NETWORK SCIENCE

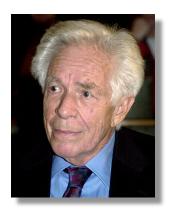
Game Theory

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What is Game Theory?



"The central problem of game theory was posed by von Neumann as early as 1926 in Göttingen. It is the following: If *n* players, P_1, \ldots, P_n , play a given game Γ , how must the *i*th player, P_i , play to achieve the most favorable result for himself?"

> Harold W. Kuhn Lectures on the Theory of Games (1953)

A few cornerstones of game theory

- **1921–1928:** Emile Borel and John von Neumann give the first modern formulation of a mixed strategy along with the idea of finding minimax solutions of normal-form games.
- **1944, 1947:** John von Neumann and Oskar Morgenstern publish *Theory of Games and Economic Behavior*.
- **1950–1953:** In four papers John Nash made seminal contributions to both non-cooperative game theory and to bargaining theory.
- **1972–1982:** John Maynard Smith applies game theory to biological problems thereby founding "evolutionary game theory."
- late 1990's -: Development of algorithmic game theory...



Normal-form Games

We shall focus on finite, non-cooperative, simultaneous-move games in **normal form**, which are characterized by:

- ✓ A set of **players**: $I = \{1, 2, ..., n\}$ (n ≥ 2)
- ✓ A set of **pure strategy profiles**: $S = S_1 \times S_2 \times ... \times S_n$ where each $S_i = \{1, 2, ..., m_i\}$ is the (finite) set of pure strategies (actions) available to the player *i*
- ✓ A **payoff function**: $\pi : S \to \Re^n$, $\pi(s) = (\pi_1(s), ..., \pi_n(s))$, where $\pi_i(s)$ (*i*=1...*n*) represents the "payoff" (or utility) that player *i* receives when strategy profile *s* is played

Each player is to choose one element from his strategy space in the absence of knowledge of the choices of the other players, and "payments" will be made to them according to the function $\pi_i(s)$.

Players' goal is to maximize their own returns.



Two Players

In the case of two players, payoffs can be represented as two $m_1 \ge m_2$ matrices (say, *A* for player 1 and *B* for player 2):

$$A = (a_{hk}) \qquad \qquad a_{hk} = \pi_1(h,k)$$

$$B = (b_{hk}) \qquad b_{hk} = \pi_2(h,k)$$

Special cases:

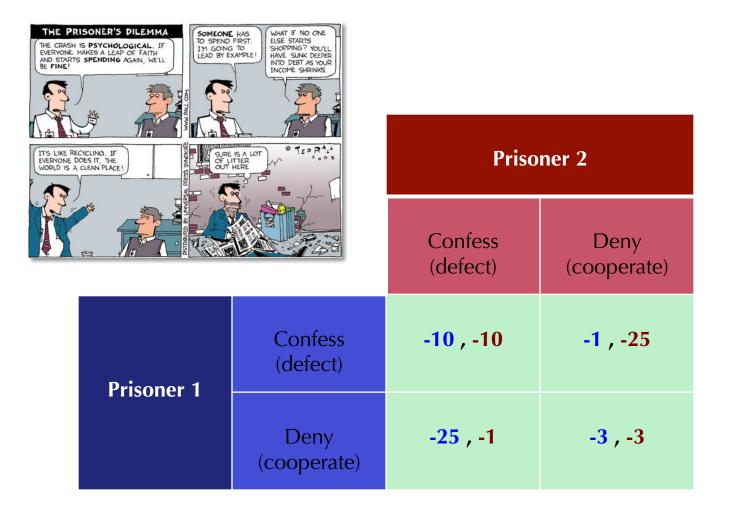
✓ Zero-sum games: A + B = 0 ($a_{hk} = -b_{hk}$ for all h and k)

✓ Symmetric games: $B^T = A$

✓ Doubly-symmetric games: $A = A^T = B^T$

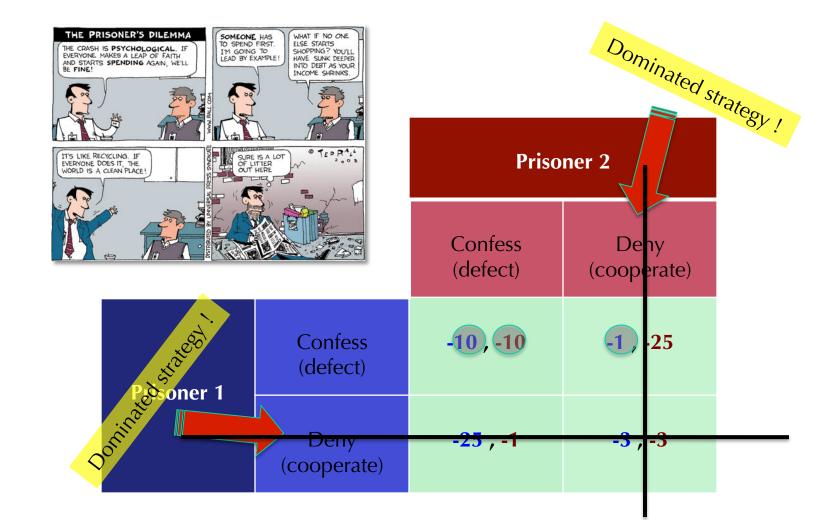


Example 1: Prisoner's Dilemma





What Would You Do?





Example 2: Battle of the Sexes

		Wife		
60	3	Soccer	Ballet	
Husband	Soccer	2,1	0,0	
	Ballet	0,0	1,2	



Example 3: Rock-Scissors-Paper

Well, this is Kind of awkward		You			
		Rock	Scissors	Paper	
		Rock	0,0	1,-1	-1 , 1
	Me	Scissors	-1,1	0,0	1,-1
		Paper	1,-1	-1 , 1	0,0



Mixed Strategies

A **mixed strategy** for player *i* is a probability distribution over his set S_i of pure strategies, which is a point in the **standard simplex**:

$$\Delta_{i} = \left\{ x_{i} \in \mathbb{R}^{m_{i}} : \forall h = 1...m_{i} : x_{ih} \ge 0, \text{ and } \sum_{h=1}^{m_{i}} x_{ih} = 1 \right\}$$

The set of pure strategies that is assigned positive probability by mixed strategy $x_i \in \Delta_i$ is called the **support** of x_i :

$$\sigma(x_i) = \left\{ h \in S_i : x_{ih} > 0 \right\}$$

A **mixed strategy profile** is a vector $x = (x_1, ..., x_n)$ where each component $x_i \in \Delta_i$ is a mixed strategy for player $i \in I$.

The **mixed strategy space** is the multi-simplex $\Theta = \Delta_1 \times \Delta_2 \times \ldots \times \Delta_n$



Standard Simplices ×3 x_{i2} 1 1 ×2 1 x_{il} $m_i = 2$ $m_i = 3$

Note: Corners of standard simplex correspond to pure strategies.



Mixed-Strategy Payoff Functions

In the standard approach, all players' randomizations are assumed to be independent.

Hence, the probability that a pure strategy profile $s = (s_1, ..., s_n)$ will be used when a mixed-strategy profile x is played is:

$$x(s) = \prod_{i=1}^{n} x_{is_i}$$

and the expected value of the payoff to player *i* is:

$$u_i(x) = \sum_{s \in S} x(s) \pi_i(s)$$

In the special case of two-players games, one gets:

$$u_1(x) = \sum_{h=1}^{m_1} \sum_{k=1}^{m_2} x_{1h} a_{hk} x_{2k} = x_1^T A x_2 \qquad u_2(x) = \sum_{h=1}^{m_1} \sum_{k=1}^{m_2} x_{1h} b_{hk} x_{2k} = x_1^T B x_2$$

where *A* and *B* are the payoff matrices of players 1 and 2, respectively.



Best Replies

Notational shortcut. If $z \in \Theta$ and $x_i \in \Delta_i$, the notation (x_i, z_{-i}) stands for the strategy profile in which player $i \in I$ plays strategy x_i , while all other players play according to z.

Player *i*'s **best reply** to the strategy profile x_{-i} is a mixed strategy $x_i^* \in \Delta_i$ such that

$$u_i(x_i^*, x_{-i}) \ge u_i(x_{i}, x_{-i})$$

for all strategies $x_i \in \Delta_i$.

Note. The best reply is not necessarily unique. Indeed, except in the extreme case in which there is a unique best reply that is a pure strategy, the number of best replies is always infinite.



Nash Equilibria

A strategy profile $x \in \Theta$ is a **Nash equilibrium** if it is a best reply to itself, namely, if:

 $u_i(x_{i'}, x_{-i}) \ge u_i(z_{i'}, x_{-i})$

for all i = 1...n and all strategies $z_i \in \Delta_i$.

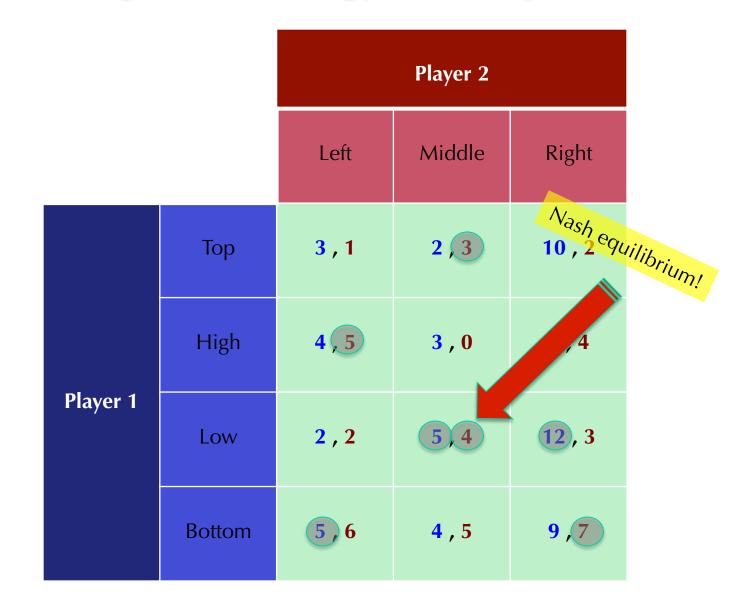
If strict inequalities hold for all $z_i \neq x_i$ then x is said to be a **strict Nash** equilibrium.

Theorem. A strategy profile $x \in \Theta$ is a Nash equilibrium if and only if for every player $i \in I$, every pure strategy in the support of x_i is a best reply to x_{-i} .

It follows that every pure strategy in the support of any player's equilibrium mixed strategy yields that player the same payoff.

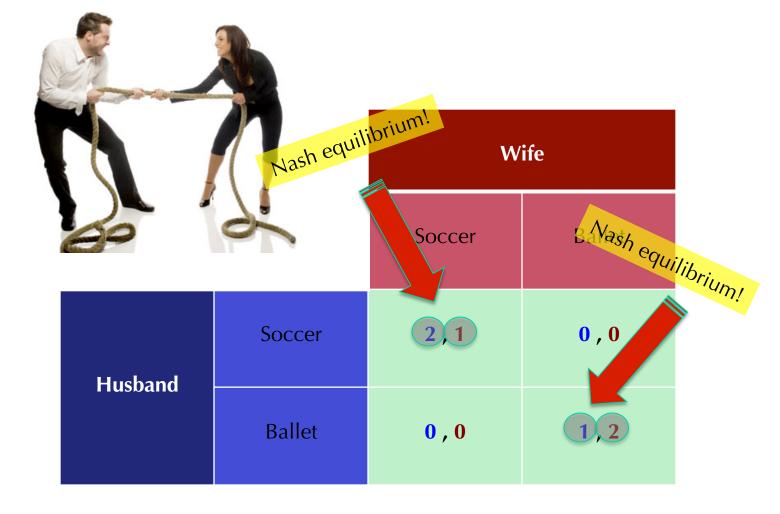


Finding Pure-strategy Nash Equilibria



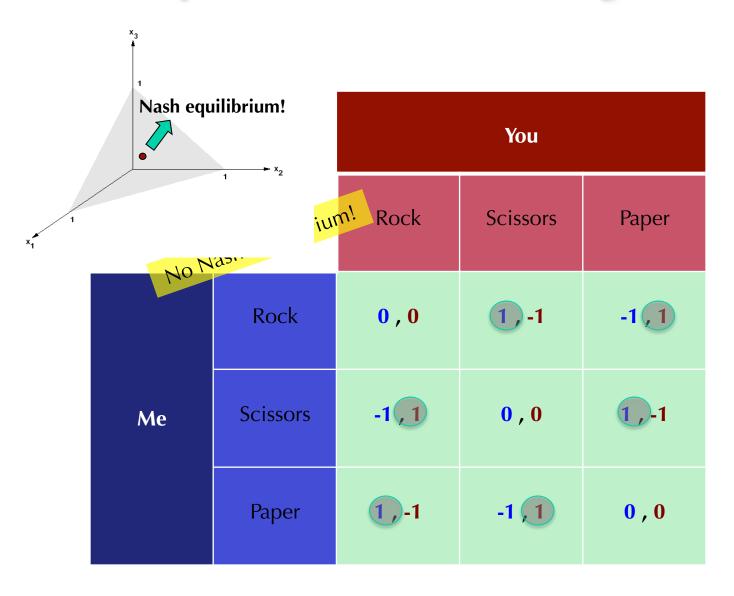


Multiple Equilibria in Pure Strategies





No Equilibrium in Pure Strategies





Existence of Nash Equilibria

Theorem (Nash, 1951). Every finite normal-form game admits a mixed-strategy Nash equilibrium.

Idea of proof.

- 1. Define a continuous map T on Θ such that the fixed points of T are in one-to-one correspondence with Nash equilibria.
- 2. Use Brouwer's theorem to prove existence of a fixed point.

Note. For symmetric games, Nash proved that there always exists a **symmetric Nash equilibrium**, namely a Nash equilibrium where all players play the same (possibly mixed) strategy.



The Complexity of Finding Nash Equilibria



"Together with factoring, the complexity of finding a Nash equilibrium is in my opinion the most important concrete open question on the boundary of P today."

> Christos Papadimitriou Algorithms, games, and the internet (2001)

At present, no known reduction exists from our problem to a decision problem that is NP-complete, nor has it been shown to be easier.

Theorem (Daskalakis *et al.* **2005; Chen and Deng, 2005, 2006).** The problem of finding a sample Nash equilibrium of a general-sum finite game with two or more players is PPAD-complete.



Variations on Theme

Theorem (Gilboa and Zemel, 1989). The following are *NP*-complete problems, even for symmetric games.

Given a two-player game in normal form, does it have:

- 1. at least two Nash equilibria?
- 2. a Nash equilibrium in which player 1 has payoff at least a given amount?
- 3. a Nash equilibrium in which the two players have a total payoff at least a given amount?
- 4. a Nash equilibrium with support of size greater than a give number?
- 5. a Nash equilibrium whose support contains a given strategy?
- 6. a Nash equilibrium whose support does not contain a given strategy?
- 7. etc.



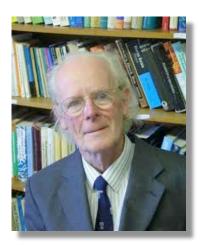
Evolution and the Theory of Games

"We repeat most emphatically that our theory is thoroughly static. A dynamic theory would unquestionably be more complete and therefore preferable.

But there is ample evidence from other branches of science that it is futile to try to build one as long as the static side is not thoroughly understood."

John von Neumann and Oskar Morgenstern Theory of Games and Economic Behavior (1944)





"Paradoxically, it has turned out that game theory is more readily applied to biology than to the field of economic behaviour for which it was originally designed."

> John Maynard Smith Evolution and the Theory of Games (1982)



Evolutionary Games

Introduced by John Maynard Smith and Price (1973) to model the evolution of behavior in animal conflicts.

Assumptions:

- ✓ A large population of individuals belonging to the same species which compete for a particular limited resource
- This kind of conflict is modeled as a symmetric two-player game, the players being pairs of randomly selected population members
- Players do not behave "rationally" but act according to a pre-programmed behavioral pattern (pure strategy)
- ✓ Reproduction is assumed to be asexual
- ✓ Utility is measured in terms of Darwinian fitness, or reproductive success



Evolutionary Stability

A strategy is **evolutionary stable** if it is resistant to invasion by new strategies.

Formally, assume:

- ✓ A small group of "invaders" appears in a large populations of individuals, all of whom are pre-programmed to play strategy $x \in \Delta$
- ✓ Let $y \in \Delta$ be the strategy played by the invaders
- ✓ Let $\boldsymbol{\varepsilon}$ be the share of invaders in the (post-entry) population (0 < $\boldsymbol{\varepsilon}$ < 1)

The payoff in a match in this bimorphic population is the same as in a match with an individual playing mixed strategy:

$$w = \mathbf{\epsilon} y + (1 - \mathbf{\epsilon}) x \in \Delta$$

and the (post-entry) payoffs got by the incumbent and the mutant strategies are u(x,w) and u(y,w), respectively.



Evolutionary Stable Strategies

Definition. A strategy $x \in \Delta$ is said to be an **evolutionary stable strategy** (ESS) if for all $y \in \Delta - \{x\}$ there exists $\delta \in (0,1)$, such that for all $\epsilon \in (0, \delta)$ we have:

$$u[x, \varepsilon y + (1 - \varepsilon)x] > u[y, \varepsilon y + (1 - \varepsilon)x]$$

incumbent mutant

Theorem. A strategy $x \in \Delta$ is an ESS if and only if it meets the following first- and second-order best-reply conditions:

- 1. $u(y,x) \le u(x,x)$ for all $y \in \Delta$ (Nash equilibrium)
- 2. $u(y,x) = u(x,x) \Rightarrow u(y,y) < u(x,y)$ for all $y \in \Delta \{x\}$

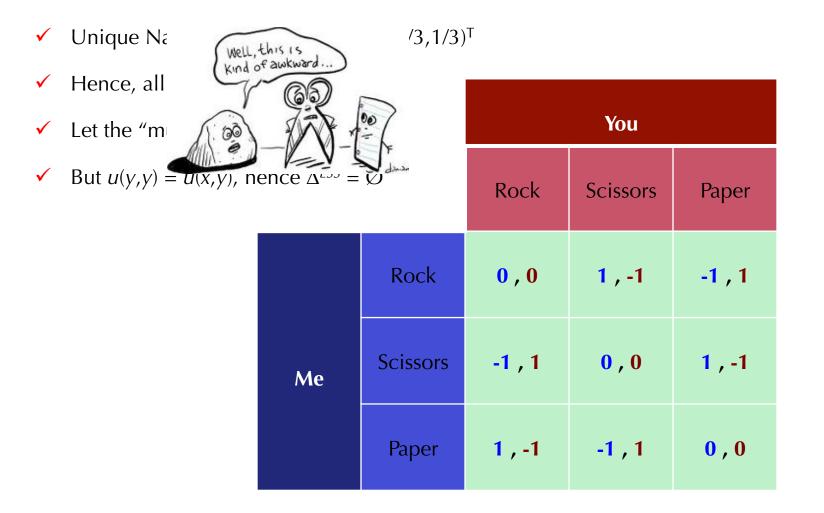
Note. From the conditions above, we have:

- $\checkmark \quad \Delta^{ESS} \subseteq \Delta^{NE}$
- ✓ If $x \in \Delta$ is a strict Nash equilibrium, then x is an ESS



Existence of ESS's

Unlike Nash equilibria existence of ESS's is not guaranteed.





Complexity Issues

Two questions of computational complexity naturally present themselves:

- ✓ What is the complexity of determining whether a given game has an ESS (and of finding one)?
- ✓ What is the complexity of recognizing whether a given *x* is an ESS for a given game?

Theorem (Etessami and Lochbihler, 2004). Determining whether a given two-player symmetric game has an ESS is both NP-hard and coNP-hard.

Theorem (Nisan, 2006). Determining whether a (mixed) strategy *x* is an ESS of a given two-player symmetric game is coNP-hard.



Replicator Dynamics

Let $x_i(t)$ the population share playing pure strategy *i* at *time t*. The **state** of the population at time *t* is: $x(t) = (x_1(t), \dots, x_n(t)) \in \Delta$.

Replicator dynamics (Taylor and Jonker, 1978) are motivated by Darwin's principle of natural selection:

 $\frac{\dot{x}_i}{x_i} \propto$ payoff of pure strategy i – average population payoff

which yields:

$$\dot{x}_i = x_i \Big[u(e^i, x) - u(x, x) \Big]$$
$$= x_i \Big[(Ax)_i - x^T Ax \Big]$$

Notes.

✓ Invariant under positive affine transformations of payoffs

✓ Standard simplex Δ is invariant under replicator dynamics, namely, $x(0) \in \Delta \Rightarrow x(t) \in \Delta$, for all t > 0 (so is its interior and boundary)



Replicator Dynamics and ESS's

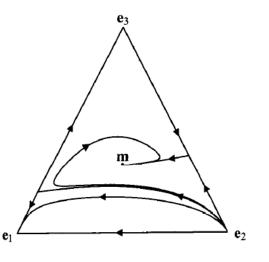
Theorem (Nachbar, 1990; Taylor and Jonker, 1978). A point $x \in \Delta$ is a Nash equilibrium if and only if x is the limit point of a replicator dynamics trajectory starting from the interior of Δ .

Furthermore, if $x \in \Delta$ is an ESS, then it is an asymptotically stable equilibrium point for the replicator dynamics.

The opposite need not be true.

$$A = \begin{bmatrix} 0 & 6 & -4 \\ -3 & 0 & 5 \\ -1 & 3 & 0 \end{bmatrix}$$

- ✓ The point $m = (1/3, 1/3, 1/3)^T$ is asymptotically stable (its eigenvalues have negative parts).
- ✓ But $e^1 = (1,0,0)^T$ is an ESS.
- Hence *m* cannot be an ESS (being in the interior, it would have to be the unique ESS).





Doubly Symmetric Games

In a doubly symmetric (or partnership) game, the payoff matrix A is symmetric ($A = A^{T}$).

Fundamental Theorem of Natural Selection (Losert and Akin, 1983).

For any doubly symmetric game, the average population payoff

 $f(x) = x^{\mathsf{T}} A x$

is strictly increasing along any non-constant trajectory of replicator dynamics, namely, $d/dtf(x(t)) \ge 0$ for all $t \ge 0$ (with equality if and only if x(t) is a stationary point).

Characterization of ESS's (Hofbauer and Sigmund, 1988)

For any doubly simmetric game with payoff matrix *A*, the following statements are equivalent:

a) $x \in \Delta^{ESS}$

- b) $x \in \Delta$ is a strict local maximizer of $f(x) = x^T A x$ over the standard simplex Δ
- c) $x \in \Delta$ is asymptotically stable under the replicator dynamics



Discrete-time Replicator Dynamics

A well-known discretization of replicator dynamics, which assumes nonoverlapping generations, is the following (assuming a non-negative *A*):

$$x_i(t+1) = x_i(t) \frac{A(x(t))_i}{x(t)^T A x(t)}$$

which inherits most of the dynamical properties of its continuous-time counterpart (e.g., the fundamental theorem of natural selection).

MATLAB implementation

distance=inf;		
while distance>epsilon		
old_x=x;		
x = x.*(A*x);		
x = x./sum(x);		
distance=pdist([x,old_x]');		
end		





Texts on classical and evolutionary game theory

- J. von Neumann and O. Morgerstern. *Theory of Games and Economic Behavior*. Princeton University Press (1944, 1953).
- D. Fudenberg and J. Tirole. *Game Theory*. MIT Press (1991).
- M. J. Osborne and A. Rubinstein. A Course in Game Theory. MIT Press (1994).
- J. Weibull. Evolutionary Game Theory. MIT Press (1995).
- J. Hofbauer and K. Sigmund. *Evolutionary Games and Population Dynamics*. Cambridge University Press (1998).

Computationally-oriented texts

- N. Nisan, T. Roughgarden, E. Tardos, and V. Vazirani (Eds.) *Algorithmic Game Theory*. Cambridge University Press (2007).
- Y. Shoham and K. Leyton-Brown. *Multiagent Systems: Algorithmic, Game-Theoretic, and Logical Foundations*. Cambridge University Press (2009).

On-line resources

http://gambit.sourceforge.net/a library of game-theoretic algorithmshttp://gamut.stanford.edu/a suite of game generators for testing game algorithms