# mputational gic 

## Game Theory

## Georg Moser

## Institute of Computer Science @ UIBK

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## Definition

let $\sigma \in \prod_{i \in N} \Delta\left(C_{i}\right)$, let $u_{i}(\sigma)$ denote the expected utility payoff for player $i$, when players choose strategies according to $\sigma$ :

$$
u_{i}(\sigma)=\sum_{c \in C}\left(\prod_{j \in N} \sigma_{j}\left(c_{j}\right)\right) u_{i}(c) \quad \text { for all } i \in N
$$

for $\tau_{i} \in \Delta\left(C_{i}\right)$, let $\left(\sigma_{-i}, \tau_{i}\right)$ denote the randomised strategy profile, where $\tau_{i}$ is substituted for $\sigma_{i}$, thus

$$
u_{i}\left(\sigma_{-i}, \tau_{i}\right)=\sum_{c \in C}\left(\prod_{j \in N \backslash\{i\}} \sigma_{j}\left(c_{j}\right)\right) \tau_{i}\left(c_{i}\right) u_{i}(c)
$$

## Definition

Nash equilibrium
a randomised strategy profile $\sigma$ is a Nash equilibrium of $\Gamma$ if the following holds for all $i \in N$, and every $\tau_{i} \in \Delta\left(C_{i}\right)$

$$
u_{i}(\sigma) \geqslant u_{i}\left(\sigma_{-i}, \tau_{i}\right)
$$

## Content

motivation, introduction to decision theory, decision theory
basic model of game theory, dominated strategies, common knowledge, Bayesian games, incomplete information, Nash equilibrium
two-person zero-sum games, Bayesian equilibria, sequential equilibra of extensive-form games, subgame-perfect equilibra
(efficient) computation of Nash equilibria, complexity class PPAD, complexity of Nash equilibria, refinements of equilibrium in strategic form, persistent equilibria, games with communication, sender-receiver games

## Existence of Nash Equilibrium

## Theorem

given a finite game $\Gamma$ in strategic form, there exists at least one (Nash)
equilibrium in $\prod_{i \in N} \Delta\left(C_{i}\right)$

## Example

|  | $C_{2}$ |  |
| :---: | :---: | :---: |
| $C_{1}$ | $M$ | $F$ |
| $R r$ | 0,0 | $1,-1$ |
| $R p$ | $0.5,-0.5$ | 0,0 |
| $P r$ | $-0.5,0.5$ | $1,-1$ |
| $P p$ | 0,0 | 0,0 |

then no pure equilibrium exists, and we can only eliminated $P p$
Fact
randomised strategies are needed for this theorem

## Definition

the outcome of a game in Pareto efficient if there is no other outcome that would make all players better of
a game may have equilibria that are inefficient, and a game may have multiple equilibria

Example
prisoner dilemma


- the only equilibrium is $\left(\left[f_{1}\right],\left[f_{2}\right]\right)$ which is inefficient

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| :---: | :---: | :---: | :---: |
| Example |  |  |  |
|  |  | $C_{2}$ |  |
|  | $C_{1}$ | $f_{2}$ | $s_{2}$ |
|  | $f_{1}$ | 3,1 | 0,0 |
|  | $s_{1}$ | 0,0 | 1,3 |

- the game as two pure equilibria

$$
\left(\left[f_{1}\right],\left[f_{2}\right]\right) \quad\left(\left[s_{1}\right],\left[s_{2}\right]\right)
$$

- and one (inefficient) mixed equilibria

$$
\left(0.75\left[f_{1}\right]+0.25\left[s_{1}\right], 0.25\left[f_{2}\right]+0.75\left[s_{2}\right]\right)
$$

## The Focal-Point Effect

## Definition

focal-point effect
anything that tends to focus the players' attention on one equilibrium may make them all expect it and hence fulfil it; this is called focal-point effect

## Example

battle of the sexes with communication

|  | $C_{2}$ |  |  |  |
| :---: | :---: | :---: | :---: | :---: |
| $C_{1}$ | $f_{2} f_{2}$ | $f_{2} s_{2}$ | $s_{2} f_{2}$ | $s_{2} s_{2}$ |
| $F f_{1}$ | 3,1 | 3,1 | 0,0 | 0,0 |
| $F s_{1}$ | 0,0 | 0,0 | 1,3 | 1,3 |
| $S f_{1}$ | 3,1 | 0,0 | 3,1 | 0,0 |
| $S s_{1}$ | 0,0 | 1,3 | 0,0 | 1,3 |

## Definition

if a game can be influence by preplay communication, the player whose words are headed is called focal arbitrator

|  | $C_{2}$ |  |
| :---: | :---: | :---: |
| $C_{1}$ | $f_{2}$ | $s_{2}$ |
| $f_{1}$ | 3,1 | 0,0 |
| $s_{1}$ | 0,0 | $\mathbf{1 , 3}$ |

Example
battle of the sexes (3)

|  | $C_{2}$ |  |
| :---: | :---: | :---: |
| $C_{1}$ | $f_{2}$ | $s_{2}$ |
| $f_{1}$ | 3,1 | 0,0 |
| $s_{1}$ | 0,0 | 1,3 |

- assumption: the man is Dr. Taub and he has recently confessed his adultery
- there are two players
- both can make demands for sum $[1,100]$ in $€$ i.e.,

$$
C_{1}=C_{2}=\{x \in \mathbb{R} \mid 0 \leqslant x \leqslant 100\}
$$

- the payoff function is defined as follows:

$$
u_{i}\left(c_{1}, c_{2}\right)= \begin{cases}0 & \text { if } c_{1}+c_{2}>100 \\ c_{i} & \text { if } c_{1}+c_{2} \leqslant 100\end{cases}
$$

## Analysis

- any pair $(x, 100-x)$ is an equilibrium, on the other hand also the pair $(100,100)$ is an equilibrium
- an impartial moderator may suggest $(50,50)$ as it is efficient
- moreover $(50,50)$ has strong incentive, it is a focal equilibrium even if the moderator is absent


## Evolution

Idea
Axelrod 1984
identify good strategies by a biological evolutionary criterion
Definition

- $L_{i} \subseteq \Delta\left(C_{i}\right)$ of promising randomised strategies
- $\forall$ player $i$
- $\exists i$-animals that implement a strategy $\sigma_{i} \in L_{i}$
- each $i$-animal plays the game repeatedly using $\sigma_{i}$
- $\forall$ player $j \neq i$
- let the $j$-animals randomly choose among the strategies in $L_{j}$
- define

$$
q_{j}^{k}\left(\sigma_{j}\right)=\frac{j \text {-animals that implement } \sigma_{j}}{\text { all } j \text {-animals }}
$$

(in generation $k$ )

## Definition

- define

$$
\bar{\sigma}_{-i}^{k}\left(c_{j}\right)=\sum_{\sigma_{j} \in L_{j}} q_{j}^{k}\left(\sigma_{j}\right) \sigma_{j}\left(c_{j}\right) \quad \forall j \in N, \forall c_{j} \in C_{j}
$$

- set $\bar{\sigma}^{k}=\left(\bar{\sigma}_{j}^{k}\right)_{j \in N}$
- and $\bar{u}_{i}^{k}\left(\sigma_{i}\right)=u_{i}\left(\bar{\sigma}_{-i}^{k}, \sigma_{i}\right)$


## Definition

the number of children in the next generation $k+1$ depends on the expected payoff:

$$
q_{i}^{k+1}\left(\sigma_{i}\right)=\frac{q_{i}^{k}\left(\sigma_{i}\right) \bar{u}_{i}^{k}\left(\sigma_{i}\right)}{\sum_{\tau_{i} \in L_{i}} q_{i}^{k}\left(\sigma_{i}\right) \bar{u}_{i}^{k}\left(\tau_{i}\right)}
$$

## "Definition"

strategies that survive in the end, are good
strategies that behave poorly can be crucial to determine which strategy reproduces best

## Risk Dominance

Idea (2)
Harsanyi, Selten 1988
overcome this dependency on poor strategies, using risk dominance of
strategies

## Definition

- $\forall$ games $\Gamma$ is strategic form
- $\forall \sigma, \tau$ equilibria in $\prod_{i \in N} \Delta\left(C_{i}\right)$ the resistance of $\sigma$ against $\tau$ is the largest $\lambda \in[0,1]$ such that $\forall j \in N$ :

$$
u_{i}\left(\left(\lambda \tau_{j}+(1-\lambda) \sigma_{j}\right)_{j \in N-\{i\}}, \sigma_{i}\right) \geqslant u_{i}\left(\left(\lambda \tau_{j}+(1-\lambda) \sigma_{j}\right)_{j \in N-\{i\}}, \tau_{i}\right)
$$

- an equilibrium $\sigma$ risk dominates another equilibrium $\tau$ if the resistance of $\sigma$ against $\tau$ is greater than the resistance of $\tau$ against $\sigma$


## Note

the resistance measure the "evolutionary" strength of an equilibrium

## Two-Person Zero-Sum Games

## Example

|  | $C_{2}$ |  |
| :---: | :---: | :---: |
| $C_{1}$ | $M$ | $F$ |
| $R r$ | 0,0 | $1,-1$ |
| $R p$ | $0.5,-0.5$ | 0,0 |
| $P r$ | $-0.5,0.5$ | $1,-1$ |
| $P p$ | 0,0 | 0,0 |

## Observation

$$
u_{1}\left(c_{1}, c_{2}\right)=-u_{2}\left(c_{1}, c_{2}\right) \quad \forall c_{1} \in\{R r, R p, \operatorname{Pr}, \operatorname{Pp}\} \quad \forall c_{2} \in\{M, F\}
$$

## Definition

a two-person zero-sum game $\Gamma$ in strategic form is a game
$\Gamma=\left(\{1,2\}, C_{1}, C_{2}, u_{1}, u_{2}\right): u_{1}\left(c_{1}, c_{2}\right)=-u_{2}\left(c_{1}, c_{2}\right) \forall c_{1} \in C_{1}, \forall c_{2} \in C_{2}$

## Min-Max Theorem

## Theorem

( $\sigma_{1}, \sigma_{2}$ ) is an equilibrium of a finite two-person zero-sum game
$\Gamma=\left(\{1,2\}, C_{1}, C_{2}, u_{1},-u_{1}\right)$ if and only if

$$
\begin{aligned}
& \sigma_{1} \in \operatorname{argmax}_{\tau_{1} \in \Delta\left(C_{1}\right)} \min _{\tau_{2} \in \Delta\left(C_{2}\right)} u_{1}\left(\tau_{1}, \tau_{2}\right) \\
& \sigma_{2} \in \operatorname{argmin}_{\tau_{2} \in \Delta\left(C_{2}\right)} \max _{\tau_{1} \in \Delta\left(C_{1}\right)} u_{1}\left(\tau_{1}, \tau_{2}\right)
\end{aligned}
$$

furthermore if ( $\sigma_{1}, \sigma_{2}$ ) an equilibrium of $\Gamma$, then

$$
u_{1}\left(\sigma_{1}, \sigma_{2}\right)=\max _{\tau_{1} \in \Delta\left(C_{1}\right)} \min _{\tau_{2} \in \Delta\left(C_{2}\right)} u_{1}\left(\tau_{1}, \tau_{2}\right)=\min _{\tau_{2} \in \Delta\left(C_{2}\right)} \max _{\tau_{1} \in \Delta\left(C_{1}\right)} u_{1}\left(\tau_{1}, \tau_{2}\right)
$$

## Proof

easy

## Observation

withouth randomised strategies, the existence of an equilibrium cannot be guranteed and the min-max theorem fail

## Example

|  | $C_{2}$ |  |
| :---: | :---: | :---: |
| $C_{1}$ | $M$ | $F$ |
| $R r$ | 0,0 | $1,-1$ |
| $R p$ | $0.5,-0.5$ | 0,0 |
| $P r$ | $-0.5,0.5$ | $1,-1$ |
| $P p$ | 0,0 | 0,0 |

- allow only the pure strategies
- we obtain

$$
\begin{aligned}
& \max _{c_{1} \in\{R r, R p, P r, P p\}} \min _{c_{2} \in\{M, F\}} u_{1}\left(c_{1}, c_{2}\right)=\max \{0,0,-0.5,0\}=0 \\
& \min _{c_{2} \in\{M, F\}} \max _{c_{1} \in\{R r, R p, P r, P p\}} u_{1}\left(c_{1}, c_{2}\right)=\min \{0.5,1\}=0.5 \neq 0
\end{aligned}
$$

- 「 doesn't admit a pure equilibrium


## Example (cont'd)

- proof of the theorem uses the existence of a Nash equilibrium, this is essential
- we need this for

$$
\max _{\tau_{1} \in \Delta\left(C_{1}\right)} \min _{\tau_{2} \in \Delta\left(C_{2}\right)} u\left(\tau_{1}, \tau_{2}\right)=\min _{\tau_{2} \in \Delta\left(C_{2}\right)} \max _{\tau_{1} \in \Delta\left(C_{1}\right)} u\left(\tau_{1}, \tau_{2}\right)
$$

## Definition

an optimisation problem is defined as

$$
\operatorname{minimise}_{x \in \mathbb{R}^{n}} f(x) \quad \text { subject to } g_{i}(x) \geqslant 0 \quad \forall i \in\{1, \ldots, m\}
$$

where $f, g_{1}, \ldots, g_{m}$ are functions from $\mathbb{R}^{n} \rightarrow \mathbb{R}$

## Observation

two-person zero-sum games and optimisation problems are closely linked

## Lemma

the optimisation problem

$$
\operatorname{minimise}_{x \in \mathbb{R}^{n}} f(x) \quad \text { subject to } g_{i}(x) \geqslant 0 \quad \forall i \in\{1, \ldots, m\}
$$

is equivalent to

$$
\begin{equation*}
\operatorname{minimise}_{x \in \mathbb{R}^{n}}\left(\max _{y \in \mathbb{R}_{+}^{m}} f(x)-\sum_{i=1}^{m} y_{i} g_{i}(x)\right) \tag{1}
\end{equation*}
$$

here $R_{+}^{m}=\left\{\left(y_{1}, \ldots, y_{m}\right) \mid y_{i} \geqslant 0\right\}$
Proof
observe that $\max _{y \in \mathbb{R}_{+}^{m}}\left(f(x)-\sum_{i=1}^{m} y_{i} g_{i}(x)\right)=f(x)$ if the constraints are met, otherwise it is $+\infty$

## Definition

the dual of (1) is defined as

$$
\operatorname{maximise}_{y \in \mathbb{R}_{+}^{m}}\left(\min _{x \in \mathbb{R}^{n}} f(x)-\sum_{i=1}^{m} y_{i} g_{i}(x)\right)
$$

## Bayesian Equlibria

consider

$$
\Gamma^{b}=\left(N,\left(C_{i}\right)_{i \in N},\left(T_{i}\right)_{i \in N},\left(p_{i}\right)_{i \in N},\left(u_{i}\right)_{i \in N}\right)
$$

such that

- $T_{i}$ is the set of types of player $i ; T=\prod_{i \in N} T_{i}$
- $p_{i}\left(\cdot \mid t_{i}\right) \in \Delta\left(T_{-i}\right)$ is the probability distribution over the types of the other players $T_{-i}$
- for each $i: u_{i}: C \times T \rightarrow \mathbb{R}$ is the expected utility payoff


## Definition

- strategy for player $i$ is a function $f: T \rightarrow C$
- randomised strategy profile $\sigma \in \prod_{i \in N} \prod_{t_{i} \in T_{i}} \Delta\left(C_{i}\right)$
Definition
Bayesian equilibrium
$\sigma_{i}\left(\cdot \mid t_{i}\right) \in \operatorname{argmax}_{\tau_{i} \in \Delta\left(C_{i}\right)} \sum_{t_{-i} \in T_{-i}} p_{i}\left(t_{-i} \mid t_{i}\right) \sum_{c \in C}\left(\prod_{j \in N \backslash\{i\}} \sigma_{j}\left(c_{j} \mid t_{j}\right)\right) \tau_{i}\left(c_{i}\right) u_{i}(c, t)$

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