## ESS and Replicator Dynamics

## Tutorial

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## Bibliography

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Capture the interplay between individual and collective behaviour in a large population of identical agents.

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- ¿ Stock market?


## Strategies

## Pure strategy

Individual behavior: choice of a phenotye or trait (evolution) or strategy (routing, sociology) $x$ in a trait or strategy space $X$ either finite or infinite.

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- individual
- population monomorphic or polymorphic


## Generating function

An individual using strategy $x$ in a population using $p$ gets $G(x, p)$

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## Generating function

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Linear case $G$ may be given via an pairwise encounter function $H(x, y)$

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## ESS

Interpret "fitness" as reproductive success. Assume most of the population behaves according to strategy $p$, while a small proportion $\varepsilon$ behaves according to $q$. The overall population then behaves according to

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q_{\varepsilon}:=(1-\varepsilon) p+\varepsilon q
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$$

and in the linear case

$$
[F(p, p)=F(q, p)] \Rightarrow F(p, q)>F(q, q) .
$$

## A menagery of concepts

Other concepts, close to ESS, have been introduced : locally superior, evolutioarily robust (ERS), uninvadable, continuously stable, etc.

Definition A strategy $p$ is locally superior if there exists a neighborhood $\mathcal{N}(p)$

$$
\forall q \in \mathcal{N}(p), q \neq p, \quad F(p, q)>F(q, q) .
$$

In the infinite cases, depends on the topology used on $\Delta(X)$.
Definition A strategy is Evolutionarily Robust if it is locally superior in the weak topology of $\Delta(X)$.

Theorem In the linear case, ERS $\Rightarrow$ ESS. In the finite linear case, ESS $\Leftrightarrow$ ERS.

## Relation with Nash

Define a non zero sum two players game on $\Delta(X) \times \Delta(X)$ by

$$
J_{1}(p, q)=F(p, q), \quad J_{2}(p, q)=F(q, p) .
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Then the first ESS condition (linear case) is equivalent to ( $p, p$ ) being a Nash point of this game. As a consequence we get

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Theorem [Von Neuman 1944, Kuhn 1950, Wardrop 1952] Let $p$ be an ESS.

1. $\forall x \in X, \quad G(x, p) \leq F(p, p)$,
2. let $N=\{x \in X \mid G(x, p)<F(p, p)\}$, then, $p(N)=0$.

## Finite linear case

Let $X=\left\{x_{i}, i=1, \ldots, n\right\} . \Delta(X)$ is the simplex of $\mathbb{R}^{n}$.
Let $H\left(x_{i}, x_{j}\right)=a_{i j}, \quad A=\left(a_{i j}\right), \quad F(p, q)=\langle p, A q\rangle$
Wardrop $\Leftrightarrow \quad p=\binom{p_{1}}{0}, \quad A p=\binom{v \mathbb{1}}{r_{2}}, \quad r_{2}<v \mathbb{1} ;$

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The second condition is a second order condition. Partition $A$ according to the partition of $p$. A sufficient condition is that the restriction of the quadratic form $\left\langle q_{1}, A_{11} q_{1}\right\rangle$ to the orthogonal subspace of $\mathbb{1}$ be negative definite.

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A necessary condition is that the same restricted quadratic form for the submatrix $A_{22}$ of $A$ corresponding to the strictly positive elements of $p_{1}$ be nonpositive definite.

## A second order criterion

Given a $2 \times 2$ matrix

$$
M=\left(\begin{array}{ll}
a & b \\
c & d
\end{array}\right)
$$

define its symmetric difference $\sigma(M)=a-b-c+d$.
Given an $n \times n$ matrix $M$, define the $n-1 \times n-1$ matrix $\sigma(M)$ by replacing each block of four adjacent numbers in $M$ by their symmetric difference.

The sufficient condition is that $\sigma\left(A_{11}\right)+\sigma\left(A_{11}\right)^{t}<0$.
The necessary condition is that $\sigma\left(A_{22}\right)+\sigma\left(A_{22}\right)^{t} \leq 0$.

## Example

A rare example of a non trivial ESS in a continuous setting.
$X=[0,1], H(x, y)=\max \{x-y, \lambda(y-x)\}, \lambda \in(0,1)$.
Theorem $p$ is a global ERS:

$$
p=\frac{\lambda}{1+\lambda} \delta_{0}+\frac{1}{1+\lambda} \delta_{1} .
$$

For a given $q \in \Delta(X)$, define $x_{0}[\ldots]$ such that $q\left(\left[0, x_{0}\right]\right)=\lambda /(1+\lambda)$, and $a=\int_{0}^{x_{0}} x q(\mathrm{~d} x), x_{1}$ and $b$ symmetrically in the neighborhood of 1 . We show

$$
F(p, q)-F(q, q) \geq(1+\lambda)\left(\frac{a^{2}}{x_{0}}+\frac{b^{2}}{1-x_{1}}\right) .
$$

## Hawk and Doves

Sharing a prey

| Opponent <br> Him | $D$ | $H$ |
| ---: | :---: | :---: |
| $D$ | $1 / 2$ | 0 |
| $H$ | 1 | $-\theta$ |

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Wardrop condition:

$$
0.5 p=p-\theta(1-p) \quad \Rightarrow \quad p=\frac{\theta}{\theta+0.5}
$$

## Replicator Dynamics

If $G(x, q)$ measures the excess of births over deaths, like begets like, and $q$ is interpreted as the distribution of strategies among the population, it follows that

$$
\dot{q}(t, x)=q(t, x)[G(x, q(t))-F(q(t), q(t))] .
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\begin{equation*}
\dot{q}(t, x)=q(t, x)[G(x, q(t))-F(q(t), q(t))] . \tag{1}
\end{equation*}
$$

This is called the replicator dynamics. "Evolutionary game ?"
Theorem In the finite linear case, every locally stable point of (1) is a Nash point. Every ESS is a locally asymptotically stable point of (1) and its attraction basin contains a neighborhood of the relative interior of the lowest dimensional face of $\Delta(X)$ it lies on.

## Proof

Use the Lyapunov function

$$
V(q)=\sum_{i} p_{i} \ln \left(\frac{p_{i}}{q_{i}}\right)
$$

In $x \leq 1+x \Rightarrow V(q)>0$ if $q \neq p$ and

$$
\dot{V}(q)=F(q, q)-F(p, q)
$$

## Stability in the infinite case

with A. J. Shaiju

We are investigating the stability of a differential equation over a space of measures. The Lyapunov function used to prove the above theorem is not continuous in the weak topology in the infinite case. We do not know whether an ERS (even global) is necessarily a stable point of (1).

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Theorem If $H(x, y)=H(y, x)$, an ERS is locally asymptotically stable.

Proof: use the Lyapunov function $F(p, p)-F(q, q)$.

## Two by two case

Players have only two pure strategies (two phenotypes in the population). The game is described by a $2 \times 2$ matrix

$$
A=\left(\begin{array}{ll}
a & b \\
c & d
\end{array}\right)
$$

Let $\sigma=a-b-c+d$, (assume here $\sigma \neq 0$.)
Candidate Nash (via Wardrop condition) $p^{\star}=\frac{d-c}{\sigma}$. Is an ESS iff $\sigma<0$.

Example : for Hawk and Dove, $\sigma=-0.5-\theta<0$.

## $2 \times 2$ ESS

In this most simple context,
Definition $p$ is an ESS if

1. $(p, p)$ is Nash, $\Longleftrightarrow$

$$
\forall q \in[0,1], \quad\left(\begin{array}{cc}
p-q & q-p
\end{array}\right) A\binom{p}{1-p} \geq 0
$$

and
2. If $p \notin\{0,1\}$, then necessarily $p=p^{\star}$ and the above is 0 ,

$$
\forall q \neq p^{\star}, \quad\left(\begin{array}{ll}
p^{\star}-q & q-p^{\star}
\end{array}\right) A\binom{q}{1-q}>0
$$

## $2 \times 2$ ESS continued

Corollary If $a \neq c$ and $b \neq d$, there always exists an ESS

1. If $\sigma=0$ or $\sigma \neq 0$, but $p^{\star} \notin(0,1)$, the unique ESS is pure,
2. If $\sigma \neq 0$ and $p^{\star} \in(0,1)$,
(a) If $\sigma<0, p^{\star}$ is the only ESS,
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Proof

$$
\left(\begin{array}{cc}
p^{\star}-q & q-p^{\star}
\end{array}\right) A\binom{p^{\star}}{1-p^{\star}}=-\sigma\left(q-p^{\star}\right)^{2} .
$$

## Replicator equation

If "fitness" measures the reproductive efficiency, i.e. the rate of the excess of births over deaths (possibly negative), $p$ evolves accordingly to the simple replicator equation which becomes here

If $\sigma \neq 0$,

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If $\sigma=0$,

$$
\dot{p}=(b-d) p(1-p)
$$

# Elementary replicator equation analysis 




## Population game

Two populations, with two behaviors each, interacting with each other. States $p_{1}$ and $p_{2}$, proportions of individuals using the first behavior.

Fitness accrued by an individual of population $i$ using its first behavior $a_{i} p_{j}+b_{i}\left(1-p_{j}\right)$, and with its second behavior $c_{i} p_{j}+d_{i}\left(1-p_{j}\right)$.

Replicator equations become

$$
\dot{p}_{i}=\sigma_{i} p_{i}\left(1-p_{i}\right)\left(p_{j}-p_{j}^{\star}\right) .
$$

## Phase portrait

1. Unique (pure) Nash equilibria are stable,
2. If $\left(p_{1}^{\star}, p_{2}^{\star}\right) \in(0,1) \times(0,1)$,
(a) If $\sigma_{1} \sigma_{2}<0$, the unique Nash equilibrium $\left(p_{1}^{\star}, p_{2}^{\star}\right)$ is a center, and all trajectories are periodic.
(b) If $\sigma_{1} \sigma_{2}>0$, the Nash equilibrium ( $p_{1}^{\star}, p_{2}^{\star}$ ) is a saddle, the two pure Nash are stable.

Proof The following function is a first integral:

$$
V\left(p_{1}, p_{2}\right)=\sum_{i=1}^{2}(-1)^{i} \sigma_{i}\left[p_{i}^{\star} \ln \frac{p_{i}^{\star}}{p_{i}}+\left(1-p_{i}^{\star}\right) \ln \frac{1-p_{i}^{\star}}{1-p_{i}}\right]
$$

## Lynxes and wolves

| $L \backslash W$ |  | cow. |  | agr. |
| :---: | :---: | :---: | :---: | :---: |
| cow. |  | $1-\lambda$ |  | 1 |
|  | $\lambda$ |  | 0 |  |
| agr. | $1-\mu$ | 0 |  | $-\theta$ |
|  |  |  | $1-\nu$ |  |
| $\lambda+\mu>1>\nu$ |  |  |  |  |

$$
\begin{array}{ll}
\sigma_{1}=\lambda+\mu-\nu, & p_{2}^{\star}=(1-\nu) /(\lambda+\mu-\nu) \\
\sigma_{2}=-\lambda-\theta, & p_{1}^{\star}=\theta /(\lambda+\theta)
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\sigma_{1}=\lambda+\mu-\nu, & p_{2}^{\star}=(1-\nu) /(\lambda+\mu-\nu) \\
\sigma_{2}=-\lambda-\theta, & p_{1}^{\star}=\theta /(\lambda+\theta)
\end{array}
$$

Draw case $\quad \lambda=\nu=0,5, \quad \mu=0,75, \quad \theta=1,5$.

Wolves and Lynxer


## Two identical populations

Two identical but distinct populations. Potentially, $p_{1} \neq p_{2}$.
But $A_{1}=A_{2}$, hence $\sigma_{1}=\sigma_{2}$. Assume further that $\sigma_{i}<0$. The single population replicator equation is stable at $p^{\star}$, but $\sigma_{1} \sigma_{2}>0$ hence ( $p^{\star}, p^{\star}$ ) is not stable.

The diagonal $p_{1}=p_{2}=p$ represents the solution of the one-population game. Highly unstable in the two-population game.

Two identical populations


