{1}{6} + 6 \cdot \frac{5}{12} = 4.5.$$

$$\begin{aligned} \text{Var}[X] &= E[X^2] - E[X]^2 = \left[ (1+4+9) \cdot \frac{1}{12} + (16+25) \cdot \frac{1}{6} + 36 \cdot \frac{5}{12} \right] - 4.5^2 = \frac{271}{12} - 4.5^2 \\ &= 2.\bar{3}. \end{aligned}$$

$$E[X_1 + E_2] = 2 \cdot E[X_1] = 9.$$

1A-6.

$$\begin{aligned} E[(X - E[X])(Y - E[Y])] &= E[XY - XE[Y] - YE[X] + E[X]E[Y]] \\ &= E[XY] - E[Y]E[X] - E[X]E[Y] + E[X]E[Y] \\ &= \underbrace{E[XY] - E[X]E[Y]}_{*} \\ &= 0. \end{aligned}$$

$$\begin{aligned} E[XY] &= \int_{y_0}^{y_1} \int_{x_0}^{x_1} xy \cdot f_{X,Y}(x,y) dx dy = \int_{y_0}^{y_1} y \int_{x_0}^{x_1} x \cdot \underbrace{f_X(x)f_Y(y)}_{\text{independence}} dx dy \\ &= \int_{y_0}^{y_1} y f_Y(y) \int_{x_0}^{x_1} x f_X(x) dx dy = \int_{y_0}^{y_1} y f_Y(y) \cdot E[X] dy = E[X] \int_{y_0}^{y_1} y f_Y(y) dy \\ &= E[X]E[Y]. \end{aligned}$$

Utilizing that  $* E[XY] = E[X]E[Y]$  if  $X, Y$  are independent.