

# Probabilities and Expectations

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# Probabilities

- Probability is a measure of uncertainty
- Being uncertain is much more than “I don’t know”
- We can make informed guesses about uncertain events

# Intelligent Systems

- An intelligent system maximizes its “chances” of success
- Intelligent systems create a favorable future
- Probabilities and expectations are tools for reasoning about uncertain future events

# Example: Monty Hall Problem

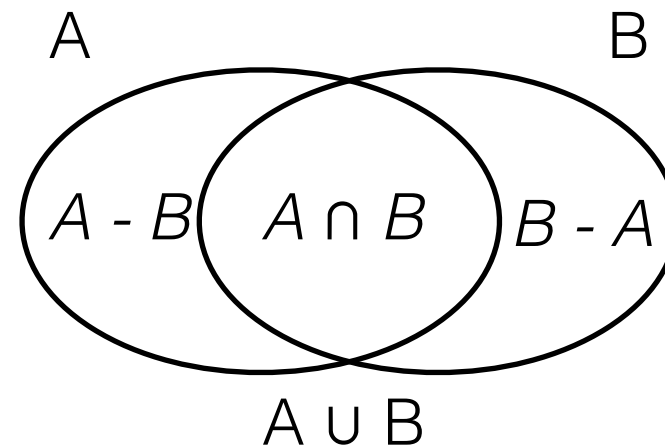


# Sets

- A set is a collection of distinct of objects
- $S = \{\text{head}, \text{tail}\}$
- **Element**:  $\text{head} \in S, \text{tail} \in S$
- **Subsets**:  $\{\text{head}\} \subset S, S \subset S, \phi = \{\} \subset S$
- Power set:  $2^S = \{\{\text{head}\}, \{\text{tail}\}, S, \phi\}$
- **Union**:  $A = \{1, 2\}, B = \{2, 3\}, A \cup B = \{1, 2, 3\}$
- **Intersection**:  $A = \{1, 2\}, B = \{2, 3\}, A \cap B = \{2\}$
- A **complement** set of A in B:  $A = \{1, 2\}, B = \{2, 3\}, B - A = \{3\}$
- The **Cartesian product** of two sets:  $A = \{1, 2\}, B = \{a, b\}, A \times B = \{(1,a), (1,b), (2,a), (2,b)\}$

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# Functions

- A function is a map from one set to another
- $S = \{\text{head}, \text{tail}\}$ ,  $V = \{+1, -1\}$ ,  $f: S \rightarrow V$
- $f(\text{head}) = 1$ ,  $f(\text{tail}) = -1$
- $f(\text{head}) = 1$  ✘
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# Sample space & Events

- An **experiment** is a repeatable process
- A **sample space** is the set of all possible outcomes of an experiment

Dice-rolling:  $S = \{1, 2, 3, 4, 5, 6\}$

- An **event** is a subset of a sample space  
the event of even number appearing:  $\{2, 4, 6\}$



# Probabilities

- **Probability** is a function that maps all possible events from a sample space to a number

$$\mathbf{Pr}: 2^S \rightarrow [0, 1]$$

- Probability is a measure of uncertain events
- **Non-negativity**: A probability is always non-negative:

$$0 \leq \mathbf{Pr}(A) \leq 1$$

- **Normalization**: Addition of probabilities of all individual outcomes of a sample space is always 1

$$\sum_{e \in S} \mathbf{Pr}(e) = 1$$

**Probability distribution** defines how the probability is distributed among the outcomes

- **Additivity**:  $\mathbf{Pr}(A \cup B) = \mathbf{Pr}(A) + \mathbf{Pr}(B)$ ;  $A \cap B = \phi$

# Random Variables

- Random variables are a convenient way to express events
- A random variable is a function that maps a sample space to a real number

$$X : S \rightarrow \mathbb{R}$$

- Dice-rolling experiment:  $[X < 4]$  stands for

$$\{ \omega \in S : X(\omega) < 4 \} = \{ 1, 2, 3 \}$$

# Examples

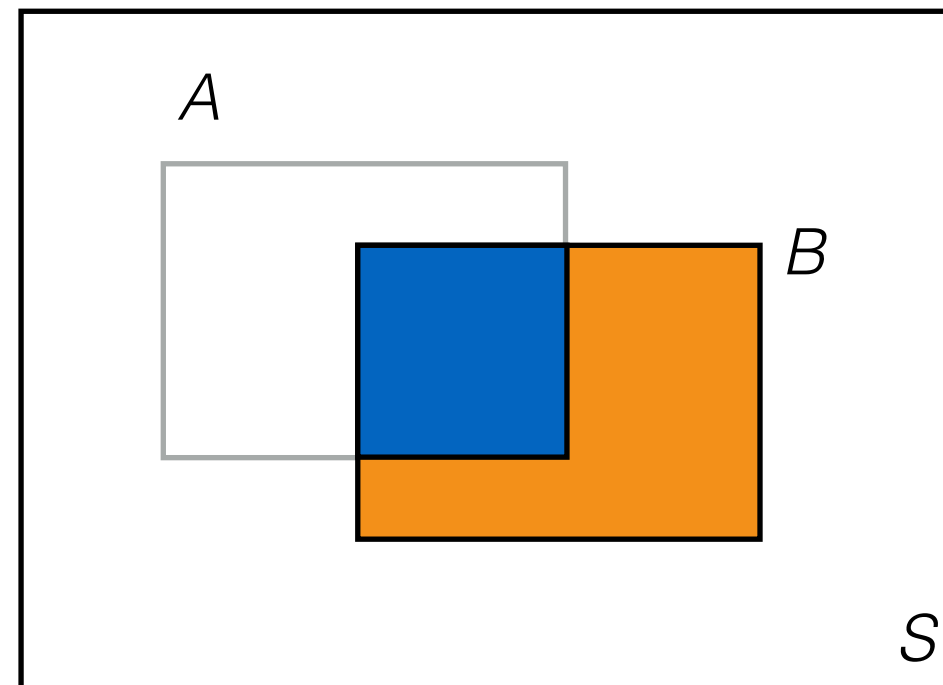
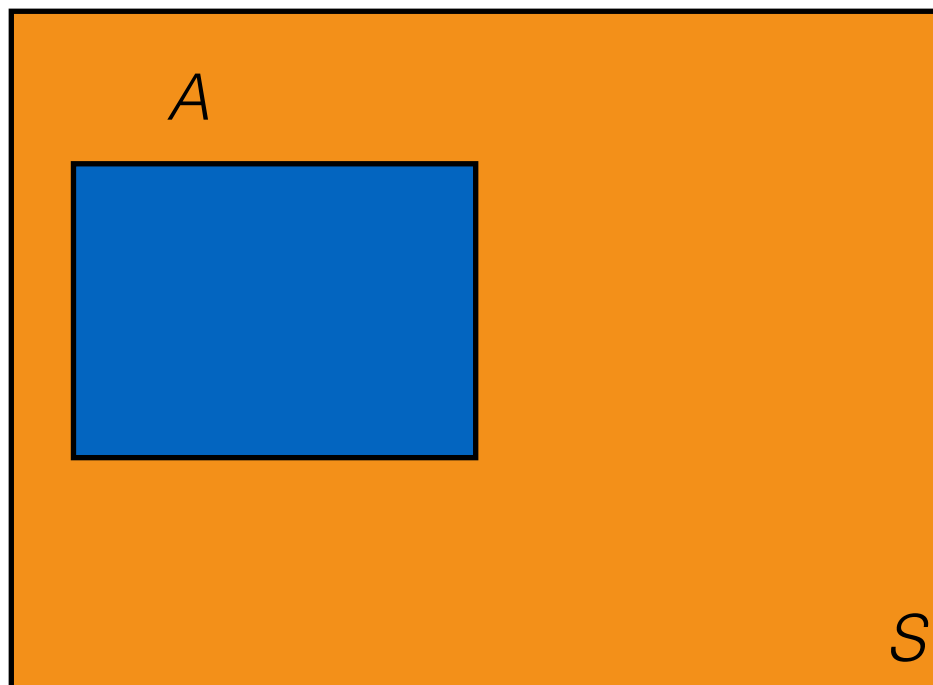
- In the dice-rolling experiment, what is the probability that the outcome is a prime number?
  - Sample space:  $S = \{1, 2, 3, 4, 5, 6\}$
  - Distribution:  $1/6$  for each outcome
  - Event in question:  $E = \{2, 3, 5\}$
  - **$\Pr(E) = \Pr(\{2, 3, 5\}) = \Pr(2) + \Pr(3) + \Pr(5) = 3 \times 1/6 = 1/2.$**

# Examples

- If we roll two dices together, what is the probability that sum of the two numbers is greater than 2?
  - Sample space:  $S = \{1, \dots, 6\} \times \{1, \dots, 6\}$  ; (compound experiment)  
 $= \{(1,1), (1,2), \dots, (1,6),$   
 $(2,1), (2,2), \dots, (2,6), \dots,$   
 $(6,1), (6,2), \dots, (6,6)\}$
  - Distribution:  $1/36$  for each outcome
  - Event in question:  $E = \{(1,2), \dots, (6,6)\}$
  - Define a random variable to be the sum of the two numbers:  $X(a, b) = a + b$
  - Event:  $E = [X > 2]$
  - $1 = \mathbf{Pr}(S) = \mathbf{Pr}([X=2] \cup [X > 2]) = \mathbf{Pr}([X=2]) + \mathbf{Pr}([X > 2])$

# Conditional Probabilities

- A **conditional probability** is a measure of an uncertain event when we know that another event has occurred
- Definition:  $\Pr(A | B) = \Pr(A \cap B) / \Pr(B) \neq \Pr(A)$

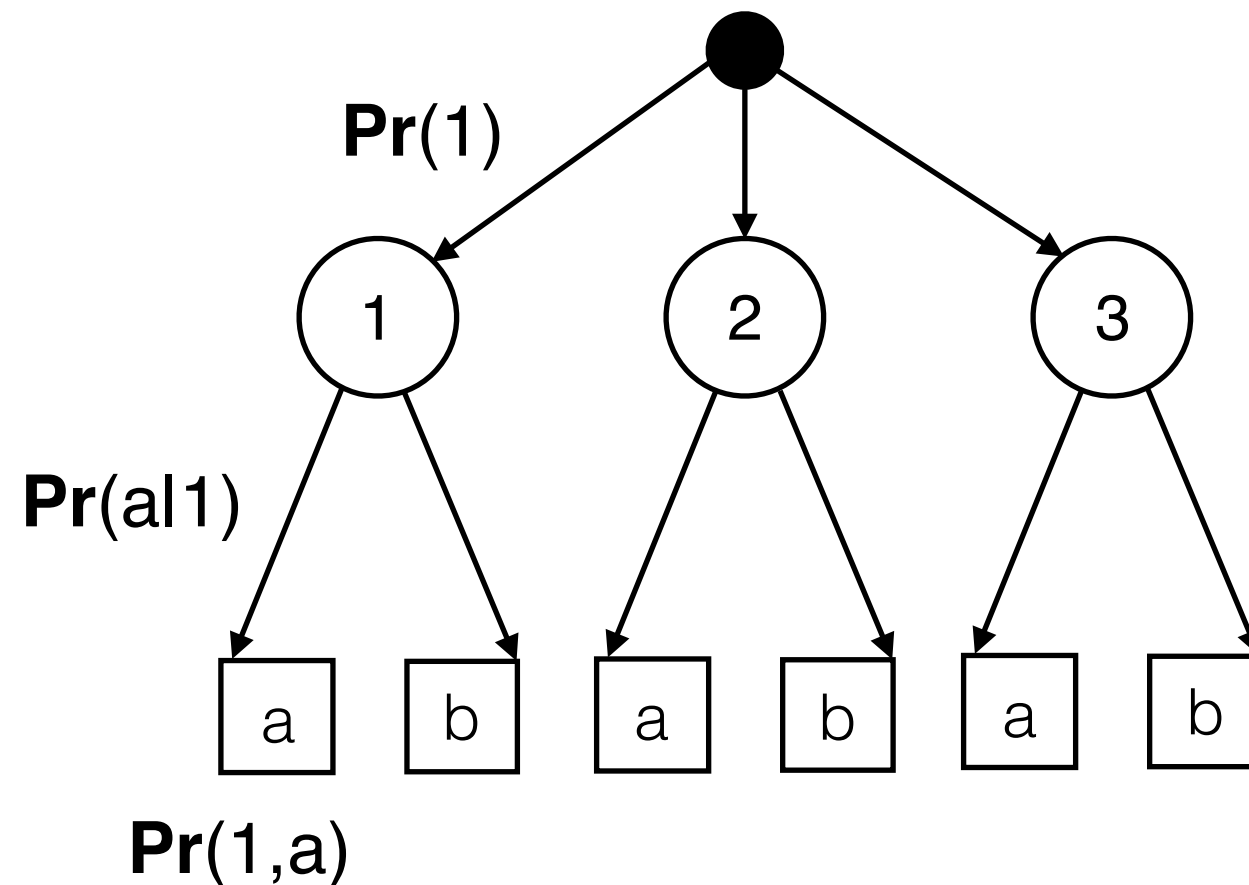


# Examples

- In the dice-rolling experiment, if a prime number appears, what is the probability that it is even?
  - Sample space:  $S = \{1, 2, 3, 4, 5, 6\}$
  - Distribution:  $1/6$  for each outcome
  - Events:  $A = \{2, 4, 6\}$ ,  $B = \{2, 3, 5\}$
- **$\Pr(A \mid B) = \Pr(A \cap B) / \Pr(B)$** 
  - $= \Pr(\{2, 4, 6\} \cap \{2, 3, 5\}) / \Pr(\{2, 3, 5\})$**
  - $= \Pr(2) / \Pr(\{2, 3, 5\})$**
  - $= (1/6)/(1/2) = 1/3$**

# Probability Trees

- Often in compound experiments, outcome of one depends on the other (unlike the double dice-rolling experiment)
- It is convenient in that case to calculate probabilities using probability trees



# Examples: Monty Hall Problem

- In the Monty Hall problem, we chose the 1st door and the Host revealed 2nd.

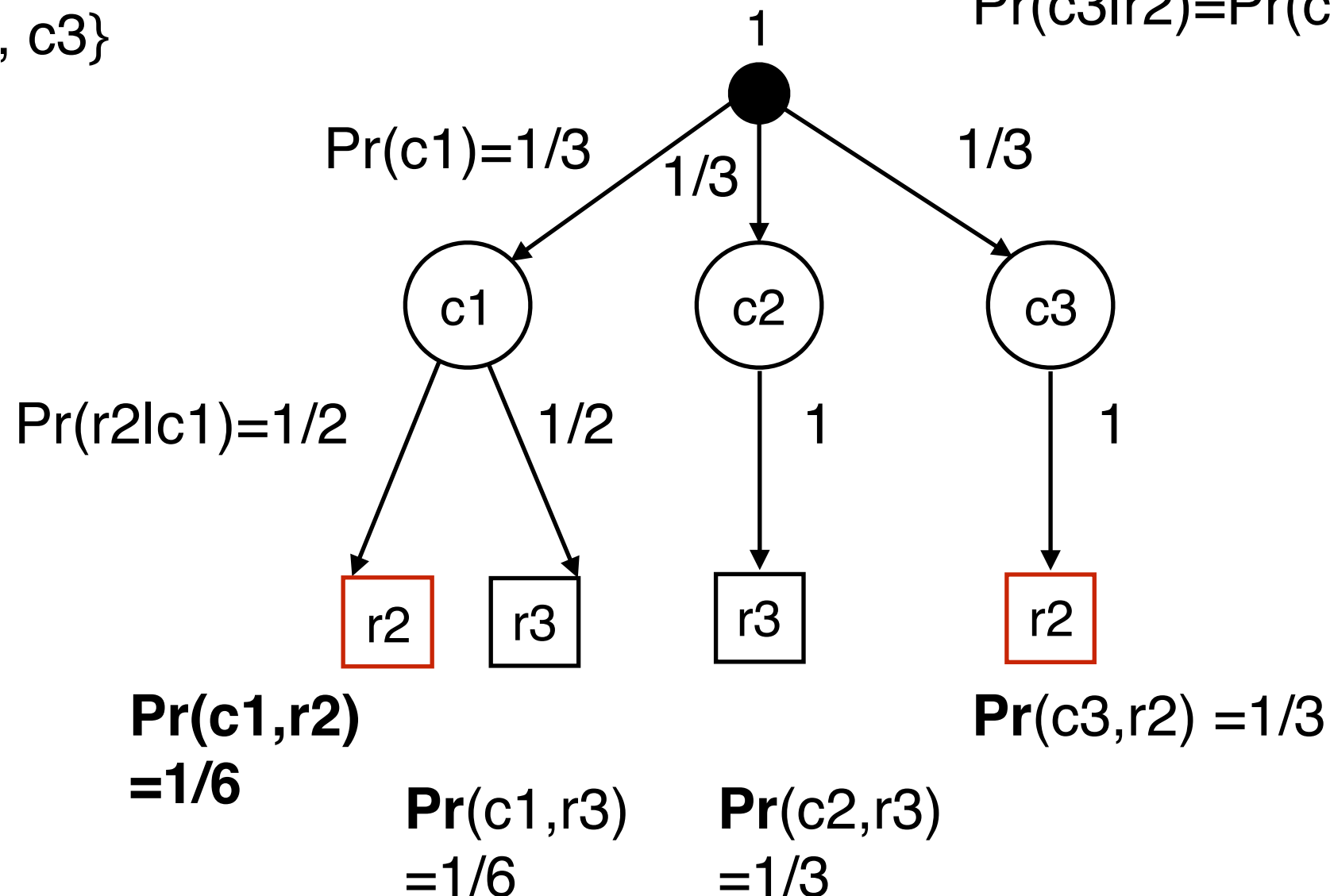
$$S_1 = \{c1, c2, c3\}$$

$$S_2 = \{r2, r3\}$$

$$S = S_1 \times S_2$$

$$\Pr(c1|r2) = \Pr(c1, r2) / \Pr(r2)$$

$$\Pr(c3|r2) = \Pr(c3, r2) / \Pr(r2)$$





# Examples: Monty Hall Problem

- In the Monty Hall problem, we chose the 1st door and the Host revealed 3rd.

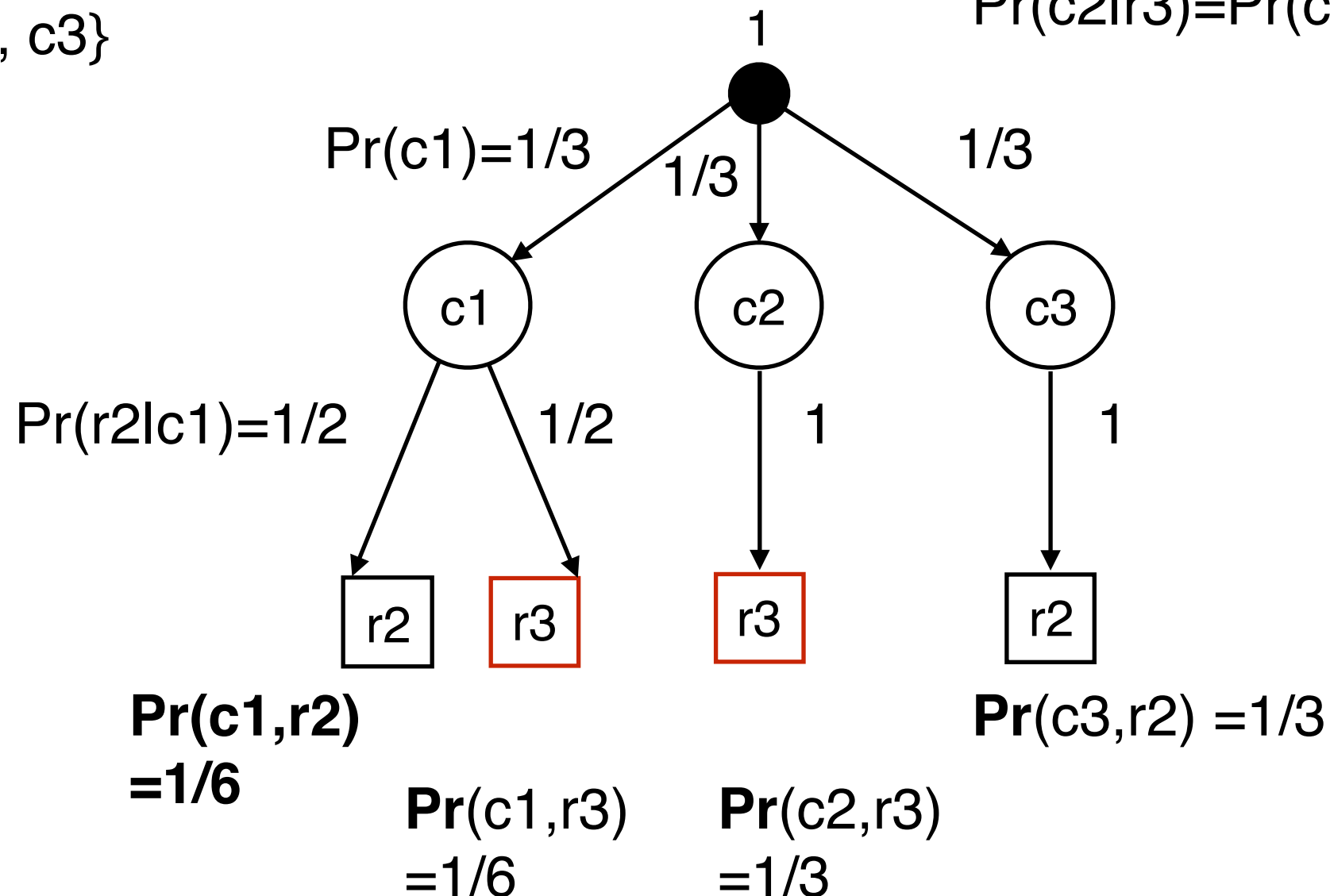
$$S_1 = \{c1, c2, c3\}$$

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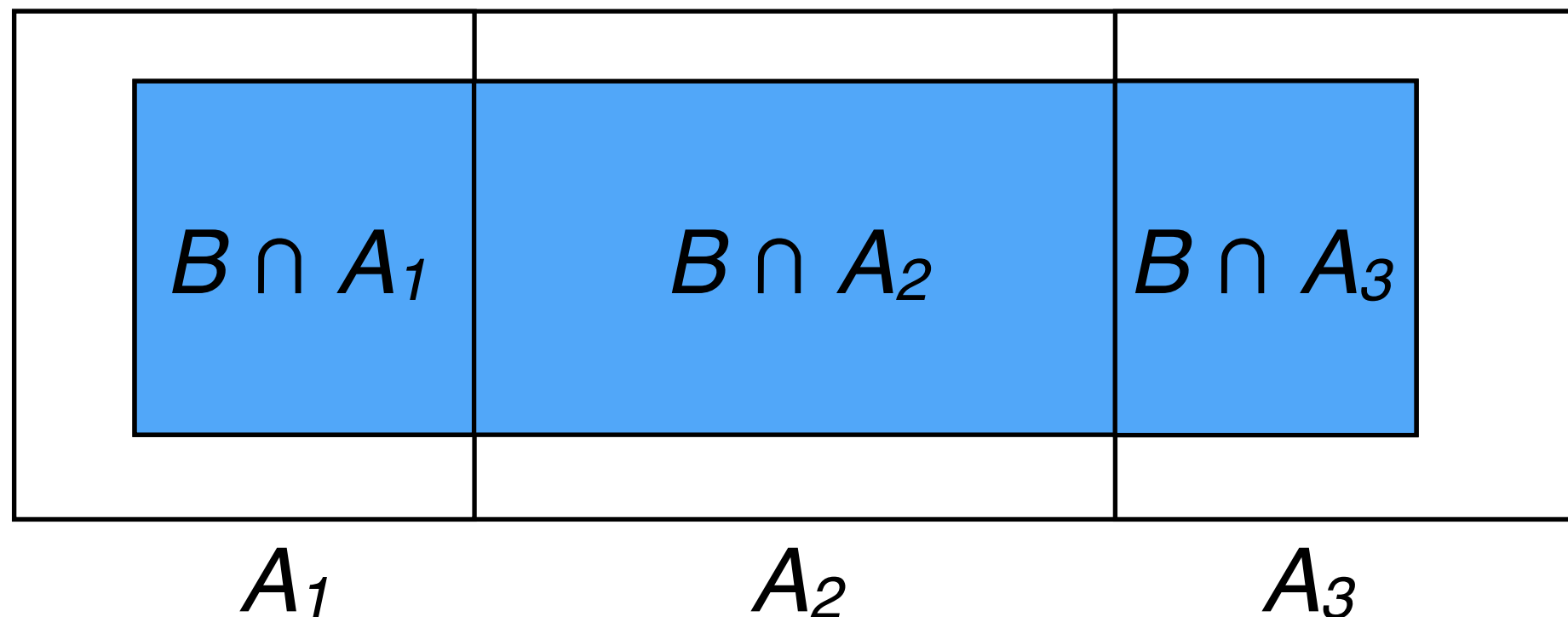
$$\Pr(c2|r3) = \Pr(c2, r3) / \Pr(r3)$$



# Law of Total Probability

$$\begin{aligned}\Pr(B) &= \sum_j \Pr(B \cap A_j) \\ &= \sum_j \Pr(B \mid A_j) \Pr(A_j)\end{aligned}$$

$$A_i \cap A_j = \phi, \quad i \neq j, \quad \bigcup_i A_i = S$$



# Bayes Theorem

$$\Pr(A_1|B) = \frac{\Pr(A_1 \cap B)}{\Pr(B)} = \frac{\Pr(B|A_1)\Pr(A_1)}{\sum_j \Pr(B|A_j)\Pr(A_j)}$$

# Examples

- A drug test returns positive for a drug user 99% of the time and returns negative for a non-user 95% of the time. Suppose that 1% of the population uses drug. Then what is the probability that an individual is a drug user given that she tests positive?
- Sample space: { user+, user-, nonuser+, nonuser- }

- $\Pr(+|user) = 0.99$

- $\Pr(-|nonuser) = 0.95$

- $\Pr(user) = 0.01$

- $\Pr(user|+) = ?$

$$\begin{aligned}\Pr(user|+) &= \frac{\Pr(+|user)\Pr(user)}{\Pr(+|user)\Pr(user) + \Pr(+|nonuser)\Pr(nonuser)} \\ &= \frac{0.99 \times 0.01}{0.99 \times 0.01 + (1 - \Pr(-|nonuser)) \times (1 - \Pr(user))} \\ &= \frac{0.0099}{0.0099 + (1 - 0.95) \times (1 - 0.01)} \\ &= \frac{0.0099}{0.0099 + 0.05 \times 0.99} \\ &= \frac{0.0099}{0.0099 + 0.0495} \\ &\approx 0.167.\end{aligned}$$

# Expectations & Conditional Expectations

- An expected value of a random variable is a weighted average of possible outcomes, where the weights are the probabilities of those outcomes

$$\mathbf{E}[X] = \sum_{x \in S} x \mathbf{Pr}(X=x)$$

- An expected value of a random variable conditional on another event is a weighted average of possible outcomes, where the weights are the conditional probabilities of those outcomes given the event

$$\mathbf{E}[X \mid Y=y] = \sum_{x \in S} x \mathbf{Pr}(X=x \mid Y=y)$$

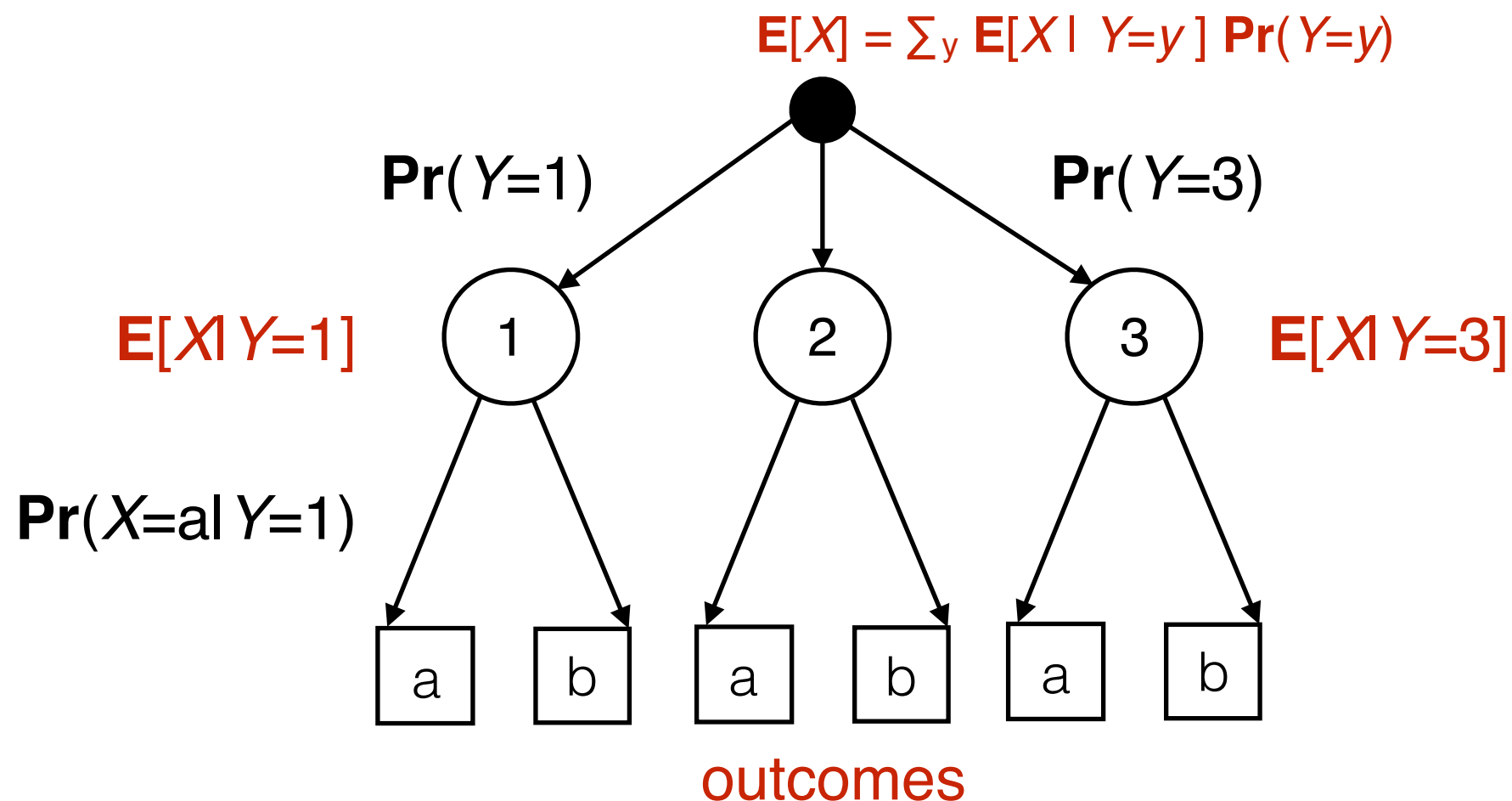
- Law of total expectation:  $\mathbf{E}[X] = \sum_y \mathbf{E}[X \mid Y=y] \mathbf{Pr}(Y=y)$

# Examples

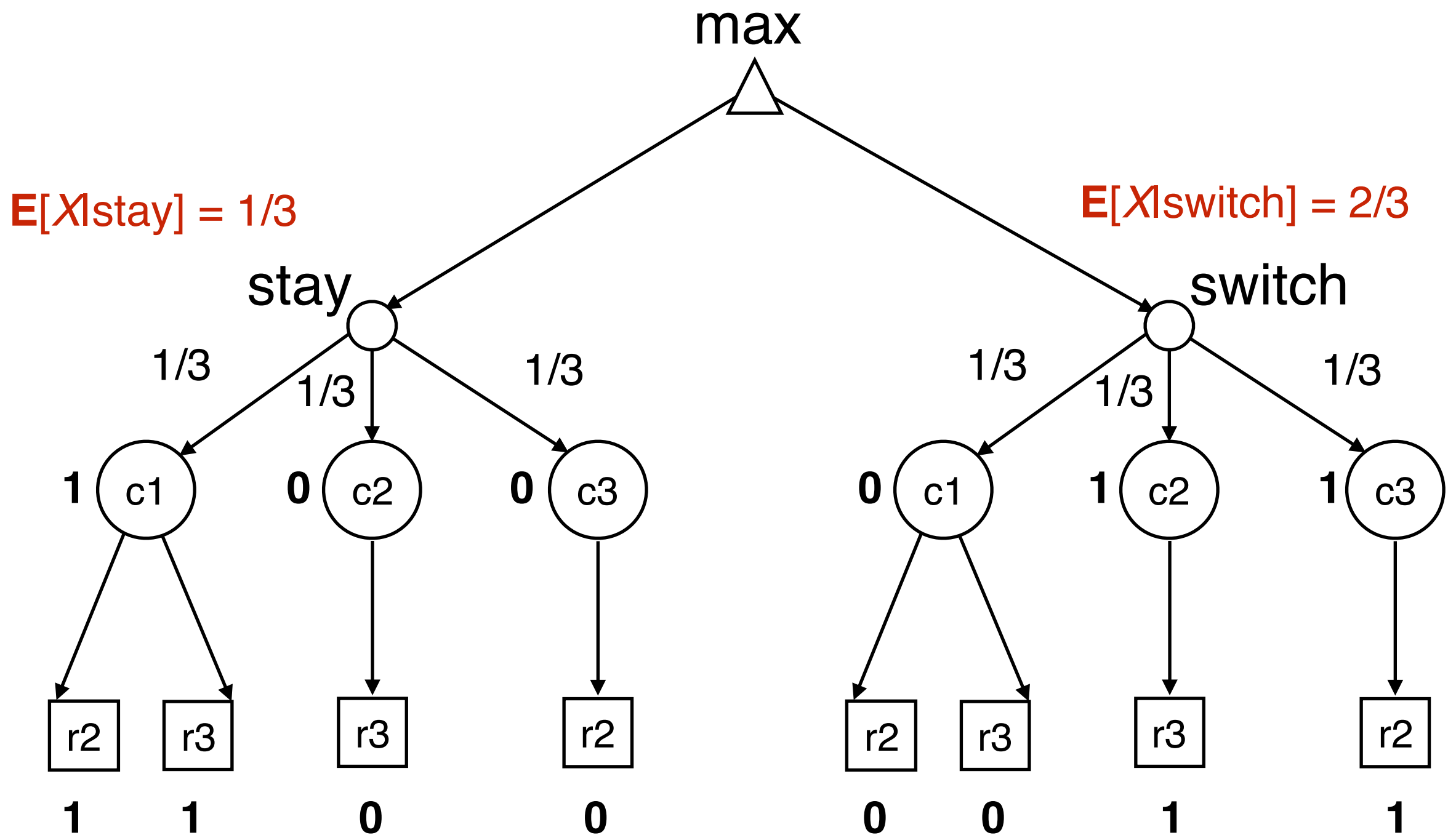
- In a certain lottery, it is 0.01% likely to win, and the prize is 1000 dollars. The ticket price is 10 dollars. What is the expected monetary gain?
- Sample space:  $S = \{ 990, -10 \}$
- Expected value: 
$$\begin{aligned} \mathbf{E}[X] &= 990 \mathbf{Pr}(X=990) + (-10) \mathbf{Pr}(X=-10) \\ &= 990 * 0.0001 + (-10) * 0.9999 \\ &= 0.099 - 9.999 \\ &= -9.9. \end{aligned}$$

# Expectation Trees

- Often in compound experiments, outcome of one depends on the other (unlike the double dice-rolling experiment)
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# Examples: Monty Hall Problem





# Two Ways of Calculation

- Model-based calculation
  - We know the probability model
- Model-free or empirical estimation
  - Learn from experience!

# Concluding Remarks

- Probabilities and expectations let us make favorable choices
- There are two ways of calculating them
- If we know the model, we can make intelligent systems by feeding them the model and automating the calculation
- If we do not know the model, we can let the intelligent try things out!
- In either case, intelligent systems can make favorable choices by dealing with probabilities and expectations