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#### **Games with Missing Information: Modelling**

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## Previously ...

- **Monte Carlo Tree Search** uses random playouts to evaluate moves and keeps statistics on which moves led to which payoffs how many times.
- A selection policy balances exploitation and exploration.
- **UCT** is an effective selection policy that applies UCB1 to trees.
- A **playout policy** steers playout simulations towards realistic play.
- MCTS and deep reinforcement learning led to expert-level Go programs.





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Example: The Monty Hall Problem

**Extensive-Form Games** 

Bayes' Theorem

Preview: Simplified Poker





# **Motivation: Missing Information**

- So far, we have considered games with perfect information:
- In every state, all players know the full history of play so far, i.e. they know their (joint) position in the game tree.
- However, e.g. in card games, players typically do not know the cards of opponents.
- This form of incomplete knowledge can be formalised by sets of indistinguishable nodes in the game tree, typically called information sets.
- In this context, we also add another element to games: chance.
- This is modelled via moves by nature and can be used to formalise dealing cards or throwing dice.
- We will see that this also allows us to model games with incomplete information, where e.g. some of the payoffs may be uncertain.
- In principle, however, chance and imperfect information are unrelated and we could model either without the other.







#### **Course Evaluation: Lecture**

#### 15min to fill out this form:



https://befragung.zqa.tu-dresden.de/uz/de/sl/uGo82Q2GBW5i



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#### **Example: The Monty Hall Problem**



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### **Example: The Monty Hall Problem**



#### **The Monty Hall Problem**

A game show participant (Guest) is shown three doors behind which there are prizes. Behind one door, there is an expensive car, behind each of the other doors there is a goat. (The participant prefers the car over a goat.) The Guest is asked to Choose one of the doors. The game show Host (the other player) now opens one of the remaining doors that has a goat behind it. The Guest then gets their final move: Stay with the door they initially picked, or Switch to the other door. What should the participant do?





### The Monty Hall Problem: Game Tree Sketch





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## The Monty Hall Problem: Analysis

- Each of the possible states  $s_1, s_2, s_3$  after Nature's move has probability  $\frac{1}{3}$ .
- For each of these states, the ensuing game is symmetric.
- If Guest chooses their door uniformly at random, then:
  - With probability  $\frac{1}{3}$ , their initial guess is correct; thus Switch has a payoff of 0.
  - With probability  $\frac{2}{3}$ , their initial guess is wrong; thus Switch has a payoff of 100.
- Thus in each  $s_i$ , Switch has an expected payoff of  $\frac{1}{3} \cdot 0 + \frac{2}{3} \cdot 100$ .
- The overall payoff of Switch is thus

$$u_{\text{Guest}}(\text{Switch}) = 3 \cdot \frac{1}{3} \cdot \left(\frac{1}{3} \cdot 0 + \frac{2}{3} \cdot 100\right) = 66\frac{2}{3}$$

- Likewise, the overall payoff of Stay is obtained as  $u_{Guest}(Stay) = \frac{1}{3} \cdot 100 = 33\frac{1}{3}$ .
- Therefore, a rational player should always choose Switch over Stay.





#### **Extensive-Form Games**



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# **Missing Information: Formalisation**

#### Definition

#### An extensive-form game consists of the following:

- 1. A set  $P = \{1, ..., n\}$  of at least two players, and possibly Nature.
- 2. An *n* + 1-tuple  $\mathbf{M} = (M_1, \dots, M_n, M_{\text{Nature}})$  of sets  $M_i$  of **moves** for all players.
- 3. A set *H* of **histories**, sequences of moves  $m_j \in M_1 \cup \ldots \cup M_n \cup M_{Nature}$ .
- 4. A subset  $Z \subseteq H$  of **terminal** histories.
- 5. A partition  $\mathcal{I}_1 \cup \ldots \cup \mathcal{I}_k = H \setminus Z$  of non-terminal histories into **information sets** such that for all  $1 \le j \le k$ , all  $h_1, h_2 \in \mathcal{I}_j$  have the same legal moves.
- 6. A **player function**  $p: \{1, ..., k\} \rightarrow P \cup \{\text{Nature}\}$  (stating whose turn it is).
- 7. An *n*-tuple  $\mathbf{u} = (u_1, \ldots, u_n)$  of utility functions  $u_i: Z \to \mathbb{R}$ .

Starting with the **empty history** [], in each history  $h = [m_1, ..., m_k] \in H \setminus Z$ , player i = p(h) chooses a move  $m \in M_i$ , leading to the history  $[m_1, ..., m_k, m]$ . Each  $\mathcal{I}_j$  with  $p(\mathcal{I}_j) = Nature$  has a probability distribution on possible moves.





### **Information Sets: Remarks**

Intuition of information sets: The player (whose turn it is) does not have the information to distinguish between states in the set (but from other sets).

- $\mathfrak{I} = {\mathfrak{I}_1, \dots, \mathfrak{I}_k}$  being a **partition** means that:
  - for all  $1 \le j \le k$ , we have  $\mathfrak{I}_j \ne \emptyset$ ,
  - $\mathfrak{I}_1 \cup \ldots \cup \mathfrak{I}_k = H \setminus Z$ , and
  - for all  $1 \leq j, \ell \leq k$ , we have  $\mathfrak{I}_j \cap \mathfrak{I}_\ell = \emptyset$ .
- Thus every  $h \in H \setminus Z$  belongs to exactly one information set  $\mathfrak{I}_j \in \mathfrak{I}$ .
- For all  $1 \le j \le k$  and  $h \in \mathcal{I}_j$ , we denote  $p(h) := p(\mathcal{I}_j)$ .
- $\mathfrak{I}$  can also be represented by an **equivalence relation**  $\sim^{G}$ , where for any  $h_1, h_2 \in H$ , we have  $h_1 \sim^{G} h_2$  iff there is a  $\mathfrak{I}_j \in \mathfrak{I}$  such that  $h_1, h_2 \in \mathfrak{I}_j$ .
- We graphically represent  $\sim^{G}$  in game trees via dashed edges -----.

#### Battleship

The initial placement of ships is private to the players and can be modelled via information sets. Some information may later be disclosed through hits.







### **Information Sets: Example**

#### The Monty Hall Problem

- The (true) initial state is represented by the information set  $\mathfrak{I}_0 = \{[]\}$ .
- The (seemingly) initial state for Guest is given by the information set  $\label{eq:car1} \Im_1 = \{[\texttt{Car1}], [\texttt{Car2}], [\texttt{Car3}]\}\,.$
- For each possible (initial) choice of door for Guest, there is one set:

$$\begin{split} & \mathbb{J}_{\text{Choose1}} = \{ [\text{Car1}, \text{Choose1}], [\text{Car2}, \text{Choose1}], [\text{Car3}, \text{Choose1}] \} \\ & \mathbb{J}_{\text{Choose2}} = \{ [\text{Car1}, \text{Choose2}], [\text{Car2}, \text{Choose2}], [\text{Car3}, \text{Choose2}] \} \\ & \mathbb{J}_{\text{Choose3}} = \{ [\text{Car1}, \text{Choose3}], [\text{Car2}, \text{Choose3}], [\text{Car3}, \text{Choose3}] \} \end{split}$$

• Some information is disclosed by the host opening a door:

 $\mathbb{J}_{[Choose1,0pen2]} = \{ [Car1, Choose1], [Car3, Choose1] \}$ 





### Perfect Recall

#### Definition

Let  $G = (P, \mathbf{M}, H, \mathcal{I}, p, \mathbf{u})$  be an extensive-form game.

For every player *i* ∈ *P* and history *h* ∈ *H*, define the sequence *h<sub>i</sub>* of pairs (J<sub>j</sub>, *m*) for J<sub>j</sub> ∈ J and *m* ∈ *M<sub>i</sub>* by induction:

$$[]_i := [] \quad \text{and} \quad [h; m]_i := \begin{cases} [h_i; (\mathfrak{I}_h, m)] & \text{if } p(h) = i, \\ h_i & \text{otherwise.} \end{cases}$$

where for  $h = [m_1, ..., m_k]$  we denote  $[h; m] := [m_1, ..., m_k, m]$ .

- Player  $i \in P$  has **perfect recall** in *G* iff for all  $\mathfrak{I}_j \in \mathfrak{I}$ , for every  $h, h' \in \mathfrak{I}_j$ , it holds that  $h_i = h'_i$ .
- *G* has **perfect recall** iff every player  $i \in P$  has perfect recall in *G*.
- $h_i$  extracts all decision points and decisions (of player *i*) from history *h*.
- $h_i = h'_i$  means that *i* made the same moves *in the same information sets*.
- With perfect recall, players remember their trajectory through the game.







#### **Perfect Recall: Examples (1)**



This game does not have perfect recall:

- Denote  $\mathbb{J}_0=\{[]\} \text{ and } \mathbb{J}_1=\{[L]\,,[R]\}.$
- We have  $[L]\in {\mathbb J}_1$  and  $[R]\in {\mathbb J}_1,$  but:

$$[L]_1 = [(\mathfrak{I}_0, L)] \neq [(\mathfrak{I}_0, R)] = [R]_1$$





#### Perfect Recall: Examples (2)



This game does not have perfect recall:

- Denote  $J_0 = \{[]\}, J_1 = \{[L]\}, J_2 = \{[R]\}, J_3 = \{[L, A], [R, A]\}, J_4 = \{[L, B], [R, B]\}.$
- Then  $[L, A]_2 = [(\mathcal{I}_1, A)] \neq [(\mathcal{I}_2, A)] = [R, A]_2$ .





### **Perfect Recall: Examples (3)**



This game has perfect recall: e.g.  $h \in \{h_2, h_3\}$  implies  $h_{Candidate2} = []$ .





## **Strategic Games and Imperfect Information**

- Uncertainty induced by simultaneous moves can be modelled in extensive-form games (that seem to be sequential by definition).
- Main idea: Sequentialise moves, model uncertainty in information sets.

Example: Recall the game penalties. One extensive-form variant is:





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### Chance Nodes (Moves by Nature)

Intuition of chance nodes: Something happens that is controlled by an entity with no strategic interest in the game's outcome.

#### Examples

- In card games, Nature controls the dealer's shuffling the cards.
- In games involving dice, Nature controls the dice throws.
- Probability distributions model uncertainty about effects of such actions.
- We typically use uniform distributions over possible atomic results.
- $\rightsquigarrow$  We need some (more) probability theory to analyse games with chance  $\dots$







#### **Bayes' Theorem**



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### **Probabilities**

#### Recall

- A probability space is a finite set  $\mathcal{E} = \{e_1, \dots, e_k\}$  of atomic events.
- A **probability distribution** is a mapping  $P: \mathcal{E} \to [0, 1]$ , where atomic event  $e_i$  occurs with probability  $P(e_i)$  and we have  $\sum_{i=1}^{k} P(e_i) = 1$ .
- An **event**  $E \subseteq \mathcal{E}$  has (total) probability  $P(E) = \sum_{e \in E} P(e)$ .
- For all events  $A, B \subseteq \mathcal{E}$  we have the following:
  - 1.  $0 \le P(A) \le 1$  with  $P(\emptyset) = 0$  and  $P(\mathcal{E}) = 1$ .
  - 2.  $P(\overline{A}) = 1 P(A)$  where  $\overline{A} := \mathcal{E} \setminus A$  is the event **complementary** to A.
  - 3.  $P(A \cup B) = P(A) + P(B) P(A \cap B)$ .

#### Example

# If all events $e_i \in \mathcal{E}$ have the same probability $\frac{1}{|\mathcal{E}|}$ , we have a **uniform distribution**.





## **Conditional Probabilities**

#### Definition

Let A and B be events with P(B) > 0.

1. The **conditional probability** for *A* to occur under the condition of *B* occurring is

$$P(A \mid B) := \frac{P(A \cap B)}{P(B)}$$

2. Events A and B are **independent** iff

 $P(A \cap B) = P(A) \cdot P(B)$ 

That events A and B are independent is equivalently characterised by each of:

- $P(A \mid B) = P(A)$
- $P(A | B) = P(A | \overline{B})$
- P(B|A) = P(B)
- $P(B|A) = P(B|\overline{A})$





### **Bayes' Theorem**

#### Theorem (Bayes)

1. If A and B are two events with P(A) > 0 and P(B) > 0, then

 $P(A) \cdot P(B | A) = P(B) \cdot P(A | B)$ 

2. If *A* and  $B_1, B_2, ..., B_\ell$  are events with P(A) > 0 and  $P(B_i) > 0$  for all  $1 \le i \le \ell$ , where  $\bigcup_{i=1}^{\ell} B_i = \mathcal{E}$  is a partition of  $\mathcal{E}$ , then for every  $1 \le i \le \ell$ :

$$P(B_i | A) = \frac{P(A | B_i) \cdot P(B_i)}{\sum_{j=1}^{\ell} (P(A | B_j) \cdot P(B_j))} = \frac{P(A | B_i) \cdot P(B_i)}{P(A)}$$

#### In the second item of the theorem, the law of total probability is used:

$$P(A) = \sum_{j=1}^{\ell} P(B_j \cap A) = \sum_{j=1}^{\ell} \left( P(A \mid B_j) \cdot P(B_j) \right)$$

Note that  $P(A) = P(\mathcal{E} \cap A) = P((\bigcup_{j=1}^{\ell} B_j) \cap A) = P((\bigcup_{j=1}^{\ell} (B_j \cap A))) = \sum_{j=1}^{\ell} P(B_j \cap A).$ 



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# Solving the Monty Hall Problem (1)

Consider the following events:

A: The Guest wins the car.

 $B_1$ : The Guest initially chooses a goat door.

B<sub>2</sub>: The Guest initially chooses the car door.

- If Guest chooses uniformly at random, then  $P(B_1) = \frac{2}{3}$  and  $P(B_2) = \frac{1}{3}$ .
- Since every door has exactly one object,  $B_1$  and  $B_2$  are complementary, and the law of total probability yields  $P(A) = P(A|B_1) \cdot P(B_1) + P(A|B_2) \cdot P(B_2)$ .
- If Guest plays Stay, then clearly  $P(A|B_1) = 0$  and  $P(A|B_2) = 1$ , whence

$$P(A) = P(A | B_1) \cdot P(B_1) + P(A | B_2) \cdot P(B_2) = 0 \cdot \frac{2}{3} + 1 \cdot \frac{1}{3} = \frac{1}{3}$$

• If Guest plays Switch, then  $P(A|B_1) = 1$  and  $P(A|B_2) = 0$ , thus

$$P(A) = P(A | B_1) \cdot P(B_1) + P(A | B_2) \cdot P(B_2) = 1 \cdot \frac{2}{3} + 0 \cdot \frac{1}{3} = \frac{2}{3}$$





## Solving the Monty Hall Problem (2)

Consider the following events:

A: The Guest wins the car.

B: The Guest initially chooses a goat door.

If the Guest plays Switch, then

- $P(B) = \frac{2}{3}$  as before,
- P(A|B) = 1 (initially choosing a goat door and switching win the car), and
- P(B|A) = 1 (initially choosing a goat door is the only way a Switch player can win the car).

According to Bayes' Theorem, we thus obtain

$$P(A) = \frac{P(A | B) \cdot P(B)}{P(B | A)} = \frac{2}{3}$$





#### **Preview: Simplified Poker**



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### **Example: Simplified Poker**

#### **Binmore's Simplified Poker**

- Two players, Ann and Bob, each put \$1 into a jackpot.
- They then draw one card from a deck of three cards: {1, 2, 3}.
- Ann can either check (pass on), or raise (put another \$1 into the jackpot).
- Next, Bob responds:
  - If Ann has checked, then Bob must call, that is, a showdown happens: Both players show their cards and the player with the higher (number) card receives the jackpot.
  - If Ann has raised, then Bob can decide between fold (withdraw from the game and let Ann get the jackpot) or call (put another \$1 into the jackpot and then have a showdown).







# **Simplified Poker: Preliminary Analysis**

Nature shuffles and deals the cards. There are six possible outcomes:



- If Ann draws a 3, she will raise; if Bob draws a 1, he will fold.
- If Bob draws a 3, he will call; if Ann draws a 2, she will check: Were she to raise, she would lose 2 if Bob has a 3 (as he would call), but still only win 1 if Bob has a 1 (as he would fold then).

#### What happens in the two remaining cases?

- 1. Should Ann raise (i.e. bluff) if she has a 1?
- 2. Should Bob call (the bluff) if he has a 2?





### Conclusion

#### Summary

- In **complete information** games, players know the rules, possible outcomes and each other's preferences over outcomes.
- In **perfect information** games, moves are sequential and all players know all previous moves.
- In **extensive-form** games, information is not necessarily complete or perfect.
- Uncertainty of players (due to missing information) can be modelled by **information sets** and **chance nodes** (moves by Nature).
- Bayes' Theorem shows how to compute with conditional probabilities.
- The **law of total probability** relates marginal to conditional probabilities.

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