

# Game Theory

## 5. Complexity

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# 1 Motivation



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Summary

**Motivation:** We already know some algorithms for finding Nash equilibria in restricted settings from the previous chapter, and **upper bounds** on their complexity.

- For **finite zero-sum games**: **polynomial-time** computation.
- For **general finite two player games**: computation in **NP**.

**Question:** What about **lower bounds** for those cases and in general?

**Approach to an answer:** In this chapter, we study the **computational complexity** of finding Nash equilibria.

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## Definition (The problem of computing a Nash equilibrium)

### NASH

**Given:** A finite two-player strategic game  $G$ .

**Find:** A mixed-strategy Nash equilibrium  $(\alpha, \beta)$  of  $G$ .

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### Remarks:

- No need to add restriction "...if one exists, else 'fail'", because existence is guaranteed by Nash's theorem.
- The corresponding **decision** problem can be trivially solved in **constant time** (always return "true"). Hence, we really need to consider the **search** problem version instead.



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In this form, NASH looks similar to other search problems, e. g.:

## SAT

**Given:** A propositional formula  $\varphi$  in CNF.

**Find:** A truth assignment that makes  $\varphi$  true, if one exists, else 'fail'.

**Note:** This is the search version of the usual decision problem.

## 2 Search Problems



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A **search problem** is given by a binary relation  $R(x, y)$ .

## Definition (Search problem)

A **search problem** is a problem that can be stated in the following form, for a given binary relation  $R(x, y)$  over strings:

### SEARCH- $R$

**Given:**  $x$ .

**Find:** Some  $y$  such that  $R(x, y)$  holds, if such a  $y$  exists, else 'fail'.

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## Some complexity classes for search problems:

- **FP**: class of search problems that can be solved by a **deterministic** Turing machine in **polynomial time**.
- **FNP**: class of search problems that can be solved by a **nondeterministic** Turing machine in **polynomial time**.
- **TFNP**: class of search problems in **FNP** where the relation **R is total**, i. e.,  $\forall x \exists y. R(x, y)$ .
- **PPAD**: class of search problems that can be **polynomially reduced to END-OF-LINE**.  
(**PPAD**: Polynomial Parity Argument in Directed Graphs)

To understand **PPAD**, we need to understand what the **END-OF-LINE** problem is.

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## Definition (END-OF-LINE instance)

Consider a **directed graph**  $\mathcal{G}$  with node set  $\{0, 1\}^n$  such that each node has **in-degree and out-degree at most one** and there are no isolated vertices. The graph  $\mathcal{G}$  is specified by two polynomial-time computable functions  $\pi$  and  $\sigma$ :

- $\pi(v)$ : returns the **predecessor of  $v$** , or  $\perp$  if  $v$  has no predecessor.
- $\sigma(v)$ : returns the **successor of  $v$** , or  $\perp$  if  $v$  has no successor.

In  $\mathcal{G}$ , there is an arc from  $v$  to  $v'$  if and only if  $\sigma(v) = v'$  and  $\pi(v') = v$ .

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## Definition (END-OF-LINE instance (ctd.))

We call a triple  $(\pi, \sigma, v)$  consisting of such functions  $\pi$  and  $\sigma$  and a node  $v$  in  $\mathcal{G}$  with in-degree zero (a “source”) an **END-OF-LINE instance**.

With this, we can define the **END-OF-LINE problem**:

## Definition (END-OF-LINE problem)

### END-OF-LINE

**Given:** An END-OF-LINE instance  $(\pi, \sigma, v)$ .

**Find:** Some node  $v' \neq v$  such that  $v'$  has out-degree zero (a “sink”) or in-degree zero (another “source”).

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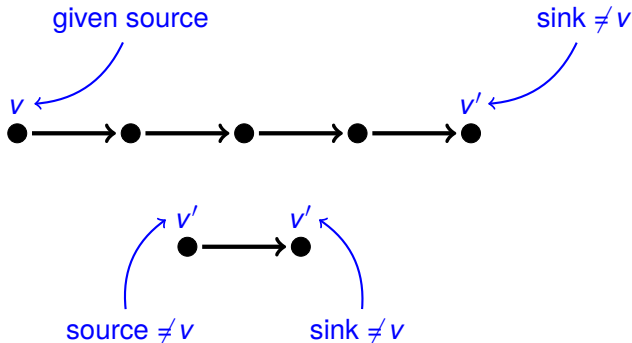
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# The END-OF-LINE Problem



## Example (END-OF-LINE)



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Relationship of different search complexity classes:

$$FP \subseteq PPAD \subseteq TFNP \subseteq FNP$$

Compare to upper runtime bound that we already know:

Lemke-Howson algorithm has **exponential** time complexity in the worst case.

# 3 Complexity Results



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## Theorem (Daskalakis et al., 2006)

NASH is **PPAD**-complete.

*The same holds for  $k$ -player instead of just two-player NASH.  $\square$*

Thus, NASH is presumably “simpler” than the SAT search problem, but presumably “harder” than any polynomial search problem.



Another search problem related to Nash equilibria is the problem of **finding a second Nash equilibrium** (given a first one has already been found). As it turns out, this is **at least as hard** as finding a first Nash equilibrium.

## Definition (2ND-NASH problem)

### 2ND-NASH

- Given:** A finite two-player game  $G$  and a mixed-strategy Nash equilibrium of  $G$ .
- Find:** A second different mixed-strategy Nash equilibrium of  $G$ , if one exists, else 'fail'.

## Theorem (Conitzer and Sandholm, 2003)

2ND-NASH is **FNP-complete**. □

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## Theorem (Conitzer and Sandholm, 2003)

For each of the following properties  $P^\ell$ ,  $\ell = 1, 2, 3, 4$ , given a finite two-player game  $G$ , it is **NP**-hard to decide whether there exists a mixed-strategy Nash equilibrium  $(\alpha, \beta)$  in  $G$  that has property  $P^\ell$ .

$P^1$  : player 1 (or 2) receives a payoff  $\geq k$  for some given  $k$ .  
("Guaranteed payoff problem")

$P^2$  :  $U_1(\alpha, \beta) + U_2(\alpha, \beta) \geq k$  for some given  $k$ .  
("Guaranteed social welfare problem")

$P^3$  : player 1 (or 2) plays some given action  $a$  with prob.  $> 0$ .

$P^4$  :  $(\alpha, \beta)$  is Pareto-optimal, i. e., there is no strategy profile  $(\alpha', \beta')$  such that

- $U_i(\alpha', \beta') \geq U_i(\alpha, \beta)$  for both  $i \in \{1, 2\}$ , and
- $U_i(\alpha', \beta') > U_i(\alpha, \beta)$  for at least one  $i \in \{1, 2\}$ .



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- **PPAD** is the complexity class for which the **END-OF-LINE problem** is complete.
- Finding a mixed-strategy Nash equilibrium in a finite two-player strategic game is **PPAD-complete**.
- **FNP** is the search-problem equivalent of the class **NP**.
- Finding a second mixed-strategy Nash equilibrium in a finite two-player strategic game is **FNP-complete**.
- Several **decision problems** related to Nash equilibria are **NP-complete**:
  - guaranteed payoff
  - guaranteed social welfare
  - inclusion in support
  - Pareto-optimality of Nash equilibria

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