Introduktion

$$\begin{aligned} & \textbf{Sample space } \Omega = \{e_1, \dots, e_n\} \\ & \textbf{Event } E \subseteq \Omega \\ & \textbf{Probability function } P(E) = x \\ & \forall E: |E| = i > 1: P(E) = \sum_{i}^{n} P(\{e_i\}), e_i \in E \\ & \textbf{Probability space } (\Omega, P) \\ & \textbf{Bayes Theorem } P(S|T) = \frac{P(T|S)P(S)}{P(T)} \\ & \textbf{Total Probability } \frac{P(B|A)P(A)}{\sum_{r} P(B|r)} \\ & \textbf{Conditional Probability } \frac{P(B|A)P(A)}{\sum_{r} P(B|r)P(r)} \\ & \textbf{Axiom 1 } P(\Omega) = 1 \\ & \textbf{Axiom 2 } 0 \leq P(E) \leq 1, \forall E \subseteq \Omega \\ & \textbf{Axiom 3 } E, F \subseteq \Omega: E \cap F = \emptyset \Rightarrow P(E \cup F) = P(E) + P(F) \\ & \textbf{In general } P(E \cup F) = P(E) + P(F) - P(E \cap F) & (P(\emptyset) = 0) \end{aligned}$$

$$\textbf{Conditional Probability } P(E|F) = \frac{P(E \cap F)}{P(F)} \text{Probability of E occurring when F occurred}$$

Independence

Product Rule
$$P(x,y) = P(x|y)P(y)$$

Chain Rule $P(x_1,x_2) = P(x_2|x_1) \cdot P(x_1) = P(x_1|x_2) \cdot P(x_2)$
Independence $I_p(A,B) \Leftrightarrow P(a|b) = P(a) \lor \Big(P(a) = 0 \lor P(b) = 0\Big)$
Conditional Independence $I_p(A,B|C) \Leftrightarrow P(a|b,c) = P(a|b)$
Symmetry $I_p(X,Y|Z) \Rightarrow I_p(Y,X|Z)$
Decomposition $I_p(X,\{Y,W\}|Z) \Rightarrow I_p(X,Y|Z)$
Weak Union $I_p(X,\{Y,W\}|Z) \Rightarrow I_p(X,Y|Z)$
Contraction $I_p(X,\{Y,W\}|Z) \Rightarrow I_p(X,Y|Z) \Rightarrow I_p(X,\{Y,W\}|Z)$
Intersection $I_p(X,Y|Z,W)$ und $I_p(X,Y|Z) \Rightarrow I_p(X,\{Y,W\}|Z)$
For positive distributions and for mutually disjoint sets X,Y,Z,W

Markov

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\mathbf{DAG}\ G = (V,E) \mathbf{Nodes}\ V := \{A,B,C,\ldots\} \mathbf{Edge}\ A \to B := (A,B) \in E \land (B,A) \not\in E \mathbf{Parent}\ \mathbf{node}\ PA_X := \mathbf{nodes}\ \mathbf{with}\ \mathbf{a}\ \mathbf{direct}\ \mathbf{path}\ \mathbf{to}\ X \mathbf{Non\text{-}Descendants}\ ND_X := \mathbf{nodes}\ \mathbf{that}\ \mathbf{have}\ \mathbf{no}\ \mathbf{path}\ \mathbf{from}\ X \mathbf{Markov}\ \mathbf{Condition}\ I_p(X,ND_X|PA_X) \mathbf{G}\ \mathbf{satisfies}\ \mathbf{MC}\ \forall X \in V: I_p(X,ND_X|PA_X) \mathbf{Chain}\ \rho = [X,...,Y] := X - ... - Y \qquad X,...,Y \in A \subseteq V \mathbf{Directed}\ \mathbf{Chain}\ X(\leftarrow \lor \to)...(\leftarrow \lor \to)Y \mathbf{Blockade}\ \mathbf{by}\ A\ \exists Z \in A : \leftarrow Z \leftarrow \exists Z \in A : \leftarrow Z \to \exists Z \in A : \forall z \in desc_Z : z \not\in A \land z \not\in \rho \land \to Z \leftarrow \mathbf{d-Separation}\ I_G(X,Y|A) := \mathbf{every}\ \mathbf{chain}\ \mathbf{between}\ X\ \mathbf{and}\ Y\ \mathbf{is}\ \mathbf{blocked}\ \mathbf{by}\ A \mathbf{Active}\ \mathbf{Chain}\ \rho\ \mathbf{is}\ \mathbf{not}\ \mathbf{blocked}\ \mathbf{by}\ A\ \mathbf{given}\ A
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Subjective / Bayesian Probabilities

Prior Probability probability of some event prior to updating its probability using new information.

Posteriori Probability probability of some event after updating its probability based on new information.

Join Prob. Distribution