

Sequential Games and Optimal Strategies

Paulo Oliva

Queen Mary University of London



LIAMF, USP

São Paulo, 20 October 2011



Outline

- 1 Player-based Games
- 2 Quantifiers and Selection Functions
- 3 Playerless Games
- 4 Computing Optimal Strategies



Outline

- 1 Player-based Games
- 2 Quantifiers and Selection Functions
- 3 Playerless Games
- 4 Computing Optimal Strategies



Single-player Games

SUDOKU 数独 HARD
Time: 19:09

8		4		2	9	4		6	
2	5	7	4	1	⁴ ₅			9	7
9			1	⁵	8			3	4
5	2	6	⁷ ₇				2	1	3
4		6		9			7		8
1	1	3	2	⁴ ₃ ⁴ ₃		7		5	
	9	2	3		4	⁵ ₇		6	
⁷ ₃	6	⁵			1	3	2	1	
⁷ ₃	1	4	7		9	4	⁷ ₃	2	



Two-player Games

Two **players**: Black and White



Two-player Games

Two **players**: Black and White

Possible **outcomes**:

- Black wins
- White wins
- Draw



Two-player Games

Two **players**: Black and White

Possible **outcomes**:

- Black wins
- White wins
- Draw



Strategy: Choice of move at round k given previous moves



Another Game

Two **players**: John and Julia



Another Game

Two **players**: John and Julia



John splits a cake. Julia chooses one of the two pieces



Another Game



Two **players**: John and Julia

John splits a cake. Julia chooses one of the two pieces

Possible **outcomes**:

- John gets $N\%$ of the cake (John's payoff)
- Julia gets $(100 - N)\%$ of the cake (Julia's payoff)



Another Game



Two **players**: John and Julia

John splits a cake. Julia chooses one of the two pieces

Possible **outcomes**:

- John gets $N\%$ of the cake (John's payoff)
- Julia gets $(100 - N)\%$ of the cake (Julia's payoff)

Best strategy for John is to split cake into half

It is not a “winning strategy” but it is an **optimal strategy**

It maximises his payoff



Traditional Game Theory

Game defined via:

- Set of **players** P
- Sets of **moves** X_i for each player $i \in P$
- Set of **outcomes** R
- **Preference relations** on R for each player $i \in P$
- **Outcome function** mapping plays to outcomes



Set of Players vs Number of Rounds

Number of players is not essential

It is important what the “goal” at each round is

Rounds with “**same goal**” mean played by “**same player**”



Set of Players vs Number of Rounds

Number of players is not essential

It is important what the “goal” at each round is

Rounds with “**same goal**” mean played by “**same player**”

How to describe the goal at a particular round?



Set of Players vs Number of Rounds

Number of players is not essential

It is important what the “goal” at each round is

Rounds with “**same goal**” mean played by “**same player**”

How to describe the goal at a particular round?

You could say: The goal is to win!

But maybe this is not possible (or might not even make sense)

Instead, the goal should be described as

a choice of outcome from each set of possible outcomes



As in...

Q: How much would you like to pay for your flight?



As in...

Q: How much would you like to pay for your flight?

A: As little as possible!



Target function

If $R =$ set of outcomes and $X =$ set of possible moves then

$$\phi \in (X \rightarrow R) \rightarrow R$$

describes the desired outcome $\phi p \in R$ given that the outcome of the game $px \in R$ for each move $x \in X$ is given.



Target function

If R = set of outcomes and X = set of possible moves then

$$\phi \in (X \rightarrow R) \rightarrow R$$

describes the desired outcome $\phi p \in R$ given that the outcome of the game $px \in R$ for each move $x \in X$ is given.

In the example:

X	=	<i>possible flights</i>
R	=	<i>real number</i>
$X \rightarrow R$	=	<i>price of each flight</i>
ϕ	=	<i>minimal value functional</i>



Outline

- 1 Player-based Games
- 2 Quantifiers and Selection Functions**
- 3 Playerless Games
- 4 Computing Optimal Strategies



Generalised quantifiers

$$\phi : (X \rightarrow R) \rightarrow R$$



Generalised quantifiers

$$\phi : (X \rightarrow R) \rightarrow R$$

For instance

Operation	$\phi : (X \rightarrow R) \rightarrow R$
Quantifiers	$\forall_X, \exists_X : (X \rightarrow \mathbb{B}) \rightarrow \mathbb{B}$
Double negation	$\neg\neg X : (X \rightarrow \perp) \rightarrow \perp$
Integration	$\int_0^1 : ([0, 1] \rightarrow \mathbb{R}) \rightarrow \mathbb{R}$
Supremum	$\sup_{[0,1]} : ([0, 1] \rightarrow \mathbb{R}) \rightarrow \mathbb{R}$
Limit	$\lim : (\mathbb{N} \rightarrow R) \rightarrow R$
Fixed point operator	$\text{fix}_X : (X \rightarrow X) \rightarrow X$



Generalised quantifiers

$$\phi : (X \rightarrow R) \rightarrow R \quad (\equiv K_R X)$$

For instance

Operation	$\phi : (X \rightarrow R) \rightarrow R$
Quantifiers	$\forall_X, \exists_X : (X \rightarrow \mathbb{B}) \rightarrow \mathbb{B}$
Double negation	$\neg\neg X : (X \rightarrow \perp) \rightarrow \perp$
Integration	$\int_0^1 : ([0, 1] \rightarrow \mathbb{R}) \rightarrow \mathbb{R}$
Supremum	$\sup_{[0,1]} : ([0, 1] \rightarrow \mathbb{R}) \rightarrow \mathbb{R}$
Limit	$\lim : (\mathbb{N} \rightarrow R) \rightarrow R$
Fixed point operator	$\text{fix}_X : (X \rightarrow X) \rightarrow X$



Theorem (Mean Value Theorem)

For any $p \in C[0, 1]$ there is a point $a \in [0, 1]$ such that

$$\int_0^1 p = p(a)$$



Theorem (Mean Value Theorem)

For any $p \in C[0, 1]$ there is a point $a \in [0, 1]$ such that

$$\int_0^1 p = p(a)$$

Theorem (Maximum Value Theorem)

For any $p \in C[0, 1]$ there is a point $a \in [0, 1]$ such that

$$\sup p = p(a)$$



Theorem (Witness Theorem)

For any $p: X \rightarrow \mathbb{B}$ there is a point $a \in X$ such that

$$\exists x^X p(x) \Leftrightarrow p(a)$$

(similar to Hilbert's ε -term).



Theorem (Witness Theorem)

For any $p: X \rightarrow \mathbb{B}$ there is a point $a \in X$ such that

$$\exists x^X p(x) \Leftrightarrow p(a)$$

(similar to Hilbert's ε -term).

Theorem (Counter-example Theorem)

For any $p: X \rightarrow \mathbb{B}$ there is a point $a \in X$ such that

$$\forall x^X p(x) \Leftrightarrow p(a)$$

(a is counter-example to p if one exists).



Let $JX \equiv (X \rightarrow R) \rightarrow X$



Let $JX \equiv (X \rightarrow R) \rightarrow X$

Definition (Selection Functions)

$\varepsilon: JX$ is called a **selection function** for $\phi: (X \rightarrow R) \rightarrow R$ if

$$p(\varepsilon p) = \phi(p)$$

holds for all $p: X \rightarrow R$



Let $JX \equiv (X \rightarrow R) \rightarrow X$

Definition (Selection Functions)

$\varepsilon: JX$ is called a **selection function** for $\phi: (X \rightarrow R) \rightarrow 2^R$ if

$$p(\varepsilon p) \in \phi(p)$$

holds for all $p: X \rightarrow R$



Let $JX \equiv (X \rightarrow R) \rightarrow X$

Definition (Selection Functions)

$\varepsilon: JX$ is called a **selection function** for $\phi: (X \rightarrow R) \rightarrow 2^R$ if

$$p(\varepsilon p) \in \phi(p)$$

holds for all $p: X \rightarrow R$

Definition (Attainable Quantifiers)

A generalised quantifier $\phi: KX$ is called **attainable** if it has a selection function $\varepsilon: JX$

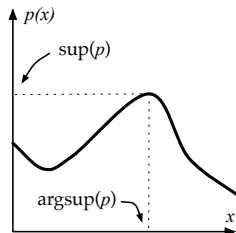


For Instance

- $\text{sup} : K_{\mathbb{R}}[0, 1]$ is an attainable quantifier
as

$$\text{sup}(p) = p(\text{argsup}(p))$$

where $\text{argsup} : J_{\mathbb{R}}[0, 1]$



For Instance

- $\text{sup}: K_{\mathbb{R}}[0, 1]$ is an attainable quantifier
as

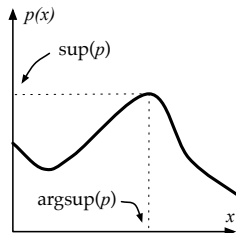
$$\text{sup}(p) = p(\text{argsup}(p))$$

where $\text{argsup}: J_{\mathbb{R}}[0, 1]$

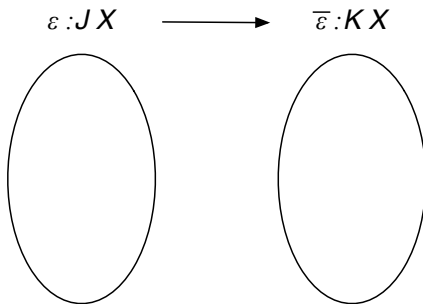
- $\text{fix}: K_X X$ is an attainable quantifier as

$$\text{fix}(p) = p(\text{fix}(p))$$

where $\text{fix}: J_X X (= K_X X)$



Selection Functions and Generalised Quantifiers

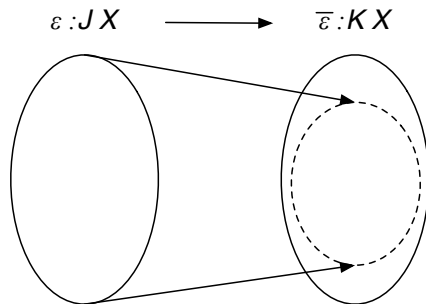


Every selection function $\varepsilon : JX$ defines a quantifier $\bar{\varepsilon} : KX$

$$\bar{\varepsilon}(p) = p(\varepsilon(p))$$



Selection Functions and Generalised Quantifiers

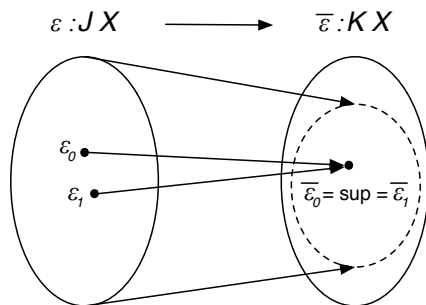


Not all quantifiers are attainable, e.g. $R = \{0, 1\}$

$$\phi(p) = 0$$



Selection Functions and Generalised Quantifiers



Different ε might define same ϕ , e.g. $X = [0, 1]$ and $R = \mathbb{R}$

$$\varepsilon_0(p) = \mu x. \text{sup } p = p(x)$$

$$\varepsilon_1(p) = \nu x. \text{sup } p = p(x)$$



Outline

- 1 Player-based Games
- 2 Quantifiers and Selection Functions
- 3 Playerless Games**
- 4 Computing Optimal Strategies



Finite Sequential Games

Definition (A tuple $(R, (X_i)_{i < n}, (\phi_i)_{i < n}, q)$ where)

- R is the set of **possible outcomes**
- X_i is the set of **available moves** at round i
- $\phi_i: (X_i \rightarrow R) \rightarrow 2^R$ is the **goal quantifier** for round i
- $q: \prod_{i=0}^{n-1} X_i \rightarrow R$ is the **outcome function**



Finite Sequential Games

Definition (A tuple $(R, (X_i)_{i < n}, (\phi_i)_{i < n}, q)$ where)

- R is the set of **possible outcomes**
- X_i is the set of **available moves** at round i
- $\phi_i: (X_i \rightarrow R) \rightarrow 2^R$ is the **goal quantifier** for round i
- $q: \prod_{i=0}^{n-1} X_i \rightarrow R$ is the **outcome function**

Definition (Strategy)

Family of mappings

$$\text{next}_k: \prod_{i=0}^{k-1} X_i \rightarrow X_k$$



Optimal Strategies

Definition (Strategic Play)

Given strategy next_k and partial play $\vec{a} = a_0, \dots, a_{k-1}$, the **strategic extension** of \vec{a} is $\vec{b}^{\vec{a}} = b_k^{\vec{a}}, \dots, b_{n-1}^{\vec{a}}$ where

$$b_i^{\vec{a}} = \text{next}_i(\vec{a}, b_k^{\vec{a}}, \dots, b_{i-1}^{\vec{a}})$$



Optimal Strategies

Definition (Strategic Play)

Given strategy next_k and partial play $\vec{a} = a_0, \dots, a_{k-1}$, the **strategic extension** of \vec{a} is $\mathbf{b}^{\vec{a}} = b_k^{\vec{a}}, \dots, b_{n-1}^{\vec{a}}$ where

$$b_i^{\vec{a}} = \text{next}_i(\vec{a}, b_k^{\vec{a}}, \dots, b_{i-1}^{\vec{a}})$$

Definition (Optimal Strategy)

Strategy next_k is **optimal** if for any partial play \vec{a}

$$q(\vec{a}, \mathbf{b}^{\vec{a}}) \in \phi_k(\lambda x_k \cdot q(\vec{a}, x_k, \mathbf{b}^{\vec{a}, x_k}))$$



Examples

Example (Nash Equilibrium with common payoff)

Moves X_i

Sets of moves

Outcomes R

Payoff \mathbb{R}

Goal quantifier ϕ_i

Maximal value function

Outcome function q

Payoff function $q: \prod_{i=0}^{n-1} X_i \rightarrow \mathbb{R}$



Examples

Example (Nash Equilibrium with common payoff)

Moves X_i

Sets of moves

Outcomes R

Payoff \mathbb{R}

Goal quantifier ϕ_i

Maximal value function

Outcome function q

Payoff function $q: \prod_{i=0}^{n-1} X_i \rightarrow \mathbb{R}$

Optimal strategy

$$\text{next}_k(x_0, \dots, x_{k-1}) = \text{argsup}_{x_k} \sup_{x_{k+1}} \dots \sup_{x_{n-1}} q(\vec{x})$$



Examples

Example (Satisfiability)

Moves X_i

Booleans \mathbb{B}

Outcomes R

Boolean \mathbb{B}

Goal quantifier ϕ_i

Existential quantifier $\exists: K_{\mathbb{B}}\mathbb{B}$

Outcome function q

Formula $q(x_0, \dots, x_{n-1})$



Examples

Example (Satisfiability)

Moves X_i	Booleans \mathbb{B}
Outcomes R	Boolean \mathbb{B}
Goal quantifier ϕ_i	Existential quantifier $\exists: K_{\mathbb{B}}\mathbb{B}$
Outcome function q	Formula $q(x_0, \dots, x_{n-1})$

Optimal strategy

$\text{next}_k(x_0, \dots, x_{k-1}) = x_k$ such that $\exists x_{k+1} \dots \exists x_{n-1} q(\vec{x})$
 (if possible)



Outline

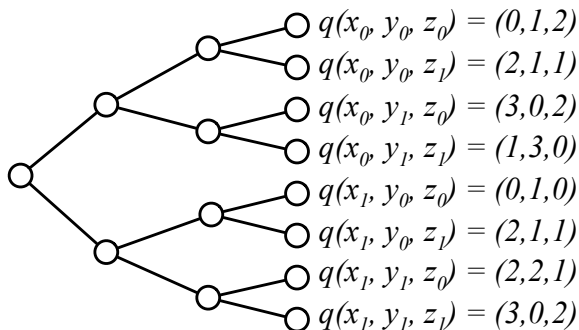
- 1 Player-based Games
- 2 Quantifiers and Selection Functions
- 3 Playerless Games
- 4 Computing Optimal Strategies**



Backward Induction (Classical Game Theory)

Three players, payoff function $q: X \times Y \times Z \rightarrow \mathbb{R}^3$

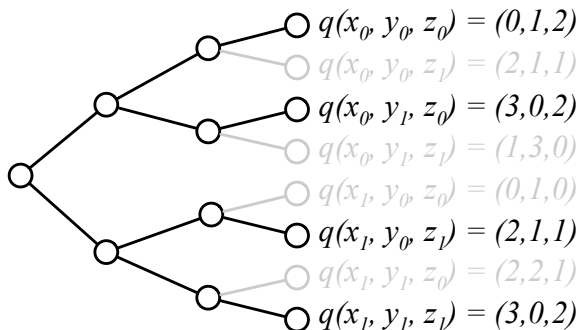
Each player is trying to maximise their own payoff



Backward Induction (Classical Game Theory)

Three players, payoff function $q: X \times Y \times Z \rightarrow \mathbb{R}^3$

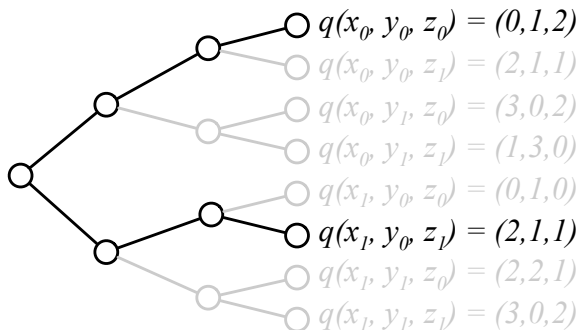
Each player is trying to maximise their own payoff



Backward Induction (Classical Game Theory)

Three players, payoff function $q: X \times Y \times Z \rightarrow \mathbb{R}^3$

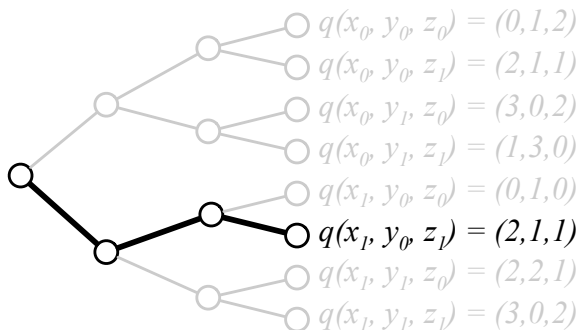
Each player is trying to maximise their own payoff



Backward Induction (Classical Game Theory)

Three players, payoff function $q: X \times Y \times Z \rightarrow \mathbb{R}^3$

Each player is trying to maximise their own payoff



Backward Induction (Classical Game Theory)

Let $\operatorname{argmax}_i: (X_i \rightarrow \mathbb{R}^n) \rightarrow X_i$ find a point $x \in X_i$
at which the function $p: X_i \rightarrow \mathbb{R}^n$ has maximal i -value

Backward Induction (Classical Game Theory)

Let $\operatorname{argmax}_i: (X_i \rightarrow \mathbb{R}^n) \rightarrow X_i$ find a point $x \in X_i$ at which the function $p: X_i \rightarrow \mathbb{R}^n$ has maximal i -value

Consider n player. Given $q: \prod_{i=0}^{n-1} X_i \rightarrow \mathbb{R}^n$, define

$$\text{BI}(s) \stackrel{\prod_{j=|s|}^{n-1} X_j}{=} \begin{cases} [] & \text{if } n = |s| \\ c_s * \text{BI}(s * c_i) & \text{otherwise} \end{cases}$$

where $c_s = \operatorname{argmax}_{|s|}(\lambda x. q(s * x * \text{BI}(s * x)))$

Backward Induction (Classical Game Theory)

Let $\operatorname{argmax}_i: (X_i \rightarrow \mathbb{R}^n) \rightarrow X_i$ find a point $x \in X_i$ at which the function $p: X_i \rightarrow \mathbb{R}^n$ has maximal i -value

Consider n player. Given $q: \prod_{i=0}^{n-1} X_i \rightarrow \mathbb{R}^n$, define

$$\operatorname{BI}(s) \stackrel{\prod_{j=|s|}^{n-1} X_j}{=} \begin{cases} [] & \text{if } n = |s| \\ c_s * \operatorname{BI}(s * c_i) & \text{otherwise} \end{cases}$$

where $c_s = \operatorname{argmax}_{|s|}(\lambda x. q(s * x * \operatorname{BI}(s * x)))$

Each player's **optimal strategy** can be described as

$$\operatorname{next}_i(s) = \operatorname{argmax}_{|s|} \underbrace{(\lambda x. q(s * x * \operatorname{BI}(s * x)))}_{p: X_{|s|} \rightarrow \mathbb{R}^n}$$

Spector's Bar Recursion (1962)

Let

$$s: X^* \quad q: X^* \rightarrow R \quad \varepsilon_s: J_R X$$

Spector's Bar Recursion (1962)

Let

$$s: X^* \quad q: X^* \rightarrow R \quad \varepsilon_s: J_R X$$

Given s, ω and ε_s define

$$\text{BR}(s) \stackrel{X^*}{=} \begin{cases} [] & \text{if } n = |s| \\ c * \text{BR}(s * c) & \text{otherwise} \end{cases}$$

where $c = \varepsilon_s(\lambda x. q(s * x * \text{BR}(s * x)))$

Spector's Bar Recursion (1962)

Let

$$s: X^* \quad q: X^* \rightarrow R \quad \boxed{\varepsilon_s: J_R X}$$

Given s, ω and ε_s define

$$\text{BR}(s) \stackrel{X^*}{=} \begin{cases} [] & \text{if } n = |s| \\ c * \text{BR}(s * c) & \text{otherwise} \end{cases}$$

where $c = \varepsilon_s(\lambda x. q(s * x * \text{BR}(s * x)))$

Spector's Bar Recursion (1962)

Let

$$s: X^* \quad q: X^* \rightarrow R \quad \boxed{\varepsilon_s: J_R X}$$

Given s, ω and ε_s define

$$\text{BR}(s) \stackrel{X^*}{=} \begin{cases} [] & \text{if } n = |s| \\ c * \text{BR}(s * c) & \text{otherwise} \end{cases}$$

where $c = \varepsilon_s(\lambda x. q(s * x * \text{BR}(s * x)))$

Spector actually defined a much more general recursion scheme where stopping condition depends on the play s

Main Theorem

Theorem (Escardó/O.'2011)

Given game (R, X_i, ϕ_i, q) , if ϕ_i are attainable with selection functions ε_i then

$$\text{next}(s) \stackrel{X}{=} (\text{BR}(s))_0$$

is an **optimal strategy**, i.e.

$$q(s * \mathbf{b}^s) \in \phi_{|s|}(\lambda x. q(s * x * \mathbf{b}^{s*x}))$$

where \mathbf{b}^s is the strategic extension of partial play s

Summary and Further Connections

- New notion of **sequential game** based on quantifiers



Summary and Further Connections

- New notion of **sequential game** based on quantifiers
- Generalisation of backward induction, based on selection functions, calculates **optimal strategies**



Summary and Further Connections

- New notion of **sequential game** based on quantifiers
- Generalisation of backward induction, based on selection functions, calculates **optimal strategies**
- Relates Nash equilibrium, backtracking, Bekič's lemma



Summary and Further Connections

- New notion of **sequential game** based on quantifiers
- Generalisation of backward induction, based on selection functions, calculates **optimal strategies**
- Relates Nash equilibrium, backtracking, Bekič's lemma
- Connection to proof theory

$KA \rightarrow A$ corresponds to **double negation elimination**

$JA \rightarrow A$ corresponds to **Peirce's law**



Summary and Further Connections

- New notion of **sequential game** based on quantifiers
- Generalisation of backward induction, based on selection functions, calculates **optimal strategies**
- Relates Nash equilibrium, backtracking, Bekič's lemma
- Connection to proof theory
 - $KA \rightarrow A$ corresponds to **double negation elimination**
 - $JA \rightarrow A$ corresponds to **Peirce's law**
- Calculation of strategies in general corresponds to Spector's bar recursion, used in the proof of **consistency of classical analysis**



References



M. Escardó and P. Oliva

Selection functions, bar recursion and backward induction

MSCS, 20(2):127-168, 2010



M. Escardó and P. Oliva

The Peirce translation and the double negation shift

LNCS, CiE'2010



M. Escardó and P. Oliva

Sequential games and optimal strategies

Proceedings of the Royal Society A, 2011

