

- **Formulas**

- Conditional Probability:  $P(e \wedge h) = P(e \mid h) * P(h)$
- Bayes' Rule:  $P(e \mid h) * P(h) = P(h \mid e) * P(e)$
- Chain Rule:  $P(a_1 \wedge a_2 \wedge a_3) = P(a_1 \mid a_2 \wedge a_3) * P(a_2 \mid a_3) * P(a_3)$

- Filtering Formula:  $P(s_i \mid o_{0..i}) = \frac{P(o_i \mid s_i) * \sum_{s_{i-1} \in S_{i-1}} P(s_i \mid s_{i-1}) * P(s_{i-1} \mid o_{0..i-1}) * P(o_{0..i-1})}{P(o_{0..i})}$   
where  $P(o_{0..i})$  means  $P(o_0 \wedge o_1 \wedge \dots \wedge o_i)$

- **Axioms of probability**

- $P(a) \geq 0$
- $P(\text{true}) = 1$
- $P(a \vee b) = P(a) + P(b)$  if  $a$  and  $b$  are mutually exclusive

- **Propositions**

- $P(\neg a) = 1 - P(a)$
- $P(a) = \sum_{x \in \text{domain}(X)} P(a \wedge X = x)$
- $P(a \vee b) = P(a) + P(b) - P(a \wedge b)$
- If  $a \leftrightarrow b$  then  $P(a) = P(b)$
- $P(a) = P(a \wedge b) + P(a \wedge \neg b) = \sum_{b \in B} (a \wedge b)$

- **Concepts**

- Random Variables:  $P(X)$  is a function from the domain of  $X$  to  $[0 \dots 1]$ , e.g.  $P_X(32) = P(X = 32)$
- Expected Value: Let  $f(w)$  be a function on worlds. Given a probability measure  $P$ , the expected value of  $f \in P$  is  $E_P(f) = \sum_{\omega \in \Omega} f(\omega) * P(\omega)$
- Expected Value Special Case:

$$f(\omega) = \begin{cases} 1 & \text{if } a \text{ is true in } \omega \\ 0 & \text{if } a \text{ is false in } \omega \end{cases}$$

$$E_P(f) = P(a) = \sum_{\omega: a \text{ true in } \omega} P(\omega)$$

- Conditional Expected Value:  $E_P(f \mid e) = \sum_{\omega \in \Omega} f(\omega) * P(\omega \mid e)$
- Independence:  $X$  is *independent* of  $Y$  if  $P(X \mid YZ) = P(X \mid Z)$
- Discounted Reward:  $V_1 = r_1 + \gamma r_2 + \gamma^2 r_3 + \dots + \gamma^{n-1} r_n$ , which implies  $V_i = r_i + \gamma V_{i+1}$ . Note that  $n$  may be  $\infty$ .  $V$  converges in most cases where  $\gamma < 1$
- Value of a Policy:  $V^\Pi(s)$  (which means value of  $\Pi$  at  $s$ )  $\implies$  what reward you can expect if you start at  $s$  and follow policy  $\Pi$

- **Belief Networks and Markov Chains**

- *Stationary Markov Chain*: dynamics are the same for every arrow
- *Ergodic Markov Chain*: any state has a non-zero probability of being reached from any state
- *Periodic Markov Chain*: states can only be reached during regularly occurring times
- All Markov Chains have an equivalent stationary chain
- Belief Networks only have states
- Markov Chains and Hidden Markov Models are special types of belief networks

- **Utility Theory**

- *completeness*: every outcome can be compared to all others
- *lottery*: probability distribution over outcomes
- *decomposability*: no preference between lotteries with equivalent outcomes and probability distributions
- *substitutability*: no preference between decomposable lotteries where an outcome in one lottery is different from its corresponding outcome in the other lottery if both outcomes have equivalent utilities
- *money pump*:  $a \succ a, b \succ c, c \succ a$  (cycle that breaks transitivity)
- Utility  $\sim$  Expected Value, math-wise