

# ΣΤΟΙΧΕΙΑ ΘΕΩΡΙΑΣ ΠΑΙΓΝΙΩΝ ΚΑΙ ΛΗΨΗΣ ΑΠΟΦΑΣΕΩΝ

## ΔΙΑΛΕΞΗ 8: ΕΞΕΛΙΚΤΙΚΕΣ ΔΥΝΑΜΙΚΕΣ

Παναγιώτης Μερτικόπουλος

Εθνικό και Καποδιστριακό Πανεπιστήμιο Αθηνών

Τμήμα Μαθηματικών



Χειμερινό Εξάμηνο, 2023–2024



## Outline

- ① Population games
- ② Exponential weights and the replicator dynamics
- ③ Asymptotic analysis and rationality



## Population games, I: Symmetric models

### Definition (Single-population games)

A *single-population game* is a collection of the following primitives:

- ▶ A continuous *population of players* modeled by  $\mathcal{N} = [0, 1]$
- ▶ A finite set of *actions / pure strategies*  $\mathcal{A} = \{1, \dots, m\}$ , common for all players in the population
- ▶ An ensemble of *payoff functions*  $v_\alpha: \mathcal{X} \equiv \Delta(\mathcal{A}) \rightarrow \mathbb{R}$ , one per  $\alpha \in \mathcal{A}$

A population game with primitives as above will be denoted by  $\mathcal{G} \equiv \mathcal{G}(\mathcal{A}, v)$ .



## Population games, I: Symmetric models

### Definition (Single-population games)

A *single-population game* is a collection of the following primitives:

- ▶ A continuous *population of players* modeled by  $\mathcal{N} = [0, 1]$
- ▶ A finite set of *actions / pure strategies*  $\mathcal{A} = \{1, \dots, m\}$ , common for all players in the population
- ▶ An ensemble of *payoff functions*  $v_\alpha: \mathcal{X} \equiv \Delta(\mathcal{A}) \rightarrow \mathbb{R}$ , one per  $\alpha \in \mathcal{A}$

A population game with primitives as above will be denoted by  $\mathcal{G} \equiv \mathcal{G}(\mathcal{A}, v)$ .

### Setup of the game:

- ▶ **Action selection** given by some  $i \mapsto \chi(i) \in \mathcal{A}$  #  $\chi: \mathcal{N} \rightarrow \mathcal{A}$  assumed measurable
- ▶ **Population state**  $x \in \mathcal{X} \equiv \Delta(\mathcal{A})$  defined as # as a measure:  $x = \lambda \circ \chi^{-1}$

$$x_\alpha = \lambda(\chi^{-1}(\alpha)) = \text{mass of players playing } \alpha \in \mathcal{A}$$

- ▶ **Anonymity:** payoffs determined by the *state* of the population, not *individual* player choices

$$v_\alpha(x) = \text{payoff to } \alpha\text{-players when the population is at state } x \in \mathcal{X}$$



## Example I: Symmetric random matching

### Example (Symmetric / Single-population random matching)

- ▶ **Given:**  $m \times m$  payoff matrix  $M$  # symmetric two-player finite game
- ▶ **Matching:** Two players are drawn randomly to play  $M$  # independent draws from  $x \in \mathcal{X}$
- ▶ If the population is at state  $x \in \mathcal{X}$ :

$$\mathbb{P}(\text{matching } \alpha \text{ against } \beta) = x_\alpha x_\beta$$

- ▶ Mean payoff to an  $\alpha$ -strategist:

$$v_\alpha(x) = \mathbb{E}_{\beta \sim x}[M_{\alpha\beta}] = \sum_{\beta \in \mathcal{A}} M_{\alpha\beta} x_\beta = (Mx)_\alpha$$

- ▶ Mean population payoff:

$$u(x) = \mathbb{E}_{\alpha, \beta \sim x}[M_{\alpha\beta}] = \sum_{\alpha, \beta \in \mathcal{A}} M_{\alpha\beta} x_\alpha x_\beta = x^\top Mx$$

### NB:

- ▶ Mean population payoff is **quadratic** in  $x$  # symmetric matching



## Population games, II: Asymmetric models

### Definition (Multi-population games)

A **multi-population game** is a collection of the following primitives:

- ▶  $N$  distinct **populations of players**:  $\mathcal{N} = \coprod_{i=1}^N [0, \rho_i]$  #  $\rho_i$  = total mass of  $i$ -th population
- ▶ A finite set of **actions / pure strategies**  $\mathcal{A}_i = \{1, \dots, m_i\}$  per population
- ▶ An ensemble of **payoff functions**  $v_{i\alpha_i}$ :  $\mathcal{X} \equiv \prod_j \Delta(\mathcal{A}_j) \rightarrow \mathbb{R}$ , one per  $\alpha_i \in \mathcal{A}_i$ ,  $i = 1, \dots, N$

A population game with primitives as above will be denoted by  $\mathcal{G} \equiv \mathcal{G}(\mathcal{N}, \mathcal{A}, v)$ .



## Population games, II: Asymmetric models

### Definition (Multi-population games)

A **multi-population game** is a collection of the following primitives:

- ▶  $N$  distinct **populations of players**:  $\mathcal{N} = \coprod_{i=1}^N [0, \rho_i]$  #  $\rho_i$  = total mass of  $i$ -th population
- ▶ A finite set of **actions / pure strategies**  $\mathcal{A}_i = \{1, \dots, m_i\}$  per population
- ▶ An ensemble of **payoff functions**  $v_{i\alpha_i}: \mathcal{X} \equiv \prod_j \Delta(\mathcal{A}_j) \rightarrow \mathbb{R}$ , one per  $\alpha_i \in \mathcal{A}_i$ ,  $i = 1, \dots, N$

A population game with primitives as above will be denoted by  $\mathcal{G} \equiv \mathcal{G}(\mathcal{N}, \mathcal{A}, v)$ .

### Setup of the game:

- ▶ **Population state**  $x \in \mathcal{X} \equiv \prod_j \Delta(\mathcal{A}_j)$ : # state of  $i$ -th population:  $x_i \in \mathcal{X}_i \equiv \Delta(\mathcal{A}_i)$   
 $x_{i\alpha_i}$  = mass of players of population  $i$  playing  $\alpha_i \in \mathcal{A}_i$
- ▶ **Anonymity**: payoffs determined by the state of the population, not *individual* player choices

$v_{i\alpha_i}(x)$  = payoff to players of population  $i$  playing  $\alpha_i \in \mathcal{A}_i$  when the population is at state  $x \in \mathcal{X}$



## Example II: Asymmetric random matching

### Example (Asymmetric / Multi-population random matching)

- ▶ **Given:** finite game  $\Gamma \equiv \Gamma(\mathcal{N}, \mathcal{A}, u)$ ;  $N$  unit mass populations
- ▶ **Matching:**  $N$  players are drawn randomly to play  $\Gamma$ , one per population # independent draws from  $x \in \mathcal{X}$
- ▶ If the population is at state  $x \in \mathcal{X}$ :

$$\mathbb{P}(\text{matching } \alpha_i \text{ against } \alpha_{-i}) = x_{i\alpha_i} \cdot x_{-i, \alpha_{-i}}$$

- ▶ Mean payoff to an  $\alpha$ -strategist of population  $i$ :

$$v_{i\alpha_i}(x) = \mathbb{E}_{\alpha_{-i} \sim x_{-i}} [u_\alpha(\alpha_i; \alpha_{-i})] = u_i(\alpha_i; x_{-i})$$

- ▶ Mean payoff of population  $i$ :

$$u_i(x) = \mathbb{E}_{\alpha \sim x} [u_i(\alpha)] = \sum_{\alpha_1 \in \mathcal{A}_1} \cdots \sum_{\alpha_N \in \mathcal{A}_N} x_{1, \alpha_1} \cdots x_{N, \alpha_N} u_i(\alpha_1, \dots, \alpha_N)$$

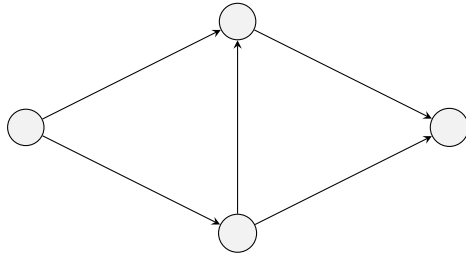
**NB:**

- ▶ Mean population payoff is **multilinear** in  $x$  # asymmetric matching





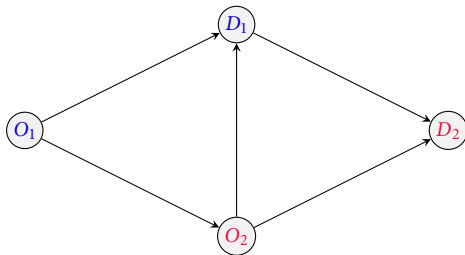
## Example III: Nonatomic congestion games



- **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$



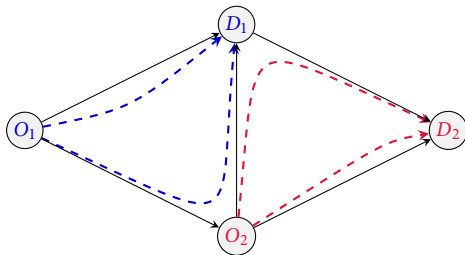
## Example III: Nonatomic congestion games



- ▶ **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- ▶ **O/D pairs  $i \in \mathcal{N}$ :** origin  $O_i$  sends  $\rho_i$  units of traffic to destination  $D_i$



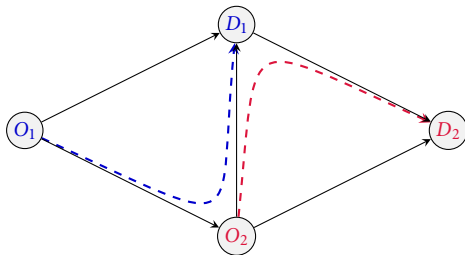
### Example III: Nonatomic congestion games



- ▶ **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- ▶ **O/D pairs**  $i \in \mathcal{N}$ : origin  $O_i$  sends  $\rho_i$  units of traffic to destination  $D_i$
- ▶ **Paths**  $\mathcal{A}_i$ : (sub)set of paths joining  $O_i \rightsquigarrow D_i$



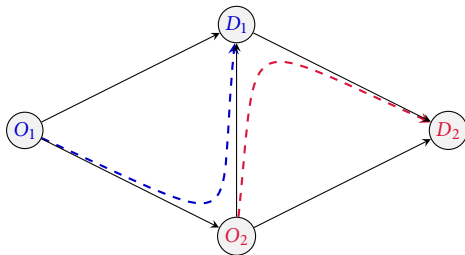
### Example III: Nonatomic congestion games



- ▶ **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- ▶ **O/D pairs  $i \in \mathcal{N}$ :** origin  $O_i$  sends  $\rho_i$  units of traffic to destination  $D_i$
- ▶ **Paths  $\mathcal{A}_i$ :** (sub)set of paths joining  $O_i \rightsquigarrow D_i$
- ▶ **Routing flow  $f_\alpha$ :** traffic along  $\alpha \in \mathcal{A} \equiv \coprod_i \mathcal{A}_i$  generated by O/D pair owning  $\alpha$

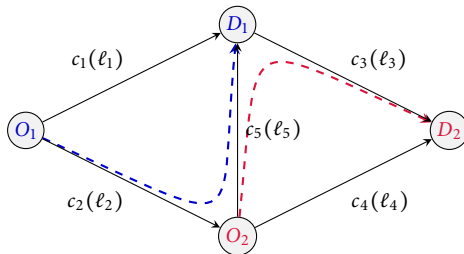


### Example III: Nonatomic congestion games



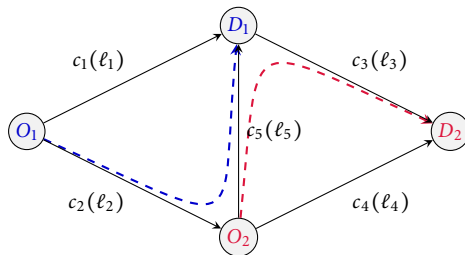
- ▶ **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- ▶ **O/D pairs  $i \in \mathcal{N}$ :** origin  $O_i$  sends  $\rho_i$  units of traffic to destination  $D_i$
- ▶ **Paths  $\mathcal{A}_i$ :** (sub)set of paths joining  $O_i \rightsquigarrow D_i$
- ▶ **Routing flow  $f_\alpha$ :** traffic along  $\alpha \in \mathcal{A} \equiv \coprod_i \mathcal{A}_i$  generated by O/D pair owning  $\alpha$
- ▶ **Load  $\ell_e = \sum_{\alpha \ni e} f_\alpha$ :** total traffic along edge  $e$

### Example III: Nonatomic congestion games



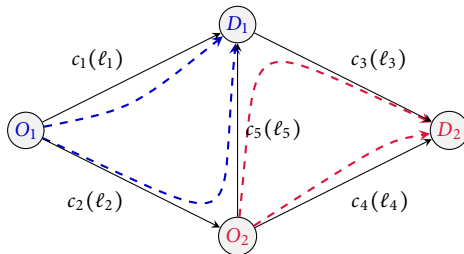
- ▶ **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- ▶ **O/D pairs  $i \in \mathcal{N}$ :** origin  $O_i$  sends  $\rho_i$  units of traffic to destination  $D_i$
- ▶ **Paths  $\mathcal{A}_i$ :** (sub)set of paths joining  $O_i \rightsquigarrow D_i$
- ▶ **Routing flow  $f_\alpha$ :** traffic along  $\alpha \in \mathcal{A} \equiv \coprod_i \mathcal{A}_i$  generated by O/D pair owning  $\alpha$
- ▶ **Load  $\ell_e = \sum_{\alpha \ni e} f_\alpha$ :** total traffic along edge  $e$
- ▶ **Edge cost function  $c_e(\ell_e)$ :** cost along edge  $e$  when edge load is  $\ell_e$

### Example III: Nonatomic congestion games



- ▶ **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- ▶ **O/D pairs  $i \in \mathcal{N}$ :** origin  $O_i$  sends  $\rho_i$  units of traffic to destination  $D_i$
- ▶ **Paths  $\mathcal{A}_i$ :** (sub)set of paths joining  $O_i \rightsquigarrow D_i$
- ▶ **Routing flow  $f_\alpha$ :** traffic along  $\alpha \in \mathcal{A} \equiv \coprod_i \mathcal{A}_i$  generated by O/D pair owning  $\alpha$
- ▶ **Load  $\ell_e = \sum_{\alpha \ni e} f_\alpha$ :** total traffic along edge  $e$
- ▶ **Edge cost function  $c_e(\ell_e)$ :** cost along edge  $e$  when edge load is  $\ell_e$
- ▶ **Path cost:**  $c_\alpha(f) = \sum_{e \in \alpha} c_e(\ell_e)$

### Example III: Nonatomic congestion games



- ▶ **Network:** multigraph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$
- ▶ **O/D pairs  $i \in \mathcal{N}$ :** origin  $O_i$  sends  $\rho_i$  units of traffic to destination  $D_i$
- ▶ **Paths  $\mathcal{A}_i$ :** (sub)set of paths joining  $O_i \rightsquigarrow D_i$
- ▶ **Routing flow  $f_\alpha$ :** traffic along  $\alpha \in \mathcal{A} \equiv \coprod_i \mathcal{A}_i$  generated by O/D pair owning  $\alpha$
- ▶ **Load  $\ell_e = \sum_{\alpha \ni e} f_\alpha$ :** total traffic along edge  $e$
- ▶ **Edge cost function  $c_e(\ell_e)$ :** cost along edge  $e$  when edge load is  $\ell_e$
- ▶ **Path cost:**  $c_\alpha(f) = \sum_{e \in \alpha} c_e(\ell_e)$
- ▶ **Nonatomic congestion game:**  $\mathcal{G} = \mathcal{G}(\mathcal{N}, \mathcal{A}, -c)$





## Mixing versus matching

⚠ **Symmetric Random matching  $\neq$  Mixed extension**

# Population matched against itself  $\implies$  *symmetric interactions*

⚠ **Asymmetric random matching = Mixed Extension**

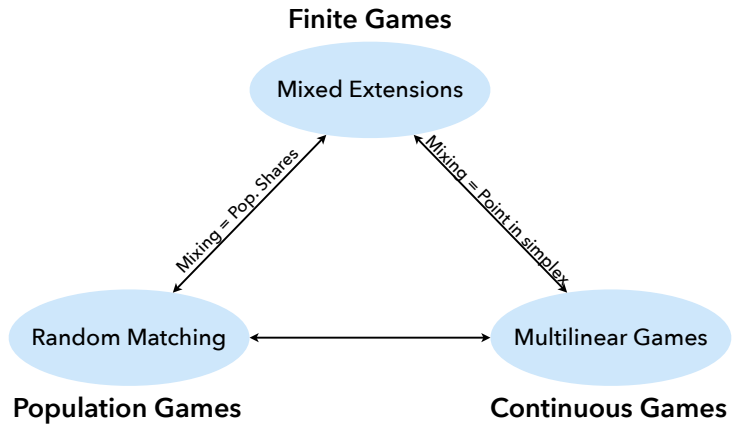
# Populations matched against each other  $\implies$  *asymmetric interactions*

⚠ **Multi-population games  $\not\equiv$  Mixed Extensions**

# Nonatomic congestion games, ...



## Relations between classes





## Nash equilibrium

### Nash equilibrium (Nash, 1950, 1951)

*“No player has an incentive to deviate from their chosen strategy if other players don’t”*

- ▶ **In finite games** (mixed extension formulation):

$$u_i(x_i^*; x_{-i}^*) \geq u_i(x_i; x_{-i}^*) \quad \text{for all } x_i \in \mathcal{X}_i, i \in \mathcal{N}$$

- ▶ **In population games:**

$$v_{i\alpha_i}(x^*) \geq v_{i\beta_i}(x^*) \quad \text{whenever } \alpha_i \in \text{supp}(x^*)$$



## Nash equilibrium

### Nash equilibrium (Nash, 1950, 1951)

*“No player has an incentive to deviate from their chosen strategy if other players don’t”*

- ▶ In finite games (mixed extension formulation):

$$u_i(x_i^*; x_{-i}^*) \geq u_i(x_i; x_{-i}^*) \quad \text{for all } x_i \in \mathcal{X}_i, i \in \mathcal{N}$$

- ▶ In population games:

$$v_{i\alpha_i}(x^*) \geq v_{i\beta_i}(x^*) \quad \text{whenever } \alpha_i \in \text{supp}(x^*)$$

### Variational formulation (Stampacchia, 1964)

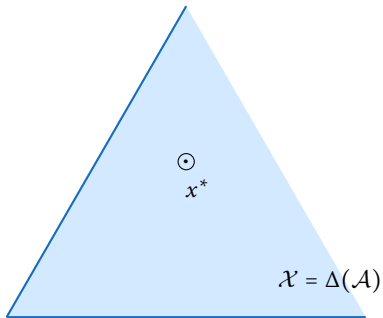
$$\langle v(x^*), x - x^* \rangle \leq 0 \quad \text{for all } x \in \mathcal{X}$$

where  $v(x) = (v_1(x), \dots, v_N(x))$  is the **payoff field** of the game



## Geometric characterization

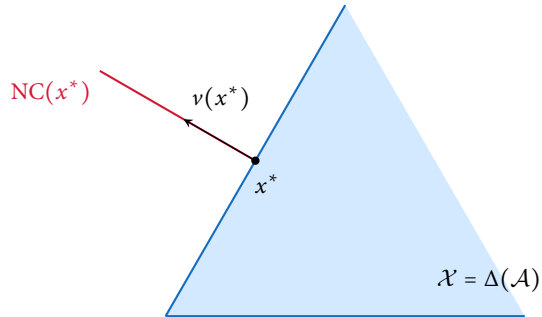
At Nash equilibrium, payoff vectors are outward-pointing





## Geometric characterization

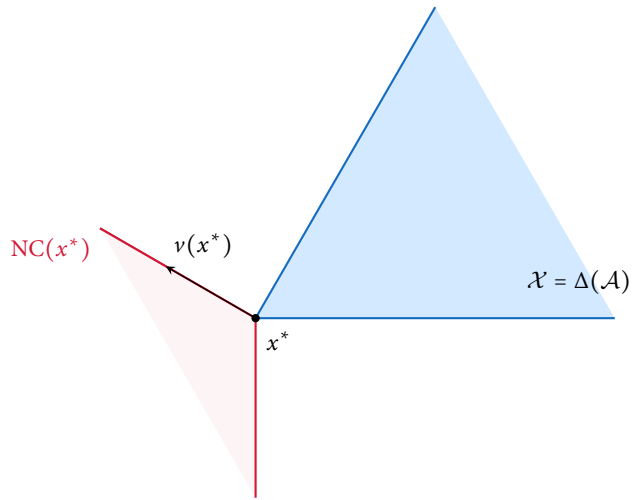
At Nash equilibrium, payoff vectors are outward-pointing





# Geometric characterization

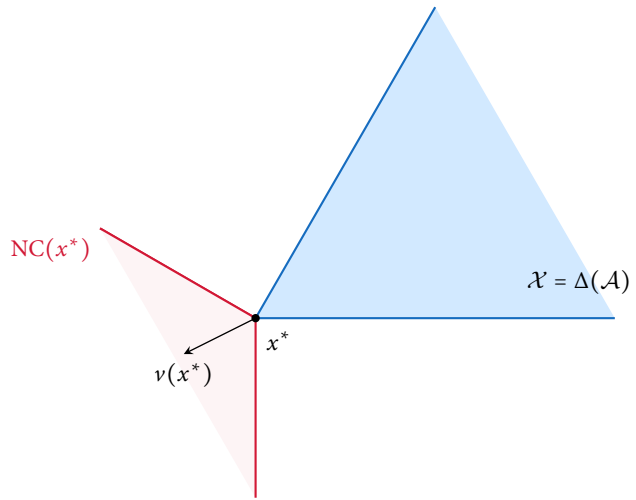
At Nash equilibrium, payoff vectors are outward-pointing





# Geometric characterization

At Nash equilibrium, payoff vectors are outward-pointing







## Outline

- ① Population games
- ② Exponential weights and the replicator dynamics
- ③ Asymptotic analysis and rationality



## Basic questions

*How do players learn from the history of play?*

*Do players end up playing a Nash equilibrium?*



## Learning, evolution and dynamics

What is “learning” in games?



## Learning, evolution and dynamics

What is “learning” in games?

### The basic process:

- ▶ Players choose strategies and receive corresponding payoffs
- ▶ Depending on outcome and information revealed, they choose new strategies and they play again
- ▶ Rinse, repeat



## Learning, evolution and dynamics

What is “learning” in games?

### The basic process:

- ▶ Players choose strategies and receive corresponding payoffs
- ▶ Depending on outcome and information revealed, they choose new strategies and they play again
- ▶ Rinse, repeat

### The basic questions:

- ▶ *How do populations evolve over time?* # Population biology
- ▶ *How do people learn in a game?* # Economics
- ▶ *What algorithms should we use to learn in a game?* # Computer science
- ▶ *Given a dynamical system on  $\mathcal{X}$ , what is its long-term behavior?* # Mathematics



## Age the First (1970's-1990's): Population Biology

- ▶ Strategies are *phenotypes* in a given species

$$z_\alpha = \text{absolute population mass of type } \alpha \in \mathcal{A}$$
$$z = \sum_\alpha z_\alpha = \text{absolute population mass}$$



## Age the First (1970's-1990's): Population Biology

- ▶ Strategies are *phenotypes* in a given species

$$z_\alpha = \text{absolute population mass of type } \alpha \in \mathcal{A}$$
$$z = \sum_\alpha z_\alpha = \text{absolute population mass}$$

- ▶ Utilities measure *fecundity / reproductive fitness*:

$$v_\alpha = \text{per capita growth rate of type } \alpha$$

- ▶ Population evolution:

$$\dot{z}_\alpha = z_\alpha v_\alpha$$



## Age the First (1970's-1990's): Population Biology

- ▶ Strategies are **phenotypes** in a given species

$$z_\alpha = \text{absolute population mass of type } \alpha \in \mathcal{A}$$
$$z = \sum_\alpha z_\alpha = \text{absolute population mass}$$

- ▶ Utilities measure **fecundity / reproductive fitness**:

$$v_\alpha = \text{per capita growth rate of type } \alpha$$

- ▶ Population evolution:

$$\dot{z}_\alpha = z_\alpha v_\alpha$$

- ▶ Evolution of population shares ( $x_\alpha = z_\alpha/z$ ):

$$\dot{x}_\alpha = \frac{d}{dt} \frac{z_\alpha}{z} = \frac{\dot{z}_\alpha z - z_\alpha \sum_\beta \dot{z}_\beta}{z^2} = \frac{z_\alpha}{z} v_\alpha - \frac{z_\alpha}{z} \sum_\beta \frac{z_\beta}{z} v_\beta$$





## Age the First (1970's-1990's): Population Biology

- ▶ Strategies are **phenotypes** in a given species

$$z_\alpha = \text{absolute population mass of type } \alpha \in \mathcal{A}$$

$$z = \sum_\alpha z_\alpha = \text{absolute population mass}$$

- ▶ Utilities measure **fecundity / reproductive fitness**:

$$v_\alpha = \text{per capita growth rate of type } \alpha$$

- ▶ Population evolution:

$$\dot{z}_\alpha = z_\alpha v_\alpha$$

- ▶ Evolution of population shares ( $x_\alpha = z_\alpha/z$ ):

$$\dot{x}_\alpha = \frac{d}{dt} \frac{z_\alpha}{z} = \frac{\dot{z}_\alpha z - z_\alpha \sum_\beta \dot{z}_\beta}{z^2} = \frac{z_\alpha}{z} v_\alpha - \frac{z_\alpha}{z} \sum_\beta \frac{z_\beta}{z} v_\beta$$

### Replicator dynamics (Taylor & Jonker, 1978)

$$\dot{x}_\alpha = x_\alpha [v_\alpha(x) - u(x)] \quad (\text{RD})$$



## Age the Second (1990's-2010's): Economics

- ▶ Agents receive **revision opportunities** to switch strategies

$$\rho_{\alpha\beta}(x) = \text{conditional switch rate from } \alpha \text{ to } \beta$$

# NB: dropping player index for simplicity



## Age the Second (1990's-2010's): Economics

- ▶ Agents receive **revision opportunities** to switch strategies

$$\rho_{\alpha\beta}(x) = \text{conditional switch rate from } \alpha \text{ to } \beta$$

# **NB**: dropping player index for simplicity

- ▶ **Pairwise proportional imitation:**

$$\rho_{\alpha\beta}(x) = x_{\beta} [v_{\beta}(x) - v_{\alpha}(x)]_+$$

# Imitate with probability proportional to excess payoff (Helbing, 1992; Schlag, 1998)



## Age the Second (1990's-2010's): Economics

- ▶ Agents receive **revision opportunities** to switch strategies

$$\rho_{\alpha\beta}(x) = \text{conditional switch rate from } \alpha \text{ to } \beta$$

# **NB**: dropping player index for simplicity

- ▶ **Pairwise proportional imitation:**

$$\rho_{\alpha\beta}(x) = x_{\beta} [v_{\beta}(x) - v_{\alpha}(x)]_+$$

# Imitate with probability proportional to excess payoff (Helbing, 1992; Schlag, 1998)

- ▶ Inflow/outflow:

$$\text{Incoming toward } \alpha = \sum_{\beta} \text{mass}(\beta \rightsquigarrow \alpha) = \sum_{\beta \in \mathcal{A}} x_{\beta} \rho_{\beta\alpha}(x)$$

$$\text{Outgoing from } \alpha = \sum_{\beta} \text{mass}(\alpha \rightsquigarrow \beta) = x_{\alpha} \sum_{\beta \in \mathcal{A}} \rho_{\alpha\beta}(x)$$



## Age the Second (1990's-2010's): Economics

- Agents receive **revision opportunities** to switch strategies

$$\rho_{\alpha\beta}(x) = \text{conditional switch rate from } \alpha \text{ to } \beta$$

# NB: dropping player index for simplicity

- Pairwise proportional imitation:**

$$\rho_{\alpha\beta}(x) = x_{\beta} [v_{\beta}(x) - v_{\alpha}(x)]_+$$

# Imitate with probability proportional to excess payoff (Helbing, 1992; Schlag, 1998)

- Inflow/outflow:

$$\text{Incoming toward } \alpha = \sum_{\beta} \text{mass}(\beta \rightsquigarrow \alpha) = \sum_{\beta \in \mathcal{A}} x_{\beta} \rho_{\beta\alpha}(x)$$

$$\text{Outgoing from } \alpha = \sum_{\beta} \text{mass}(\alpha \rightsquigarrow \beta) = x_{\alpha} \sum_{\beta \in \mathcal{A}} \rho_{\alpha\beta}(x)$$

- Detailed balance:

$$\dot{x}_{\alpha} = \text{inflow}_{\alpha}(x) - \text{outflow}_{\alpha}(x) = \dots = x_{\alpha} [v_{\alpha}(x) - u(x)] \quad (\text{RD})$$



## Age the Third (2000's-present): Computer Science

---

### Learning in finite games

---

**Require:** finite game  $\Gamma \equiv \Gamma(\mathcal{N}, \mathcal{A}, u)$

**repeat**

At each epoch  $t \geq 0$  **do simultaneously** for all players  $i \in \mathcal{N}$

# continuous time

Choose **mixed strategy**  $x_i(t) \in \mathcal{X}_i := \Delta(\mathcal{A}_i)$

# mixing

Encounter **mixed payoff vector**  $v_i(x(t))$  and get **mixed payoff**  $u_i(x(t)) = \langle v_i(t), x(t) \rangle$

# feedback phase

**until** end

---

### Defining elements

- ▶ **Time:** continuous
- ▶ **Players:** finite
- ▶ **Actions:** finite
- ▶ **Mixing:** yes
- ▶ **Feedback:** mixed payoff vectors



## Exponential weights

### Exponential reinforcement mechanism:

- ▶ Score each action based on its cumulative payoff over time:

$$y_{i\alpha_i}(t) = \int_0^t v_{i\alpha_i}(x(s)) ds$$

- ▶ Play an action with probability exponentially proportional to its score

$$x_{i\alpha_i}(t) \propto \exp(y_{i\alpha_i}(t))$$

### Exponential weights in continuous time

$$\dot{y}_{i\alpha_i} = v_{i\alpha_i}(x)$$

$$x_{i\alpha_i} = \frac{\exp(y_{i\alpha_i})}{\sum_{\beta_i \in \mathcal{A}_i} \exp(y_{i\beta_i})} \quad (\text{EW})$$



## Replicator dynamics

How do mixed strategies evolve under (EW)?





## Replicator dynamics

How do mixed strategies evolve under (EW)?

### Replicator dynamics (Taylor & Jonker, 1978)

$$\begin{aligned}\dot{x}_{i\alpha_i} &= x_{i\alpha_i} \left[ v_{i\alpha_i}(x) - \sum_{\beta_i \in \mathcal{A}_i} x_{i\beta_i} v_{i\beta_i}(x) \right] \\ &= x_{i\alpha_i} [u_i(\alpha_i; x_{-i}) - u_i(x)]\end{aligned}\tag{RD}$$

*“The per capita growth rate of a strategy is proportional to its payoff excess”*

- ◆ Hofbauer & Sigmund (1998); Weibull (1995); Hofbauer & Sigmund (2003); Sandholm (2010)



## Replicator dynamics

How do mixed strategies evolve under (EW)?

### Replicator dynamics (Taylor & Jonker, 1978)

$$\begin{aligned}\dot{x}_{i\alpha_i} &= x_{i\alpha_i} \left[ v_{i\alpha_i}(x) - \sum_{\beta_i \in \mathcal{A}_i} x_{i\beta_i} v_{i\beta_i}(x) \right] \\ &= x_{i\alpha_i} [u_i(\alpha_i; x_{-i}) - u_i(x)]\end{aligned}\tag{RD}$$

*“The per capita growth rate of a strategy is proportional to its payoff excess”*

- ◆ Hofbauer & Sigmund (1998); Weibull (1995); Hofbauer & Sigmund (2003); Sandholm (2010)

### Proposition

*Solution orbits of (EW)  $\iff$  Interior orbits of (RD)*



## Basic properties

### Replicator dynamics

$$\dot{x}_{i\alpha_i} = x_{i\alpha_i} [v_{i\alpha_i}(x) - u_i(x)] \quad (\text{RD})$$



## Basic properties

### Replicator dynamics

$$\dot{x}_{i\alpha_i} = x_{i\alpha_i} [v_{i\alpha_i}(x) - u_i(x)] \quad (\text{RD})$$

### Structural properties

↔ Weibull, 1995; Hofbauer & Sigmund, 1998

- ▶ **Well-posed:** every initial condition  $x \in \mathcal{X}$  admits unique solution trajectory  $x(t)$  that exists for all time  
# Assuming  $v$  Lipschitz
- ▶ **Consistent:**  $x(t) \in \mathcal{X}$  for all  $t \geq 0$   
# Assuming  $x(0) \in \mathcal{X}$
- ▶ **Faces are forward invariant** (“strategies breed true”):

$$x_{i\alpha_i}(0) > 0 \iff x_{i\alpha_i}(t) > 0 \quad \text{for all } t \geq 0$$

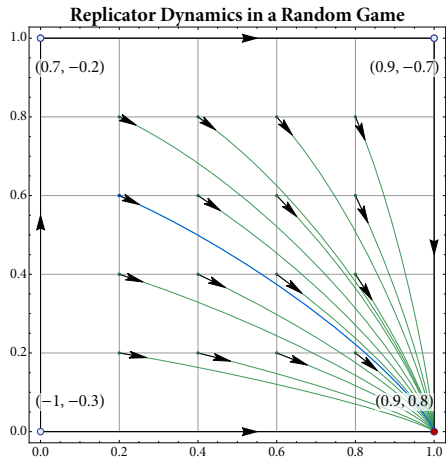
$$x_{i\alpha_i}(0) = 0 \iff x_{i\alpha_i}(t) = 0 \quad \text{for all } t \geq 0$$



## Evolution of mixed strategies I: $2 \times 2$ games

What do the dynamics look like?

# phase portraits

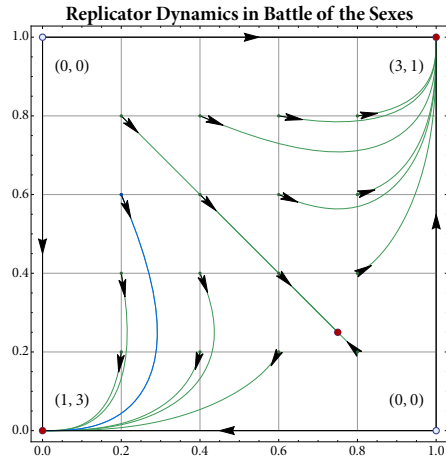




## Evolution of mixed strategies I: $2 \times 2$ games

What do the dynamics look like?

# phase portraits

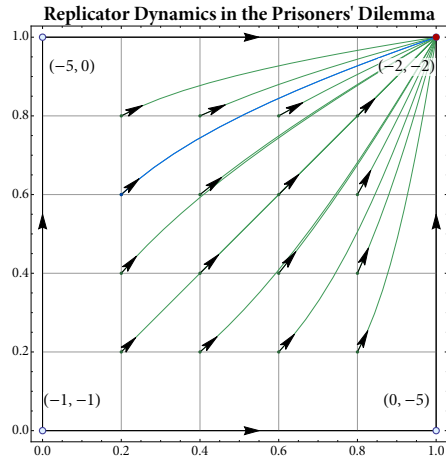




## Evolution of mixed strategies I: $2 \times 2$ games

What do the dynamics look like?

# phase portraits

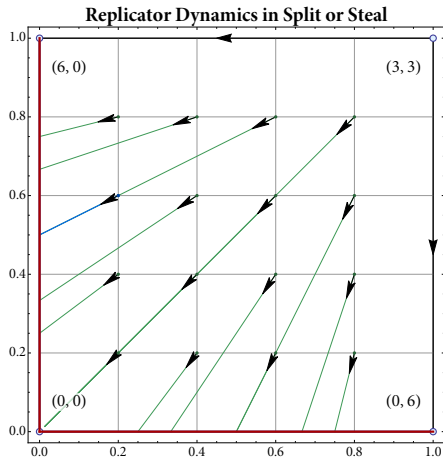




## Evolution of mixed strategies I: $2 \times 2$ games

What do the dynamics look like?

# phase portraits



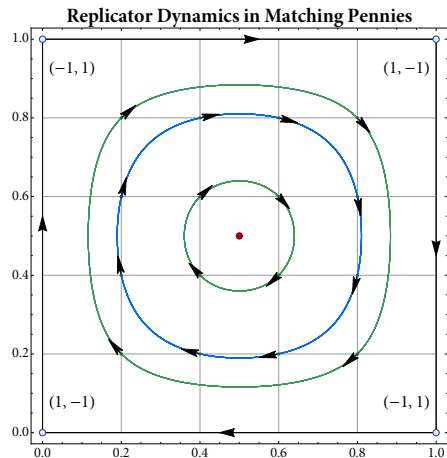




## Evolution of mixed strategies I: $2 \times 2$ games

What do the dynamics look like?

# phase portraits

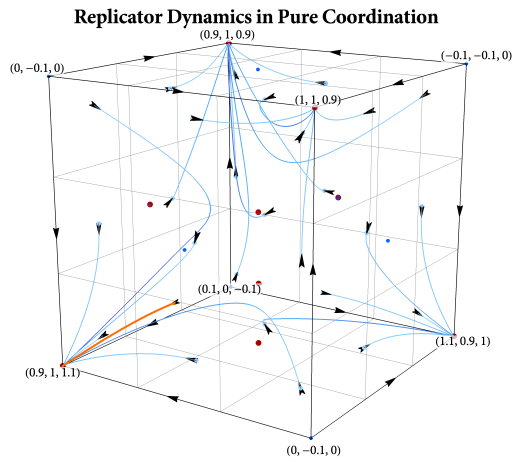




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits

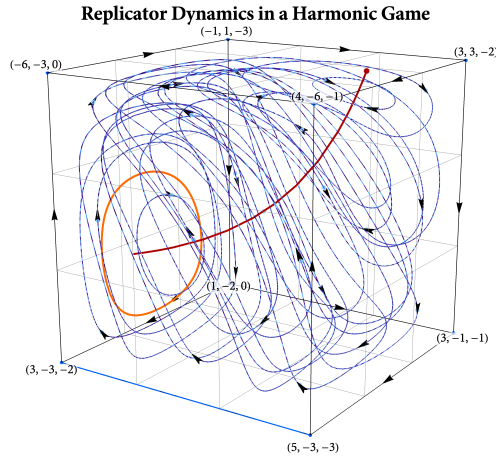




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits

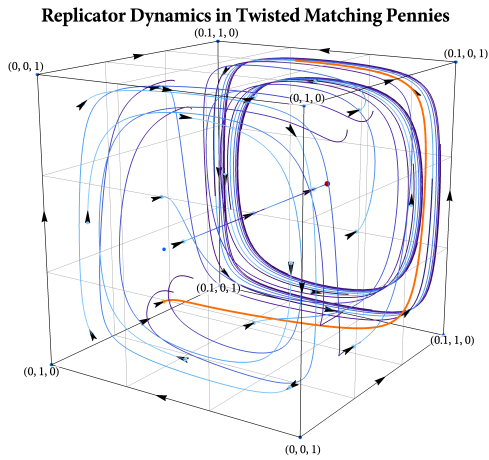




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits

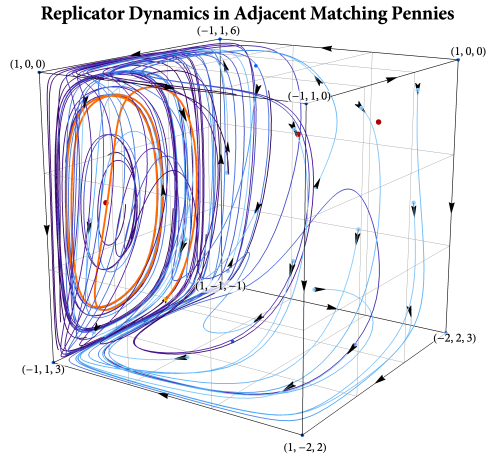




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits

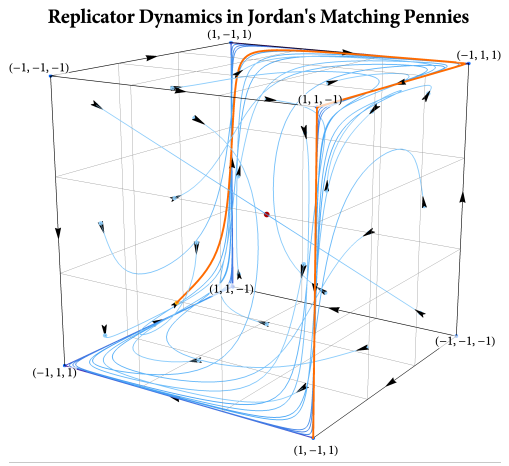




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits

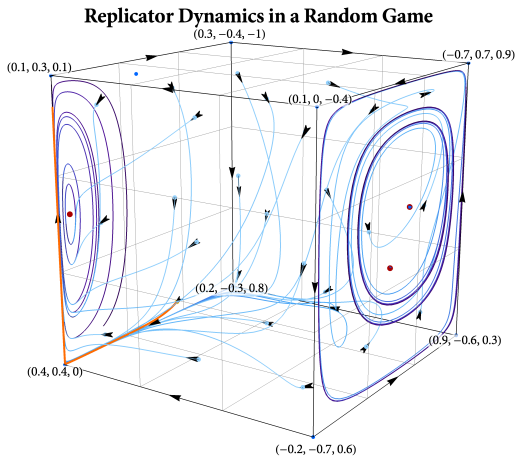




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits

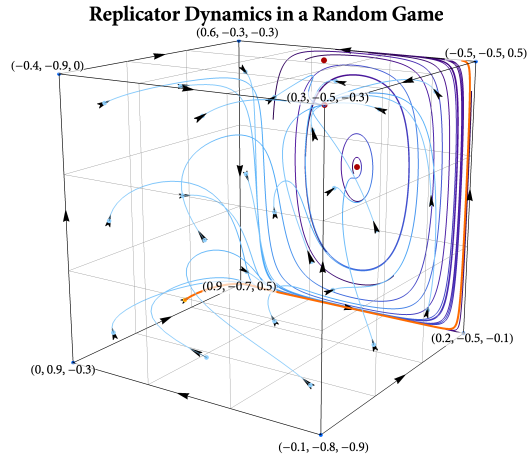




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits



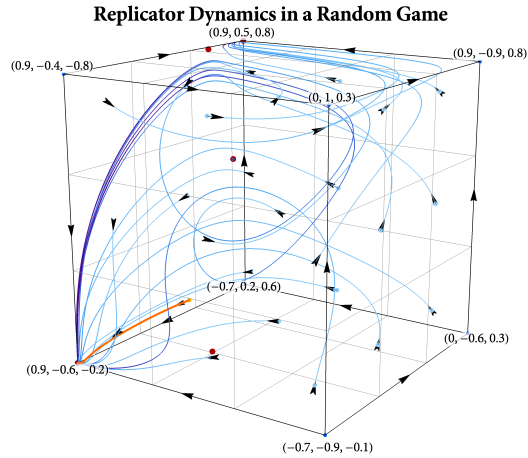




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits

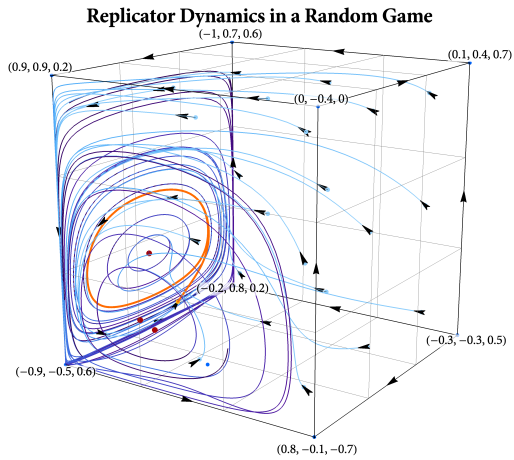




## Evolution of mixed strategies II: $2 \times 2 \times 2$ games

What do the dynamics look like?

# phase portraits





## Outline

- ① Population games
- ② Exponential weights and the replicator dynamics
- ③ Asymptotic analysis and rationality



## Dynamics and rationality

Are game-theoretic solution concepts consistent with the players' dynamics?

- ▶ Do dominated strategies die out in the long run?
- ▶ Are Nash equilibria stationary?
- ▶ Are they *stable*? Are they *attracting*?
- ▶ Do the replicator dynamics always converge?
- ▶ What other behaviors can we observe?
- ▶ ...



## Dominated strategies

Suppose  $\alpha_i \in \mathcal{A}_i$  is **dominated** by  $\beta_i \in \mathcal{A}_i$

- ▶ Consistent payoff gap:

$$v_{i\alpha_i}(x) \leq v_{i\beta_i}(x) - \varepsilon \quad \text{for some } \varepsilon > 0$$



## Dominated strategies

Suppose  $\alpha_i \in \mathcal{A}_i$  is **dominated** by  $\beta_i \in \mathcal{A}_i$

- ▶ Consistent payoff gap:

$$v_{i\alpha_i}(x) \leq v_{i\beta_i}(x) - \varepsilon \quad \text{for some } \varepsilon > 0$$

- ▶ Consistent difference in scores:

$$y_{i\alpha_i}(t) = \int_0^t v_{i\alpha_i}(x) ds \leq \int_0^t [v_{i\beta_i}(x) - \varepsilon] ds = y_{i\beta_i}(t) - \varepsilon t$$



## Dominated strategies

Suppose  $\alpha_i \in \mathcal{A}_i$  is **dominated** by  $\beta_i \in \mathcal{A}_i$

- ▶ Consistent payoff gap:

$$v_{i\alpha_i}(x) \leq v_{i\beta_i}(x) - \varepsilon \quad \text{for some } \varepsilon > 0$$

- ▶ Consistent difference in scores:

$$y_{i\alpha_i}(t) = \int_0^t v_{i\alpha_i}(x) ds \leq \int_0^t [v_{i\beta_i}(x) - \varepsilon] ds = y_{i\beta_i}(t) - \varepsilon t$$

- ▶ Consistent difference in choice probabilities

$$\frac{x_{i\alpha_i}(t)}{x_{i\beta_i}(t)} = \frac{\exp(y_{i\alpha_i}(t))}{\exp(y_{i\beta_i}(t))} \leq \exp(-\varepsilon t)$$



## Dominated strategies

Suppose  $\alpha_i \in \mathcal{A}_i$  is **dominated** by  $\beta_i \in \mathcal{A}_i$

- ▶ Consistent payoff gap:

$$v_{i\alpha_i}(x) \leq v_{i\beta_i}(x) - \varepsilon \quad \text{for some } \varepsilon > 0$$

- ▶ Consistent difference in scores:

$$y_{i\alpha_i}(t) = \int_0^t v_{i\alpha_i}(x) ds \leq \int_0^t [v_{i\beta_i}(x) - \varepsilon] ds = y_{i\beta_i}(t) - \varepsilon t$$

- ▶ Consistent difference in choice probabilities

$$\frac{x_{i\alpha_i}(t)}{x_{i\beta_i}(t)} = \frac{\exp(y_{i\alpha_i}(t))}{\exp(y_{i\beta_i}(t))} \leq \exp(-\varepsilon t)$$

### Theorem (Samuelson & Zhang, 1992)

Let  $x(t)$  be a solution orbit of (EW)/(RD). If  $\alpha_i \in \mathcal{A}_i$  is dominated, then

$$x_{i\alpha_i}(t) = \exp(-\Theta(t)) \quad \text{as } t \rightarrow \infty$$

In words: under (EW)/(RD), dominated strategies become extinct at an exponential rate.





## Dominated strategies

Suppose  $\alpha_i \in \mathcal{A}_i$  is **dominated** by  $\beta_i \in \mathcal{A}_i$

- ▶ Consistent payoff gap:

$$v_{i\alpha_i}(x) \leq v_{i\beta_i}(x) - \varepsilon \quad \text{for some } \varepsilon > 0$$

- ▶ Consistent difference in scores:

$$y_{i\alpha_i}(t) = \int_0^t v_{i\alpha_i}(x) ds \leq \int_0^t [v_{i\beta_i}(x) - \varepsilon] ds = y_{i\beta_i}(t) - \varepsilon t$$

- ▶ Consistent difference in choice probabilities

$$\frac{x_{i\alpha_i}(t)}{x_{i\beta_i}(t)} = \frac{\exp(y_{i\alpha_i}(t))}{\exp(y_{i\beta_i}(t))} \leq \exp(-\varepsilon t)$$

### Theorem (Samuelson & Zhang, 1992)

Let  $x(t)$  be a solution orbit of (EW)/(RD). If  $\alpha_i \in \mathcal{A}_i$  is dominated, then

$$x_{i\alpha_i}(t) = \exp(-\Theta(t)) \quad \text{as } t \rightarrow \infty$$

In words: under (EW)/(RD), dominated strategies become extinct at an exponential rate.

❖ **Self-check:** extend to iteratively dominated strategies



## Stationarity of equilibria

**Nash equilibrium:**  $v_{i\alpha_i}(x^*) \geq v_{i\beta_i}(x^*)$  for all  $\alpha_i, \beta_i \in \mathcal{A}_i$  with  $x_{i\alpha_i}^* > 0$

- ▶ Supported strategies have equal payoffs:

$$v_{i\alpha_i}(x^*) = v_{i\beta_i}(x^*) \quad \text{for all } \alpha_i, \beta_i \in \text{supp}(x_i^*)$$



## Stationarity of equilibria

**Nash equilibrium:**  $v_{i\alpha_i}(x^*) \geq v_{i\beta_i}(x^*)$  for all  $\alpha_i, \beta_i \in \mathcal{A}_i$  with  $x_{i\alpha_i}^* > 0$

- ▶ Supported strategies have equal payoffs:

$$v_{i\alpha_i}(x^*) = v_{i\beta_i}(x^*) \quad \text{for all } \alpha_i, \beta_i \in \text{supp}(x_i^*)$$

- ▶ Mean payoff equal to equilibrium payoff:

$$u_i(x^*) = v_{i\alpha_i}(x^*) \quad \text{for all } \alpha_i \in \text{supp}(x_i^*)$$



## Stationarity of equilibria

**Nash equilibrium:**  $v_{i\alpha_i}(x^*) \geq v_{i\beta_i}(x^*)$  for all  $\alpha_i, \beta_i \in \mathcal{A}_i$  with  $x_{i\alpha_i}^* > 0$

- ▶ Supported strategies have equal payoffs:

$$v_{i\alpha_i}(x^*) = v_{i\beta_i}(x^*) \quad \text{for all } \alpha_i, \beta_i \in \text{supp}(x_i^*)$$

- ▶ Mean payoff equal to equilibrium payoff:

$$u_i(x^*) = v_{i\alpha_i}(x^*) \quad \text{for all } \alpha_i \in \text{supp}(x_i^*)$$

- ▶ Replicator field vanishes at Nash equilibria:

$$x_{i\alpha_i}^* [v_{i\alpha_i}(x^*) - u_i(x^*)] = 0 \quad \text{for all } \alpha_i \in \mathcal{A}_i$$



## Stationarity of equilibria

**Nash equilibrium:**  $v_{i\alpha_i}(x^*) \geq v_{i\beta_i}(x^*)$  for all  $\alpha_i, \beta_i \in \mathcal{A}_i$  with  $x_{i\alpha_i}^* > 0$

- ▶ Supported strategies have equal payoffs:

$$v_{i\alpha_i}(x^*) = v_{i\beta_i}(x^*) \quad \text{for all } \alpha_i, \beta_i \in \text{supp}(x_i^*)$$

- ▶ Mean payoff equal to equilibrium payoff:

$$u_i(x^*) = v_{i\alpha_i}(x^*) \quad \text{for all } \alpha_i \in \text{supp}(x_i^*)$$

- ▶ Replicator field vanishes at Nash equilibria:

$$x_{i\alpha_i}^* [v_{i\alpha_i}(x^*) - u_i(x^*)] = 0 \quad \text{for all } \alpha_i \in \mathcal{A}_i$$

### Proposition (Stationarity of Nash equilibria)

Let  $x(t)$  be a solution orbit of (RD). Then:

$$x(0) \text{ is a Nash equilibrium} \implies x(t) = x(0) \text{ for all } t \geq 0$$



## Stationarity of equilibria

**Nash equilibrium:**  $v_{i\alpha_i}(x^*) \geq v_{i\beta_i}(x^*)$  for all  $\alpha_i, \beta_i \in \mathcal{A}_i$  with  $x_{i\alpha_i}^* > 0$

- ▶ Supported strategies have equal payoffs:

$$v_{i\alpha_i}(x^*) = v_{i\beta_i}(x^*) \quad \text{for all } \alpha_i, \beta_i \in \text{supp}(x_i^*)$$

- ▶ Mean payoff equal to equilibrium payoff:

$$u_i(x^*) = v_{i\alpha_i}(x^*) \quad \text{for all } \alpha_i \in \text{supp}(x_i^*)$$

- ▶ Replicator field vanishes at Nash equilibria:

$$x_{i\alpha_i}^* [v_{i\alpha_i}(x^*) - u_i(x^*)] = 0 \quad \text{for all } \alpha_i \in \mathcal{A}_i$$

### Proposition (Stationarity of Nash equilibria)

Let  $x(t)$  be a solution orbit of (RD). Then:

$$x(0) \text{ is a Nash equilibrium} \implies x(t) = x(0) \text{ for all } t \geq 0$$

**X The converse does not hold!**

◆ **Self-check:** All vertices of  $\mathcal{X}$  are stationary. General statement?



## Stability

Are all stationary points created equal?

### Definition (Lyapunov stability)

$x^*$  is **(Lyapunov) stable** if, for every neighborhood  $\mathcal{U}$  of  $x^*$  in  $\mathcal{X}$ , there exists a neighborhood  $\mathcal{U}'$  of  $x^*$  such that

$$x(0) \in \mathcal{U}' \implies x(t) \in \mathcal{U} \quad \text{for all } t \geq 0$$

• Trajectories that start close to  $x^*$  remain close for all time



## Stability and equilibrium

### Proposition (Folk)

Suppose that  $x^*$  is Lyapunov stable under (EW)/(RD). Then  $x^*$  is a Nash equilibrium.





## Stability and equilibrium

### Proposition (Folk)

Suppose that  $x^*$  is Lyapunov stable under (EW)/(RD). Then  $x^*$  is a Nash equilibrium.

**Proof.** Argue by contradiction:

- ▶ **Suppose that  $x^*$  is not Nash.** Then

$$v_{i\alpha_i^*}(x^*) = u_i(\alpha_i^*; x_{-i}^*) < u_i(\alpha_i; x_{-i}^*) = v_{i\alpha_i}(x^*)$$

for some  $\alpha_i^* \in \text{supp}(x_i^*)$ ,  $\alpha_i \in \mathcal{A}_i$ ,  $i \in \mathcal{N}$



## Stability and equilibrium

### Proposition (Folk)

Suppose that  $x^*$  is Lyapunov stable under (EW)/(RD). Then  $x^*$  is a Nash equilibrium.

**Proof.** Argue by contradiction:

- ▶ **Suppose that  $x^*$  is not Nash.** Then

$$v_{i\alpha_i^*}(x^*) = u_i(\alpha_i^*; x_{-i}^*) < u_i(\alpha_i; x_{-i}^*) = v_{i\alpha_i}(x^*)$$

for some  $\alpha_i^* \in \text{supp}(x_i^*)$ ,  $\alpha_i \in \mathcal{A}_i$ ,  $i \in \mathcal{N}$

- ▶ There exist  $\varepsilon > 0$  and neighborhood  $\mathcal{U}$  of  $x^*$  such that  $v_{i\alpha_i}(x) - v_{i\alpha_i^*}(x) > \varepsilon$  for  $x \in \mathcal{U}$



## Stability and equilibrium

### Proposition (Folk)

Suppose that  $x^*$  is Lyapunov stable under (EW)/(RD). Then  $x^*$  is a Nash equilibrium.

**Proof.** Argue by contradiction:

- ▶ **Suppose that  $x^*$  is not Nash.** Then

$$v_{i\alpha_i^*}(x^*) = u_i(\alpha_i^*; x_{-i}^*) < u_i(\alpha_i; x_{-i}^*) = v_{i\alpha_i}(x^*)$$

for some  $\alpha_i^* \in \text{supp}(x_i^*)$ ,  $\alpha_i \in \mathcal{A}_i$ ,  $i \in \mathcal{N}$

- ▶ There exist  $\varepsilon > 0$  and neighborhood  $\mathcal{U}$  of  $x^*$  such that  $v_{i\alpha_i}(x) - v_{i\alpha_i^*}(x) > \varepsilon$  for  $x \in \mathcal{U}$
- ▶ If  $x(t)$  is contained in  $\mathcal{U}$  for all  $t \geq 0$  (**Lyapunov property**), then:

$$y_{i\alpha_i^*}(t) - y_{i\alpha_i}(t) = c + \int_0^t [v_{i\alpha_i^*}(x(s)) - v_{i\alpha_i}(x(s))] ds < c - \varepsilon t$$



## Stability and equilibrium

### Proposition (Folk)

Suppose that  $x^*$  is Lyapunov stable under (EW)/(RD). Then  $x^*$  is a Nash equilibrium.

**Proof.** Argue by contradiction:

- ▶ **Suppose that  $x^*$  is not Nash.** Then

$$v_{i\alpha_i^*}(x^*) = u_i(\alpha_i^*; x_{-i}^*) < u_i(\alpha_i; x_{-i}^*) = v_{i\alpha_i}(x^*)$$

for some  $\alpha_i^* \in \text{supp}(x_i^*)$ ,  $\alpha_i \in \mathcal{A}_i$ ,  $i \in \mathcal{N}$

- ▶ There exist  $\varepsilon > 0$  and neighborhood  $\mathcal{U}$  of  $x^*$  such that  $v_{i\alpha_i}(x) - v_{i\alpha_i^*}(x) > \varepsilon$  for  $x \in \mathcal{U}$
- ▶ If  $x(t)$  is contained in  $\mathcal{U}$  for all  $t \geq 0$  (**Lyapunov property**), then:

$$y_{i\alpha_i^*}(t) - y_{i\alpha_i}(t) = c + \int_0^t [v_{i\alpha_i^*}(x(s)) - v_{i\alpha_i}(x(s))] ds < c - \varepsilon t$$

- ▶ We conclude that  $x_{i\alpha_i^*}(t) \rightarrow 0$ , contradicting the Lyapunov stability of  $x^*$ . □



## Asymptotic stability

Are Nash equilibria attracting?

### Definition

- ▶  $x^*$  is **attracting** if  $\lim_{t \rightarrow \infty} x(t) = x^*$  whenever  $x(0)$  is close enough to  $x^*$
- ▶  $x^*$  is **asymptotically stable** if it is stable and attracting



## Asymptotic stability

### Are Nash equilibria attracting?

#### Definition

- ▶  $x^*$  is **attracting** if  $\lim_{t \rightarrow \infty} x(t) = x^*$  whenever  $x(0)$  is close enough to  $x^*$
- ▶  $x^*$  is **asymptotically stable** if it is stable and attracting

#### Proposition (Folk)

Strict Nash equilibria are asymptotically stable under (RD).



## Asymptotic stability

### Are Nash equilibria attracting?

#### Definition

- ▶  $x^*$  is **attracting** if  $\lim_{t \rightarrow \infty} x(t) = x^*$  whenever  $x(0)$  is close enough to  $x^*$
- ▶  $x^*$  is **asymptotically stable** if it is stable and attracting

#### Proposition (Folk)

Strict Nash equilibria are asymptotically stable under (RD).

**Proof.** Compare scores:

- ▶ If  $\alpha^* = (\alpha_1^*, \dots, \alpha_N^*)$  is strict Nash  $\implies v_{i\alpha_i^*}(x^*) > v_{i\alpha_i}(x^*)$  for all  $\alpha_i \in \mathcal{A}_i \setminus \{\alpha_i^*\}$
- ▶ There exist  $\varepsilon > 0$  and a nhd  $\mathcal{U}$  of  $x^*$  such that  $v_{i\alpha_i^*}(x) - v_{i\alpha_i}(x) > \varepsilon$  for  $x \in \mathcal{U}$



## Asymptotic stability

### Are Nash equilibria attracting?

#### Definition

- ▶  $x^*$  is **attracting** if  $\lim_{t \rightarrow \infty} x(t) = x^*$  whenever  $x(0)$  is close enough to  $x^*$
- ▶  $x^*$  is **asymptotically stable** if it is stable and attracting

#### Proposition (Folk)

Strict Nash equilibria are asymptotically stable under (RD).

**Proof.** Compare scores:

- ▶ If  $\alpha^* = (\alpha_1^*, \dots, \alpha_N^*)$  is strict Nash  $\implies v_{i\alpha_i^*}(x^*) > v_{i\alpha_i}(x^*)$  for all  $\alpha_i \in \mathcal{A}_i \setminus \{\alpha_i^*\}$
- ▶ There exist  $\varepsilon > 0$  and a nhd  $\mathcal{U}$  of  $x^*$  such that  $v_{i\alpha_i^*}(x) - v_{i\alpha_i}(x) > \varepsilon$  for  $x \in \mathcal{U}$
- ▶ If  $x(t)$  remains in  $\mathcal{U}$  for all  $t \geq 0$ , then

$$y_{i\alpha_i}(t) - y_{i\alpha_i^*}(t) = c + \int_0^t [v_{i\alpha_i}(x(s)) - v_{i\alpha_i^*}(x(s))] ds < c - \varepsilon t$$

i.e.,  $\lim_{t \rightarrow \infty} x_{i\alpha_i}(t) = 0$





## Asymptotic stability

### Are Nash equilibria attracting?

#### Definition

- ▶  $x^*$  is **attracting** if  $\lim_{t \rightarrow \infty} x(t) = x^*$  whenever  $x(0)$  is close enough to  $x^*$
- ▶  $x^*$  is **asymptotically stable** if it is stable and attracting

#### Proposition (Folk)

Strict Nash equilibria are asymptotically stable under (RD).

**Proof.** Compare scores:

- ▶ If  $\alpha^* = (\alpha_1^*, \dots, \alpha_N^*)$  is strict Nash  $\implies v_{i\alpha_i^*}(x^*) > v_{i\alpha_i}(x^*)$  for all  $\alpha_i \in \mathcal{A}_i \setminus \{\alpha_i^*\}$
- ▶ There exist  $\varepsilon > 0$  and a nhd  $\mathcal{U}$  of  $x^*$  such that  $v_{i\alpha_i^*}(x) - v_{i\alpha_i}(x) > \varepsilon$  for  $x \in \mathcal{U}$
- ▶ If  $x(t)$  remains in  $\mathcal{U}$  for all  $t \geq 0$ , then

$$y_{i\alpha_i}(t) - y_{i\alpha_i^*}(t) = c + \int_0^t [v_{i\alpha_i}(x(s)) - v_{i\alpha_i^*}(x(s))] ds < c - \varepsilon t$$

i.e.,  $\lim_{t \rightarrow \infty} x_{i\alpha_i}(t) = 0$

- ▶ Proof complete by showing Lyapunov stability

◆ Left as self-check exercise





## The "folk theorem" of evolutionary game theory

### Theorem ("folk"; Hofbauer & Sigmund, 2003)

Let  $\Gamma$  be a finite game. Then, under (RD), we have:

1.  $x^*$  is a Nash equilibrium  $\implies x^*$  is stationary
2.  $x^*$  is the limit of an interior trajectory  $\implies x^*$  is a Nash equilibrium
3.  $x^*$  is stable  $\implies x^*$  is a Nash equilibrium
4.  $x^*$  is asymptotically stable  $\iff x^*$  is a strict Nash equilibrium

### Notes:

- ▶ Single-population case similar **except**  $\implies$  of (4)
- ✗ **Converse to (1), (2) and (3) does not hold!**
- ✓ Proof of (2) similar to (3)
- ▶ Proof of " $\iff$ " in (4): requires different techniques

• Do as self-check



## References I

- [1] Helbing, D. A mathematical model for behavioral changes by pair interactions. In Haag, G., Mueller, U., and Troitzsch, K. G. (eds.), *Economic Evolution and Demographic Change: Formal Models in Social Sciences*, pp. 330–348. Springer, Berlin, 1992.
- [2] Hofbauer, J. and Sigmund, K. *Evolutionary Games and Population Dynamics*. Cambridge University Press, Cambridge, UK, 1998.
- [3] Hofbauer, J. and Sigmund, K. Evolutionary game dynamics. *Bulletin of the American Mathematical Society*, 40(4):479–519, July 2003.
- [4] Nash, J. F. Equilibrium points in  $n$ -person games. *Proceedings of the National Academy of Sciences of the USA*, 36:48–49, 1950.
- [5] Nash, J. F. Non-cooperative games. *The Annals of Mathematics*, 54(2):286–295, September 1951.
- [6] Samuelson, L. and Zhang, J. Evolutionary stability in asymmetric games. *Journal of Economic Theory*, 57:363–391, 1992.
- [7] Sandholm, W. H. *Population Games and Evolutionary Dynamics*. MIT Press, Cambridge, MA, 2010.
- [8] Schlag, K. H. Why imitate, and if so, how? A boundedly rational approach to multi-armed bandits. *Journal of Economic Theory*, 78(1): 130–156, 1998.
- [9] Stampacchia, G. Formes bilineaires coercitives sur les ensembles convexes. *Comptes Rendus Hebdomadaires des Séances de l'Académie des Sciences*, 1964.
- [10] Taylor, P. D. and Jonker, L. B. Evolutionary stable strategies and game dynamics. *Mathematical Biosciences*, 40(1-2):145–156, 1978.
- [11] Weibull, J. W. *Evolutionary Game Theory*. MIT Press, Cambridge, MA, 1995.