# 高级算法 Advanced Topics in Algorithms

# 陈旭

# 数据科学与计算机学院

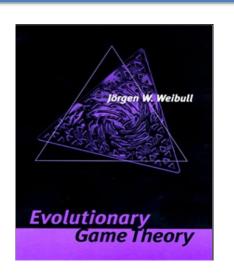


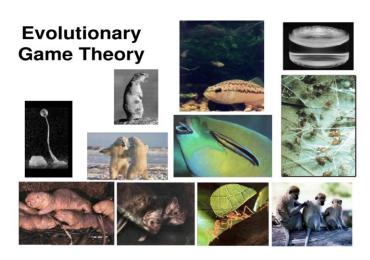
# Chapter 8 Evolutionary Game Theory: Population Games

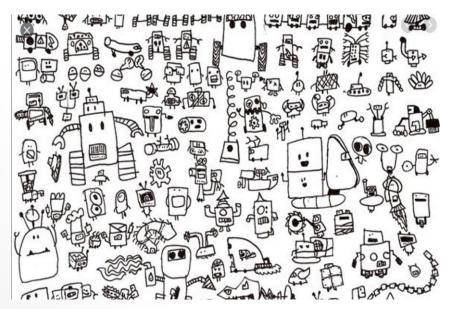
# Introduction

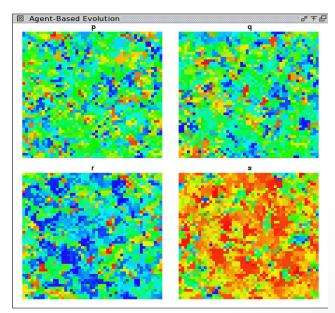
- So far we have considered "classical game theory", where outcome depends on the choice of rational individuals, and each individual uses a strategy that is the "best response" to other players' choice.
- If we map the same concept to "population", symmetric Nash equilibria (σ<sub>i</sub>\*, σ<sub>i-1</sub>\*), has an alternative interpretation. In a population where everyone uses σ\*, the best thing to do is to follow the crowd. So if everyone using σ\*, it will remain that way.
- Some interesting (and important) questions:
  - What happens if the population is close to, but not at, the NE?
  - Will the population evolve toward the equilibrium?
  - Will the population move away from the equilibrium?
- Evolutionary Game Theory considers a population decision makers wherein the *frequency* with which a particular decision is made can be time varying. It is a theory started from Biology.

# **Evolutionary Game Theory**









Under the evolutionary game, one type of end point (if any) is called an evolutionary stable strategy (ESS).

#### Definition

Consider an infinite population of individuals that can use a set of pure strategies, S. A population profile is a vector x that gives a probability x(s) with which each strategy  $s \in S$  is played in the population.

Note that the population profile needs not correspond to a strategy adopted by any members of the population !!

# Example

- A population can use  $S = \{s_1, s_2\}$ .
- If every member of the population randomizes by playing each of the pure strategies with probability  $\frac{1}{2}$ , then  $\mathbf{x} = (\frac{1}{2}, \frac{1}{2})$ . In this case, the population profile  $\mathbf{x}$  is identical to the mixed strategy adopted by all members.
- If half of the population adopt the strategy  $s_1$  and other half adopt strategy  $s_2$ . We have  $\mathbf{x} = (\frac{1}{2}, \frac{1}{2})$ , and this is NOT the same as the strategy adopted by **any** member of the population.

## **Definition**

Consider a population where initially all the individuals adopt some strategy  $\sigma^*$ . Suppose a mutation occurs and a small proportion  $\epsilon$  of individuals use some other strategy  $\sigma$ . The new population is called the post-entry population and will be denoted as  $\mathbf{x}_{\epsilon}$ .

# Example

Consider a population with  $\mathbf{S} = \{s_1, s_2\}$  and  $\sigma^* = (\frac{1}{2}, \frac{1}{2})$ . Suppose the mutant strategy is  $\sigma = (\frac{3}{4}, \frac{1}{4})$ , then

$$\boldsymbol{x}_{\epsilon} = (1 - \epsilon)\sigma^* + \epsilon\sigma = (1 - \epsilon)\left(\frac{1}{2}, \frac{1}{2}\right) + \epsilon\left(\frac{3}{4}, \frac{1}{4}\right) = \left(\frac{1}{2} + \frac{\epsilon}{4}, \frac{1}{2} - \frac{\epsilon}{4}\right).$$

# Stability of ESS

A mixed strategy  $\sigma^*$  is an evolutionary stable strategy (ESS) if there exists an  $\bar{\epsilon}$  such that for every  $0 < \epsilon < \bar{\epsilon}$  and every  $\sigma \neq \sigma^*$ 

$$\pi(\sigma^*, \mathbf{X}_{\epsilon}) > \pi(\sigma, \mathbf{X}_{\epsilon}).$$

**Physical meaning:** a strategy  $\sigma^*$  is an ESS if mutants that adopt any other strategy  $\sigma$  leave fewer offspring in the post-entry population, provided that the proportion of mutants is sufficiently small.

# Types of Population Games

In general, there are two types of population game: (1) games against the field and (2) games with pairwise contests.

#### **Definition**

A game against the field is one in which there is no specific "opponent" for a given individual - their payoff depends on what everyone in the population is doing.

#### **Definition**

A pairwise contest game describes a situation in which a given individual plays against an opponent that has been randomly selected (by nature) from the population and the payoff depends just on what both individual do.

- In an remote island, inhabitants have to decide to use either "beads" or "shells" as tokens of money in commerce.
- A transaction is only successful if both parties use the same form of token.
- Assume that a trader gets a utility increment of 1 if the transaction is successful and 0 if it fails.



Shells

**Beads** 

**Beads** 



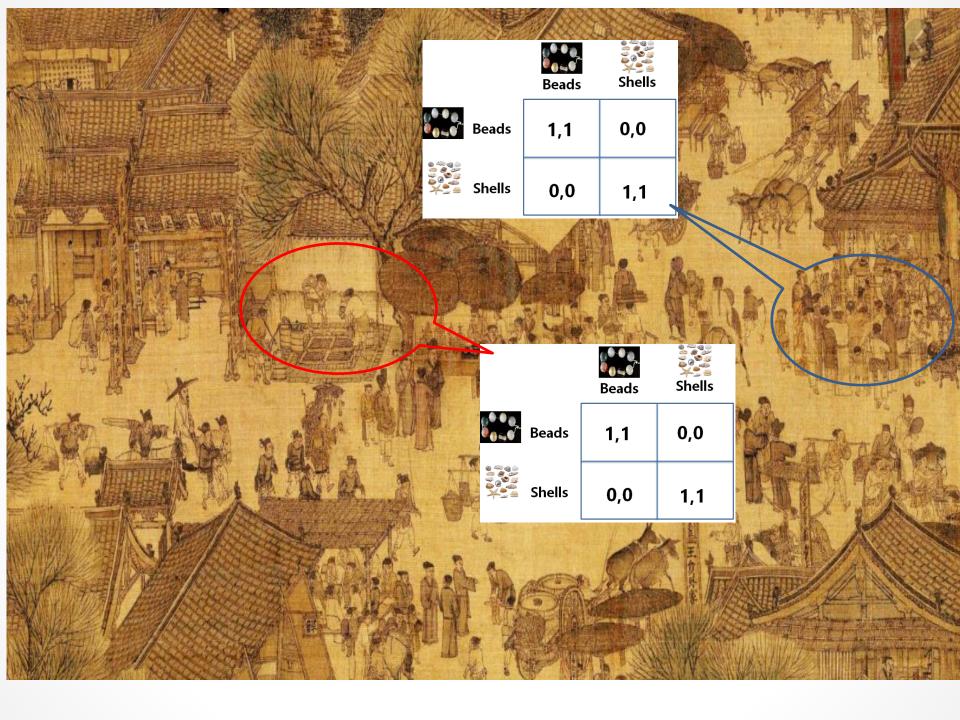
**Shells** 

1,1

0,0

0,0

1,1



- In an remote island, inhabitants have to decide to use either "beads" or "shells" as tokens of money in commerce.
- A transaction is only successful if both parties use the same form of token.
- Assume that a trader gets a utility increment of 1 if the transaction is successful and 0 if it fails.
- The general strategy to an individual is to use beads with p, i.e.,  $\sigma = (p, 1 p)$ . The population profile  $\mathbf{x} = (x, 1 x)$ .
- What is an ESS?

#### Solution

 An individual attempts to trade with a randomly selected member of the population, his payoff

$$\pi(\sigma, \mathbf{x}) = px + (1-p)(1-x) = (1-x) + p(2x-1).$$

We see that

$$x > \frac{1}{2} \longrightarrow \hat{p} = 1$$
 and  $p = 1 \longrightarrow x = 1$ .

So  $\sigma_b^* = (1,0)$  is a potential ESS with  $\mathbf{x} = (1,0)$ .

The post-entry population is:

$$\mathbf{x}_{\epsilon} = (1-\epsilon)(1,0) + \epsilon(p,1-p) = (1-\epsilon(1-p),\epsilon(1-p)).$$

## Solution: continue

In this population, the payoff for an arbitrary strategy is

$$\pi(\sigma, \mathbf{x}_{\epsilon}) = \epsilon(1-p) + p(1-2\epsilon(1-p)).$$

• The payoff for the candidate ESS is  $\pi(\sigma_b^*, \mathbf{x}_{\epsilon}) = 1 - \epsilon(1 - p)$ , so

$$\pi(\sigma_b^*, \mathbf{x}_{\epsilon}) - \pi(\sigma, \mathbf{x}_{\epsilon}) > 0,$$

$$\iff (1 - p)(1 - 2\epsilon(1 - p)) > 0.$$

• Now,  $\forall p \neq p^*$ , we have (1-p) > 0, so  $\sigma_b^*$  is an ESS if and only iff  $\epsilon(1-p) < \frac{1}{2}$ . That is  $\bar{\epsilon} = \frac{1}{2}$ .

## Solution: continue

• The strategy  $\sigma_s^* = (0, 1)$  is another ESS because the post-entry population,

$$\mathbf{x}_{\epsilon} = (\epsilon \mathbf{p}, 1 - \epsilon \mathbf{p}),$$

the payoff for an arbitrary strategy is

$$\pi(\sigma, \mathbf{x}_{\epsilon}) = (1 - \epsilon p) - p(1 - 2\epsilon p),$$

and the payoff for the candidate ESS is

$$\pi(\sigma_b^*, \mathbf{x}_{\epsilon}) = 1 - \epsilon p.$$

We have:

$$\pi(\sigma_b^*, \boldsymbol{x}_{\epsilon}) - \pi(\sigma, \boldsymbol{x}_{\epsilon}) > 0 \Longleftrightarrow p(1 - 2\epsilon p) > 0.$$

• Now,  $\forall p \neq p^*$ , we have p > 0, so  $\sigma_s^*$  is an ESS if and only if  $\epsilon p < \frac{1}{2}$ , i.e.,  $\bar{\epsilon} = \frac{1}{2}$ .

# ESSs and Nash Equilibria

- In this section, we show that ESSs in a pairwise contest population game correspond to a (possibly empty) subset of the set of Nash equilibria for an associated two-player game.
- In a pairwise contest population game, the payoff to a focal individual using  $\sigma$  in a population with profile  $\mathbf{x}$  is

$$\pi(\sigma, \mathbf{X}) = \sum_{s \in \mathbf{S}} \sum_{s' \in \mathbf{S}} p(s) \chi(s') \pi(s, s'). \tag{1}$$

Note that the above payoff is the same as a two-player game against an opponent using a strategy σ' that assigns p'(s) = x(s) ∀s ∈ S. So there is an association between a two-player game with a population game involving pairwise contests.

# Definition

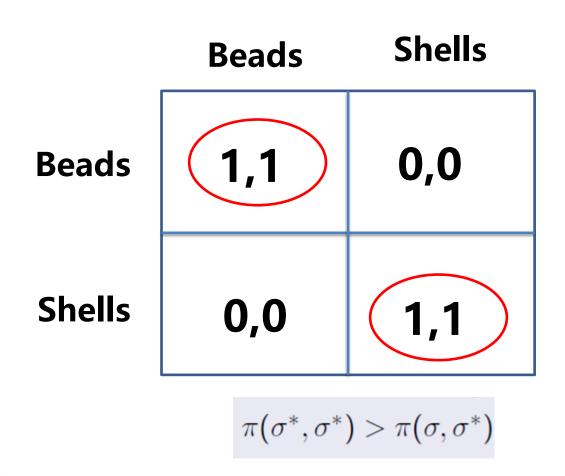
In a pairwise contest population game has payoffs given by Eq. (1), then the associated two-player game is the game with the payoffs given by the numbers  $\pi_1(s, s') = \pi(s, s') = \pi_2(s', s)$ .

#### Theorem

Let  $\sigma^*$  be an ESS in a pairwise contest, then  $\forall \sigma \neq \sigma^*$ , either

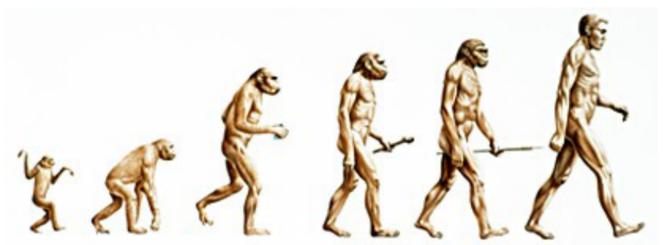
- $\bullet$   $\pi(\sigma^*, \sigma^*) > \pi(\sigma, \sigma^*)$ , or
- ②  $\pi(\sigma^*, \sigma^*) = \pi(\sigma, \sigma^*)$  and  $\pi(\sigma^*, \sigma) > \pi(\sigma, \sigma)$ .

Conversely, if either (1) or (2) holds for each  $\sigma \neq \sigma^*$  in a two-player game, then  $\sigma^*$  is an ESS in the corresponding population game.



# **Evolutionary Simulations**

- An evolutionary simulation is a stochastic game whose structure is intended to model certain aspects of evolutionary environments
  - > At each stage (or generation) there is a large set (e.g., hundreds) of agents
- Different agents may use different strategies
  - A strategy s is represented by the set of all agents that use strategy s
  - Over time, the number of agents using s may grow or shrink depending on how well s performs
- s's **reproductive success** is the fraction of agents using s at the end of the simulation,
  - $\triangleright$  i.e., (number of agents using s)/(total number of agents)



# **Replicator Dynamics**

• **Replicator dynamics** works as follows:

$$p_i^{new} = p_i^{curr} r_i / R,$$

#### where

- $\triangleright p_i^{new}$  is the proportion of agents of type i in the next stage
- $\triangleright p_i^{curr}$  is the proportion of agents of type i in the current stage
- $ightharpoonup r_i$  = average payoff received by agents of type *i* in the current stage
- $ightharpoonup R_i$  = average payoff received by all agents in the current stage
- Under the replicator dynamics, an agent's numbers grow (or shrink) proportionately to how much better it does than the average
- Probably the most popular reproduction dynamics
  - > e.g., does well at reflecting growth of animal populations

# **Example: A Simple Lottery Game**

- A repeated lottery game
- At each stage, agents make choices between two lotteries
  - "Safe" lottery: guaranteed reward of 4
  - "Risky" lottery: [0, 0.5; 8, 0.5],
    - i.e., probability ½ of 0, and probability ½ of 8
- Let's just look at stationary strategies
- Two pure strategies:
  - > S: always choose the "safe" lottery
  - > R: always choose "risky" lottery
- Many mixed strategies, one for every p in [0,1]
  - $R_p$ : probability p of choosing the "risky" lottery, and probability 1-p of choosing the "safe" lottery

# **Lottery Game with Replicator Dynamics**

- At each stage, each strategy's average payoff is 4
  - ➤ Thus on average, each strategy's population size should stay roughly constant
- Verified by simulation for *S* and *R*
- Would get similar behavior with any of the  $R_p$  strategies

