

Noncooperative Games

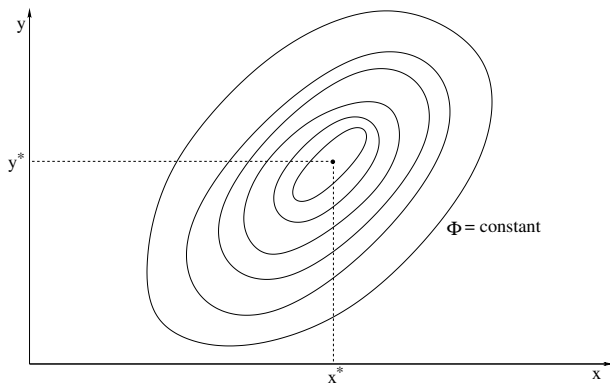
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- introduction to non-cooperative games: solution concepts
- differential games in continuous time and economic models
- a game theoretical model of debt and bankruptcy

Optimal decision problem

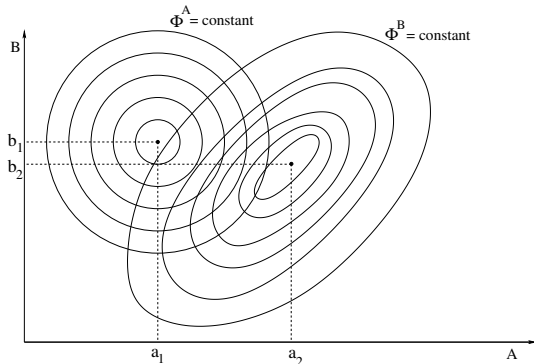
maximize: $\Phi(x, y)$



The choice $(x^*, y^*) \in \mathbb{R}^2$ yields the maximum payoff

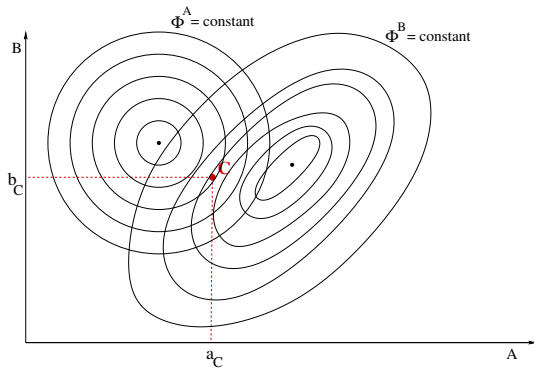
A game for two players

- Player A wishes to maximize his payoff $\Phi^A(a, b)$
- Player B wishes to maximize his payoff $\Phi^B(a, b)$



- Player A chooses the value of $a \in A$
- Player B chooses the value of $b \in B$

A cooperative solution



- maximize the **sum** of payoffs $\Phi^A(a, b) + \Phi^B(a, b)$
- split the total payoff **fairly** among the two players (how ???)

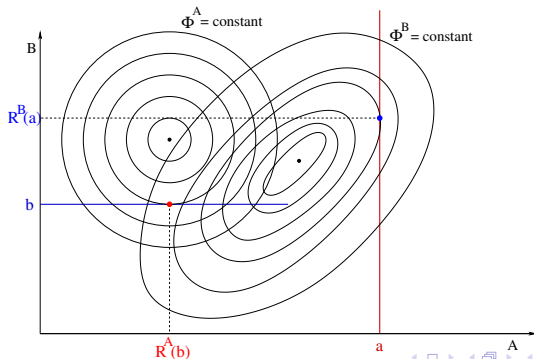
The best reply map

If Player A adopts the strategy a , the set of best replies for Player B is

$$R^B(a) = \left\{ b; \Phi^B(a, b) = \max_{s \in B} \Phi^B(a, s) \right\}$$

If Player B adopts the strategy b , the set of best replies for Player A is

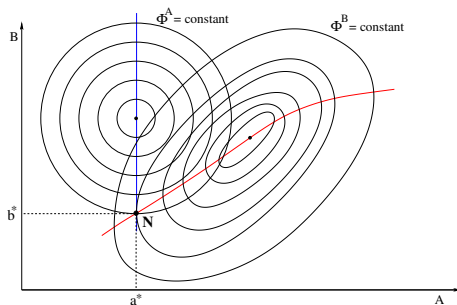
$$R^A(b) = \left\{ a; \Phi^A(a, b) = \max_{s \in A} \Phi^A(s, b) \right\}$$



Nash equilibrium solutions

A couple of strategies (a^*, b^*) is a **Nash equilibrium** if

$$a^* \in R^A(b^*) \quad \text{and} \quad b^* \in R^B(a^*)$$



Antoin Augustin Cournot (1838)
John Nash (1950)

Existence of Nash equilibria

Theorem. *Assume*

- *Sets of available strategies for the two players: $A, B \subset \mathbb{R}^n$ are compact and convex*
- *Payoff functions: $\Phi^A, \Phi^B : A \times B \mapsto \mathbb{R}$ are continuous*
- *For each $a \in A$, the set of best replies $R^B(a) \subset B$ is **convex***
- *For each $b \in B$, the set of best replies $R^A(b) \subset A$ is **convex***

Then the game admits at least one Nash equilibrium.

Proof. If the best reply maps are single valued, the map

$$(a, b) \mapsto (R^A(b), R^B(a))$$

is a continuous map from the compact convex set $A \times B$ into itself.

By Brouwer's fixed theorem, it has a fixed point (a^*, b^*) .

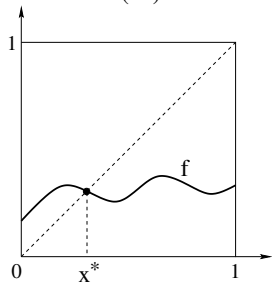
If R^A, R^B are convex-valued, by Kakutani's fixed point theorem there exists

$$(a^*, b^*) \in (R^A(b^*), R^B(a^*))$$

One-dimensional version of Brouwer's and Kakutani's theorems

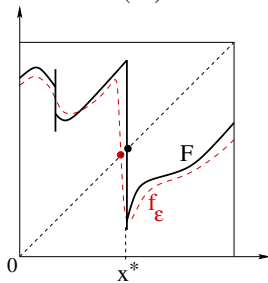
Brouwer 1910

$$x^* = f(x^*)$$

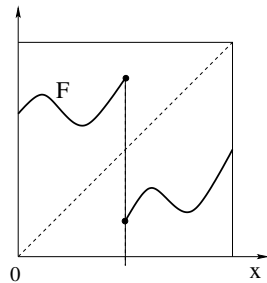


Kakutani 1941

$$x^* \in F(x^*)$$



no fixed point



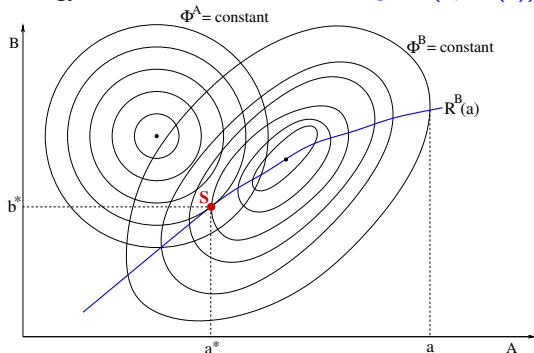
Luitzen Egbertus Jan Brouwer (1910)
Shizuo Kakutani (1941)
Arrigo Cellina (1969)

Stackelberg equilibrium

- Player A (the **leader**) announces his strategy $a \in A$ in advance
- Player B (the **follower**) adopts his best reply: $b \in R^B(a) \subseteq B$

What is the best strategy for the leader?

$$\max_{a \in A} \Phi^A(a, R^B(a))$$



A couple of strategies (a^*, b^*) is a **Stackelberg equilibrium** if $b^* \in R^B(a^*)$ and

$$\Phi^A(a^*, b^*) \geq \Phi^A(a, b^*) \quad \text{for all } a \in A, b \in R^B(a)$$

Game theoretical models in Economics and Finance

- Sellers (choosing prices charged) vs. buyers (choosing quantities bought)
- Companies competing for market share (choosing production level, prices, amount spent on research & development or advertising)
- Auctions, bidding games
- Economic growth. Leading player: central bank (choosing prime rate)
followers: private companies (choosing investment levels)
- Debt management. Lenders (choosing interest rate) vs. borrower (choosing repayment strategy)
- . . .

Differential games in finite time horizon

$x(t) \in \mathbb{R}^n =$ state of the system

Dynamics: $\dot{x}(t) = f(x(t), u_1(t), u_2(t)), \quad x(t_0) = x_0$

$u_1(\cdot), u_2(\cdot) =$ controls implemented by the two players

Goal of i -th player:

$$\begin{aligned} \text{maximize: } J_i &\doteq \psi_i(x(T)) - \int_{t_0}^T L_i(x(t), u_1(t), u_2(t)) dt \\ &= [\text{terminal payoff}] - [\text{running cost}] \end{aligned}$$

Differential games in infinite time horizon

$$\text{Dynamics: } \dot{x} = f(x, u_1, u_2), \quad x(0) = x_0$$

Goal of i -th player:

$$\text{maximize: } J_i \doteq \int_0^{+\infty} e^{-\gamma t} \Psi_i(x(t), u_1(t), u_2(t)) dt$$

(running payoff, exponentially discounted in time)

Example 1: an advertising game

- Two companies, competing for market share

state variable: $x(t) \in [0, 1]$ = market share of company 1, at time t

$$\dot{x} = (1 - x) u_1 - x u_2$$

controls: u_1, u_2 = advertising rates

$$\text{payoffs: } J_i = N x_i(T) p_i - \int_0^T c_i u_i(t) dt \quad i = 1, 2$$

N = expected number of items purchased by consumers

p_i = profit made by player i on each sale

c_i = advertising cost

x_i = market share of player i

$$(x_1 = x, \quad x_2 = 1 - x)$$

Example 2: harvesting of marine resources

$x(t)$ = amount of fish in a lake, at time t

dynamics: $\dot{x} = \alpha x(M - x) - xu_1 - xu_2$

controls: u_1, u_2 = harvesting efforts by the two players

payoffs: $J_i = \int_0^{+\infty} e^{-\gamma t} (p x u_i - c_i u_i) dt$

p = selling price of fish

c_i = harvesting cost

Example 3: a producer vs. consumer game

$$\text{State variables: } \begin{cases} p & = \text{price} \\ q & = \text{size of the inventory} \end{cases}$$

$$\text{Controls: } \begin{cases} a(t) & = \text{production rate} \\ b(t) & = \text{consumption rate} \end{cases}$$

$$\text{The system evolves in time according to } \begin{cases} \dot{p} & = p \ln(q_0/q) \\ \dot{q} & = a - b \end{cases}$$

Here q_0 is an “appropriate” inventory level

$$\text{Payoffs: } \begin{cases} J^{\text{producer}} & \doteq \int_0^{+\infty} e^{-\gamma t} [p(t) \cdot b(t) - c(a(t))] dt \\ J^{\text{consumer}} & \doteq \int_0^{+\infty} e^{-\gamma t} [\phi(b(t)) - p(t)b(t)] dt \end{cases}$$

$c(a)$ = production cost, $\phi(b)$ = utility to the consumer

- No outcome can be optimal simultaneously for all players

Different outcomes may arise, depending on

- information available to the players
- their ability and willingness to cooperate

Nash equilibria (in infinite time horizon)

Seek: **feedback strategies**: $u_1^*(x)$, $u_2^*(x)$ with the following properties

- Given the strategy $u_2 = u_2^*(x)$ adopted by the second player, for every initial data $x(0) = y$, the assignment $u_1 = u_1^*(x)$ provides a solution to the **optimal control problem for the first player**:

$$\max_{u_1(\cdot)} \int_0^{\infty} e^{-\gamma t} \Psi_1(x, u_1, u_2^*(x)) dt$$

subject to

$$\dot{x} = f(x, u_1, u_2^*(x)), \quad x(0) = y$$

- Similarly, given the strategy $u_1 = u_1^*(x)$ adopted by the first player, the feedback control $u_2 = u_2^*(x)$ provides a solution to the optimal control problem for the second player.

Solving an optimal control problem by PDE methods

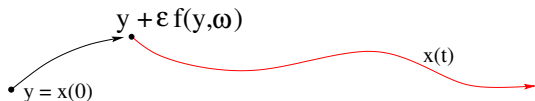
$$V(y) = \inf_{u(\cdot)} \int_0^{+\infty} e^{-\gamma t} L(x(t), u(t)) dt$$

subject to:

$$\dot{x}(t) = f(x(t), u(t)) \quad x(0) = y \quad u(t) \in U$$

$V(y)$ = minimum cost, if the system is initially at y

A PDE for the value function (by Bellman's dynamic programming)



If we use the constant control $u(t) = \omega$ for $t \in [0, \varepsilon]$
then we play optimally for $t \in [\varepsilon, \infty[$, the total cost is

$$\begin{aligned} J^{\varepsilon, \omega} &= \left(\int_0^{\varepsilon} + \int_{\varepsilon}^{+\infty} \right) e^{-\gamma t} L(x(t), u(t)) dt \\ &= \varepsilon L(y, \omega) + e^{-\gamma \varepsilon} V(y + \varepsilon f(y, \omega)) + o(\varepsilon) \\ &= \varepsilon L(y, \omega) + (1 - \gamma \varepsilon) V(y) + \nabla V(y) \cdot \varepsilon f(y, \omega) + o(\varepsilon) \\ &\geq V(y) \end{aligned}$$

Minimize w.r.t. ω :

$$V(y) = V(y) - \gamma \varepsilon V(y) + \varepsilon \cdot \min_{\omega \in U} \left\{ L(y, \omega) + \nabla V(y) \cdot f(y, \omega) \right\} + o(\varepsilon)$$

The Hamilton-Jacobi PDE for the value function

$$V(y) = V(y) - \gamma\varepsilon V(y) + \varepsilon \cdot \min_{\omega \in U} \left\{ L(y, \omega) + \nabla V(y) \cdot f(y, \omega) \right\} + o(\varepsilon)$$

Letting $\varepsilon \rightarrow 0$ we obtain

$$\gamma V(y) = \min_{\omega \in U} \left\{ L(y, \omega) + \nabla V(y) \cdot f(y, \omega) \right\} \doteq H(y, \nabla V(y))$$

If $V(\cdot)$ is known, the optimal feedback control can be recovered by

$$u^*(y) = \operatorname{argmin}_{\omega \in U} \left\{ L(y, \omega) + \nabla V(y) \cdot f(y, \omega) \right\}$$

An example

$$\text{minimize: } \int_0^{\infty} e^{-\gamma t} \left(\phi(x(t)) + \frac{u^2(t)}{2} \right) dt$$

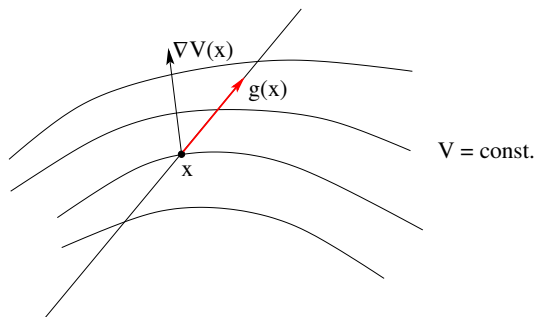
$$\text{subject to: } \dot{x} = f(x) + g(x)u, \quad x(0) = y, \quad u(t) \in \mathbb{R}$$

The value function $V(y) = \text{minimum cost starting at } y$ satisfies the PDE

$$\gamma V(x) = \phi(x) + \nabla V(x) \cdot f(x) - \frac{1}{2} (\nabla V(x) \cdot g(x))^2$$

Finding the optimal feedback control

$$\dot{x} = f(x) + g(x) u$$



If $V(\cdot)$ is known, the optimal control can be recovered by

$$u^*(x) = \operatorname{argmin}_{u \in \mathbb{R}} \left\{ \nabla V(x) \cdot g(x) u + \frac{u^2}{2} \right\} = -\nabla V(x) \cdot g(x)$$

Solving a differential game by PDE methods

$$\text{Dynamics: } \dot{x} = f(x) + g_1(x)u_1 + g_2(x)u_2$$

$$\text{Player } i \text{ seeks to minimize: } J_i = \int_0^\infty e^{-\gamma t} \left(\phi_i(x(t)) + \frac{u_i^2(t)}{2} \right) dt$$

The value functions V_1, V_2 for the two players satisfy the system of H-J equations

$$\begin{cases} \gamma V_1 = (f \cdot \nabla V_1) - \frac{1}{2}(g_1 \cdot \nabla V_1)^2 - (g_2 \cdot \nabla V_1)(g_2 \cdot \nabla V_2) + \phi_1 \\ \gamma V_2 = (f \cdot \nabla V_2) - \frac{1}{2}(g_2 \cdot \nabla V_2)^2 - (g_1 \cdot \nabla V_1)(g_1 \cdot \nabla V_2) + \phi_2 \end{cases}$$

$$\text{Optimal feedback controls: } u_i^*(x) = -\nabla V_i(x) \cdot g_i(x) \quad i = 1, 2$$

highly nonlinear, implicit !

Linear - Quadratic games

Assume that the dynamics is linear:

$$\dot{x} = (Ax + \mathbf{b}_0) + \mathbf{b}_1 u_1 + \mathbf{b}_2 u_2, \quad x(0) = y$$

and the cost functions are quadratic:

$$J_i = \int_0^{+\infty} e^{-\gamma t} \left(\mathbf{a}_i \cdot x + x^T P_i x + \frac{u_i^2}{2} \right) dt$$

Then the system of PDEs has a special solution of the form

$$V_i(x) = \alpha_i + \beta_i \cdot x + x^T \Gamma_i x \quad i = 1, 2 \quad (*)$$

$$\text{optimal controls: } u_i^*(x) = -(\beta_i + 2x^T \Gamma_i) \cdot \mathbf{b}_i$$

To find this solution, it suffices to

determine the coefficients $\alpha_i, \beta_i, \Gamma_i$ by solving a system of algebraic equations

Validity of linear-quadratic approximations ?

Assume the dynamics is almost linear

$$\dot{x} = f_0(x) + g_1(x)u_1 + g_2(x)u_2 \approx (Ax + \mathbf{b}_0) + \mathbf{b}_1u_1 + \mathbf{b}_2u_2, \quad x(0) = y$$

and the cost functions are almost quadratic

$$J_i = \int_0^{+\infty} e^{-\gamma t} \left(\phi_i(x) + \frac{u_i^2}{2} \right) dt \approx \int_0^{+\infty} e^{-\gamma t} \left(\mathbf{a}_i \cdot x + x^T P_i x + \frac{u_i^2}{2} \right) dt$$

Is it true that the nonlinear game has a feedback solution close to the linear-quadratic game?

On optimal control:

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- A.B., Noncooperative differential games. A tutorial. *Milan J. Math.*, 2011.
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- E. Dockner, S. Jorgensen, N. Van Long, and G. Sorger, *Differential games in economics and management science*, Cambridge University Press, 2000.