## Probabilistic Artificial Intelligence

Problem Set 1
September 28, 2018

## 1. Conditional Independence

Consider the following joint distribution for three random variables, $a, b, c \in\{0,1\}$.

| $a$ | $b$ | $c$ | $p(a, b, c)$ |
| :---: | :---: | :---: | :---: |
| 0 | 0 | 0 | 0.192 |
| 0 | 0 | 1 | 0.144 |
| 0 | 1 | 0 | 0.048 |
| 0 | 1 | 1 | 0.216 |
| 1 | 0 | 0 | 0.192 |
| 1 | 0 | 1 | 0.064 |
| 1 | 1 | 0 | 0.048 |
| 1 | 1 | 1 | 0.096 |

Show that $a$ and $b$ are dependent, namely $p(a, b) \neq p(a) p(b)$. But, they are marginally independent given $c$, namely $p(a, b \mid c)=p(a \mid c) p(b \mid c)$. (c.f. Bishop Pattern Recognition and Machine Learning, Exercise 8.3)

## 2. Bayes Rule

A routine breast cancer mammography screening is performed on a group of people of age fourty. $1 \%$ of the participants in the screening actually have breast cancer. $80 \%$ of the people in the screening with breast cancer received positive results (has breast cancer) on the mammmography test. $9.6 \%$ of people without breast cancer received a positive result on their mammographies. Suppose a person of this age receives a positive result on their mammography. Given the information in this screening, what is the probability that he has breast cancer?

## 3. Chain rule

Derive the chain rule from the basic rules of probability. (Hint: by the definition of conditional probability $P(A, B)=P(A \mid B) P(B)$ ). How many factorizations are possible for a distribution on $n$ random variables?

