

*AA 598B Special Topics*

# Decision-Making & Control for Safe Interactive Autonomy

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*Autumn 2024*

<https://faculty.washington.edu/kymleung/aa598/>



# Announcements

- Guest lecture by Prof. David Fridovich-Keil from UT Austin today
- Proposal feedback
- Homework 2 due (recommended)
- Start your project!

# Last time

- Socially-aware planning approaches
  - Including the human agent's reward in the robot's reward
  - Social Value Orientation
  - Courtesy / counterfactual reasoning
  - Proactive and legibility
- Methods to solve an optimal control problem
  - Sequential quadratic programming
    - Assumes ability to get (good) gradient information

# Today

- Brief introduction to game theory
- Sampling-based methods for planning

# Game theory

“No dynamics”

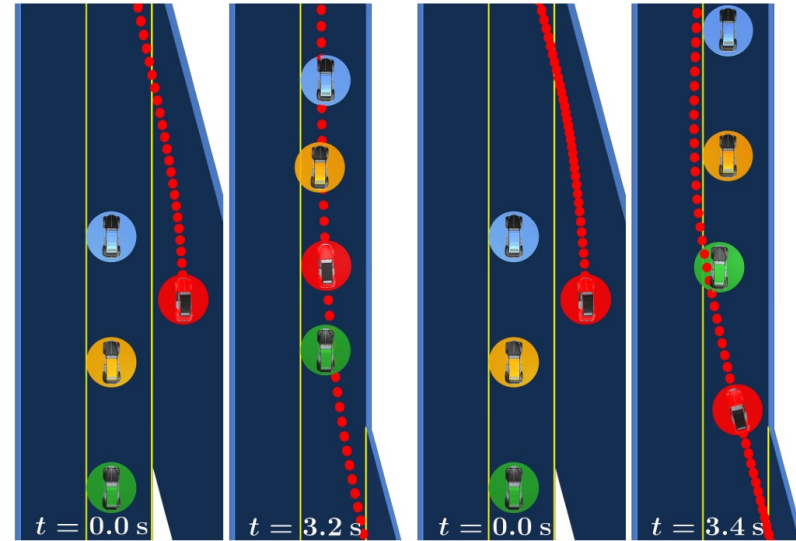
Prisoners' dilemma

		prisoner B			
		confess	remain silent	confess	remain silent
prisoner A	confess	 5 years 5 years	 0 year 20 years		
	remain silent	 20 years 0 year	 1 year 1 year		

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<https://www.britannica.com/science/game-theory/The-von-Neumann-Morgenstern-theory>

Dynamic games



ALGAMES: A Fast Augmented Lagrangian Solver for Constrained Dynamic Games

Autonomous Robots (AuRo 2021),

S. Le Cleac'h, M. Schwager, Z. Manchester

# Game theory

**Definition:** A mathematical framework for modeling scenarios in which multiple decision-makers (agents) interact, with each agent's outcome depending not only on its own actions but also on the actions of others.

**Relevance:** In human-robot interaction, game theory helps model how robots can make decisions while considering the possible actions of human agents.

# Problem formulation

$$\begin{array}{ll} \min_{X, U^1} J^1(X, U^1) & \min_{X, U^M} J^M(X, U^M) \\ \text{s.t. } D(X, U) = 0, & \cdots \quad \text{s.t. } D(X, U) = 0, \\ C(X, U) \leq 0, & C(X, U) \leq 0, \end{array}$$

# Problem formulation

$$\forall i \in [N] \left\{ \begin{array}{l} \min_{X^i, U^i} J^i(\mathbf{X}, U^i; \theta^i) \\ \text{s.t. } x_{t+1}^i = f^i(x_t^i, u_t^i), \forall t \in [T - 1] \\ x_1^i = \hat{x}_1^i \\ p g^i(X^i, U^i) \geq 0 \\ {}^s g(\mathbf{X}, \mathbf{U}) \geq 0. \end{array} \right.$$



# Payoff structure

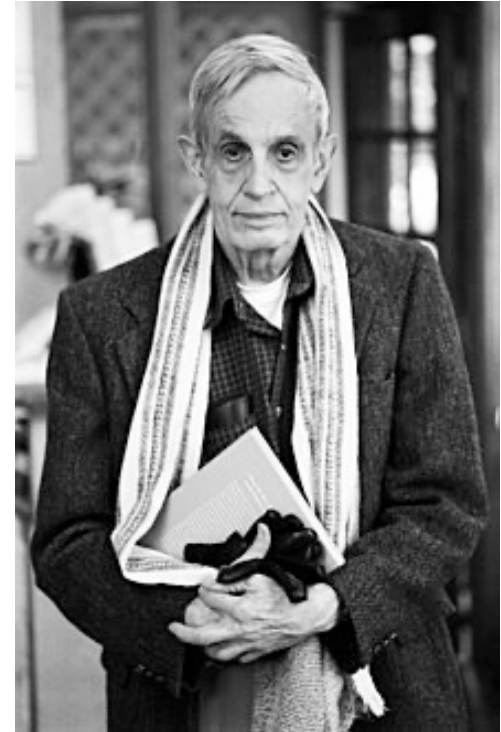
- Zero sum two player game
  - Total payoff always sums to zero
  - One player's gain is exactly equal to the other player's loss
  - E.g., tic-tac-toe
- General sum games
  - Payoff does not need to sum to zero
  - No strong sense of win or lose
  - More representative of human-robot interactions(?)

# Types of solutions: Nash equilibrium

- At Nash equilibrium, every player is playing optimally given the choices of others,
  - No player has an incentive to deviate from their chosen strategy.

$$J_i(u_i^*, u_{-i}^*) \leq J_i(u_i, u_{-i}^*) \quad \forall u_i \in U_i$$

- Other types:
  - Subgame Perfect Equilibrium (Nash over multiple steps)
  - Correlated Equilibrium (follow recommendation from external source)
  - Bayesian Nash Equilibrium (Nash with incomplete information, have beliefs over others)



Wikipedia

# Finding Nash equilibria

Generally difficult to find

- Find *local open-loop Nash equilibria* via KKT conditions

$$\begin{aligned} \min_x \quad & f(x) \\ \text{s.t.} \quad & \mathbf{g}(x) \leq 0 \\ & \mathbf{h}(x) = 0 \end{aligned} \quad \mathcal{L}(x, \mu, \lambda) = f(x) + \mu^T \mathbf{g}(x) + \lambda^T \mathbf{h}(x)$$

$$\nabla_x \mathcal{L}(x^*, \mu^*, \lambda^*) = 0$$

$$\nabla_\mu \mathcal{L}(x^*, \mu^*, \lambda^*) = 0$$

$$\nabla_\lambda \mathcal{L}(x^*, \mu^*, \lambda^*) = 0$$

$$\mu^* \geq 0$$

“First-order optimality conditions with constraints”

**KKT conditions**

# Inferring costs and objectives

$$\forall i \in [N] \begin{cases} \min_{\mathbf{x}, \mathbf{u}^i} J^i(\mathbf{u}; x_1) & (1a) \\ \text{s.t. } x_{t+1} = f_t(x_t, u_t^1, \dots, u_t^N), \forall t \in [T-1]. & (1b) \end{cases}$$

have data on  $y$ .

$$\begin{aligned} \max_{\theta, \mathbf{x}, \mathbf{u}} \quad & p(\mathbf{y} | \mathbf{x}, \mathbf{u}) & (4a) \\ \text{s.t.} \quad & (\mathbf{x}, \mathbf{u}) \text{ is an OLNE of } \Gamma(\theta) & (4b) \\ & (\mathbf{x}, \mathbf{u}) \text{ is dynamically feasible under } f, & (4c) \end{aligned}$$

KKT conditions

$$\mathbf{G}(\mathbf{x}, \mathbf{u}, \boldsymbol{\lambda}) := \begin{bmatrix} \nabla_{\mathbf{x}} J^i + \boldsymbol{\lambda}^{i\top} \nabla_{\mathbf{x}} \mathbf{F}(\mathbf{x}, \mathbf{u}) \\ \nabla_{\mathbf{u}^i} J^i + \boldsymbol{\lambda}^{i\top} \nabla_{\mathbf{u}^i} \mathbf{F}(\mathbf{x}, \mathbf{u}) \\ \mathbf{F}(\mathbf{x}, \mathbf{u}) \end{bmatrix} \forall i \in [N] = \mathbf{0}. \quad (5)$$

KKT

$$\begin{aligned} \max_{\theta, \mathbf{x}, \mathbf{u}, \boldsymbol{\lambda}} \quad & p(\mathbf{y} | \mathbf{x}, \mathbf{u}) \\ \text{s.t.} \quad & \mathbf{G}(\mathbf{x}, \mathbf{u}, \boldsymbol{\lambda}; \theta) = \mathbf{0}. \end{aligned}$$