### Algorithmic game theory

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Today, we describe strategies for such games and how to compute Nash equilibria.

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- In perfect-information games all information sets are singletons. Otherwise, we have an imperfect-information game where players have only partial knowledge of the states that they are in.

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• An example of an imperfect-information game in extensive form (part (a)) and its normal-form (part (b)).



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  - Unlike in mixed strategy, here a player might play different moves in different encounters of *h*.

• An example of a perfect-information game in extensive form (part (a)) and its normal-form (part (b)).

(b)



	(C, E)	(C, F)	(D, E)	(D,F)
(A,G)	(3,8)	(3,8)	(8,3)	(8,3)
(A, H)	(3,8)	(3,8)	(8,3)	(8,3)
(B,G)	(5,5)	(2,10)	(5,5)	(2,10)
(B,H)	(5,5)	(1,0)	(5,5)	(1,0)

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- The mixed strategy (<sup>3</sup>/<sub>5</sub>(A, G), <sup>2</sup>/<sub>5</sub>(B, H)) is not a behavioral strategy for 1 as the choices made by him at the two nodes are not independent.

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- Player *i* has perfect recall if and only if, for every *h* ∈ *H<sub>i</sub>* and any nodes *t*, *t'* ∈ *h*, we have σ<sub>i</sub>(*t*) = σ<sub>i</sub>(*t'*).

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Figure: Harold William Kuhn (1925–2014). Sources: https://alchetron.com/Harold-W-Kuhn and https://www.cantorsparadise.com/

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  - It follows that  $|S_i| = 1 + \sum_{h \in H_i} |C_h|$ ,

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  - $\circ$  *P* is a set of *n* players,
  - $S = (S_1, \ldots, S_n)$ , where  $S_i$  is a set of sequences of player i,
  - $u = (u_1, \ldots, u_n)$ , where  $u_i \colon S \to \mathbb{R}$  is the payoff function of player *i*, and
  - $C = (C_1, ..., C_n)$  is a set of linear constraints on the realization probabilities of player *i*.
- Now, we will define all these terms properly. It will take some time...
- First, we explain the set of sequences *S* in more detail.
  - Any sequence  $\sigma$  from  $S_i$  is either the empty sequence  $\emptyset$  or it is uniquely determined by the last move c at the information set h, that is,  $\sigma = \sigma_h c$ .
  - Thus,  $S_i = \{\emptyset\} \cup \{\sigma_h c \colon h \in H_i, c \in C_h\}.$
  - It follows that  $|S_i| = 1 + \sum_{h \in H_i} |C_h|$ , which is linear in the size of the tree of G.
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  - If there are only two players, then we capture their payoffs with matrices *A* and *B*.

# Example: sequence form payoff matrices

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    - $\diamond$  We let  $C_i$  be the set of constraints of the second type.

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NE of a 2-player zero-sum extensive game of perfect recall are solutions of the following LP:  $\min_{u,y} e^{\top}u \text{ subject to } Fy = f, E^{\top}u - Ay \ge \mathbf{0}, y \ge \mathbf{0}.$ 

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$$x^{\top}(E^{\top}u - Ay) = 0, \quad y^{\top}(F^{\top}v - B^{\top}x) = 0,$$
  

$$Ex = e, x \ge \mathbf{0}, \quad Fy = f, y \ge \mathbf{0},$$
  

$$E^{\top}u - Ay \ge \mathbf{0}, \quad F^{\top}v - B^{\top}x \ge \mathbf{0}.$$

• A similar idea based on the Duality theorem and complementary slackness, we can also compute NE in 2-player extensive games. See the lecture notes for more details.

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A pair (x, y) of realization plans in a 2-player game in the extensive form of perfect recall is NE iff there are vectors u and v such that:

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- Still, it is exponentially faster than to run the Lemke–Howson algorithm on the induced normal-form game.

• More about games in extensive form + implementation of the algorithms is taught in a lecture by Martin Schmid.

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# Thank you for your attention.