

Evolution & Learning in Games

Econ 243B

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Lecture 2.

Foundations of Evolution & Learning in Games II

Outline

In this lecture, we shall:

- ▶ Take a first look at local stability. In particular, we shall define an *evolutionary stable state* and explore its relationship to Nash equilibrium.
- ▶ Apply what we have learned so far to analyze iterated play of the prisoners' dilemma.

Evolutionary Stable States (ESS)

Maynard Smith and Price (1973) defined the notion of an *evolutionary stable strategy*:

- ▶ Their focus was on monomorphic populations: every member plays the same strategy, which can be a mixed strategy.
- ▶ We are concerned with a polymorphic population of agents each programmed with a pure strategy.

Evolutionary Stable States (ESS)

Mathematically these problems are identical. For example:

$$\begin{aligned}F_1(x) &= x_1u(1, 1) + x_2u(1, 2) \\ &= u(1, x).\end{aligned}$$

Being randomly matched with a population of agents proportion x_1 of which are programmed with pure strategy 1 and x_2 of which are programmed with pure strategy 2 is the same as playing a single individual who plays strategy 1 with probability x_1 and strategy 2 with probability x_2 .

Evolutionary Stable States (ESS)

Hence we can adapt the concept of an evolutionary stable strategy to a population setting:

- ▶ The term we shall use is **evolutionary stable state** (ESS).

As we shall eventually see, ESS provides a sufficient condition for local stability under a wide range of evolutionary dynamics, including the replicator dynamic.

Invasion

Let the state be $x = \begin{pmatrix} x_1 \\ x_2 \\ \dots \\ x_n \end{pmatrix}$.

Consider a game F , where $F(x) = \begin{pmatrix} F_1(x) \\ F_2(x) \\ \dots \\ F_n(x) \end{pmatrix}$.

Consider an invasion of mutants who make up a fraction ε of the post-entry population.

The shares of each strategy in the mutant population are

represented by $y = \begin{pmatrix} y_1 \\ y_2 \\ \dots \\ y_n \end{pmatrix}$.

Invasion

Therefore, the post-entry population state is:

$$x_\varepsilon = (1 - \varepsilon)x + \varepsilon y = \begin{pmatrix} (1 - \varepsilon)x_1 + \varepsilon y_1 \\ (1 - \varepsilon)x_2 + \varepsilon y_2 \\ \dots \\ (1 - \varepsilon)x_n + \varepsilon y_n \end{pmatrix}.$$

The average payoff in the incumbent population in the post-entry state is $x'F((1 - \varepsilon)x + \varepsilon y)$.

The average payoff in the mutant population in the post-entry state is $y'F((1 - \varepsilon)x + \varepsilon y)$.

Uniform Invasion Barrier

The average payoff in the incumbent population is higher if:

$$(y - x)'F((1 - \varepsilon)x + \varepsilon y) < 0. \quad (1)$$

State x is said to admit a **uniform invasion barrier** if there exists an $\bar{\varepsilon} > 0$ such that (1) holds for all $y \in X - \{x\}$ and $\varepsilon \in (0, \bar{\varepsilon})$.

That is, for all possible mutations y , as long as the mutant population is less than fraction $\bar{\varepsilon}$ of the postentry population, the incumbent population receives a higher average payoff.

Invasion Barriers

Define the invasion barrier of x against y as:

$$b_x(y) = \inf (\{\varepsilon \in (0, 1) : (y - x)'F((1 - \varepsilon)x + \varepsilon y) \geq 0\} \cup \{1\}). \quad (2)$$

If x admits a **uniform invasion barrier**, then there exists an $\bar{\varepsilon} > 0$ such that $b_x(y) \geq \bar{\varepsilon} > 0$ for all $y \in X - \{x\}$.

ESS

DEFINITION. State $x \in X$ is an **evolutionary stable state** (ESS) of F if there exists a neighborhood O of x such that:

$$(y - x)'F(y) < 0 \quad \text{for all } y \in O - \{x\}. \quad (3)$$

In other words, if x is an ESS, then for any state y sufficiently close to x , a population playing x will receive a larger average payoff in state y than a population playing y (i.e. x is a better reply to y than y is to itself).

Note that this considers invasions of other states y by x rather than invasions of x by other states. Hence it is not clear, at present, why this should be a stability condition.

ESS and Invasion Barriers

Theorem 2.1. State $x \in X$ is an **evolutionary stable state (ESS)** if and only if it admits a uniform invasion barrier.

Thus if x is stable in the face of an arbitrarily large population of entrants who mutate to a nearby state, then it is stable in the face of a sufficiently small population of entrants who mutate to an arbitrary state.

ESS and NE

What is the relationship between ESS and NE?

DEFINITION. Suppose that $x \in X$ is a NE. Then $(y - x)'F(x) \leq 0$ for all $y \in X$.

In addition, suppose there exists a neighborhood of x that does not contain any other NE.

Then x is an **isolated NE**.

Proposition 2.1. Every ESS is an isolated NE.

Proof

Let x be an ESS of F , O be the nhd posited in (3) and $y \in X - \{x\}$ (not necessarily in O).

Then for all $\varepsilon > 0$ sufficiently small, the postentry state $x_\varepsilon = \varepsilon y + (1 - \varepsilon)x$ is in O .

This implies that:

$$\begin{aligned}(x_\varepsilon - x)'F(x_\varepsilon) &< 0 \\(\varepsilon y + (1 - \varepsilon)x - x)'F(x_\varepsilon) &< 0 \\ \varepsilon(y - x)'F(x_\varepsilon) &< 0 \\(y - x)'F(x_\varepsilon) &< 0.\end{aligned}$$

(4)

Proof

Taking $\varepsilon \rightarrow 0$ yields:

$$(y - x)'F(x) \leq 0.$$

That is, x is a NE.

To establish that x is isolated, note that if $w \in O - \{x\}$ were a NE then $(w - x)'F(w) \geq 0$, contradicting the supposition that x is an ESS [by (3)]. \square

The converse of Proposition 2.1 is not true.

- ▶ The mixed equilibrium of a two-strategy coordination game is a counterexample.

More on ESS and Nash

Therefore, *ESS is stronger than NE*.

In particular, an ESS satisfies the additional property:

Suppose there exists a state y which is an alternative best reply to x , i.e. $(y - x)'F(x) = 0$.

—Then $(y - x)'F(y) < 0$, i.e. x is a better reply to y than y is to itself.

Therefore:

- ▶ A strict NE is an ESS.
- ▶ A polymorphic population state (equivalent to a mixed NE) cannot be strict and hence must satisfy the additional property.

More on ESS and Nash

In the case in which agents are matched uniformly at random to play a normal form game (the case we have been focussing on), then it is easy to see why the additional property is required.

Suppose $(y - x)'F(x) = 0$, i.e. y is an alternative best reply to x .

Then:

$$\begin{aligned}(y - x)'F(\varepsilon y + (1 - \varepsilon)x) &= \varepsilon(y - x)'F(y) + (1 - \varepsilon) \underbrace{(y - x)'F(x)}_{=0} \\ &= \varepsilon(y - x)'F(y).\end{aligned}\tag{5}$$

Therefore, $(y - x)'F(y)$ must be negative for (1) to hold and hence, by Theorem 2.1, for x to be an ESS.

Example: Hawk Dove

	<i>Hawk</i>	<i>Dove</i>
<i>Hawk</i>	-2	0
<i>Dove</i>	4	0

ESS: $x = (\frac{2}{3}, \frac{1}{3})$. **ESS payoff** = 0.

Example: Hawk Dove

- ▶ Consider a mutation y such that $y_1 > x_1 = \frac{2}{3}$.
- ▶ Check that $(y - x)'F((1 - \varepsilon)x + \varepsilon y) < 0$ for all such y :

$$(y_1 - x_1)[-2((1 - \varepsilon)x_1 + \varepsilon y_1) + 4((1 - \varepsilon)(1 - x_1) + \varepsilon(1 - y_1))].$$

- ▶ This equals:

$$(y_1 - x_1)\varepsilon[-2y_1 + 4(1 - y_1)]$$

because $-2 \times \frac{2}{3} + 4 \times (1 - \frac{2}{3}) = 0$. This in turn equals:

$$(y_1 - x_1)\varepsilon[4 - 6y_1]$$

which is negative because $y_1 > \frac{2}{3}$ by hypothesis.

- ▶ A similar argument can be applied to the case $y_1 < x_1$. Hence x is an ESS.

The Prisoners' Dilemma

	<i>C</i>	<i>D</i>
<i>C</i>	3, 3	0, 5
<i>D</i>	5, 0	1, 1

NE/ESS: $x = (0, 1)$.

Therefore, an ESS is not necessarily efficient.

Not Every Game has an ESS

	A	B	C
A	1	0	2
B	2	1	0
C	0	2	1

$x = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$ is the unique NE and therefore the only possible ESS.

Not Every Game has an ESS

Note that x is a polymorphic population state (equivalent to a mixed strategy), so any basis vector (pure strategy) is an alternative best reply to x .

Check that the additional property holds: $(e_1 - x)'F(e_1) < 0$, where $e_1 = (1, 0, 0)$, i.e. the pure-strategy A .

This is not the case: $x'F(e_1) = e_1'F(e_1) = 1$.

The Iterated Prisoners' Dilemma

- ▶ Two players engage in a series of PD games.
- ▶ The engagement ends after the current round with probability $\delta < \frac{1}{2}$. We call this the *stopping probability*.
- ▶ The expected number of rounds per engagement is:

$$1 + (1 - \delta) + (1 - \delta)^2 + (1 - \delta)^3 + \dots = \frac{1}{1 - (1 - \delta)} = \frac{1}{\delta}.$$

- ▶ Consider a population in which three strategies are present:
 - ▶ *C*—always cooperate,
 - ▶ *D*—always defect,
 - ▶ *T*—tit-for-tat, i.e. start by cooperating, thenceforth cooperate in period t if partner cooperated in $t - 1$.

Expected Payoffs Within Each Pairing

	<i>C</i>	<i>D</i>	<i>T</i>
<i>C</i>	$\frac{3}{\delta}$	0	$\frac{3}{\delta}$
<i>D</i>	$\frac{5}{\delta}$	$\frac{1}{\delta}$	$4 + \frac{1}{\delta}$
<i>T</i>	$\frac{3}{\delta}$	$\frac{1}{\delta} - 1$	$\frac{3}{\delta}$

Note:

Payoff from playing *T* against *D* is $0 + (1 - \delta)\frac{1}{\delta} = \frac{1}{\delta} - 1$.

Payoff from playing *D* against *T* is $5 + (1 - \delta)\frac{1}{\delta} = 4 + \frac{1}{\delta}$.

Expected Payoffs Over All Pairings

$$F_C(x) = (x_C + x_T)^{\frac{3}{\delta}}$$

$$F_D(x) = x_C^{\frac{5}{\delta}} + x_D^{\frac{1}{\delta}} + x_T(4 + \frac{1}{\delta})$$

$$F_T(x) = (x_C + x_T)^{\frac{3}{\delta}} + x_D(\frac{1}{\delta} - 1)$$

Replicator Dynamics

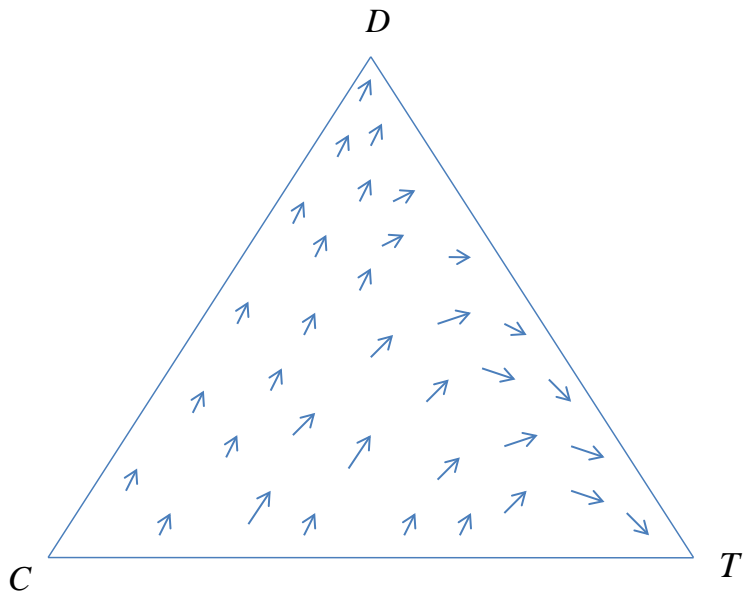
$$\begin{aligned}\frac{d}{dt} \begin{bmatrix} x_T \\ x_C \end{bmatrix} &= \frac{x_T}{x_C} (F_T(x) - F_C(x)) \\ &= \frac{x_T}{x_C} [x_D (\frac{1}{\delta} - 1)],\end{aligned}$$

which is positive because $\delta < \frac{1}{2}$.

$$\begin{aligned}\frac{d}{dt} \begin{bmatrix} x_T \\ x_D \end{bmatrix} &= \frac{x_T}{x_D} (F_T(x) - F_D(x)) \\ &= \frac{x_T}{x_D} \left[-x_C \frac{2}{\delta} - x_D + \underbrace{x_T \left(\frac{2}{\delta} - 4 \right)}_{>0} \right],\end{aligned}$$

which is positive for x_T sufficiently large.

Vector Field



All- T is not an ESS

Let $x = (x_D, x_C, x_T) = (0, 0, 1)$.

Consider any alternative state y such that $y_D = 0$.

$$\begin{aligned}(y - x)'F(y) &= (0 \ y_C \ y_T - 1) \begin{pmatrix} F_D(y) \\ F_C(y) \\ F_T(y) \end{pmatrix} \\ &= y_C F_C(y) + (y_T - 1) F_T(y) \\ &= [y_C + (y_T - 1)] \frac{3}{\delta} \quad (\text{recall that } y_D = 0) \\ &= [y_C + (1 - y_C - 1)] \frac{3}{\delta} \quad (\text{because } y_T = 1 - y_C) \\ &= 0.\end{aligned}$$

This violates (3). Hence all- T is not an ESS.

This is a case of evolutionary drift.