



Population Game

— Theory, Examples, and Applications

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Outline

- **Population Game Theory**
- **Real World Examples**
- **Our Applications**
 - **TV White Space Information Market**
 - **User-Provided Network**
 - **Wi-Fi Community Network**
 - **Peer-to-Peer Mobile Crowd Sensing**

Theory

Evolutionary Game Theory

- **Evolutionary Game Theory** considers a **population** decision makers (players), wherein the **frequency** with which a particular decision is made can be time varying. It is a theory started from Biology.
- Some interesting and important questions:
 - What is the stable population (equilibrium)?
 - Will the population evolve toward the equilibrium?
 - Will the population move away from the equilibrium (stability of equilibrium)?

Population Game

- Elements in **Population Game**
 - **Player**: An infinite population of individuals;
 - **Strategy**: Each player can use a set of pure strategies S , or a mixed strategy σ over S ;
 - **Payoff**: Each player's payoff depends on **population profile** X .

Population Game

- **Population Profile** — \mathbf{x}

Definition

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

- **Example**
 - Suppose that a population can choose $S = \{s_1, s_2\}$
 - Case 1: Half population chooses s_1 , and the other half chooses s_2 , then $X = (0.5, 0.5)$;
 - Case 2: All population chooses the mixed strategy $(0.5, 0.5)$, then $X = (0.5, 0.5)$;

Population Game

- **Payoff** of Individual

Definition

Consider a particular individual in the population with profile \mathbf{x} . If that individual uses a strategy σ , the individual's payoff is denoted as $\pi(\sigma, \mathbf{x})$. The payoff of this strategy is

$$\pi(\sigma, \mathbf{x}) = \sum_{s \in \mathbf{S}} p(s) \pi(s, \mathbf{x}).$$

Population Game

- Type 1: **Games against the field**

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.

- Type 2: **Games with pairwise contests**

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.

Equilibrium

- An **equilibrium** of a population game is the end points of the evolution of the population (also called **stable population**, **evolutionary stability**).

Theorem

Let σ^ be a strategy that generates a population profile \mathbf{x}^* . Let \mathbf{S}^* be the support for σ^* . If the population is stable, then*

$$\sigma^* \in \operatorname{argmax}_{\sigma \in \Sigma} \{\pi(\sigma, \mathbf{x}^*)\}.$$

$$\text{and } \pi(s, \mathbf{x}^*) = \pi(\sigma^*, \mathbf{x}^*), \forall s \in \mathbf{S}^*.$$

- At the equilibrium, the strategy adopted by each individual must be the **best response** to the population profile.

Stability of Equilibrium

- **Stability of Equilibrium**
 - The robustness of an equilibrium to a small change on population profile.

Comment on the above theorem

If σ^* is unique best response to \mathbf{x}^* , then the evolution of the population clearly stops. But if it is not unique, so there are some other strategies that do equally well in the population with profile \mathbf{x}^* , then the population could drift in the direction of other strategy and its corresponding population profile. We want to understand this situation.

Stability of Equilibrium

- **Post-Entry Population**

Definition

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

- **Example**

- Consider a population with $S = \{s_1, s_2\}$ and $\mathbf{X}^* = (0.5, 0.5)$
- Suppose the mutant strategy is $(0.75, 0.25)$, then

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

Stability of Equilibrium

- **Evolutionary Stable Strategy (ESS)**
 - Also called **stable equilibrium**.

Stability of ESS

A mixed strategy σ^* 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 is an ESS if mutants that adopt any other strategy **achieve a worst payoff** in the post-entry population, provided that the proportion of mutants is sufficiently small.

Examples

Example A

- Real World Example — Population Ratio

Example of a Game Against the field

Why the population ratio (male and female) is 50 : 50?

- The proportion of males (females) in the population is μ ($1 - \mu$).
- Each female mates once and produce n children.
- Males mates, on average, $(1 - \mu)/\mu$ times.
- Only female genes affect the sex ratio of offspring.
- Assume females available strategies are (a) s_1 : produce male offspring; (b) s_2 : produce female offspring. The general strategy $\sigma = (p, 1 - p)$ produces a proportion p of male offspring.
- The current population profile is $\mathbf{x} = (\mu, 1 - \mu)$
- What is the Evolutionary Stable Strategy (ESS)?

Example A

- Real World Example — Population Ratio

Solution

- Since payoff of children is n , let's consider the number of grandchildren. Given the population profile $\mathbf{x} = (\mu, 1 - \mu)$, the payoffs are

$$\pi(s_1, \mathbf{x}) = n^2 \left(\frac{1 - \mu}{\mu} \right) \quad ; \quad \pi(s_2, \mathbf{x}) = n^2.$$

- The expected payoff for the strategy σ is

$$\pi(\sigma, \mathbf{x}) = n^2 \left(\frac{1 - \mu}{\mu} \right) p + n^2(1 - p).$$

Because n is independent of the strategy chosen, we can set $n = 1$ (since we are only interested in the population ratio).

Example A

- Real World Example — Population Ratio

Solution: continue

To find an ESS, consider the following cases:

- If $\mu < 1/2$, then using s_1 will have more grandchildren which eventually cause μ to increase. So s_1 is not an ESS.
- If $\mu > 1/2$, then using s_2 have more grandchildren causing μ to fall. So s_2 is not an ESS.
- Is $\sigma^* = (\frac{1}{2}, \frac{1}{2})$ a potential ESS? Let use the **ESS Stability Theorem**, which states that

$$\pi(s_1, \mathbf{x}^*) = \pi(s_2, \mathbf{x}^*) = \pi(\sigma^*, \mathbf{x}^*).$$

So if the population profile is $\mathbf{x}^* = (\frac{1}{2}, \frac{1}{2})$, then $\sigma^* = (\frac{1}{2}, \frac{1}{2})$ is an ESS (note that this is only a **necessary condition**).

Example A

- Real World Example — Population Ratio

Solution: continue

Let us show the "sufficient" condition that $\sigma^* = (\frac{1}{2}, \frac{1}{2})$ is indeed an ESS.

- Let $\sigma = (p, 1 - p)$ be another strategy, then

$$\mathbf{x}_\epsilon = (1 - \epsilon)\sigma^* + \epsilon\sigma$$

$$\text{so, } \mu_\theta = (1 - \epsilon)\frac{1}{2} + \epsilon p = \frac{1}{2} + \epsilon \left(p - \frac{1}{2} \right).$$

- The ESS condition is $\pi(\sigma^*, \mathbf{x}_\epsilon) > \pi(\sigma, \mathbf{x}_\epsilon)$ where

$$\pi(\sigma^*, \mathbf{x}_\epsilon) = \frac{1}{2} + \frac{1}{2} \left(\frac{1 - \mu_\epsilon}{\mu_\epsilon} \right)$$

$$\pi(\sigma, \mathbf{x}_\epsilon) = (1 - p) + p \left(\frac{1 - \mu_\epsilon}{\mu_\epsilon} \right).$$

Example A

- Real World Example — Population Ratio

Solution: continue

- The difference

$$\begin{aligned}\pi(\sigma^*, \mathbf{x}_\epsilon) - \pi(\sigma, \mathbf{x}_\epsilon) &= \left(p - \frac{1}{2}\right) + \left(\frac{1}{2} - p\right) \left(\frac{1 - \mu_\epsilon}{\mu_\epsilon}\right) \\ &= \left(\frac{1}{2} - p\right) \left[\frac{1 - \mu_\epsilon}{\mu_\epsilon} - 1\right] \\ &= \left(\frac{1}{2} - p\right) \left[\frac{1 - 2\mu_\epsilon}{\mu_\epsilon}\right].\end{aligned}$$

If the difference is positive for $\sigma = (p, 1 - p)$, with $p \neq 1/2$, then σ^* is an ESS.

Example A

- Real World Example — Population Ratio

**In the population ratio game, $(0.5, 0.5)$
is the unique ESS.**

Example B

- Real World Example — Social Network

Example

Consider a "simplified" Internet. There are two operating systems available: L and W . A user of window W has a basic utility of 1, but L is a better operating system so a user of L has a basic utility of 2. If two computers have the same operating system, then they can communicate over the network. A user's utility rises linearly with the proportion of computers that can be communicated with, up to a maximum increment of 2. Let x be the proportion of W -users, then $\pi(W, x) = 1 + 2x$ and $\pi(L, x) = 2 + 2(1 - x)$. What are the ESSs in this population game?

Example B

- Real World Example — Social Network

Solution

- Potential ESSs are:
 - σ_W : everyone uses W , then $x = 1$ and $\pi(W, 1) > \pi(L, 1)$.
 - σ_L : everyone uses L , then $x = 0$ and $\pi(L, 0) > \pi(W, 0)$.
 - σ_m : mixed strategy in which W is used 3/4 of the time, then $x = 3/4$ and $\pi(W, 3/4) = \pi(L, 3/4)$.
- Now $\mathbf{x}_\epsilon = (p^* + \epsilon(p - p^*), 1 - p^* - \epsilon(p - p^*))$. So

$$\begin{aligned}\delta_\pi &= \pi(\sigma^*, \mathbf{x}_\epsilon) - \pi(\sigma, \mathbf{x}_\epsilon) \\ &= p^* \pi(W, \mathbf{x}_\epsilon) + (1 - p^*) \pi(L, \mathbf{x}_\epsilon) - p \pi(W, \mathbf{x}_\epsilon) - (1 - p) \pi(L, \mathbf{x}_\epsilon) \\ &= (p^* - p) (\pi(W, \mathbf{x}_\epsilon) - \pi(L, \mathbf{x}_\epsilon)) \\ &= (p^* - p) (4p^* - 3 - 4\epsilon(p^* - p))\end{aligned}$$

Example B

- Real World Example — Social Network

Solution: continue

Taking each candidate ESSs in turn:

- σ_W : $p^* = 1$, so

$$\delta_\pi = (1 - p)(1 - 4\epsilon(1 - p)) > 0, \forall p \neq 1, \text{ and } \epsilon < \bar{\epsilon} = 1/4.$$

So it is an ESS.

Example B

- Real World Example — Social Network

Solution: continue

Taking each candidate ESSs in turn:

- $\sigma_L: p^* = 0$, so

$$\delta\pi = p(3 - 4\epsilon p) > 0, \forall p \neq 0, \text{ and } \epsilon < \bar{\epsilon} = 3/4.$$

So σ_L is an ESS.

Example B

- Real World Example — Social Network

Solution: continue

Taking each candidate ESSs in turn:

- σ_m : $p^* = 3/4$, so

$$\delta_\pi = -4\epsilon \left(\frac{3}{4} - p \right)^2 < 0, \forall p \neq \frac{3}{4} \text{ and } \epsilon > 0.$$

So it is **not** an ESS.

Example B

- Real World Example — Social Network

In the social network game, there are three equilibrium points: $(1, 0)$, $(0, 1)$, and $(3/4, 1/4)$. The first two equilibria are ESS.

Example C

- Real World Example — Currency War

Example: The evolution of money

- 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 ?

Example C

- Real World Example — Currency War

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

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

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

- The post-entry population is:

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

Example C

- Real World Example — Currency War

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

$$\begin{aligned} \pi(\sigma_b^*, \mathbf{x}_\epsilon) - \pi(\sigma, \mathbf{x}_\epsilon) &> 0, \\ \iff (1 - p)(1 - 2\epsilon(1 - p)) &> 0. \end{aligned}$$

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

Example C

- Real World Example — Currency War

Solution: continue

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

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

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:

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

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

Example C

- Real World Example — Currency War

Solution: continue

The final candidate for an ESS is $\sigma_m^* = (\frac{1}{2}, \frac{1}{2})$ because

$$x = \frac{1}{2} \implies \hat{p} \in [0, 1] \implies x \in [0, 1].$$

(including, of course, $x = 1/2$). Consider the post-entry population

$$\mathbf{x}_\epsilon = (1 - \epsilon) \left(\frac{1}{2}, \frac{1}{2} \right) + \epsilon(p, 1 - p) = \left(\frac{1}{2} - \frac{1}{2}\epsilon(1 - 2p), \frac{1}{2} + \frac{1}{2}\epsilon(1 - 2p) \right).$$

The payoff for an arbitrary strategy is $\pi(\sigma, \mathbf{x}_\epsilon) = \frac{1}{2} + \frac{1}{2}\epsilon(1 - 2p)^2$, and the payoff for the candidate ESS is $\pi(\sigma_m^*, \mathbf{x}_\epsilon) = \frac{1}{2}$. So

$$\pi(\sigma_m^*, \mathbf{x}_\epsilon) - \pi(\sigma, \mathbf{x}_\epsilon) > 0 \iff -\frac{1}{2}\epsilon(1 - 2p)^2 > 0.$$

Because $\epsilon > 0$ and $p \neq \frac{1}{2}$, this condition **cannot be satisfied**; so σ_m^* is **not** an ESS. So whether to use beads or shells depends on the initial condition.

Example C

- Real World Example — Currency War

In the currency war game, there are three equilibrium points: $(1, 0)$, $(0, 1)$, and $(1/2, 1/2)$. The first two equilibria are ESS.

Example D

- Example of Game with Pairwise Contests

The Hawk-Dove Game

- Individuals can use one of two possible pure strategies: (a) *H*: be aggressive, (b) *D*: be non-aggressive.
- In general, individual can use a randomized strategy $\sigma = (p, 1 - p)$ with probability p of using *H*.
- A population consists of individuals that are aggressive with probability x , i.e., $\mathbf{x} = (x, 1 - x)$, this can arise because
 - a monomorphic population, everyone uses $\sigma = (x, 1 - x)$, or
 - a polymorphic population, a fraction x of population use $\sigma_H = (1, 0)$ and a fraction $1 - x$ use $\sigma_D = (0, 1)$. Let consider only monomorphic population.

Example D

- Example — The Hawk-Dove Game

The Hawk-Dove Game: continue

- There is a resource (e.g., food, breeding site,...etc) with value v . The outcome of a conflict depends on the types of two individuals that meet.
- Possible combinations:
 - 1 a hawk and a dove: hawk wins,
 - 2 a dove and a dove: they "share" the resource evenly,
 - 3 a hawk and a hawk: they fight with one winner gets the resource and the other loser pays a cost (i.e., injury) of c .
- What is the outcome of the game? What is the ESS?

Example D

- Example — The Hawk-Dove Game

Solution

- The payoff of an individual:

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

Example D

- Example — The Hawk-Dove Game

Solution

- Assume $v < c$, there is no pure-strategy ESS. Why?

- In a population of Doves ($x = 0$),

$$\pi(\sigma, \mathbf{x}_D) = pv + (1 - p)\frac{v}{2} = (1 + p)\frac{v}{2}.$$

It is best to set $p = 1$ (play hawk). As a consequence, the proportion of more aggressive individual will increase.

- In a population of Hawks ($x = 1$),

$$\pi(\sigma, \mathbf{x}_H) = p\frac{v - c}{2}.$$

It is best to set $p = 0$ because $(v - c) < 0$. As a consequence, the proportion of less aggressive individual will increase.

Example D

- Example — The Hawk-Dove Game

Solution: continue

- Is there a mixed strategy ESS, $\sigma^* = (p^*, 1 - p^*)$? For σ^* to be ESS, it must be a best response to the population $\mathbf{x}^* = (p^*, 1 - p^*)$ that it generates.
- If $p^* = v/c$, then any choice of p (including p^*) gives the same payoff, so we have

$$\sigma^* = \left(\frac{v}{c}, 1 - \frac{v}{c} \right),$$

as a **candidate ESS** when $v < c$.

Example D

- Example — The Hawk-Dove Game

- To confirm σ^* is an ESS, we must show that for $\sigma = (p, 1 - p) \neq \sigma^*$, $\pi(\sigma^*, \mathbf{x}_\epsilon) > \pi(\sigma, \mathbf{x}_\epsilon)$, where

$$\begin{aligned}\mathbf{x}_\epsilon &= ((1 - \epsilon)p^* + \epsilon p, ((1 - \epsilon)(1 - p^*) + \epsilon(1 - p))) \\ &= (p^* + \epsilon(p - p^*), 1 - p^* + \epsilon(p^* - p)).\end{aligned}$$

- We have

$$\begin{aligned}\pi(\sigma^*, \mathbf{x}_\epsilon) &= p^*(p^* + \epsilon(p - p^*))\frac{v - c}{2} + p^*(1 - p^* + \epsilon(p^* - p))v + \\ &\quad (1 - p^*)(1 - p^* + \epsilon(p^* - p))\frac{v}{2},\end{aligned}$$

$$\begin{aligned}\pi(\sigma, \mathbf{x}_\epsilon) &= p(p^* + \epsilon(p - p^*))\frac{v - c}{2} + p(1 - p^* + \epsilon(p^* - p))v + \\ &\quad (1 - p)(1 - p^* + \epsilon(p^* - p))\frac{v}{2}.\end{aligned}$$

- Substituting $p^* = v/c$, we have

$$\pi(\sigma^*, \mathbf{x}_\epsilon) - \pi(\sigma, \mathbf{x}_\epsilon) = \frac{\epsilon c}{2}(p^* - p)^2 > 0.$$

so $\sigma^* = (p^*, 1 - p^*)$ is an ESS.

Example D

- Example — The Hawk-Dove Game

In the hawk-dove game with $V < C$, there is a unique ESS: $(V/C, 1 - V/C)$.

Our Applications

Model Extension

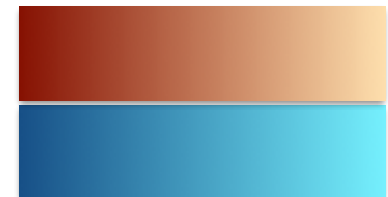
- In the basic population game model, all of the game players are **homogeneous** (i.e., with the same strategy set and payoff function).
- In a population game with homogeneous players, we can focus on the **symmetric equilibrium**, wherein all players choose the same strategy (hence equals the population profile X).
 - For any asymmetric equilibrium, we can always find an **equivalent** symmetric equilibrium.

Model Extension

- In practice, a population game may consist of a population of **heterogeneous** players.
- **Example: Modified Social Network**
 - Two social networks: L and W
 - The whole population is divided into **two types**: Q1 and Q2, depending on their evaluations for the population profile:
 - For type Q1: Payoff = $1 + Q1 * x$, or $2 + Q1 * (1 - x)$
 - For type Q2: Payoff = $1 + Q2 * x$, or $2 + Q2 * (1 - x)$
 - Suppose the population profile of type-Q1 players is $(x_1, 1-x_1)$, and the population profile of type-Q2 players is $(x_2, 1-x_2)$;
 - Then, the entire population profile $(x, 1-x)$ is given by:
 - $$x = n_1 * x_1 + n_2 * x_2$$

Model Extension

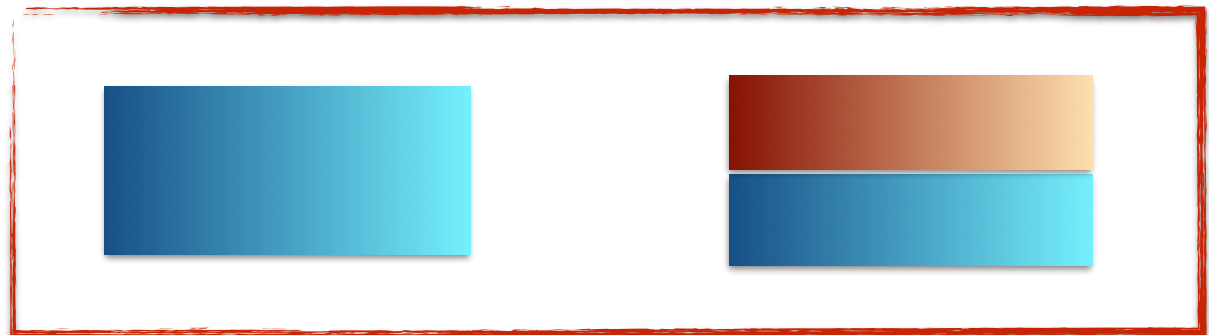
- Population Game Model Extensions
 - Homogeneous Players (One Type)
 - Heterogeneous Players with Finite Types
 - Heterogeneous Players with Infinite Types (Single-Dimension)
 - Heterogeneous Players with Infinite Types (Multi-Dimension)



Model Extension

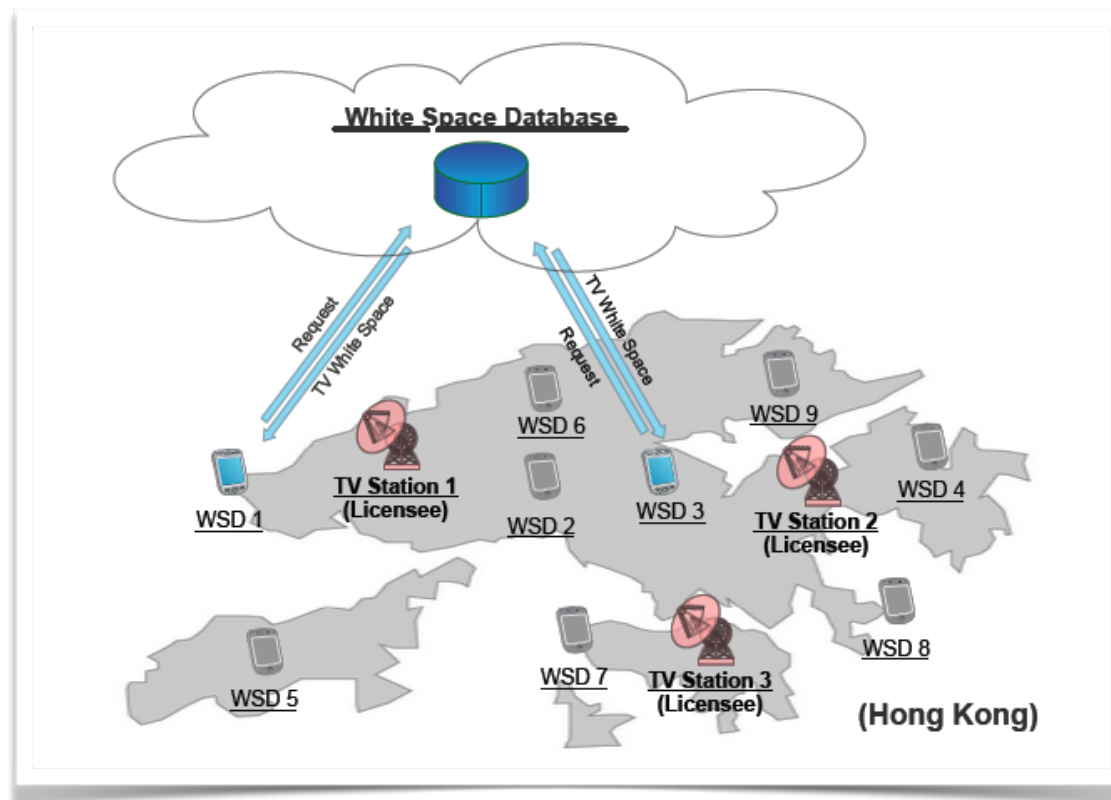
- Population Game Model
 - Homogeneous Players (One Type)
 - Heterogeneous Players with Finite Types
 - Heterogeneous Players with Infinite Types (Single-Dimension)
 - Heterogeneous Players with Infinite Types (Multi-Dimension)

- Our Focus



Application 1

- TV White Space Information Market
 - An infinite population of **heterogenous** WSDs want to access Internet through TV channels;
 - Choose to purchase information from database or not.

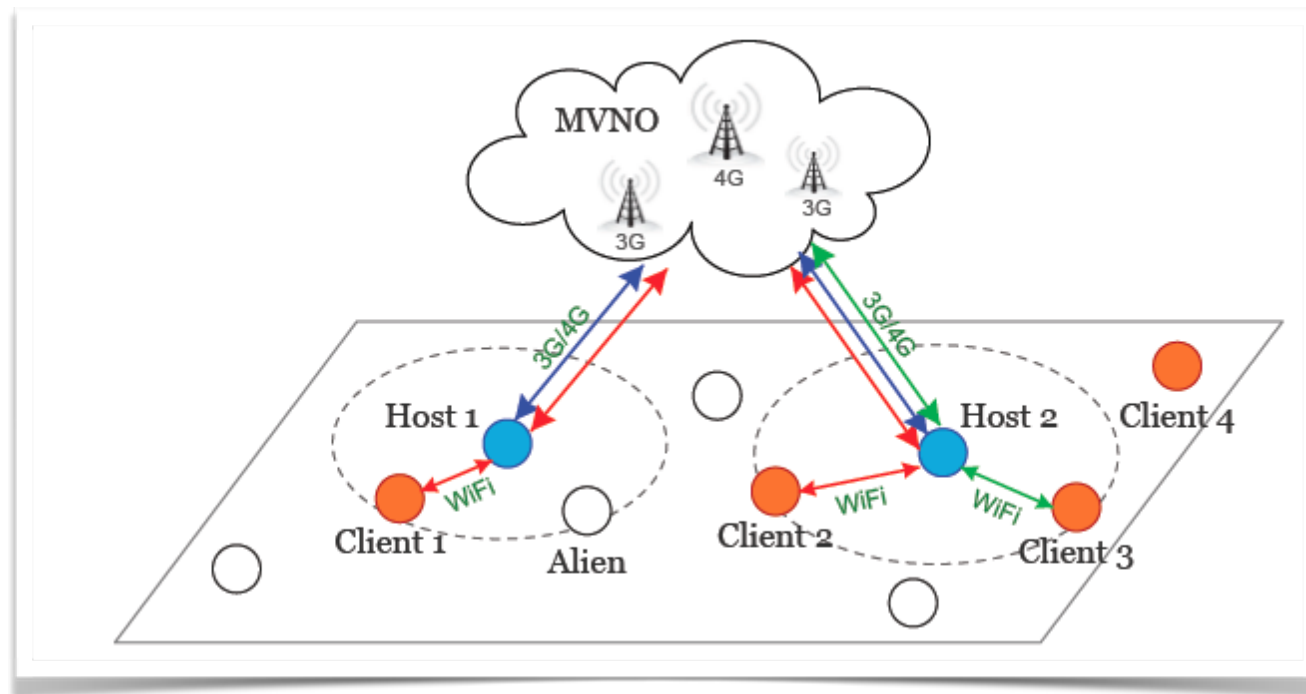


Application 1

- TV White Space Information Market
 - **WSDs are heterogenous**, with different information evaluations Q in $[Q_{\text{low}}, Q_{\text{up}}]$; (infinite and one-dimensional type)
 - **Population profile**: $(x, 1-x)$
 - A population x of WSDs choose to purchase, and the remaining population $1-x$ choose not to purchase.
 - **Payoff** of each WSD:
 - (1) Information valuation minus price, if purchasing information,
 - (2) zero, if not purchasing the information.
- Key Problem: **Equilibrium Analysis**
 - What is the equilibrium population?
 - Is the equilibrium stable?

Application 2

- User-Provided Network
 - An infinite population of **heterogenous** mobile users want to access Internet through MVNO network;
 - Choose to become hosts, clients, or aliens.

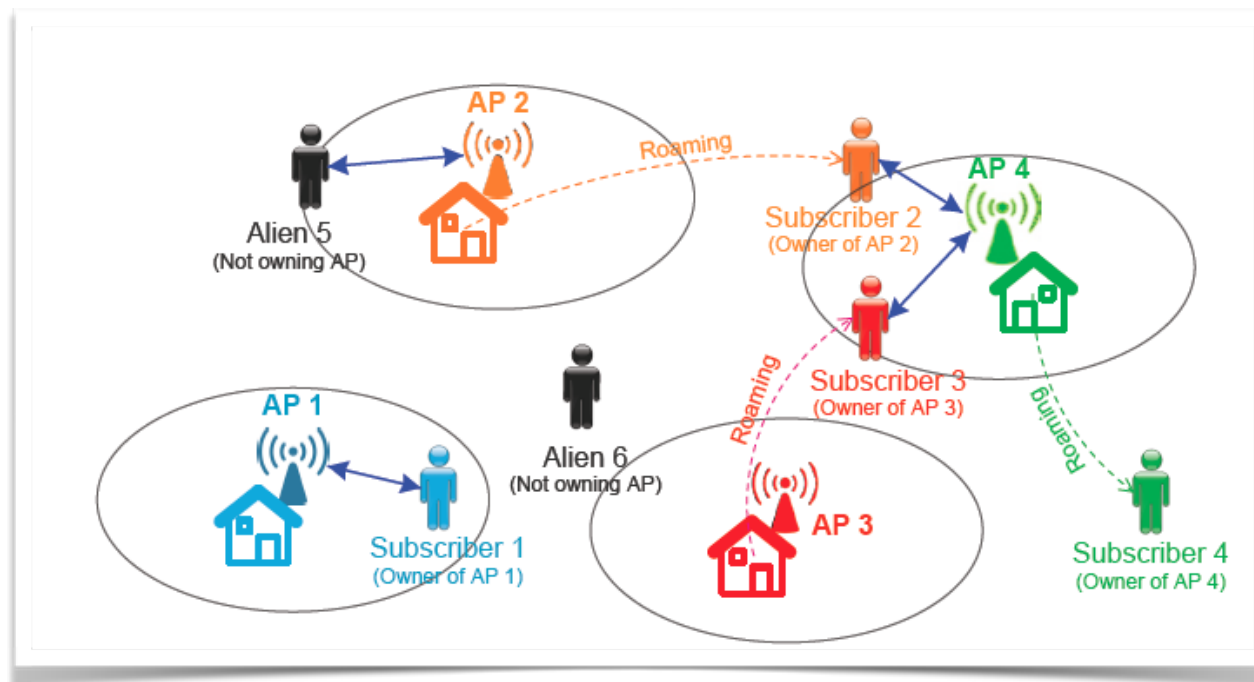


Application 2

- **User-Provided Network**
 - **Users are heterogenous**, with different service request probability Q in $[0, 1]$; (infinite and one-dimensional type)
 - **Population profile**: $(x_h, x_c, 1-x_h-x_c)$
 - A population x_h (x_c) of users choose to be hosts (clients), and the remaining population $1-x_h-x_c$ choose to be aliens.
 - **Payoff** of each WSD:
 - (1) Payoff of a host increases with x_c , and decreases with x_h ;
 - (2) Payoff of a client increases with x_h , and decreases with x_c ;
 - Key Problem: **Equilibrium Analysis**
 - What is the equilibrium population?
 - Is the equilibrium stable?

Application 3

- **Wi-Fi Community Network**
 - An infinite population of **heterogenous** WiFi AP owners share their APs with each others;
 - Choose different sharing schemes, i.e., be Bills or Linus.

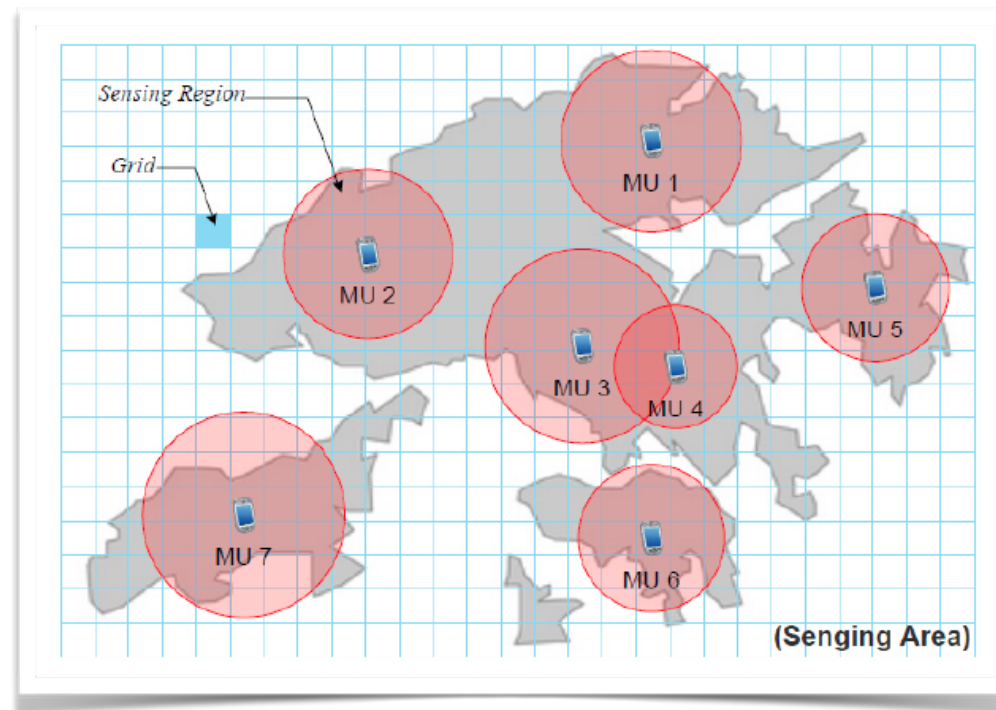


Application 3

- **Wi-Fi Community Network**
 - **Users are heterogenous**, with different data usage evaluations and roaming probabilities; (infinite and multi-dimensional type)
 - **Population profile:** $(x, 1-x)$
 - A population x of users choose to be Bills, and the remaining population $1-x$ choose to be Linus.
 - **Payoff** of each player:
 - (1) Payoff of a Bill increases with the population of Bills x ;
 - (2) Payoff of a Linus is constant.
 - **Key Problem: Equilibrium Analysis**
 - What is the equilibrium population?
 - Is the equilibrium stable?

Application 4

- Peer-to-Peer Mobile Crowd Sensing
 - An infinite population of **heterogenous** mobile users sense and share data with each other;
 - Choose to sense data (and sell to others), or purchase data from sensing users.



Application 4

- **Peer-to-Peer Mobile Crowd Sensing**
 - **Users are heterogenous**, with different data values and sensing costs; (infinite and multi-dimensional type)
 - **Population profile**: $(x, 1-x)$
 - A population x of users choose to sense, and the remaining population $1-x$ choose to purchase data from sensing users;
 - **Payoff** of each player:
 - (1) Payoff of a sensing user decreases with the population x (sensing), and increases with the population $1-x$ (purchasing);
 - (2) Payoff of a purchasing user is constant.
- **Key Problem: Equilibrium Analysis**
 - What is the equilibrium population?
 - Is the equilibrium stable?

Summary

- **Introduce Population Game Theory;**
- **Discuss Real World Examples;**
- **Discuss Our Applications;**

Thank You

