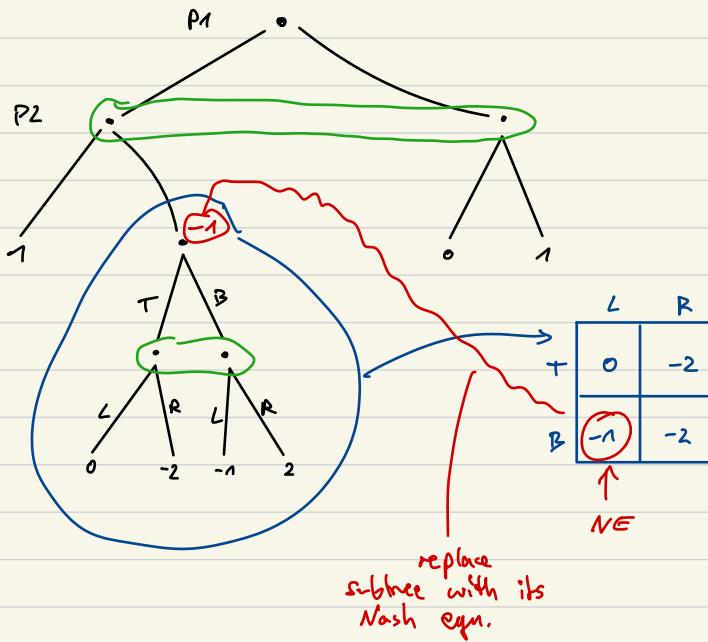


LECTURE 5

PART 1: EXTENSIVE FORM GAMES

Example for backward induction:

Consider the zero-sum game: (player 1: min, player 2: max)



We can conclude that the overall value of the game is 0.

Subgame Perfect Equilibria

Consider a feedback game (i.e. inf. sets don't span over stages) with K stages. Let I/J be the set of all inf. sets of player 1 / player 2.

Feedback game $\Rightarrow I = \bigcup_{i=1}^k I_i$ where I_i are inf. set at stage i

Define $\gamma: I \rightarrow A$ with $\gamma = \{\gamma_1, \dots, \gamma_k\}$
where $\gamma_i: I_i \rightarrow A$

and similarly define $\sigma: I \rightarrow A$ for player 2.

A pair of strategies (γ^*, σ^*) is a subgame perfect eq. if for all stages $t = 1, \dots, k$ and every $\gamma_1, \gamma_2, \dots, \gamma_{t-1}$ and $\sigma_1, \sigma_2, \dots, \sigma_{t-1}$ prior to stage t , for all γ_t, σ_t , we have

$$J(\gamma_{1:t-1}, \gamma_t^*, \gamma_{t+1:k}^*, \sigma_{1:t-1}, \sigma_t^*, \sigma_{t+1:k})$$

$$= J(\gamma_{1:t-1}, \gamma_t^*, \gamma_{t+1:k}^*, \sigma_{1:t-1}, \sigma_t^*, \sigma_{t+1:k})$$

$$= J(\gamma_{1:t-1}, \gamma_t^*, \gamma_{t+1:k}^*, \sigma_{1:t-1}, \sigma_t^*, \sigma_{t+1:k})$$

where we use the notation $\gamma_{1:t-1} := (\gamma_1, \dots, \gamma_{t-1})$, meaning

$$J(\gamma_{1:t-1}, \gamma_t^*, \gamma_{t+1:k}^*, \sigma_{1:t-1}, \sigma_t^*, \sigma_{t+1:k}) :=$$

$$J(\gamma_1, \gamma_2, \dots, \gamma_{t-1}, \gamma_t^*, \gamma_{t+1}^*, \dots, \gamma_k^*, \sigma_1, \sigma_2, \dots, \sigma_{t-1}, \sigma_t^*, \sigma_{t+1}^*, \dots, \sigma_k^*)$$

Nash Equilibria

(γ^*, σ^*) is a Nash Eqn. if $\forall \gamma_{1:k}, \sigma_{1:k}$ we have

$$J(\gamma_{1:k}^*, \sigma_{1:k}) \leq J(\gamma_{1:k}^*, \sigma_{1:k}^*) \leq J(\gamma_{1:k}, \sigma_{1:k}^*)$$

Observe that: subgame perfect equilibrium
 (see Lemma 7.1)
 in Hespanha book) \Downarrow
 Nash equilibrium

Proof idea:

For stage $t=1$ by definition of subgame perfect equilibrium,

$$\forall y^*, J(y_1^*, y_2^* :_{\mathbb{K}}, \sigma^*) \leq J(y_1, y_2^* :_{\mathbb{K}}, \sigma^*)$$

$$\begin{aligned} & \xrightarrow{\text{by applying def. for stage 2}} J(y_1, y_2, y_3^* :_{\mathbb{K}}, \sigma^*) \\ & \vdots \\ & \leq J(y_1 :_{\mathbb{K}}, \sigma^*) \end{aligned}$$

PART 2: Review (for Quiz)

① General-sum games with finite action spaces

Ex 1 (see Thm. 10.1 in Hespanha book)

Consider a 2-pl. general-sum game with costs $A, B \in \mathbb{R}^{m \times n}$

To find mixed-strategy NE $y^* \in \mathcal{Y}$, $z^* \in \mathcal{Z}$ we can solve the quadratic optimization problem

$$\begin{aligned} & \text{quadratic objective fct.} \quad \left\{ \begin{array}{l} \min_{y \in \mathbb{R}^m, z \in \mathbb{R}^n, q, p \in \mathbb{R}} \\ \quad y^T (A + B) z - p - q \end{array} \right. \\ & \text{linear constraints} \quad \left\{ \begin{array}{l} \text{s.t.} \\ \quad A z \geq 1 \cdot p, \quad 1^T z = 1, \quad z \geq 0 \\ \quad B^T y \geq q \cdot 1, \quad 1^T y = 1, \quad y \geq 0 \end{array} \right. \end{aligned}$$

coordinate-wise

To get an idea for why this is hard to solve, we want to show that the objective function is non-convex.

Recall: For $f: \mathbb{R}^d \rightarrow \mathbb{R}$, $f(x) = x^T Q x$,

f is convex \iff all eigenvalues of $Q + Q^T$ are ≥ 0

How to show this?

\hookrightarrow let $x_1, x_2 \in \mathbb{R}^n$, $\lambda \in [0, 1]$; we need to show that

$$f(\lambda x_1 + (1-\lambda)x_2) \leq \lambda f(x_1) + (1-\lambda)f(x_2) \iff \text{all ev's of } Q + Q^T \text{ are } \geq 0$$

\hookrightarrow some algebra ...

$$(\lambda x_1 + (1-\lambda)x_2)^T Q (\lambda x_1 + (1-\lambda)x_2)$$

$$= \lambda x_1^T Q x_1 + \dots$$

⋮

$$\text{more calculations yield: } 0 \leq (x_1 - x_2)^T \left(\frac{Q + Q^T}{2} \right) (x_1 - x_2) \quad \forall x_1, x_2$$

$$\iff 0 \leq y^T \left(\frac{Q + Q^T}{2} \right) y, \quad \forall y$$

\iff all ev's of $Q + Q^T$ are ≥ 0

In our case we want to show non-convexity:

$$f: \mathbb{R}^{n+m} \rightarrow \mathbb{R}$$

$$f(y, z) := y^T (A + B) z = \begin{bmatrix} y \\ z \end{bmatrix}^T \underbrace{\begin{bmatrix} 0 & \frac{A+B}{2} \\ \frac{A+B^T}{2} & 0 \end{bmatrix}}_{:= Q} \begin{bmatrix} y \\ z \end{bmatrix}$$

For any z , set $y = (A+B)z$. Then $z^T(A+B)^T(A+B)z \geq 0$.

But if we instead pick $y = -(A+B)z$, then $-z^T(A+B)^T(A+B)z = 0$.

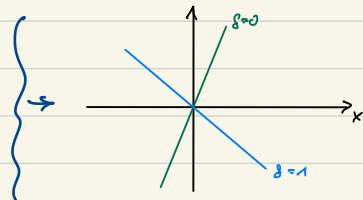
\Rightarrow eigenvalues of $Q+Q^T$ not all ≥ 0

\Rightarrow non-convexity.

② Review of 0-sum games

Ex 2: Consider $f: [-1, 1] \times \{0, 1\} \rightarrow \mathbb{R}$

$$f(x, \delta) := \begin{cases} 2x, & \text{if } \delta=0 \\ -x, & \text{if } \delta=1. \end{cases}$$

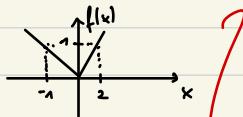


Does it hold that

$$\min_{x \in [-1, 1]} \max_{\delta \in \{0, 1\}} f(x, \delta) = \max_{\delta \in \{0, 1\}} \min_{x \in [-1, 1]} f(x, \delta)$$

(Recall: We know that LHS \geq RHS from inf. sup. theorem.)

LHS: $\max_{\delta \in \{0, 1\}} f(x, \delta)$ looks like



$$\text{RHS: } \min_{x \in [-1, 1]} f(x, \delta) = \begin{cases} -2 & \text{if } \delta=0 \\ 1 & \text{if } \delta=1 \end{cases}$$

We conclude that LHS \neq RHS!

(recall that equality is only guaranteed to hold in certain cases, e.g. convex-concave objective)

③ Potential games

Ex 3

Let $J_i: T_1 \times T_2 \times \dots \times T_N \rightarrow \mathbb{R}$. Suppose the game with payoffs $\{J_i\}_{i=1}^N$ admits a potential function.

TRUE or FALSE?

1) $\gamma^* \in T$ is a NE $\Rightarrow \gamma^*$ is a maximizer of the potential function
 \rightarrow FALSE (e.g. stag hunt: only one of NEs maximizes potential function)

2) $\gamma^* \in T$ is maximizer of potential $\Rightarrow \gamma^*$ is NE
 \rightarrow TRUE (proven in lecture)

3) best reply dynamics converges \Rightarrow game is potential

\rightarrow FALSE (we proved " \Leftarrow " direction; as counterexample consider

$\begin{bmatrix} (1,0) & (2,0) \\ (2,0) & (0,1) \end{bmatrix}$] NE (players are maximizers)

↓
best reply converges to NE but
game is not potential \rightarrow verify this!)