

# A Whirlwind Tour of Game Theory

(Mostly from Fudenberg & Tirole)

Players choose actions, receive rewards based on their own actions and those of the other players.

Example, the Prisoner's Dilemma:

	Cooperate	Defect
Cooperate	+3, +3	0, +5
Defect	+5, 0	+1, +1

# Strategies and Nash Equilibrium

A **strategy** is a specification for how to play the game for a player. A **pure strategy** defines, for every possible choice a player could make, which action the player picks. A **mixed strategy** is a probability distribution over strategies.

A **Nash equilibrium** is a profile of strategies for all players such that each player's strategy is an optimal response to the other players' strategies. Formally, a mixed-strategy profile  $\sigma_*^i$  is a Nash equilibrium if for all players  $i$ :

$$u^i(\sigma_*^i, \sigma_*^{-i}) \geq u^i(s^i, \sigma_*^{-i}) \forall s^i \in S^i$$

Nash equilibrium of Prisoner's Dilemma: Both players defect!

## Matching Pennies

	H	T
H	+1, -1	-1, +1
T	-1, +1	+1, -1

No pure strategy equilibria

Nash equilibrium: Both players randomize half and half between actions.

## More on Equilibria

Dominated strategies: Strategy  $s_i$  (strictly) dominates strategy  $s'_i$  if, for all possible strategy combinations of opponents,  $s_i$  yields a (strictly) higher payoff than  $s'_i$  to player  $i$ .

Iterated elimination of strictly dominated strategies: Eliminate all strategies which are dominated, relative to opponents' strategies which have not yet been eliminated.

If iterated elimination of strictly dominated strategies yields a unique strategy  $n$ -tuple, then this strategy  $n$ -tuple is the unique Nash equilibrium (and it is strict).

Every Nash equilibrium survives iterated elimination of strictly dominated strategies.

## Multiple Equilibria

A coordination game:

	L	R
U	9, 9	0, 8
D	8, 0	7, 7

$U, L$  and  $D, R$  are both Nash equilibria. What would be reasonable to play? With and without coordination?

While  $U, L$  is pareto-dominant, playing  $D$  and  $R$  are “safer” for the row and column players respectively...

## Existence of Equilibria

Nash's theorem, translated: every game with a finite number of actions for each player where each player's utilities are consistent with the (previously discussed) axioms of utility theory has an equilibrium in mixed strategies.

Idea 1: Reaction correspondences. Player  $i$ 's reaction correspondence  $r_i$  maps each strategy profile  $\sigma$  to the set of mixed strategies that maximize player  $i$ 's payoff when her opponents play  $\sigma_{-i}$ . Note that  $r_i$  depends only on  $\sigma_{-i}$ , so we don't really need all of  $\sigma$ , but it will be useful to think of it this way. Let  $r$  be the Cartesian product of all  $r_i$ . A fixed point of  $r$  is a  $\sigma$  such that  $\sigma \in r(\sigma)$ , so that for each player,  $\sigma_i \in r_i(\sigma)$ . Thus a fixed point of  $r$  is a Nash equilibrium.

Kakutani's FP theorem says that the following are sufficient conditions for  $r : \Sigma \rightarrow \Sigma$  to have a FP.

1.  $\Sigma$  is a compact, convex, nonempty subset of a finite-dimensional Euclidean space.

Satisfied, because it's a simplex

2.  $r(\sigma)$  is nonempty for all  $\sigma$

Each player's payoffs are linear, and therefore continuous, in her own mixed strategy. Continuous functions on compact sets attain maxima.

3.  $r(\sigma)$  is convex for all  $\sigma$

Suppose not. Then  $\exists \sigma', \sigma''$  such that  $\lambda \sigma' + (1 - \lambda) \sigma'' \notin r(\sigma)$  But for each player  $i$ ,

$$u_i(\lambda \sigma'_i + (1 - \lambda) \sigma''_i, \sigma_{-i}) =$$

$$\lambda u_i(\sigma'_i, \sigma_{-i}) + (1 - \lambda) u_i(\sigma''_i, \sigma_{-i})$$

so that if both  $\sigma'$  and  $\sigma''$  are best responses to  $\sigma_{-i}$ , then so is their weighted average.

#### 4. $r(\cdot)$ has a closed graph

The correspondence  $r(\cdot)$  has a closed graph if the graph of  $r(\cdot)$  is a closed set. Whenever the sequence  $(\sigma^n, \hat{\sigma}^n) \rightarrow (\sigma, \hat{\sigma})$ , with  $\hat{\sigma}^n \in r(\sigma^n) \forall n$ , then  $\hat{\sigma} \in r(\sigma)$  (same as upper hemicontinuity)

Suppose that there is a sequence  $(\sigma^n, \hat{\sigma}^n) \rightarrow (\sigma, \hat{\sigma})$  such that  $\hat{\sigma}^n \in r(\sigma^n)$  for every  $n$ , but  $\hat{\sigma} \notin r(\sigma)$ . Then there exists  $\epsilon > 0$  and  $\sigma'$  such that

$$u_i(\sigma'_i, \sigma_{-i}) > u_i(\hat{\sigma}_i, \sigma_{-i}) + 3\epsilon$$

Then, for sufficiently large  $n$ ,

$$\begin{aligned} u_i(\sigma'_i, \sigma_{-i}^n) &> u_i(\sigma'_i, \sigma_{-i}) - \epsilon > u_i(\hat{\sigma}_i, \sigma_{-i}) + 2\epsilon \\ &> u_i(\hat{\sigma}_i^n, \sigma_{-i}^n) + \epsilon \end{aligned}$$

which means that  $\sigma'_i$  does strictly better against  $\sigma_{-i}^n$  than  $\hat{\sigma}_i^n$  does, contradicting our assumption.



# Learning in Games\*

How do players reach equilibria?

What if I don't know what payoffs my opponent will receive?

I can try to learn her actions when we play repeatedly (consider 2-player games for simplicity).

**Fictitious play** in two player games. Assumes stationarity of opponent's strategy, and that players do not attempt to influence each others' future play. Learn *weight functions*

$$\kappa_t^i(s^{-i}) = \kappa_{t-1}^i(s^{-i}) + \begin{cases} 1 & \text{if } s_{t-1}^{-i} = s^{-i} \\ 0 & \text{otherwise} \end{cases}$$

\*Fudenberg & Levine, *The Theory of Learning in Games*, 1998

Calculate probabilities of the other player playing various moves as:

$$\gamma_t^i(s^{-i}) = \frac{\kappa_t^i(s^{-i})}{\sum_{\tilde{s}^{-i} \in S^{-i}} \kappa_t^i(\tilde{s}^{-i})}$$

Then choose the best response action.

## Fictitious Play (contd.)

If fictitious play converges, it converges to a Nash equilibrium.

If the two players ever play a (strict) NE at time  $t$ , they will play it thereafter. (Proofs omitted)

If *empirical marginal distributions* converge, they converge to NE. But this doesn't mean that play is similar!

t	Player1 Action	Player2 Action	$\kappa_T^1$	$\kappa_T^2$
1	T	T	(1.5, 3)	(2, 2.5)
2	T	H	(2.5, 3)	(2, 3.5)
3	T	H	(3.5, 3)	(2, 4.5)
4	H	H	(4.5, 3)	(3, 4.5)
5	H	H	(5.5, 3)	(4, 4.5)
6	H	H	(6.5, 3)	(5, 4.5)
7	H	T	(6.5, 4)	(6, 4.5)

**Cycling of actions in fictitious play in the matching pennies game**

# Universal Consistency

**Persistent miscoordination:** Players start with weights of  $(1, \sqrt{2})$

	A	B
A	0, 0	1, 1
B	1, 1	0, 0

A rule  $\rho^i$  is said to be  $\epsilon$ -**universally consistent** if for any  $\rho^{-i}$

$$\lim_{T \rightarrow \infty} \sup_{\sigma^i} \max_{\gamma_t^i} u^i(\sigma^i, \gamma_t^i) - \frac{1}{T} \sum_t u^i(\rho_t^i(h_{t-1})) \leq \epsilon$$

almost surely under the distribution generated by  $(\rho^i, \rho^{-i})$ , where  $h_{t-1}$  is the history up to time  $t - 1$ , available for the decision-making algorithm at time  $t$ .

## Back to Experts

Bayesian learning cannot give good payoff guarantees.

- Suppose the true way your opponent's actions are being generated is not in the support of the prior – want protection from unanticipated play, which can be endogenously determined.
- The Bayesian optimal method guarantees a measure of learning something close to the true model, but provides no guarantees on received utility.
- Can use the notion of experts to bound regret!

Define *universal expertise* analogously to universal consistency, and bound regret (lost utility) with respect to the best expert, which is a strategy.

The best response function is derived by solving the optimization problem

$$\max_{\mathcal{I}^i} \mathcal{I}^i \vec{u}_t^i + \lambda v^i(\mathcal{I}^i)$$

$\vec{u}_t^i$  is the vector of average payoffs player  $i$  would receive by using each of the experts

$\mathcal{I}^i$  is a probability distribution over experts

$\lambda$  is a small positive number.

Under technical conditions on  $v$ , satisfied by the entropy:

$$-\sum_s \sigma(s) \log \sigma(s)$$

we retrieve the exponential weighting scheme, and for every  $\epsilon$  there is a  $\lambda$  such that our procedure is  $\epsilon$ -universally expert.