

February, 28th

Differential games

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Dynamic game - Players - Information structure

$[0, T]$ Programming interval $T < +\infty$, $T = +\infty$

N players $i \in \{1, 2, \dots, N\}$

$u^i(t) : [0, T] \rightarrow \mathbb{R}$

$u^i \in \mathcal{U}^2(x(t), u^{-i}(t), t) \subset \mathcal{R}^{mj}$ Player's strategies

Strategies based on the **information** revealed during all times $t \in [0, T]$ when the game takes place.

At any t players have the knowledge of all the previous actions.

Perfect information

Differential game

$\underline{x}(t) : [0, T] \rightarrow \mathbb{R}$ state function

The system varies according to a differential equation

$$\begin{aligned}\dot{\underline{x}}(t) &= \underline{f}(\underline{x}(t), \underline{u}(t), t) \\ \underline{x}(t_0) &= \underline{x}^0 \\ \underline{x}(t) &\in X \subset \mathcal{R}^2\end{aligned}$$

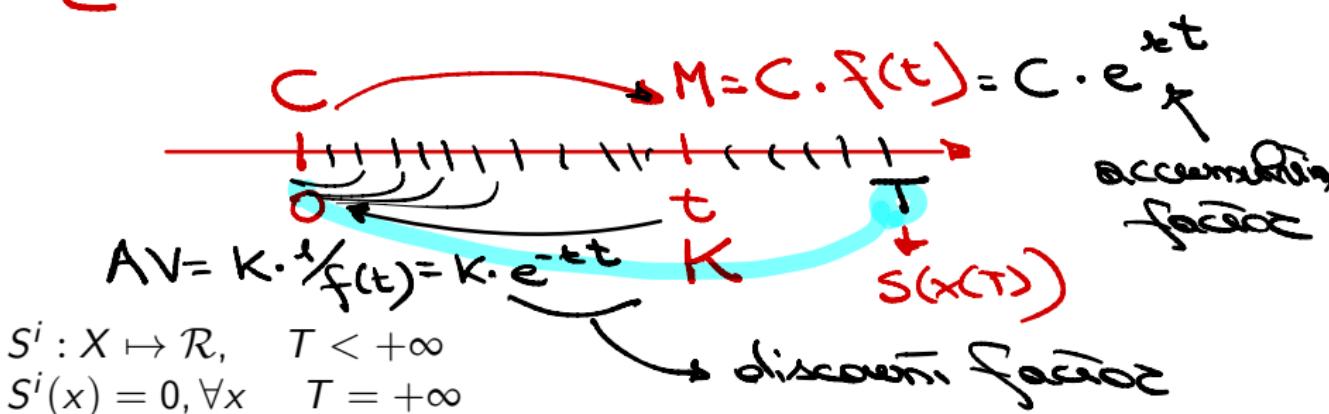
Payoffs

$$J^i(u^i(\cdot)) = \int_0^T e^{-r^i t} F^i(x(t), u^1(t), \dots, u^N(t), t) dt + e^{-r^i T} S^i(x(T))$$

C

• Scrap Value
• Salvage

discount factor



Problem for each player i

$$\forall i \in \{1, \dots, N\}$$

(Dockner et al p.86)

$$\max J_{\Phi^{-i}}^i(u^i(\cdot)) = \int_0^T e^{-r^i t} F_{\Phi^{-i}}^i(x(t), u^i(t), t) dt + e^{-r^i T} S^i(x(T))$$

subject to $\begin{cases} \dot{x}(t) = f_{\Phi^{-i}}^i(x(t), u^i(t), t) \\ x(0) = x^0 \\ u^i(t) \in \mathcal{U}_{\Phi^{-i}}^i(x(t), t) \end{cases}$

$$i \in \{1, \dots, N\}$$

$F_{\Phi^{-i}}^i(x, u^i, t) = F^i(x(t), \Phi^1(\cdot), \dots, \Phi^{i-1}(\cdot), u^i, \Phi^{i+1}(\cdot), \dots, \Phi^N(\cdot), t)$

$f_{\Phi^{-i}}^i$
 $\mathcal{U}_{\Phi^{-i}}^i$

player "i" cannot control x on $J_{\Phi^{-i}}$

Different types of strategies Φ_i

- A) OPEN-LOOP $\Phi_i = \Phi_i(t)$
- B) CLOSED-LOOP WITH MEMORY $\Phi_i = \Phi_i(t, x(\tau), 0 \leq \tau < t)$ future depends on the past
- C) CLOSED-LOOP WITHOUT MEMORY (no-memory) MARKOVIAN STRATEGY $\Phi_i = \Phi_i(t, x(t)), \forall t$ all payoffs depend on the present
 - C1) $T = +\infty$: STATIONARY MARKOVIAN (autonomous) $\Phi_i = \Phi_i(x(t))$

Observations

- INFORMATION: Different information is required for the implementation
Markovian strategy is more demanding from the informative point of view
If Information is either Irrelevant or Inaccessible \Rightarrow OPEN-LOOP
- COMMITMENT: Players can deviate from the declared strategy \Rightarrow MARKOVIAN consistent strategy

Nash equilibrium for a differential game

The N -tuple

$$(\Phi^1, \Phi^2, \dots, \Phi^i, \dots, \Phi^N)$$

constitutes a Nash equilibrium iff for all players, $i \in \{1, 2, \dots, N\}$

$$J^i(\Phi^1, \Phi^2, \dots, \Phi^i, \dots, \Phi^N) \geq J^i(\Phi^1, \Phi^2, \dots, u^i, \dots, \Phi^N) \quad \forall u^i \in \mathcal{U}^i$$

FIXED Φ^{-i} we need to compute the **best response strategy** of player i

$$\max_{\Phi^{-i}} J^i_{\Phi^{-i}}(u^i(\cdot)) = \int_0^T e^{-r^i t} F^i_{\Phi^{-i}}(x(t), u^i(t), t) dt + e^{-r^i T} S^i(x(T))$$

$$\text{subject to } \dot{x}(t) = f^i_{\Phi^{-i}}(x(t), u^i(t), t)$$

$$x(0) = x^0$$

$$u^i(t) \in \mathcal{U}_{\Phi^{-i}}^i(x(t), t)$$

$$i \in \{1, \dots, N\}$$

Nash equilibrium for a differential game

Finding a **Nash equilibrium** in a differential game with N players is equivalent to **solve N Optimal Control problems**.

- OPEN-LOOP NASH EQUILIBRIUM (OLNE)
Pontryagin's Maximum Principle (1962) **PMP**

$$(\Phi^1(t), \Phi^2(t), \dots, \Phi^i(t), \dots, \Phi^N(t))$$

- MARKOVIAN NASH EQUILIBRIUM (MNE)
Hamilton Jacobi Bellman ('50s) **HJB**

$$(\Phi^1(t, x(t)), \Phi^2(t, x(t)), \dots, \Phi^i(t, x(t)), \dots, \Phi^N(t, x(t)))$$

OPEN-LOOP NASH EQUILIBRIUM (OLNE)

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OPEN-LOOP NASH EQUILIBRIUM (OLNE)

Definition 4.2 The N -tuple $(\phi^1, \phi^2, \dots, \phi^N)$ of functions $\phi^i : [0, T] \rightarrow \mathbb{R}^{m^i}$, $i \in \{1, 2, \dots, N\}$, is called an open-loop Nash equilibrium if, for each $i \in \{1, 2, \dots, N\}$, an optimal control path $u^i(\cdot)$ of the problem (4.1) exists and is given by the open-loop strategy $u^i(t) = \phi^i(t)$.

OPEN-LOOP NASH EQ. (OLNE)

Pontryagin Maximum Principle (1962)

Hamiltonian function

$$\dot{x}(t) = f^i(x)$$

$$H^i(x, u^I, p, t) = e^{-rt} F^i(\) + p \cdot f^i(\)$$

$p(t) : [0, T] \mapsto \mathbb{R}^n$ co-state function (piece wise C')
m = # of state functions. *Co-state for associated to*

Current Value Hamiltonian

(discount factor) is not included in the *each state*

Dodfmer

$$H^{iC}(x, u^I, \lambda, t) = F^i(\) + \lambda \cdot f^i(\)$$

$$H = e^{-\alpha t} F + p f = \underbrace{e^{-\alpha t}}_{\lambda} (F + \underbrace{e^{\alpha t} p f}_{\lambda}) = e^{-\alpha t} H^C$$

$\arg\max_u H \approx \arg\max H^C$

$$\lambda = \underbrace{e^{-\alpha t} p f}_{\lambda} \rightarrow -$$

Differential game Example: OPEN-LOOP NASH EQ.

(Dockner p.87)

$$\begin{aligned}
 \max_{u \geq 0} J^1(u(\cdot)) &= \int_0^T e^{-rt} \left[v(t) - x(t) - \frac{\alpha}{2} u^2(t) \right] dt \\
 \max_{v \in [0,1]} J^2(v(\cdot)) &= \int_0^T e^{-rt} [v(t) - x(t)] dt \\
 \text{s.t. } \dot{x}(t) &= 1 + v(t) - u(t) \sqrt{x(t)}, \\
 x(0) &= x^0
 \end{aligned}$$

Assume $v(t) = \psi(t)$ for P2, find best response strategy for P1

$$H^{C1}(x, u, \lambda, t) = P_0 \left(\psi - x - \frac{\alpha}{2} u^2 \right) + \lambda (1 + \psi - u \sqrt{x})$$

(P₀=1)

Assume $u(t) = \Phi(t)$ for P1, find best response strategy for P2

$$H^{C2}(x, v, \lambda, t) = P_0 (v - x) + \lambda (1 + v - \Phi \sqrt{x})$$

(P₀=1)

Pontryagin's Maximum Principle approach (discount factor) $\rho \geq 0$

learning ~ inner loop

$$\begin{aligned} & \text{maximize } J(\underline{u}) = \int_0^T e^{-\rho t} F_0(\underline{x}(t), \underline{u}(t), t) dt + e^{-\rho T} S(\underline{x}(T)) \\ & \text{subject to } \dot{\underline{x}}(t) = \underline{f}(\underline{x}(t), \underline{u}(t), t) \\ & \quad \underline{x}(0) = \underline{x}^0 \\ & \quad x_i(T) = x_i^1 \quad i = 1, \dots, l \\ & \quad x_i(T) \geq x_i^1 \quad i = l+1, \dots, m \\ & \quad x_i(T) \in \mathbb{R} \quad i = m+1, \dots, n \\ & \quad \underline{u}(t) \in \Omega \end{aligned}$$

Terminal Cost!

associated **current value** Hamiltonian function

$$H^C(\underline{x}, \underline{u}, \underline{\lambda}, t) = p_0 F_0(x(t), u(t), t) + \underline{\lambda} \cdot \underline{f}(\underline{x}(t), \underline{u}(t), t)$$

$$H^C(\underline{x}, \underline{u}, \underline{\lambda}, t) = p_0 F_0(x(t), u(t), t) + \sum_{i=1}^{i=n} \lambda_i f_i(\underline{x}(t), \underline{u}(t), t)$$

Pontryagin Maximum Principle (discount factor)

Theorem

Let $u^*(t)$ be a piecewise continuous control defined on $[0, T]$ which solves problem (DOC) and let $x^*(t)$ be the associated optimal path. Then $\exists n+1$ constants $p_0, \gamma_1, \dots, \gamma_n \in \mathbb{R}$ and a continuous and piecewise continuously differentiable function $\lambda(t) = (\lambda_1(t), \dots, \lambda_n(t))$ such that $\forall t \in [0, T]$

- $(p_0, \gamma_1, \dots, \gamma_n) \neq (0, 0, \dots, 0)$
- $u^*(t)$ maximizes $H^C(x^*(t), u, \lambda(t), t)$ for all $u \in \Omega$
- Excepts at the points of discontinuities of $u^*(t)$ co-state equation (ODE)

$$\dot{\lambda}_i(t) = -\frac{\partial H^C(x^*(t), u^*(t), \lambda(t), t)}{\partial x_i} + \rho \lambda_i(t)$$

- $p_0 \in \{0, 1\}$ for $i = 1, \dots, n$
- Transversality conditions (\rightarrow next page)

Theorem

- *Transversality conditions*

$$\lambda_i(T) = p_0 \frac{\partial S(x^*(T))}{\partial x_i} + \gamma_i, \quad i = 1, \dots, n$$

where

$$\gamma_i \in \Re, \quad i = 1, \dots, l \quad \text{if } x_i^*(T) = x'_i$$

$$\gamma_i \geq 0 \quad i = l+1, \dots, m \quad \text{if } x_i^*(T) \geq x'_i$$

~~complementary
conditions~~

$$\sim \gamma_i(x_i^*(T) - x'_i) = 0$$

$$\gamma_i = 0 \quad i = m+1, \dots, n \quad \text{if } x_i^*(T) \in \Re$$

Player 1

$$H^{C_1}(x, u, \lambda, t) = u - x - \frac{d}{2} u^2 + \lambda (1 + u - u\sqrt{x})$$

ii) $u \in \arg \max_{u \geq 0} H^{C_1}$

$$\frac{\partial H^{C_1}}{\partial u} = -u - \lambda \sqrt{x} = 0 \Rightarrow u(t) = -\frac{\lambda(t) \sqrt{x(t)}}{d}$$

$$\frac{\partial^2 H^{C_1}}{\partial u^2} = -2 < 0 \quad \text{concave}$$

 $\lambda \in [0, \infty)$

iii) COSTATE EQ

$$\dot{\lambda}(t) = -\frac{\partial H^{C_1}}{\partial x} + \kappa \lambda(t) = 1 + \frac{\lambda(t)u}{2\sqrt{x}} + \kappa \lambda(t)$$

$$\left\{ \begin{array}{l} \dot{\lambda}(t) = 1 - \frac{\lambda^2(t)}{2\sqrt{x}} \cdot \frac{\sqrt{x}}{d} + \kappa \lambda(t) \\ \lambda(T) = 0 \end{array} \right.$$

Riccati
eq

$x(T) \in \mathbb{R}$ free $S(x(T)) \approx 0$

Player 1

$$H^{C^1}(x, u, \lambda, t) = 0 - x - \frac{a}{2}u^2 + \lambda(1 + \varepsilon - u\sqrt{x})$$

$$\bullet u^* = \underset{u \geq 0}{\operatorname{argmax}} H^{C^1}$$

$$\frac{\partial H^{C^1}}{\partial u} = -du - \lambda\sqrt{x} = 0$$

$$\frac{\partial^2 H^{C^1}}{\partial u^2} = -d < 0 \quad \text{concave}$$

$$u^*(t) = -\frac{\lambda(t)\sqrt{x(t)}}{d}$$

$$\bullet \dot{\lambda}(t) = -\frac{\partial H^{C^1}}{\partial x} + \varepsilon \lambda(t) = 1 + \frac{\lambda(t)u}{2\sqrt{x}} + \varepsilon \lambda(t)$$

CO-STATE SYSTEM PDE backward

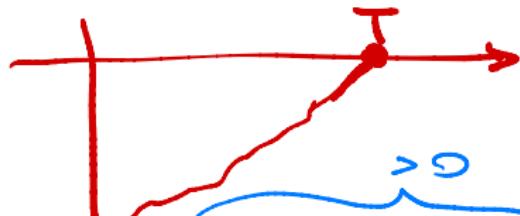
$$\begin{cases} \dot{\lambda}(t) = 1 - \frac{\lambda^2(t)}{2\sqrt{x}} + \varepsilon \lambda(t) \\ \lambda(T) = 0 \end{cases} \quad \text{Picard eq}$$

$$\left| \begin{array}{l} S(x(T)) = 0 \\ x(T) \in \mathbb{R} \end{array} \right. \Rightarrow \xi = 0$$

Player 1

$$\dot{\lambda}(\tau) = 1$$

$$\dot{\lambda} > 0$$



$$\exists! \lambda(t) = \frac{e^{(1-e^{-\sqrt{x^2+2/d}(T-t)})}}{(x - \sqrt{x^2+2/d}) e^{-c(T-t)} - (x + \sqrt{x^2+2/d})} \leq 0$$

$\underbrace{(x - \sqrt{\quad})}_{<(x-\sqrt{\quad})} < \underbrace{(x + \sqrt{\quad})}_{<(x+\sqrt{\quad})}$

$$\phi(t) = u^* = -\frac{\lambda \sqrt{x}}{d} \geq 0$$

best response

$$\boxed{\lambda(t) = 0 \Leftrightarrow t = T}$$

Player 1

$$\begin{cases} \dot{x}(t) = u + \psi(t) + \frac{\lambda(t)x(t)}{d} \\ x(0) = x_0 \end{cases} \text{ Linear ODE}$$

strategy of P2 ?
 $x(t) = \underline{\quad} - ?$

Such a ψ is only a candidate

Apply ARROW's Sufficiency Theorem

$$(H^{c_1})^* = H^{c_2}(x, u^*, \lambda, t) = v - x - \frac{d}{2} \left(-\frac{\lambda \nabla x}{d} \right)^2 + \lambda(u + v + \frac{\lambda \nabla x}{d} \cdot \nabla v)$$
$$= v - x - \frac{\lambda^2 x^2}{2d} + \lambda + \lambda v + \frac{\lambda^2 x^2}{d}$$

linear in x
⇒ concave in x

$\Rightarrow \psi$ is optimal strategy. \Rightarrow ARROW ✓

Player 2

$$u(t) = \phi(t) \text{ player 2}$$

$$H^{C2}(x, \sigma, \mu, t) = \sigma - x + \mu(x + \sigma - \frac{1}{\phi} \sqrt{\alpha}) \text{ linear in } \sigma$$

- $\underset{\sigma \in [0, 1]}{\operatorname{argmax}} H^{C2} = (\sigma + \mu)x - x + \mu(x - \phi \sqrt{x})$

$$\phi = \sigma = \begin{cases} 0, & \mu < -1 \\ 1, & \mu > -1 \end{cases}$$

bang-bang
sol

$$\phi = -\frac{\lambda \sqrt{x}}{\alpha}$$

- $\dot{x}(t) = -\frac{\partial H^{C2}}{\partial x} + x \phi(t) = 1 + \frac{\mu \phi}{2\sqrt{x}} + 10\mu e(t)$

$$= 1 + x \mu(t) - \frac{\mu \lambda}{2\alpha}$$

- $\mu(\tau) = 0$

Player 2

$$\begin{cases} \dot{\lambda} = \lambda + \varepsilon \lambda - \frac{1}{2d} \lambda^2 & \lambda(\tau) = 0 \\ \dot{\mu} = \lambda + \varepsilon \mu - \frac{1}{2d} \mu \lambda & \mu(\tau) = 0 \end{cases}$$

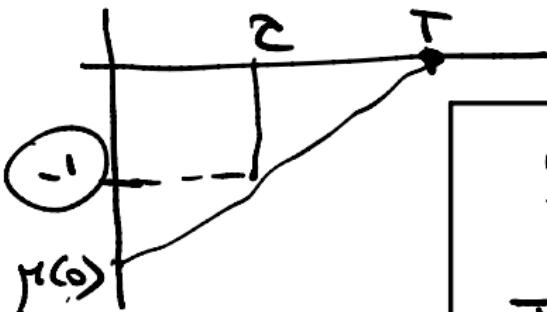
$$\dot{\lambda} - \dot{\mu} = \frac{1}{2d} \lambda (\mu - \lambda) + \varepsilon (\lambda - \mu)$$

$$y(t) = \lambda - \mu$$

$$\begin{cases} \dot{y}(t) = \frac{\lambda y}{2d} + \varepsilon y = \left(\frac{\lambda}{2d} + \varepsilon \right) y \Rightarrow y(t) = 0 \\ \underline{y(\tau) = 0} \end{cases}$$

$\lambda(t) = \mu(t)$

Player 2



$$\begin{aligned} u(0) &< -1 \\ \Updownarrow \\ T &> \frac{1}{c} \ln \left(\frac{2 - (x - \sqrt{x^2 + 2/c})}{2 - (x + \sqrt{x^2 + 2/c})} \right) = c \end{aligned}$$

$$u(0) > -1 \Rightarrow c = 0$$

$$u(t) = \psi(t) = \begin{cases} 0, & t \in [0, c] \\ 1, & t \in [c, T] \end{cases}$$

*degenerate strategy
not depending
on the initial state*

Check optimality

$$H^{CC}(\dot{x}, \dot{\sigma}, \dot{y}, t) = \dot{\sigma}^2 - \dot{x} + \mu (\dot{x} + \dot{\sigma}^2 - \phi \sqrt{x})$$

$\frac{-2\sqrt{x}}{\alpha}$

$\overset{\Delta}{\underset{0}{\overset{1}{\Delta}}} \quad \overset{\Delta}{\underset{0}{\overset{1}{\Delta}}}$

Linear in $\dot{x} \Rightarrow$ Kakutani's Theorem holds

Open-Loop Nash EQ

$$(\phi^*, \psi^*) \quad \underline{\phi^*(t)} = -\frac{\gamma(t)}{\alpha} \sqrt{x(t)}$$

$$\underline{\psi^*(t)} = \begin{cases} 0 & t \in [0, \bar{t}) \\ 1 & t \in [\bar{t}, T] \end{cases}$$

\bar{x}, \bar{t} see above

Player 2

$\dot{x}^*(t)$ State functions

$$\begin{cases} \dot{x}(t) = 1 + \psi(t) - \alpha(t) \sqrt{x(t)} \\ x(0) = x_0 \end{cases}$$

$$\alpha^* = -\frac{\lambda \sqrt{x}}{\alpha}$$

$$\text{In } [0, \infty) \quad \psi = 0 = \psi^*$$

$$\dot{x}(t) = 1 + \cancel{\lambda x}(t)$$

$$x(t) = e^{\int_0^t \frac{\lambda(s)}{\alpha} ds} \left\{ x_0 + \int_0^t (1 + \psi) e^{-\int_0^s \frac{\lambda(\zeta)}{\alpha} d\zeta} ds \right\}$$

$x(t)$

(x_0)

Player 2

$$\ln [z, \tau] \quad \dot{z} = 1 = \psi^*$$

$$\begin{cases} \dot{x}(t) = 2 + \frac{\gamma(t)x(t)}{d} \\ x(0) = x_0 \end{cases}$$

$$x(t) = e^{\int_0^t \frac{\gamma(s)}{d} ds} \left[x_0 + \int_0^t 2 e^{-\int_s^t \frac{\gamma(\zeta)}{d} d\zeta} ds \right]$$

//