

# Stability Analysis on Replicator Games with Perturbations

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# ToG

1. Replicator Dynamics
2. Stability Analysis
3. Single Population Game
4. Multi-Population Game
5. Perturbations

# 1. Replicator Dynamics

# Replicator Dynamics

## Game Settings

- A game within a population.
- All have the same game matrix  $\Pi$  and strategy:  
$$S = \{s_1, \dots, s_n\}.$$
- Ratio of players playing strategy  $s_k$  at an instance is  $x_k$ . Let the state of game be  
$$\mathbf{x} = [x_1, \dots, x_n].$$
- The expected payoff playing  $s_k$  is  
$$\pi_{k\sigma} = (\Pi \mathbf{x})_k.$$

# Replicator Dynamics

## Replicator Dynamics:

- At any instance, players of the population are paired up to play the game.
- If the enemy has a higher payoff, one might replicate the enemy's strategy.
- The game dynamics:

$$\dot{x}_k = \alpha x_k [(\Pi x)_k - x^T \Pi x]$$

(replicator dynamics has  $n - 1$  DEs)

## 2. Stability Analysis

# Stability Analysis

For game described as

$$\dot{x} = f(x)$$

It can be approximated linearly as

$$\dot{x} = \frac{\partial f}{\partial x}(x - a) := A(x - a)$$

Where  $f(a) = 0$  is a stationary point.

- If  $A < 0$ , it is negative definite, then it is asymptotically stable.
- Imaginary E-val. of  $A$  creates oscillatory orbits.

# Stability Analysis

- Nash Equilibrium

$x^N$  is Nash(pure/mixed) of  $\Pi$



$$f(x^N) = \mathbf{0} \text{ and } \frac{\partial f(x^N)}{\partial x} \neq 0$$

This definition comes straightly from the dynamics of the phase portrait.



### 3. Single Population Game

# Single Population RPS Game

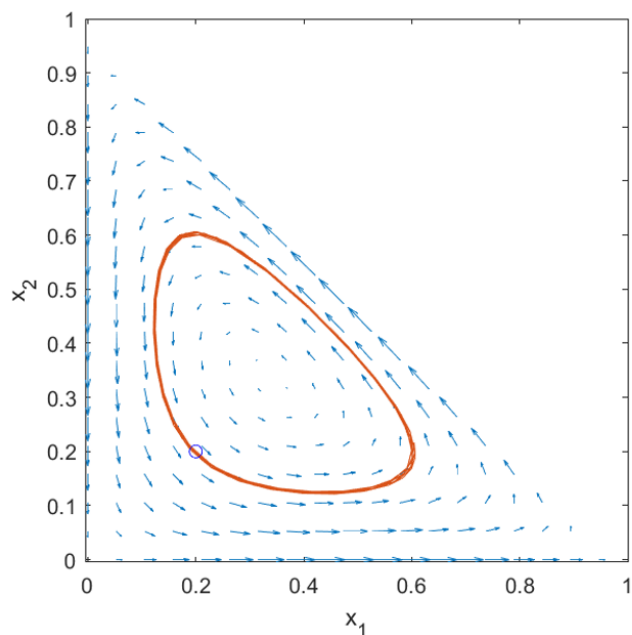
(Rock-Paper-Scissors Game)

$P \backslash \text{Opponent}$	Rock $x_1$	Paper $x_2$	Scissor $1 - x_1 - x_2$
Rock $x_1$	0	-1	a
Paper $x_2$	a	0	-1
Scissor $1 - x_1 - x_2$	-1	a	0

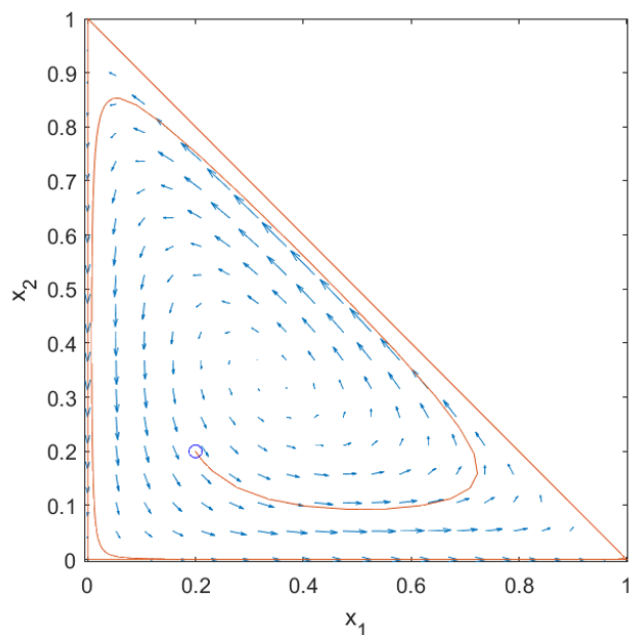
$$\Pi = \begin{bmatrix} 0 & -1 & a \\ a & 0 & -1 \\ -1 & a & 0 \end{bmatrix} \quad \mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ 1 - x_1 - x_2 \end{bmatrix}$$

# Single Population RPS Game

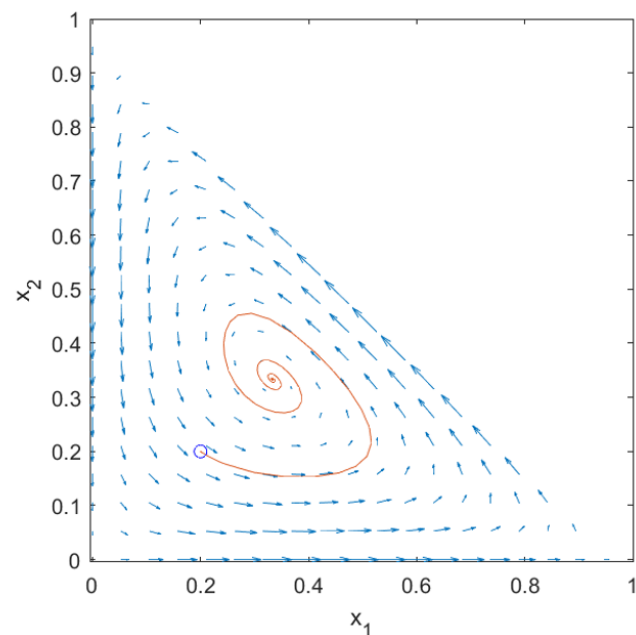
(Rock-Paper-Scissors Game)



$$a = 1$$



$$a = 0.5$$



$$a = 2$$

## 4. Multi-Population Game

# Extension to Multi-Population

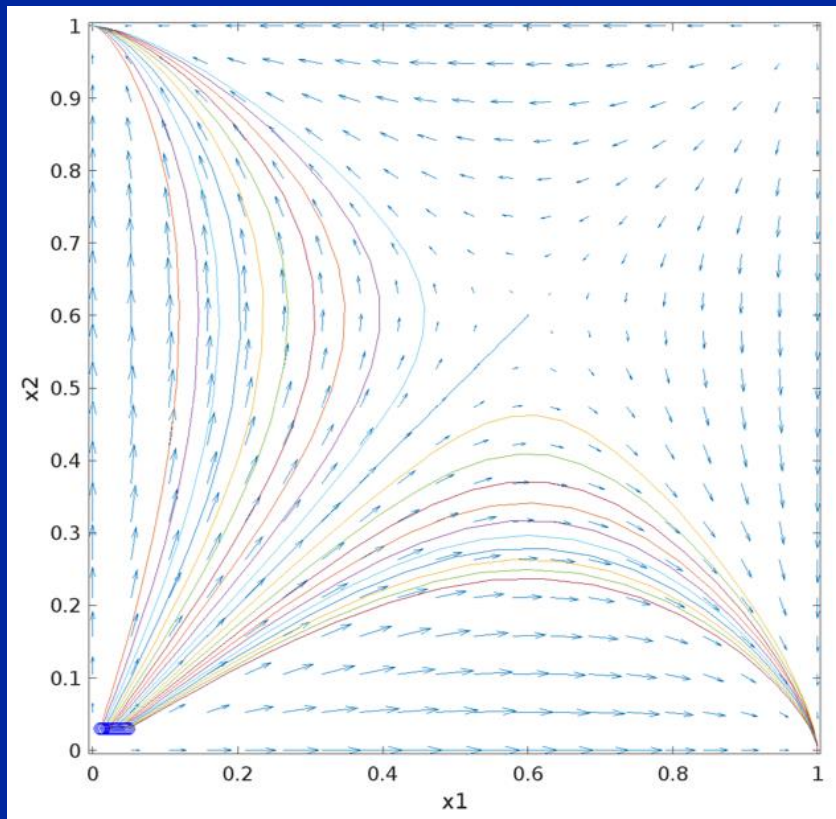
- Population  $\mathbf{x}$  and  $\mathbf{y}$  compete with each other.
- Same strategy set  $S$  and payoff matrix  $\Pi$ .
- The Dynamics:

$$\begin{aligned}\dot{x}_k &= \alpha x_k [(\Pi \mathbf{y})_k - \mathbf{x}^T \Pi \mathbf{y}] \\ \dot{y}_k &= \beta y_k [(\Pi \mathbf{x})_k - \mathbf{y}^T \Pi \mathbf{x}]\end{aligned}$$

- The same idea of changing strategy based on difference of payoff.

# Extension to Multi-Population

## Game of Chicken (Cold War)



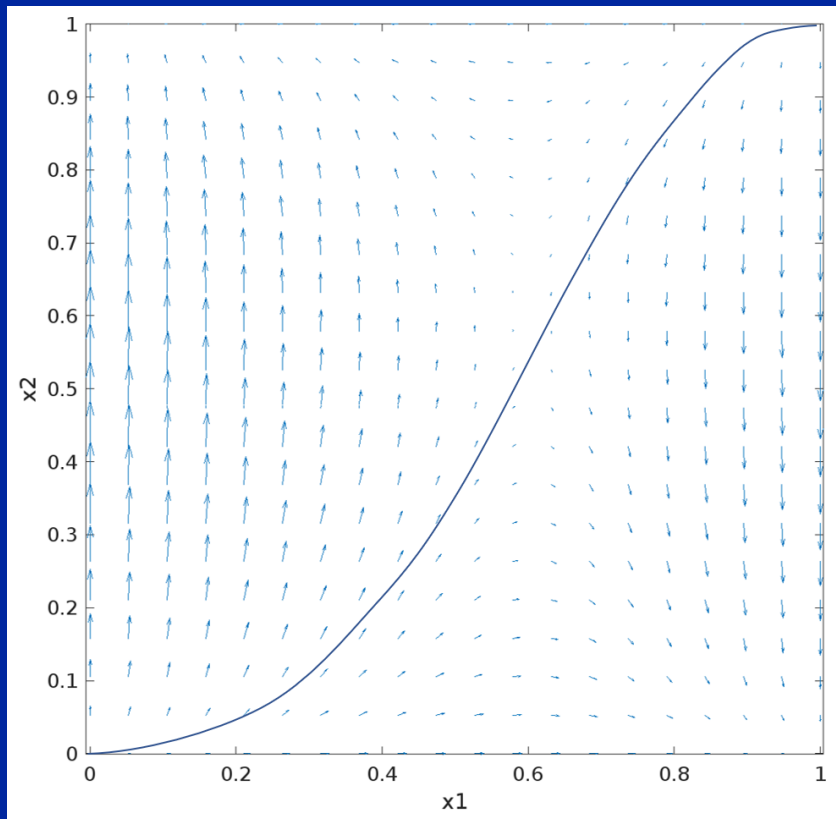
US \ USSR	Launch $x_2$	Don't $1 - x_2$
Launch $x_1$	-5,-5	3,-3
Don't $1 - x_1$	-3,3	0,0

$$\dot{x}_1 = x_1(1 - x_1)(3 - 5x_2)$$

$$\dot{x}_2 = x_2(1 - x_2)(3 - 5x_1)$$

# Extension to Multi-Population

## Game of Chicken (Cold War)



US \ USSR	Launch $x_2$	Don't $1 - x_2$
	Launch $x_1$	Don't $1 - x_1$
	-5,-5	3,-3
	-3,3	0,0

$$\dot{x}_1 = x_1(1 - x_1)(3 - 5x_2)$$

$$\dot{x}_2 = 5x_2(1 - x_2)(3 - 5x_1)$$

# Extension to Multi-Population

## Problems:

1. No derivation, formed from viewing the population as a whole.
2. The dynamical equations neglect the individual choices.
3. The dynamics of  $x_n$  is not compatible.



# 5. Perturbations

# Perturbations

- Perturbed Dynamics:

$$\dot{x}_k = \alpha x_k [(\Pi \mathbf{x})_k - \mathbf{x}^T \Pi \mathbf{x} - \varepsilon \cdot v(\mathbf{x})]$$

where

$$v(\mathbf{x}) = - \sum_{k=1}^n x_k \ln x_k$$

**“The Gibbs entropy is a measure of unexpectedness.”**

# Perturbations

## Surprise:

- Expectedness:  $\mathcal{S}(1) = 0$
- Unexpectedness:  $\mathcal{S}(0) = \infty$
- Additivity:  $\mathcal{S}(x_1 \cdot x_2) = \mathcal{S}(x_1) + \mathcal{S}(x_2)$
- Continuity:  $\mathcal{S} \in \mathcal{C}$   
 $\rightarrow \mathcal{S}(x) = -\ln x$

## Expected Surprise:

$$v(x) = - \sum_k x_k \ln x_k$$

# Perturbations

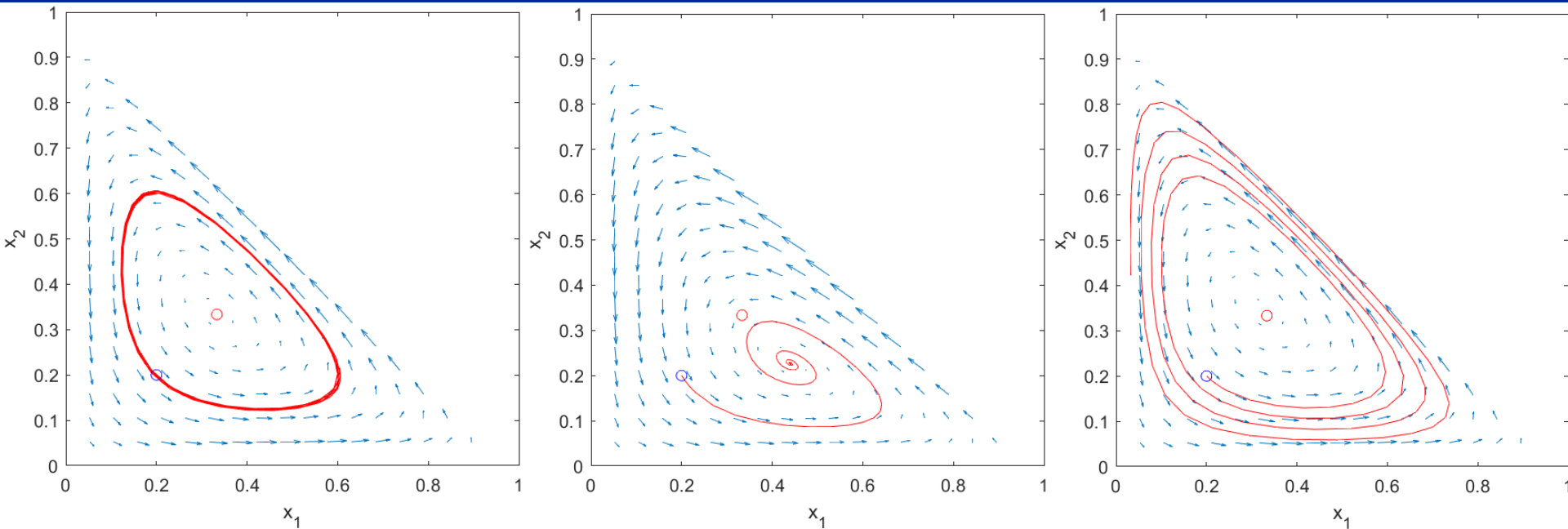
$$\dot{x}_k = \alpha x_k [(\Pi x)_k - x^T \Pi x - \varepsilon \cdot v(x)]$$

The  $\varepsilon$  term measures how the population cope with unexpectedness.

- $\varepsilon > 0$ : stick to ones strategy even if lost.
- $\varepsilon < 0$ : change strategy even if won.

# Perturbations

## Perturbed Rock-Paper-Scissors Game ( $a = 1$ )



$$\varepsilon = 0$$

$$\varepsilon = 0.1$$

$$\varepsilon = -0.01$$

# Perturbations

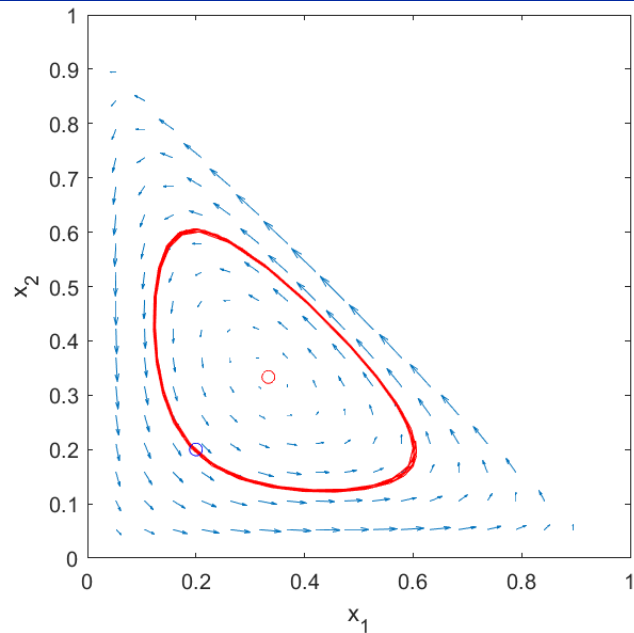
- A better choice:

$$\dot{x}_k = \alpha x_k [(\Pi x)_k - x^T \Pi x - \varepsilon \cdot (v(x) + \ln x_k)]$$

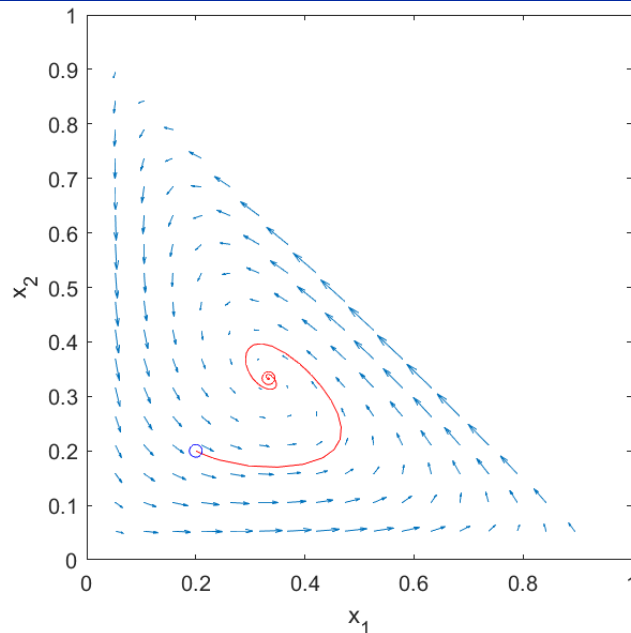
1. Preserves equilibrium of uniform distribution.
2. Compatible with dynamics of  $x_n$  ( $n - 1$  degrees of freedom as before).

# Perturbations

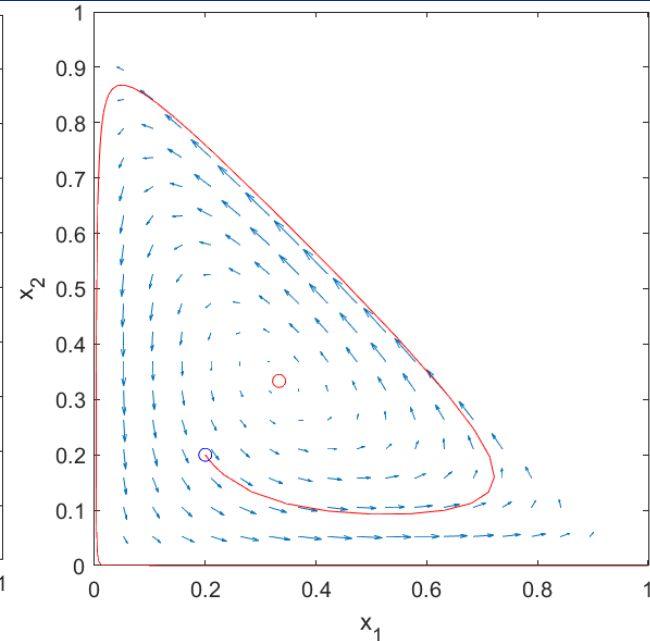
## Perturbed Rock-Paper-Scissors Game (new) ( $a = 1$ )



$$\varepsilon = 0$$



$$\varepsilon = 0.1$$



$$\varepsilon = -0.1$$

# Conclusion

Replicator Dynamics  
of Single-Population

$$\dot{x}_k = \alpha x_k [(\Pi x)_k - x^T \Pi x]$$

Stability Analysis

$$\dot{x} = \frac{\partial f}{\partial x}(x - a) := A(x - a)$$

See eigenvalues of  $A$ .

Replicator Dynamics  
of Multi-Populations

Divergent

Game of  
Chicken

Divergent  
Oscillatory  
Convergent

Rock-Paper-  
Scissors

$$v(x) = - \sum_{k=1}^n x_k \ln x_k$$

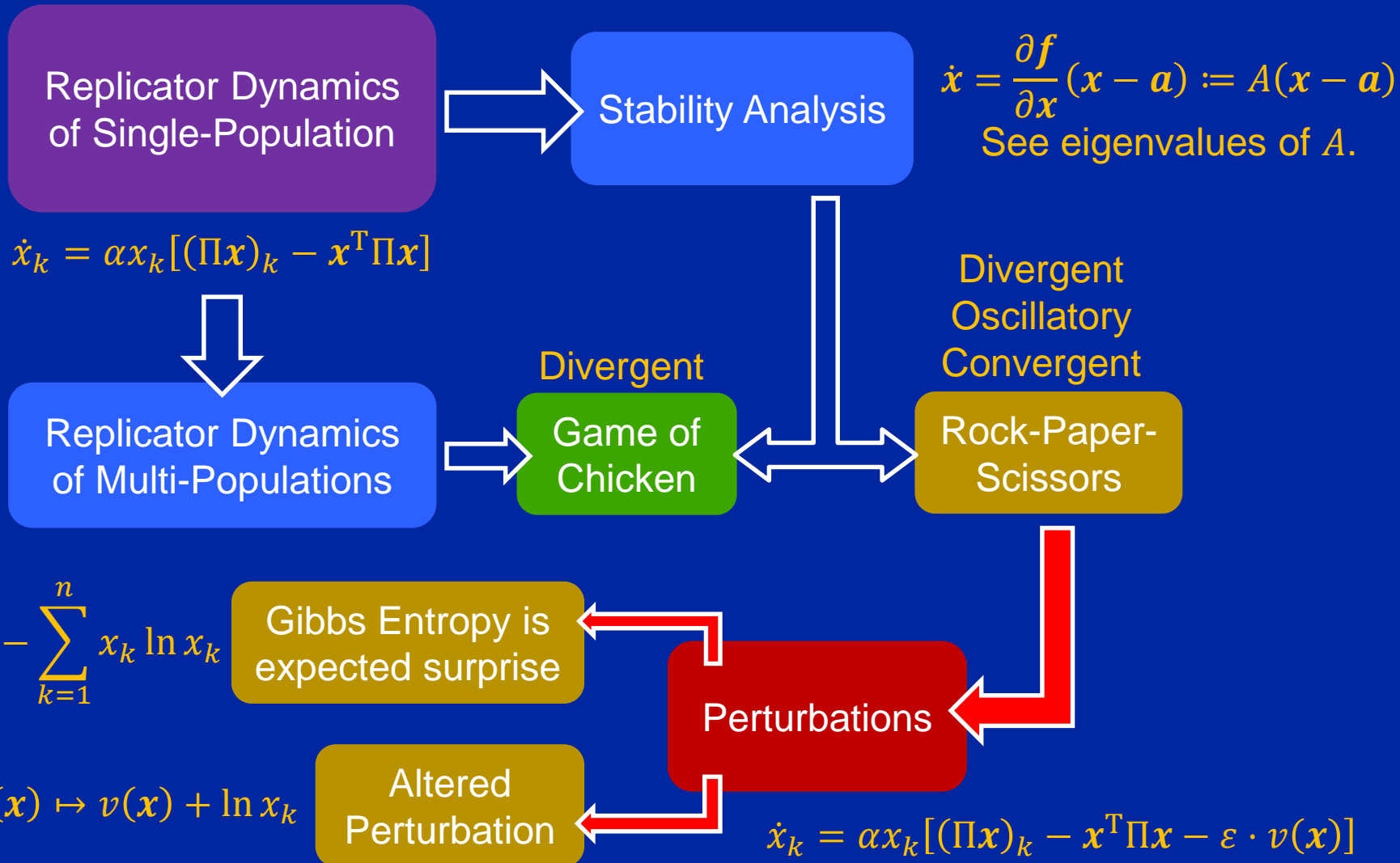
Gibbs Entropy is  
expected surprise

Perturbations

$$v(x) \mapsto v(x) + \ln x_k$$

Altered  
Perturbation

$$\dot{x}_k = \alpha x_k [(\Pi x)_k - x^T \Pi x - \varepsilon \cdot v(x)]$$





# QnA and Feedback