

Task 1

Part a):

$$P(\text{color is not green} \mid \text{vehicle is Truck}) = ?$$

Calculating probability of "color is green"

$$\begin{aligned} P(\text{color is Green}) &= 0.0195 + 0.0252 + \\ &\quad 0.0160 + 0.0465 \\ &= 0.1072 \end{aligned}$$

$$\begin{aligned} \therefore P(\text{color is not green}) &= 1 - P(\text{color is green}) \\ &= 0.8928 \end{aligned}$$

$$\begin{aligned} P(\text{vehicle is Truck}) &= 0.1070 + 0.0160 \\ &\quad + 0.0265 \\ &= 0.1495 \end{aligned}$$

$$\begin{aligned} P(\text{color is not green} \mid \text{vehicle is Truck}) &= P(\text{not green} \cap \text{Truck}) \\ &\quad P(\text{Truck}) \end{aligned}$$

$$\begin{aligned} P(\text{not green} \cap \text{Truck}) &= 0.1070 + 0.0265 \\ &= 0.1335 \end{aligned}$$

$$\begin{aligned} \therefore P(\text{color not green} \mid \text{Truck}) &= \frac{0.1335}{0.1495} \\ &= 0.8923 \end{aligned}$$

b) we want to prove that
 $P(A)^* P(B) = P(A \cap B)$

$$P(\text{vehicle})^* P(\text{color})$$

consider, car as SUV &
color as green

$$P(\text{Green}) = 0.1072$$

$$P(\text{SUV}) = 0.4337$$

$$P(\text{Green})^* P(\text{SUV}) = 0.0465$$

$$\text{From table, } P(\text{Green} \cap \text{SUV}) = 0.0465$$

As both values are same,

they are independent from each other.

Task 2)

a) Given that, there are 11 variables & variable A has 7 values & variable B₁..B₁₀ has 8 values.

As variable B is conditionally independent, it'll give 8^{10} values.

As variable A has 7 values.

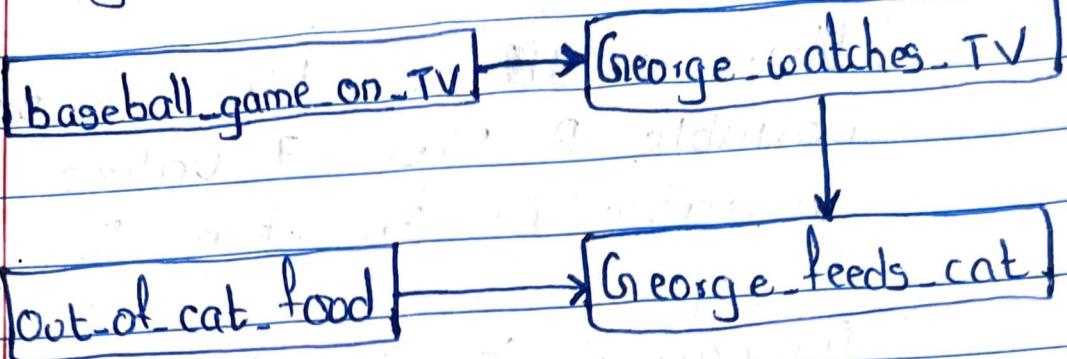
So, total $7 * 8^{10}$ many numbers is needed to store in joint distribution table

b) we can do one thing is that by storing only 8^{10} values. so there will be $7 + 8^{10}$ values which is way less than given in a).

c) yes, scenario does not follow the naive-bayes model

Task 3)

Bayesian network design.



Baseball.game.on.TV has no parent

0	0.5	0.70
1	0.5	0.30

(P) from code

out-of-cat-food has no parent

Value	Prob
0	0.5
1	0.5

from code

baseball.game.on.TV is parent of George.watches.TV

bb.g.o.TV	G.w.TV	P(bb.g.o.TV G.w.TV)
0	0	0.88
0	1	0.12
1	0	0.03
1	1	0.93

George-watches-Tv & out-of-cat-food
are parents of George-feeds-cat.

G-W-TV O-O-C-F G-F-C ~~with 0.04~~
~~0.04~~ P(G-F-C | G-W-TV)

O	O	O	O	O-O-C-F
O	O	O	1	0.96
O	1	O	O	0.68
O	1	1	1	0.32
1	O	O	O	0.29
1	O	G-F-C	O	0.71
1	1	O	O	0.96
1	1	1	1	0.04

I've drawn tables independently,
not in diagram because of space issue.

Task 5)

using inference by enumeration.

$$P(B \text{---} g \text{---} o \text{---} TV \mid \neg(G \text{---} f \text{---} c))$$

$$= \frac{P(B \text{---} g \text{---} o \text{---} TV \text{---} \neg(G \text{---} f \text{---} c))}{P(G \text{---} f \text{---} o)}$$

$$P(B \text{---} g \text{---} o \text{---} TV \text{---} \neg(G \text{---} f \text{---} c)) =$$

$$(0.3041 * 0.879 * 0.1698 * 0.9588) + \\ (0.3041 * 0.07207 * 0.1698 * 0.6842) + \\ 0.3041 * 0.9279 * 0.8301 * 0.2936) + \\ 0.3041 * 0.07207 * 0.8301 * 0.0421)$$

$$= \underline{\underline{0.118}}$$

$$P(G \text{---} f \text{---} o) = 0.7569$$

$$P(\neg G \text{---} f \text{---} o) = 1 - 0.7569 \\ = 0.2438$$

$$\therefore P(B \text{---} g \text{---} o \text{---} TV \mid \neg(G \text{---} f \text{---} c))$$

$$= \underline{\underline{0.118}} \\ 0.2438$$

$$= \underline{\underline{0.484}}$$

Task 6) Information content

$$\text{Entropy}(\text{class}) = -P(x) \cdot \log_2 P(x) - P(y) \cdot \log_2 P(y)$$

$$= -\frac{5}{10} \cdot \log_2 \left(\frac{5}{10}\right) - \frac{5}{10} \cdot \log_2 \left(\frac{5}{10}\right)$$

Information Gain of A

$$\text{Entropy}(\text{class} | A=1) = -\frac{3}{3} \log_2 \left(\frac{3}{3}\right) = 0$$

$$\text{Entropy}(\text{class} | A=2) = -\frac{1}{4} \log_2 \left(\frac{1}{4}\right) - \frac{3}{4} \log_2 \left(\frac{3}{4}\right) \\ = 0.813$$

$$\text{Entropy}(\text{class} | A=3) = -\frac{1}{3} \log_2 \left(\frac{1}{3}\right) - \frac{2}{3} \log_2 \left(\frac{2}{3}\right) \\ = 0.922$$

$$\text{Gain}(\text{class} | A) = \text{in} [0.813 + 0.922]$$

$$= 1 - \left[\frac{3}{10} (0) + \frac{4}{10} (0.813) + \left(\frac{3}{10}\right) 0.922 \right]$$

$$= \underline{\underline{0.39}}$$

Information Gain of B

$$\text{Entropy}(\text{class } B=1) = -\frac{1}{4} \log_2\left(\frac{1}{4}\right) - \frac{3}{4} \log_2\left(\frac{3}{4}\right)$$
$$= 0.8113$$

$$\text{Entropy}(\text{class } B=2) = -\frac{3}{4} \log_2\left(\frac{3}{4}\right) - \frac{1}{4} \log_2\left(\frac{1}{4}\right)$$
$$= 0.8113$$

$$\text{Entropy}(\text{class } B=3) = -1 \log_2\left(\frac{1}{2}\right) - 1 \log_2\left(\frac{1}{2}\right)$$
$$= 1$$

$$\text{Gain}(\text{class } B) = 1 - \left[\frac{4}{10} \times 0.8113 + \frac{2}{10} \times 0.8113 + \frac{4}{10} \times 1 \right]$$
$$= 0.151$$

Information Gain of C
= 0.151

Information Gain of D
= 0.151

Information Gain of E
= 0.151

Information Gain of F
= 0.151

Information Gain of C:

$$\text{Entropy}(\text{class } c=1) = -\frac{3}{5} \log_2 \left(\frac{3}{5}\right) - \frac{2}{5} \log_2 \left(\frac{2}{5}\right)$$
$$= 0.97$$

$$\text{Entropy}(\text{class } c=2) = -\frac{2}{4} \log_2 \left(\frac{2}{4}\right) - \frac{2}{4} \log_2 \left(\frac{2}{4}\right)$$
$$= 1$$

$$\text{Entropy}(\text{class } c=3) = -1 \log_2 (1)$$
$$= 0$$

$$\text{Gain}(C) = 1 - \left[\frac{5}{10} (0.97) + \frac{4}{10} (1) + 0 \right]$$
$$= \underline{\underline{0.88}}$$