Algorithmic game theory

Martin Balko

1st lecture

September 30th 2025



Basic info

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- Webpage: https://kam.mff.cuni.cz/~balko/ath2526/ATH.html
 - $\circ\,$ lecture info, topics covered, presentations, lecture notes \dots

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- Recommended literature:
 - M. Balko: Algorithmic game theory: lecture notes.
 - The notes are still under construction. Comments are welcome.

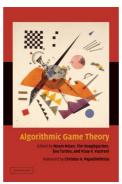
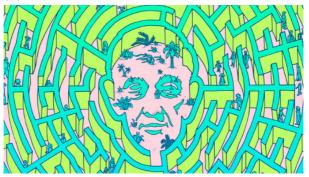


Figure: Algorithmic game theory by Nisan et al.

Source: https://amazon.com

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Zdroj: https://quantamagazine.org

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- Several real-word applications.
- More than ten game theorists have won the Nobel Prize in economics.

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- Advanced Modern Algorithmic Game Theory (NOPT022) (long title)
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 - a continuation of the previous class that focuses on approximation methods and large games.

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Češi vytvořili umělou inteligenci, která drtí v pokeru jednoho hráče za druhým

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Vědci z Matematicko-fyzikální fakulty Univerzity Karlovy a Fakulty elektrotechnické ČVUT v Praze pracovali několik posledních měsíců na vývoji umělé inteligence, jejímž hlavním úkolem bude stát se špičkou v karetní hře Poker Texas Hold'em. A to se skutečně podařilo, program porazil hned několik profesionálních hráčů.





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- Preliminary plan:
 - Finding Nash equilibria
 - Nash equilibria and Nash's Theorem,
 - o zero-sum games,
 - o bimatrix games and the Lemke-Howson algorithm,
 - o other notions of equilibria,
 - o regret minimization.
 - o Mechanism design,
 - o auctions (Vickrey),
 - o Myerson's lemma and its applications,
 - revenue maximization.

Finding Nash equilibria

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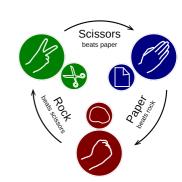
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- The *i*th coordinate $u_i(a)$ of u(a) is the gain of player *i* on *a*.

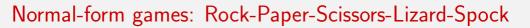
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Scissors	(-1,1)	(<mark>1,-1</mark>)	(0,0)



Sources: https://en.wikipedia.org/



Normal-form games: Rock-Paper-Scissors-Lizard-Spock

	Rock	Paper	Scissors	Lizard	Spock
Rock	(<mark>0,0</mark>)	(-1,1)	(<mark>1,-1</mark>)		
Paper	(<mark>1,-1</mark>)	(0,0)	(-1,1)	(-1,1)	(1,-1)
Scissors			(0,0)	(<mark>1,-1</mark>)	(-1,1)
Lizard	(-1,1)	(<mark>1,-1</mark>)		(0,0)	(1,-1)
Spock	(1,-1)	(-1,1)	(1,-1)	(-1,1)	(0,0)



 $Sources:\ https://bigbangtheory.fandom.com/$

 "Scissors cuts Paper, Paper covers Rock, Rock crushes Lizard, Lizard poisons Spock, Spock smashes Scissors, Scissors decapitates Lizard, Lizard eats Paper, Paper disproves Spock, Spock vaporizes Rock (and as it always has) Rock crushes Scissors."

Normal-form games: Rock-Paper-Scissors-Lizard-Spock



ROCK PAPER SCISSORS LIZARD SPOCK SPIDER-MAN BATMAN WIZARD GLOCK

Scissors cuts paper. Paper covers rock. Rock crushes lizard. Lizard poisons Spock. Spock zaps wizard. Wizard stuns Batman. Batman scares Spider-Man. Spider-Man disarms glock. Glock breaks rock. Rock interrupts wizard. Wizard burns paper. Paper disproves Spock. Spock befuddles Spider-Man. Spider-Man defeats lizard. Lizard confuses Batman (because he looks like Killer Croc). Batman dismantles scissors. Scissors cut wizard. Wizard transforms lizard. Lizard eats paper. Paper jams glock. Glock kills Batman's mom. Batman explodes rock. Rock crushes scissors. Scissors decapitates lizard. Lizard is too small for glock. Glock shoots Spock. Spock vaporizes rock. Rock knocks out Spider-Man. Spider-Man rips paper. Paper delays Batman. Batman hangs Spock. Spock smashes scissors. Scissors cut Spider-Man. Spider-Man annoys wizard. Wizard melts glock. Glock dents scissors

ROCK PAPER SCISSORS SPOCK LIZARD by Sam Kass and Karen Bryla, and then, Brian Yan messed it up into this.

Source: https://www.naturphilosophie.co.uk/

Normal-form games: Chess

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Source: https://edition.cnn.com/

• Chess as a normal-form game: Each action of player $i \in \{\text{black}, \text{white}\}$ is a list of all possible situations that can happen on the board together with the move player i would make in that situation. Then we can simulate the whole game of chess in one round.

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- that is, $u_i(s)$ is the expected value of u_i under the product distribution $\prod_{i=1}^n s_i$.
- It satisfies the linearity of the expected payoff (Exercise):

$$u_i(s) = \sum_{a_i \in A_i} s_i(a_i) \cdot u_i(a_i; s_{-i}),$$

where
$$s_{-i} = (s_1, \dots, s_{i-1}, s_{i+1}, \dots, s_n)$$
 and $(s'_i; s_{-i}) = (s_1, \dots, s_{i-1}, s'_i, s_{i+1}, \dots, s_n)$ for any $s'_i \in S_i$.

• Consider the Rock-Paper-Scissors game where each player i uses a strategy s_i that assigns each action the probability 1/3.

	Rock	Paper	Scissors
Rock	(0,0)	(- <mark>1</mark> ,1)	(1,-1)
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• By linearity, $u_1(s) = \frac{1}{3} \cdot 0 + \frac{1}{3} \cdot 0 + \frac{1}{3} \cdot 0 = 0$.

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- Player 1 will be the "row player" while player 2 will be the "column player".

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Sources: Serena Maylon (MtG)

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Paradoxically, the only stable solution is when both testify.

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Sources: https://www.fourstateshomepage.com/

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• Like Rock-Papers-Scissors, this is a zero-sum game (whatever one player gets, the other one loses). Prisoner's dilemma is not.

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	Football	Opera
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Opera	(0,0)	(1,2)



Sources: https://media.istockphoto.com/

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Football	(2,1)	(0,0)
Opera	(0,0)	(1,2)



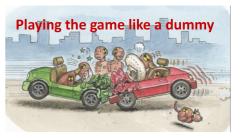
Sources: https://media.istockphoto.com/

This game displays both cooperation and competition.

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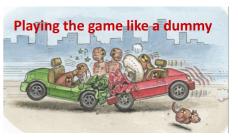
	Turn	Go straight
Turn	(0,0)	(-1,1)
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• What is the best strategy for the players?

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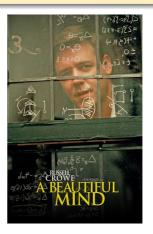


Figure: John Forbes Nash Jr. (1928–2015) and his depiction in the movie A Beautiful mind.

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- For n affinely independent points $x_1, \ldots, x_n \in \mathbb{R}^d$, an (n-1)-simplex Δ_n on x_1, \ldots, x_n is the set of convex combinations of the points x_1, \ldots, x_n .

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Lemma (Lemma 2.18)

For $n, d_1, \ldots, d_n \in \mathbb{N}$, let K_1, \ldots, K_n be compact sets, each K_i lying in \mathbb{R}^{d_i} . Then, $K_1 \times \cdots \times K_n$ is a compact set in $\mathbb{R}^{d_1 + \cdots + d_n}$.

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Figure: L. E. J. Brouwer (1881-1966).

Source: https://arxiv.org/pdf/1612.06820.pdf





Figure: John Forbes Nash Jr. receiving a Nobel prize for economics.

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Thank you for your attention.