# The Complexity of Nash Equilibria and Fixed Points of Algebraic Functions

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### Appetizer

What is the complexity of each of the following search problems:

- a. (Nash, 1950) Given a finite game, and  $\epsilon > 0$ , compute a vector x' (a mixed strategy profile) within distance  $\epsilon$  of some (exact) Nash Equilibrium.
- b. (Shapley, 1953) Given an instance of Shapley's stochastic game, and  $\epsilon > 0$ , approximate the *value* of the game to within distance  $\epsilon$ .

Note:

c. (Kolmogorov, 1947) Given a multi-type Branching Process, and  $\epsilon > 0$ , approximate its *extinction probabilities* within distance  $\epsilon$ .

Question: What do these three problems have to do with each other?

<u>Hint:</u> They are all <u>fixed point</u> problems for <u>algebraically defined</u> functions. Respectively:

- a. Brouwer
- b. Banach
- c. *Tarski*

But are they related in terms of computational complexity? Yes.

# **Outline of talk**

- Background: Games, Nash Equilibria, Brouwer Fixed Points.
- Weak vs. Strong approximation of Fixed Points.
- Scarf's classic algorithm, and its complexity implications.
- The complexity class PPAD, and weak approximation.
- Hardness of strong approximation: square-root-sum & arithmetic circuits.
- A new complexity class: **FIXP**. Nash is FIXP-complete.
- linear-FIXP = PPAD.
- Other FIXP problems: price equilibria, stochastic games, branching processes...
- Conclusions and future challenges.

#### **Finite Games**

A finite (normal form) game,  $\Gamma$ , consists of:

- 1. A set  $N = \{1, \ldots, n\}$  of players.
- 2. Each player  $i \in N$  has a finite set  $S_i = \{1, \dots, m_i\}$  of (pure) strategies. Let  $S = \prod_{i=1}^n S_i$ .
- 3. Each player  $i \in N$ , has a payoff (utility) function  $u_i : S \mapsto \mathbb{Q}$ .

### mixed strategies, expected payoffs, etc.

• A <u>mixed strategy</u>,  $x_i = (x_{i,1}, \ldots, x_{i,m_i})$ , for player i is a probability distribution over  $S_i$ .

A <u>profile</u> of mixed strategies:  $x = (x_1, \ldots, x_n)$ Let X denote the set of all profiles.

• The *expected payoff* for player *i*:

$$U_i(x) = \sum_{s=(s_1,...,s_n)\in S} (\prod_{k=1}^n x_{k,s_k}) u_i(s)$$

Let x<sub>-i</sub> denote everybody's strategy in x except player i's.
 Let (x<sub>-i</sub>; y<sub>i</sub>) denote the new profile: (x<sub>1</sub>,..., x<sub>i-1</sub>, y<sub>i</sub>, x<sub>i+1</sub>,..., x<sub>n</sub>).

### Nash Equilibria

A mixed strategy profile x is called:

• a *Nash Equilibrium* if:

 $\forall i$ , and all mixed strategies  $y_i$ :  $U_i(x) \ge U_i(x_{-i}; y_i)$ 

I.e.: No player can increase its own payoff by unilaterally switching its strategy.

• a  $\epsilon$ -Nash Equilibrium, for  $\epsilon > 0$ , if:

 $\forall i$ , and all mixed strategies  $y_i$ :  $U_i(x) \ge U_i(x_{-i}; y_i) - \epsilon$ 

I.e.: No player can increase its own payoff by more than  $\epsilon$  by unilaterally switching its strategy.

**Theorem** (Nash 1950) Every finite game has a Nash Equilibrium.

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### Nash's proof

**Brouwer's fixed point theorem:** A continuous function  $F : D \mapsto D$  from a compact convex set  $D \subseteq \mathbb{R}^m$  to itself has a fixed point:  $x^* \in D$ , s.t.  $F(x^*) = x^*$ .

Nash showed the NEs of a finite game,  $\Gamma$ , are precisely the fixed points of the following Brouwer function  $F_{\Gamma} : X \mapsto X$ :

$$F_{\Gamma}(x)_{(i,j)} = \frac{x_{i,j} + \max\{0, g_{i,j}(x)\}}{1 + \sum_{k=1}^{m_i} \max\{0, g_{i,k}(x)\}}$$

where  $g_{i,j}(x) \doteq U_i(x_{-i}; j) - U_i(x)$ .

<u>Note</u>:  $g_{i,j}(x)$  are polynomials in the variables in x, and they measure: "how much better off would player i be if it switched to pure strategy j?"

#### A basic computational question

What is the complexity of the following search problem:

#### ("Strong") *e*-approximation of a Nash Equilibrium:

Given a finite (normal form) game,  $\Gamma$ , with 3 or more players, and given  $\epsilon > 0$ , compute a rational vector x' such that there is some (exact!) Nash Equilibrium  $x^*$  of  $\Gamma$  so that:

$$\|x^* - x'\|_{\infty} < \epsilon$$

**Note:** This is <u>NOT</u> the same thing as asking for an  $\epsilon$ -Nash Equilibrium.

### Weak vs. Strong approximation of Fixed Points

- 2-player finite games always have <u>rational</u> NEs, and there are algorithms for computing an exact rational NE in a 2-player game (Lemke-Howson'64).
- For games with ≥ 3 players, all NEs can be <u>irrational</u> (Nash,1951).
  So we can't hope to compute one "*exactly*".

Two different notions of  $\epsilon$ -approximation of fixed points:

- (Weak) Given  $F : \Delta_n \mapsto \Delta_n$ , compute x' such that:  $||F(x') x'|| < \epsilon$ .
- (Strong) Given  $F : \Delta_n \mapsto \Delta_n$ , compute x' such that there exists  $x^*$  where  $F(x^*) = x^*$  and  $||x^* x'|| < \epsilon$ .

# Scarf's classic algorithm

Scarf (1967) gave a beautiful algorithm (refined by Kuhn and others) for computing (weak!)  $\epsilon$ -fixed points of a given Brouwer function  $F : \Delta_n \mapsto \Delta_n$ :

- 1. Subdivide the simplex  $\Delta_n$  into "small" subsimplices of diameter  $\delta > 0$  (depending on the "modulus of continuity" of F, and on  $\epsilon > 0$ ).
- 2. Color every vertex, z, of every subsimplex with a color i such that  $z_i > 0$  &  $F(z)_i \leq z_i$ .
- 3. By **Sperner's Lemma** there must exist a panchromatic subsimplex. (And the proof provides a way to "navigate" toward such a simplex.)
- 4. <u>Fact</u>: If  $\delta > 0$  is chosen such that  $\delta \le \epsilon/2n$  and  $\forall x, y \in \Delta_n$ ,  $||x - y||_{\infty} < \delta \implies ||F(x) - F(y)||_{\infty} < \epsilon/2n$ , then all the points in a panchromatic subsimplex are <u>weak</u>  $\epsilon$ -fixed points. (They need <u>NOT</u> in general be anywhere near an actual fixed point.)







**The underlying "directed lines" parity argument in Scarf's algorithm** (Same combinatorial argument used by [Lemke-Howson'64] for 2-player Nash.)



#### Implicit assumptions: when is Scarf's algorithm applicable?

To use Scarf's algorithm for computing a weak  $\epsilon$ -fixed point (in a reasonably efficient way) we are making several implicit assumptions. Suppose  $F : \Delta_n \mapsto \Delta_n$  is given to us in a (unspecified) form that requires m bits to describe.

- 1. F(x) should be *polynomial-time computable* for given rational vector x. I.e., the time to compute F(x) should be polynomial in both m and the encoding size of x. (Otherwise, how can we compute colors of vertices efficiently?)
- We should have a "tractable" simplicial subvidivision of Δ<sub>n</sub>. In particular, the subsimplices and their vertices, z, must have encoding size polynomial in m and size(ε). (Otherwise, again, how can we compute F(z) efficiently?) And the simplicial subdivision must yield efficient algorithms (P-time in m and size(ε)) for both starting at the extra bogus endpoint subsimplex, and for traversing "on the fly" a single directed edge of the (implicit) line graph whose nodes are subsimplices.

#### another key assumption.....

3. Finally, F(x) should be <u>polynomially continuous</u>, meaning, that there is a polynomial q(r) such that for a given  $\epsilon > 0$ , we can choose  $\delta = 1/2^{q(m+size(\epsilon))}$ , such that  $\forall x, y \in \Delta_n$ ,  $||x - y|| < \delta \Rightarrow ||F(x) - F(y)|| < \epsilon$ .

(Note: since F is continuous on a compact set  $\Delta_n$ , it is *uniformly* continuous.)

These assumptions (1. - 3.) do not guarantee that Scarf's algorithm will run in P-time. They just guarantee that each step (each edge traversal) of Scarf's algorithm can be carried out in P-time, and that it will eventually halt (after potentially exponentially many traversal steps in the encoding size m and in  $size(\epsilon)$ , because there can be exponentially many subsimplices), and will produce a panchromatic subsimplex such that every point inside that subsimplex is a weak  $\epsilon$ -fixed point of F.

### $\epsilon\text{-NEs}$ are weak $\epsilon\text{-fixed}$ points

**Fact:** For finite games,  $\Gamma$ , computing an  $\epsilon$ -NE is P-time equivalent to computing a <u>Weak</u>  $\epsilon$ -fixed point of Nash's function  $F_{\Gamma}$ .

Thus, to compute an  $\epsilon$ -NE, we can simply apply Scarf's algorithm to  $F_{\Gamma}$ .

The functions  $F_{\Gamma}$  satisfy all the implicit assumptions for applicability of Scarf's algorithm: they are polynomially continuous, polynomial-time computable, and furthermore appropriate "tractable" simplicial subdivisions are well known for the compact convex domain X of mixed strategy profiles (i.e., for cartesian products of n-simplices).

**Question:** What does this tell us about the complexity of computing an  $\epsilon$ -NE?

### The complexity class PPAD

Papadimitriou (1992) defined **PPAD**, based on the "directed line" parity argument, to capture (approximate) Nash and (approximate) Brouwer, etc...

<u>Definition</u>: **PPAD** is the class of search problems polynomial-time reducible to:

<u>Directed line endpoint problem</u>: Given two boolean circuits, S ("Successor") and P ("Predecessor"), each with n input bits and n output bits, such that  $P(0^n) = 0^n$ , and  $S(0^n) \neq 0^n$ , find a n-bit vector,  $\mathbf{z}$ , such that either:  $P(S(\mathbf{z})) \neq \mathbf{z}$  or  $S(P(\mathbf{z})) \neq \mathbf{z} \neq 0^n$ . (By the directed line parity argument such a  $\mathbf{z}$  exists (for inconsistent P and S it exists trivially).)

PPAD lies somewhere between (the search problem versions of) P and NP.

By Scarf's algorithm, computing a  $\epsilon$ -NE is in PPAD.

Can we do better?

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Computing  $\epsilon$ -NEs is already as hard as all of PPAD:

#### Theorem:

1. [Daskalakis-Goldberg-Papadimitriou'06][Chen-Deng'06]:

Computing a  $\epsilon$ -NE for a 3 player game is PPAD-complete.

2. [Chen-Deng'06]:

Computing an <u>exact</u> (rational) NE for a 2 player game is PPAD-complete.

But what if we want to approximate <u>exact</u> NEs for games with  $\geq 3$  players and to approximate <u>exact</u> fixed points?

I.e., what if we want to do *strong* approximation of fixed points?

(Warning: Scarf's algorithm does not in general yield Strong  $\epsilon$ -fixed points.)

### Why care about strong approximation of fixed points?

- It can be argued (as Scarf (1973) implicitly did) that for many applications in economics weak  $\epsilon$ -fixed points of Brouwer functions are sufficient.
- However, many important problems boil down to a fixed point computation for which weak  $\epsilon$ -FPs are useless, unless they also happen to be strong  $\epsilon$ -FPs.

#### Examples:

–Shapley's Stochastic Games; –Condon's (1992) Simple Stochastic Games; –Kolmogorov's multi-type Branching Processes;

(and Recursive Markov Chains, and Recursive Stochastic Games, .....)

#### A basic upper bound for Strong $\epsilon\text{-approximation of Nash}$

**Fact:** Given game  $\Gamma$  and  $\epsilon > 0$ , we can Strong  $\epsilon$ -approximate a NE in **PSPACE**. *Proof:* For Nash's functions,  $F_{\Gamma}$ , the expression

$$\exists \mathbf{x} (\mathbf{x} = F_{\Gamma}(\mathbf{x}) \land \mathbf{a} \le \mathbf{x} \le \mathbf{b})$$

can be expressed as a formula in the *Existential Theory of Reals* (ETR). So we can Strong  $\epsilon$ -approximate an NE,  $x^* \in \overline{\Delta_n}$ , in **PSPACE**, using  $\log(1/\epsilon)n$  queries to a PSPACE decision procedure for ETR ([Canny'89],[Renegar'92]). (These are deep, but thusfar impractical algorithms.)

Can we do better than **PSPACE**?

### The Square-Root Sum problem

The square-root sum problem (**Sqrt-Sum**) is the following decision problem:

Given  $(d_1, \ldots, d_n) \in \mathbb{N}^n$  and  $k \in \mathbb{N}$ , decide whether  $\sum_{i=1}^n \sqrt{d_i} \leq k$ .

It is known to be solvable in PSPACE.

[Allender, Bürgisser, Kjeldgaard-Pedersen, Miltersen, 2006] improved this to:

the 4th level of the Counting Hierarchy:  $P^{PP^{PP}^{PP}}$ 

Open problem ([GareyGrahamJohnson'76]) whether it is solvable even in NP.

(In particular, whether exact Euclidean-TSP is in NP hinges on this.)

#### Sqrt-Sum $\leq_p$ approximation of actual NE

**Theorem:** <u>Any non-trivial approximation</u> of an <u>actual</u> NE solves **Sqrt-Sum**. More precisely:

For every  $\epsilon > 0$ , **Sqrt-Sum** is *P*-time reducible to the following problem: Given a 3-player (normal form) game,  $\Gamma$ , with the property that:

1. in every NE, player 1 plays exactly the same mixed strategy,  $x_1^*$ , and

2. the probability,  $x_{1,1}^*$ , with which player 1 plays it first pure strategy is either:

(a.) 
$$0$$
 , or (b.)  $\geq (1-\epsilon)$  ,

decide which of (a.) or (b.) is the case.

### A harder arithmetic circuit decision problem

[Allender et al'06] reduced **Sqrt-Sum** to the following (which they showed lies in the *Counting Hierarchy*):

**PosSLP:** Given an <u>arithmetic circuit</u> (Straight Line Program) over basis  $\{+, *, -\}$  with integer inputs, decide whether the output is > 0.

Every discrete decision problem solvable in P-time in the unit-cost arithmetic RAM model in P-time, i.e., in the discrete, rational Blum-Shub-Smale class  $P_{\mathbb{R}}$ , is P-time (Turing) reducible to **PosSLP**. So, **PosSLP** captures discrete problems in  $P_{\mathbb{R}}$ .

(<u>Note</u>: testing = 0 for such arithmetic circuits (much easier than PosSLP) is already a well-known open problem. It is equivalent ([ABKM'06]) to *polynomial identity testing* (known to be in coRP). )



#### **Theorem:**

**PosSLP** is *P*-time reducible to Strong approximation of 3-player NEs.

More precisely, it reduces to the same 0 vs.  $(1 - \epsilon)$  choice problem as before.

**Question:** How far can an  $\epsilon$ -NE be from an actual NE?

Answer: Very far.

A seemingly contrary fact:

**Fact:** For every continuous function  $F : \Delta \mapsto \Delta$ , and every  $\epsilon > 0$ , there exists a  $\delta > 0$ , such that a weak  $\delta$ -fixed point of F is a strong  $\epsilon$ -fixed point of F.

But this is a non-constructive fact! It uses a compactness argument. (Bolzano-Weierstrass.)

From a quantitative, computational perspective, it is certainly not the full story:

**Theorem:** For every n, there exists a 4-player game  $\Gamma_n$  of size O(n) with an  $\epsilon$ -NE, x', where  $\epsilon = \frac{1}{2^{2^{\Omega(n)}}}$ , and yet x' has distance 1 (in  $l_{\infty}$ ) to any actual NE.

Same holds for 3 players, but with distance 1 replaced by distance  $(1 - 2^{-poly})$ .

### A new complexity class: FIXP

Consider the following class of fixed point problems:

**Input:** algebraic circuit (straight-line program) over basis  $\{+, *, -, /, \max, \min\}$  with rational constants, having n input variables and n outputs, such that the circuit represents a continuous function  $F : [0, 1]^n \mapsto [0, 1]^n$ .

(The domain  $[0,1]^n$  can be allowed to be much more general. See our full paper.)

**Output:** Compute (or strong  $\epsilon$ -approximate) a fixed point of F.

We close these problems under suitable P-time reductions.

Call the resulting class **FIXP**.



### Nash is FIXP-complete

**Theorem** *Computing a 3-player Nash Equilibrium is* **FIXP***-complete.* It is complete in several senses:

- In terms of "exact" (real valued) computation;
- In terms of strong  $\epsilon$ -approximation,
- An appropriate "decision" version of the problem: Given a game, Γ, rational value q ∈ Q, and coordinate i: if for all NEs x\*, x<sub>i</sub>\* ≥ q, then "Yes"; if for all NEs x\*, x<sub>i</sub>\* < q, then "No". Otherwise, any answer is fine.</li>

Completeness holds under very restrictive polynomial-time (real valued) search problem reductions where the "solution recovery" function g is linear.

#### Very brief sketch of some proof ingredients

- Suppose we could create a (3-player) game such that, in any NE, Player 1 plays strategy A with probability > 1/2 iff  $\sum_i \sqrt{d_i} > k$  and with probability < 1/2 iff  $\sum_i \sqrt{d_i} < k$ . (Suppose equality can't happen.)
- Add an extra player with 2 strategies, who gets high payoff if it "guesses correctly" whether player 1 plays pure strategy A, and low payoff otherwise.
  In any NE, the new player will play one of its two strategies with probability 1.
  Deciding which solves SqrtSum.
- What about equality? We don't have to worry about it because  $\sum_i \sqrt{d_i} = k$  is P-time decidable ([Borodin-Fagin-Hopcroft-Tompa'85]).

#### A key ingredient in our proofs

Two beautiful gems by Bubelis:

1. (Bubelis, 1979) Every real algebraic number can be "encoded" in a precise sense as the payoff to player 1 in a unique NE of a 3-player game.

2. (Bubelis, 1979) There is a general polynomial-time reduction from n-player games to 3-player games.

Such that you can easily recover a (real valued) NE of the n-player game as a linear function of a given NE in the resulting 3-player game.

Many details in the proof of FIXP-completeness:

- A series of transformations to get circuits into a "normal form" with additional "conditional assignment gates".
- Transform circuit to a game with a large (but bounded) number of players, using suitable *gadgets*.

Some key gadgets derived from [Bubelis'79]'s construction.

(Alternatively, the gadgets in [Gol-Pap'06], [Das-Gol-Pap'06] can also be used.)

• Reduce to 3-players: again uses [Bubelis '79].

#### Another FIXP-complete problem: Price Equilibria

- An idealized exchange economy with n agents and m commodities.
- For a given price vector, p, each agent l has an excess demand function  $g_i^l(p)$ . The total excess demand for commodity i is  $g_i(p) = \sum_l g_i^l(p)$ . Excess demands are continuous and satisfy economically justified axioms:

- (Homogeneous): For all 
$$\alpha > 0$$
,  $p \ge 0$ ,  $g_i^l(\alpha p) = g_i^l(p)$ .

- (Walras's law):  $\sum_i p_i g_i(p) = 0.$
- Price Equilibrium: prices  $p^* \ge 0$  such that  $g_i(p^*) \le 0$  for all  $i \ (= 0 \text{ if } p_i^* > 0)$ .
- Fact: Every exchange economy has a price equilibrium. (Proof via Brouwer.)
- **Proposition** Computing Price Equilibria in exchange economies where excess demands are given by algebraic circuits over {+, \*, -, /, max, min} is FIXP-complete. (Follows from Uzawa (1962).)

# A new characterization of PPAD

Let **linear-FIXP** denote the subclass of FIXP where the algebraic circuits are restricted to basis  $\{+, \max\}$  and multiplication by rational constants only.

**Theorem** *The following are all equivalent:* 

- 1. PPAD
- 2. linear-FIXP
- 3. exact fixed point problems for "polynomial piecewise-linear functions"

( Corollary: Simple-Stoch-Games (and Parity Games) are in PPAD.)

# sketch proof that $PPAD \leq Iinear-FIXP$

Computing a 2-player NE (exactly) is PPAD-complete, so we only need to give a reduction from two player NE to linear-FIXP.

Nash's functions  $F_{\Gamma}$  are already non-linear even for 2 players.

Is there a different,  $\{+, \max\}$  function for 2-player NEs??

Yes!

[Gul-Pearce-Staccetti'93] describe a fixed point approach for NEs.

By examining carefully what they do, one can derive the follow function for NEs:

1. First, let 
$$x'_{i,j} := x_{i,j} + U_i(x_{-i};j)$$
.



2. Second, "*project*" the vector  $x'_i$  onto the simplex  $\Delta_{m_i}$ , for every player *i*.

**Fact:** The fixed points of this function are the NEs.

Can "projection" be computed with a linear-FIXP function? Yes, ... with the help of *sorting networks*.

From this revised function for n-player NEs we also obtain:

**Theorem:** Basis  $\{+, *, \max\}$  (and rational constants) suffices to capture **FIXP**.

# **Shapley's Stochastic Games**

2-player, zero-sum, imperfect information, discounted stochastic games.

- 1. finite state space, finite move alphabet.
- 2. Starting in a given state, at each round both players (independently), choose a move, or a probability distribution on moves. Their joint move determines a probability distribution on the next state, and a reward to player 1.
- 3. The rewards after each round are <u>discounted</u> by given factor  $0 < \beta < 1$ , and the total discounted reward to player 1 is sum  $\sum_i \beta^i r_i$ .

The value of Shapley's games (which can be irrational) can be characterized by fixed point equations,  $\mathbf{x} = P(\mathbf{x})$ , where  $P(\mathbf{x})$  is a <u>contraction</u> map.

There is a <u>unique</u> *Banach* fixed point (which can be irrational), which yields the game value starting at each state.

**Theorem** For Shapley's stochastic games:

1. Computing the game value is in FIXP.

2. The (strong) approximation problem for the game value is in PPAD.

3. The decision problem (is the game value  $\geq r$ ?) is SqrtSum-hard.

*Proof of part (2.):*  $P(\mathbf{x})$  is a "fast enough" contraction mapping. For such mappings, Weak  $\epsilon$ -fixed points are "close enough" to the actual Banach fixed point.  $P(\mathbf{x})$  is a Brouwer function on a "not too big" domain. Thus: apply Scarf's algorithm to  $P(\mathbf{x})$ .

<u>Note:</u> this implies Condon's Simple Stochasic Games are also in PPAD.

# multi-type Branching Processes

*Branching processes*, originally studied in the 19th century by Galton and Watson. Kolmogorov (1947) defined and studied *multi-type Branching Processes* (mt-BPs) with Sevastyanov and others.

Huge literature in probability theory, population genetics, and many other areas.

- 1. A population of *individuals*. Each individual has one of a fix set of *types*.
- 2. In each generation, every individual of a given type "gives birth" to a number of (a multi-set of) individuals of different types, according to a probability distribution on multi-sets, based on its type.

**Question:** Starting from one entity of a given type, will the population eventually go extinct with probability  $\geq 1/2$  ?

(Whether it will *almost surely* go extinct is decidable in P-time ([EY05]).)

#### The extinction problem for mt-BPs is in FIXP

The extinction probabilities are the *Least Fixed Point* (LFP) solution of a *monotone* system of nonlinear polynomial equations,  $\mathbf{x} = P(\mathbf{x})$ .

(The LFP exists, by Tarski's (Tarski-Knaster) fixed point theorem.)

The LFP can be irrational, and the associated decision problems are SqrtSum-hard and PosSLP-hard ([EY05,EY07]).

The LFP can be "*isolated*" as the unique fixed point of FIXP function.

**Theorem:** The mt-BP extinction problem is in FIXP.

<u>Note</u>: mt-BP extinction  $\equiv$  <u>1-exit</u> *Recursive Markov Chain* termination

**Theorem** Any non-trivial approximation of the general multi-exit RMC termination problem is SqrtSum-hard and PosSLP-hard.

### Conclusions

A very rich landscape with many, many, open questions:

- Can strong approximation of NEs be done in anything better than **PSPACE**?
- Is strong approximation of NEs hard for a standard complexity class like NP? (Not likely to be easy. Would imply the *"rational fragment of"* the BSS class NP<sub> $\mathbb{R}$ </sub> contains both NP and coNP. That's an open problem.)
- <u>A basic practical question</u>: Is there any algorithm that, given a game & ε > 0:
  1. is guarranteed to output a point x within distance ε of some actual NE, and
  - 2. performs "reasonably well" in practice?

K. Etessami and M. Yannakakis, "On the complexity of Nash Equilibria and Other Fixed Points", FOCS'07.

(See full version of paper at: http://homepages.inf.ed.ac.uk/kousha)