

# Sequential Games and Optimal Strategies

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Logic Colloquium  
Paris, 25 July 2010



# Single-player Games

SUDOKU 数独 HARD  
Time: 19:09

8		4		2	9	4		6
2	5	7	4	1	<sup>4</sup> <sub>5</sub>		9	7
9			1	<sup>5</sup> <sub>6</sub>	8		3	4
5	2	6	<sup>7</sup> <sub>7</sub>			2	1	3
4		6		9		7		8
1	1	3	2	<sup>4</sup> <sub>3</sub>	<sup>4</sup> <sub>3</sub>	7		5
	9	2	3		4	<sup>5</sup> <sub>7</sub>	6	
<sup>7</sup> <sub>3</sub>	6	<sup>5</sup> <sub>5</sub>			1	3	2	1
<sup>7</sup> <sub>3</sub>	1	4	7		9	4	<sup>7</sup> <sub>3</sub>	2

Time: 0:54 Moves: 21



# Two-player Games

Two **players**: Black and White



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Possible **outcomes**:

- Black wins
- White wins
- Draw



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**Strategy**: Choice of move at round  $k$  given previous moves



## Another Game

Two **players**: John and Julia



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



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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



## Number of Player 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**”



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## Number of Player 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 play for your flight?**



As in...

**Q: How much would you like to play 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.



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**In the example:**

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





# Outline

- 1 Selection Functions
- 2 Sequential Games – Fixed Length
- 3 Sequential Games – Unbounded Length



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## Generalised quantifiers

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### 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

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

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$$\sup_x \int_0^1 p(x, y) dy \quad \stackrel{\mathbb{R}}{\equiv} \quad (\sup \otimes \int)(p^{[0,1]^2 \rightarrow \mathbb{R}})$$



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### Definition (Product of Generalised Quantifiers)

Given  $\phi: KX$  and  $\psi: KY$  define  $\phi \otimes \psi : K(X \times Y)$

$$(\phi \otimes \psi)(p) \stackrel{R}{\equiv} \phi(\lambda x^X. \psi(\lambda y^Y. p(x, y)))$$

where  $p: X \times Y \rightarrow R$ .



## Theorem (Mean Value Theorem)

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

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



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

### 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  $\varepsilon$ -term).



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(similar to Hilbert's  $\varepsilon$ -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  $JX \equiv (X \rightarrow R) \rightarrow X$ .



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### Definition (Selection Functions)

$\varepsilon: JX$  is called a **selection function** for  $\phi: KX$  if

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

holds for all  $p: X \rightarrow R$ .





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### Definition (Attainable Quantifiers)

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

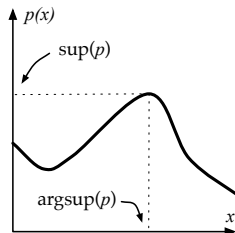


## For Instance

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

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

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



## For Instance

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

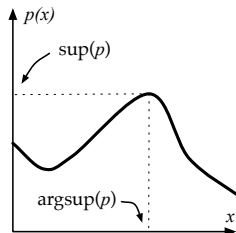
$$\text{sup}(p) = p(\text{argsup}(p))$$

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

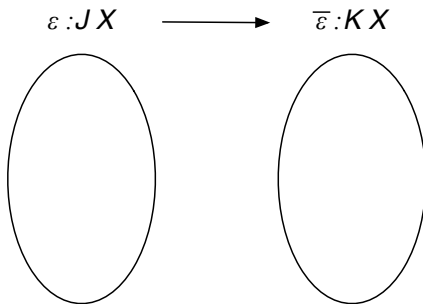
- $\text{fix}: K_X X$  is an attainable quantifier as

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

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



# Selection Functions and Generalised Quantifiers

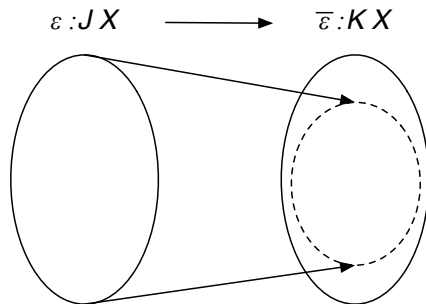


Every selection function  $\varepsilon : JX$  defines a quantifier  $\bar{\varepsilon} : KX$

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



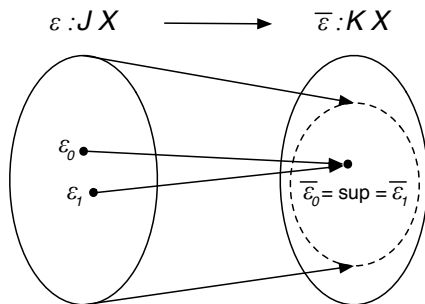
# Selection Functions and Generalised Quantifiers



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

$$\phi(p) = 0$$

# Selection Functions and Generalised Quantifiers



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

$$\varepsilon_0(p) = \mu x. \sup p = p(x)$$

$$\varepsilon_1(p) = \nu x. \sup p = p(x)$$



# Quantifier Elimination

Suppose

$$\exists x q(x) = q(\varepsilon q)$$

$$\forall y q(y) = q(\delta q).$$



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Then

$$\exists x \forall y p(x, y) = \exists x p(x, b(x))$$

where

$$b(x) = \delta(\lambda y. p(x, y))$$





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$$= p(a, b(a))$$

where

$$b(x) = \delta(\lambda y. p(x, y))$$

$$a = \varepsilon(\lambda x. p(x, b(x))).$$



# Quantifier Elimination

Suppose

$$\exists x q(x) = q(\varepsilon q)$$

$$\forall y q(y) = q(\delta q).$$

Then

$$\begin{aligned} (\exists_X \otimes \forall_Y)(p) &= \exists x p(x, b(x)) \\ &= p(a, b(a)) \end{aligned}$$

where

$$b(x) = \delta(\lambda y.p(x, y))$$

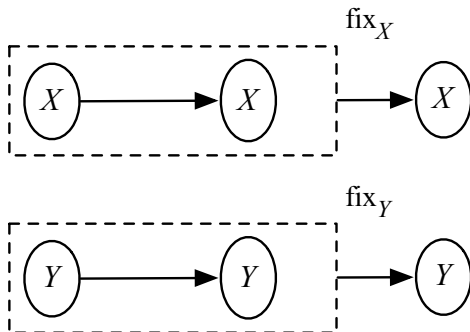
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## Bekič's Lemma

## Lemma

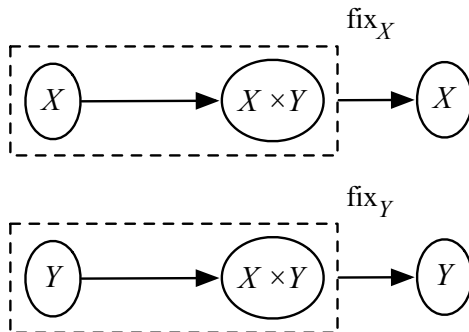
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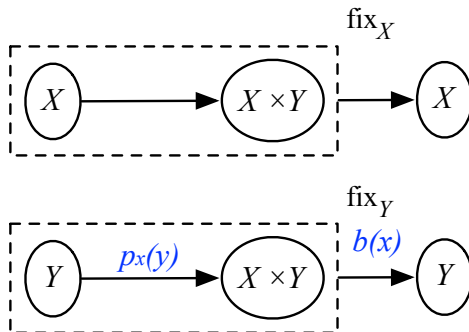
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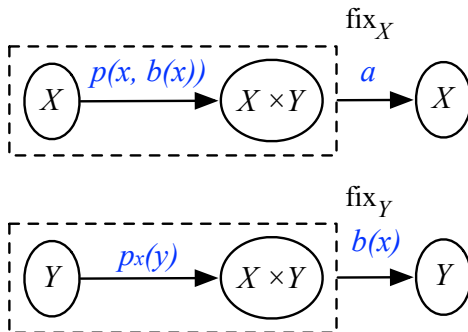
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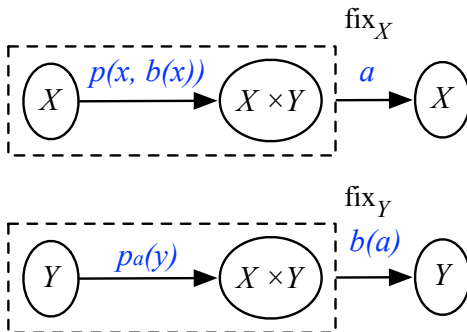
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## Definition (Product of Selection Functions)

Given  $\varepsilon: JX$  and  $\delta: JY$  define  $\varepsilon \otimes \delta: J(X \times Y)$  as

$$(\varepsilon \otimes \delta)(p^{X \times Y \rightarrow R}) \stackrel{X \times Y}{:=} ( \quad , \quad )$$

where





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## Theorem

$$\overline{\varepsilon \otimes \delta} = \overline{\varepsilon} \otimes \overline{\delta}$$



# Iterated Product of Selection Functions

## Finite iteration

$$\bigotimes_{i=k}^n \varepsilon_i \stackrel{J\Pi X_i}{=} \varepsilon_k \otimes \left( \bigotimes_{i=k+1}^n \varepsilon_i \right)$$



# Iterated Product of Selection Functions

**Infinite iteration** ( $R$  discrete,  $R^{\prod X_i}$  continuous)

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## Iterated Product of Selection Functions

**Infinite iteration I** ( $R$  discrete,  $R^{\prod X_i}$  continuous)

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**Infinite iteration II** ( $l: R \rightarrow \mathbb{N}$ ,  $\mathbb{N}^{\prod X_i}$  continuous)

$$\left( \bigotimes_{i=k}^{\infty} \varepsilon_i \right) (q) \stackrel{\prod X_i}{=} \begin{cases} \mathbf{c} & \text{if } k < l(q(\mathbf{c})) \\ (\varepsilon_k \otimes \left( \bigotimes_{i=k+1}^{\infty} \varepsilon_i \right)) (q) & \text{otherwise} \end{cases}$$



## Iterated Product of Selection Functions

**Infinite iteration I** ( $R$  discrete,  $R^{\Pi X_i}$  continuous) = MBR

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- 1 Selection Functions
- 2 Sequential Games – Fixed Length**
- 3 Sequential Games – Unbounded Length





## Finite Games ( $n$ rounds)

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  $i$
- $\phi_i: K_R X_i$  is the **goal quantifier** for round  $i$
- $q: \prod_{i=0}^{n-1} X_i \rightarrow R$  is the **outcome function**



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**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  $\text{next}_k$  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

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### Definition (Optimal Strategy)

Strategy  $\text{next}_k$  is **optimal** if for any partial play  $\vec{a}$

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



# Examples

## Example (Nash Equilibrium with common payoff)

Moves  $X_i$

Sets of moves

Outcomes  $R$

Payoff  $\mathbb{R}$

Goal quantifier  $\phi_i$

Maximal value function

Outcome function  $q$

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



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### 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  $\phi_i$

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

Outcome function  $q$

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



# Examples

## Example (Satisfiability)

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Outcome function  $q$

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### 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)





## Theorem (Main Theorem for Finite Games)

If  $\phi_k$  are attainable with selection functions  $\varepsilon_k$  then

$$\text{next}_k(x_0, \dots, x_{k-1}) \stackrel{X_k}{=} \left( \left( \bigotimes_{i=k}^{n-1} \varepsilon_i \right) (q_{x_0, \dots, x_{k-1}}) \right)_0$$

is an **optimal strategy** for the game  $(R, (X_i)_{i < n}, (\phi_i)_{i < n}, q)$ .

Moreover,

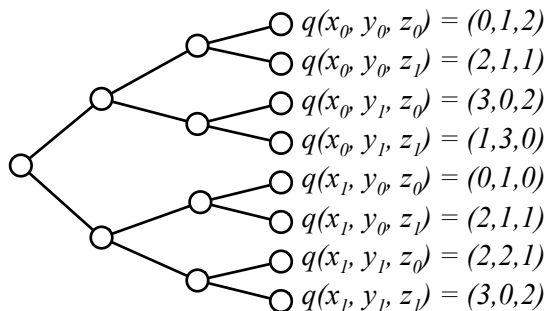
$$\vec{a} = \left( \bigotimes_{i=0}^{n-1} \varepsilon_i \right) (q)$$

is the **strategic play**.



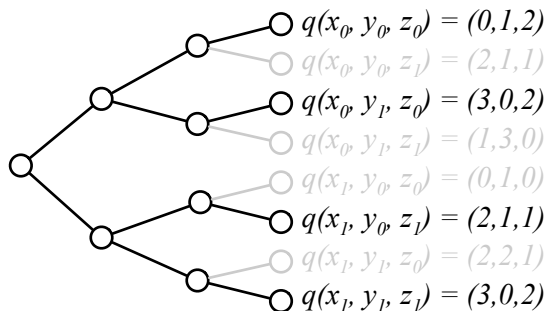
## Nash equilibrium (sequential games)

$$q: X \times Y \times Z \rightarrow \mathbb{R}^3$$



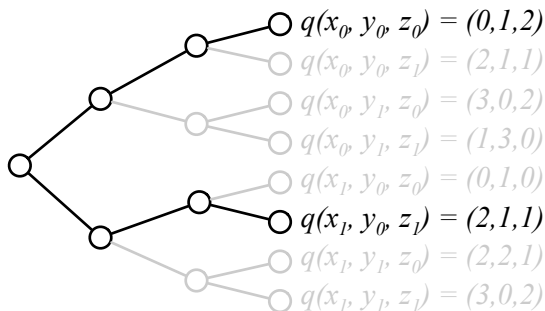
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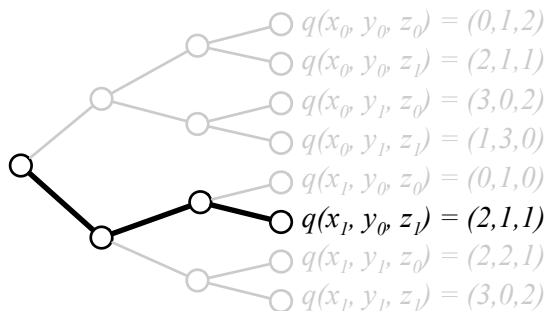
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## Backward Induction

Let  $q: \prod_{i=1}^n X_i \rightarrow \mathbb{R}^n$  be a payoff function

$\operatorname{argmax}_i(p) \left\{ \begin{array}{l} [\operatorname{argmax}_i: (X_i \rightarrow \mathbb{R}^n) \rightarrow X_i] \\ \text{return } x \in X_i \text{ such that } p(x) \text{ has maximal } i\text{-coordinate} \end{array} \right\}$



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$\operatorname{sol}_i(x_1, \dots, x_{i-1}) \{$   $[\operatorname{sol}_i: \prod_{k=1}^{i-1} X_k \rightarrow \prod_{k=i}^n X_k]$   
 if  $i = n + 1$  return  $\langle \rangle$   
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 $y := \operatorname{argmax}_i(\lambda x. q(\operatorname{sol}_{i+1}(x_1, \dots, x_{i-1}, x)))$   
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## Backward Induction

Let  $q: \prod_{i=1}^n X_i \rightarrow \mathbb{R}^n$  be a payoff function

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$\langle x_1, \dots, x_n \rangle := \operatorname{sol}_1()$





## Backward Induction

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

Each selection function

$$\operatorname{argmax}_i: (X_i \rightarrow \mathbb{R}^n) \rightarrow X_i$$

finds a point where the argument is  $i$ -maximal

Product

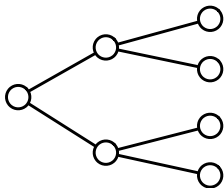
$$\operatorname{sol}_1(\cdot) = \left( \bigotimes_{i=1}^n \operatorname{argmax}_i \right) (q)$$

calculates a **strategy profile in Nash equilibrium**.



# Backtracking

good:  $X \times Y \rightarrow \mathbb{B}$

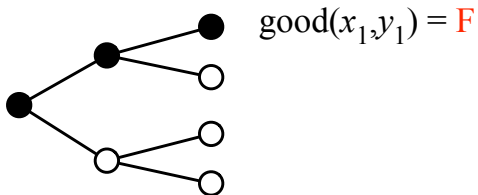


Generic algorithm has type  $(X \times Y \rightarrow \mathbb{B}) \rightarrow X \times Y$ .



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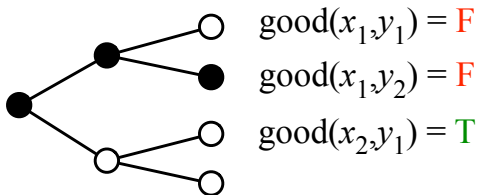


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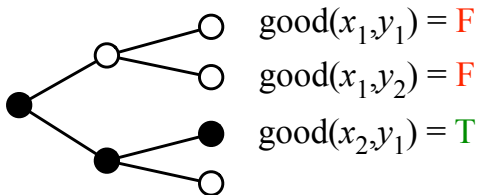


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# Backtracking

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## For Instance – Eight Queens Problem

```
 $\varepsilon(p)$  { [  $\varepsilon: (8 \rightarrow \mathbb{B}) \rightarrow 8$  ]  
  for ( $i := 1; i \leq 8; i++$ ) do  
    if  $p(i)$  return  $i$   
  return 1  
}
```



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if  $i > 8$  return  $\langle \rangle$

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calculates a solution to 8 queen problem.



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Equivalent (dialectica) to

$$\forall c^{\mathbb{N} \rightarrow n}, \forall \varepsilon^{\mathbb{J}_{\mathbb{N}} \mathbb{N}} \exists b^n, p^{\mathbb{N} \rightarrow \mathbb{N}} (\bar{\varepsilon}_b p \geq \varepsilon_b p \wedge c(\bar{\varepsilon}_b p) = b)$$



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Witnessed by

$$b = c\left(\overline{\left(\bigotimes_{i=0}^{n-1} \varepsilon_i\right)}(\max)\right)$$



# Outline

- 1 Selection Functions
- 2 Sequential Games – Fixed Length
- 3 Sequential Games – Unbounded Length





# Finite but Unbounded Games



## Finite but Unbounded Games

### Example (Chess)

Moves $X_i$	Valid chess moves
Outcomes $R$	White, black, draw, e.g. $\{-1, 0, 1\}$
Goal quantifier $\phi_{2i}$	Maximisation function
Goal quantifier $\phi_{2i+1}$	Minimisation function
Outcome function $q$	Adjudication on a given play $\alpha$



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*The game is drawn, upon a correct claim by the player having the move, if*

- he writes on his scoresheet, and declares to the arbiter his intention to make a move which shall result in the last 50 moves having been made by each player without the movement of any pawn and without the capture of any piece, or*
- the last 50 consecutive moves have been made by each player without the movement of any pawn and without the capture of any piece.*

With this rule, it can be shown that the game is finite, assuming that given the option to call for a draw, at least one player will do so.



## Finite but Unbounded Games

Definition (Tuple  $(R, (X_i)_{i \in \mathbb{N}}, (\phi_i)_{i \in \mathbb{N}}, q)$  where...)

- $R$  is the set of possible **discrete** outcomes
- $X_i$  is the set of available moves  $X_i$  at round  $i \in \mathbb{N}$
- $\phi_i: K_R X_i$  are goal quantifiers for round  $i \in \mathbb{N}$
- $q: \prod_{i=0}^{\infty} X_i \rightarrow R$  is a **continuous** outcome function



## Optimal Strategies

### Definition (Strategic Play)

Given strategy  $\text{next}_k$  and partial play  $\vec{a} = a_0, \dots, a_{k-1}$ , the **strategic extension** of  $\vec{a}$  is  $\beta^{\vec{a}} = \beta^{\vec{a}}(k), \beta^{\vec{a}}(k+1), \dots$  where

$$\beta^{\vec{a}}(i) = \text{next}_i(\vec{a}, \beta^{\vec{a}}(k), \dots, \beta^{\vec{a}}(i-1)).$$



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### Definition (Optimal Strategy)

Strategy  $\text{next}_k$  is **optimal** if for any partial play  $\vec{a}$

$$q(\vec{a} * \beta^{\vec{a}}) = \phi_k(\lambda x_k. q(\vec{a} * x_k * \beta^{\vec{a}, x_k})).$$



## Finite but Unbounded Games

### Theorem (Main Theorem for Finite but Unbounded Games)

If  $\phi_k$  are attainable with selection functions  $\varepsilon_k$  then

$$\text{next}_k(x_0, \dots, x_{k-1}) \stackrel{X_k}{=} \left( \left( \bigotimes_{i=k}^{\infty} \varepsilon_i \right) (q_{x_0, \dots, x_{k-1}}) \right)_0$$

is an **optimal strategy** for the game  $(R, (X_i)_{i \in \mathbb{N}}, (\phi_i)_{i \in \mathbb{N}}, q)$ .  
Moreover,

$$\alpha = \left( \bigotimes_{i=0}^{\infty} \varepsilon_i \right) (q)$$

is the **strategic play**.



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Countable choice is classically computational up to **DNS**

$$\forall n^{\mathbb{N}} \neg\neg A_n \rightarrow \neg\neg \forall n^{\mathbb{N}} A_n.$$



## Double negation shift

The double negation shift **DNS**

$$\forall n \neg \neg A_n \rightarrow \neg \neg \forall n A_n$$

corresponds to the type

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$$\prod_n J_{\perp} A_n \rightarrow J_{\perp} \prod_n A_n.$$

The type of the **countable product** of selection functions!



## Bar recursion

Not a coincidence!

**Modified bar recursion** is equivalent to

$$\bigotimes_{i=k}^{\infty} \varepsilon_i \stackrel{J\Pi X_i}{=} \varepsilon_k \otimes \left( \bigotimes_{i=k+1}^{\infty} \varepsilon_i \right)$$

**Spector's bar recursion** is equivalent to

$$\left( \bigotimes_{i=k}^{\infty} \varepsilon_i \right) (q) \stackrel{\Pi X_i}{=} \begin{cases} \mathbf{c} & \text{if } k < l(q(\mathbf{c})) \\ (\varepsilon_k \otimes \left( \bigotimes_{i=k+1}^{\infty} \varepsilon_i \right)) (q) & \text{otherwise} \end{cases}$$



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- E.g. Nash equilibrium, backtracking, Bekič's lemma



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- New “**negative translation**” based on  $J$  strong monad
- Functional interpretations (**proof mining**)

Theorems  $\mapsto$  games

Proofs  $\mapsto$  winning strategies



## A Few Open Questions

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GT: Standard Nash equilibrium



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5. Other places where  $\otimes$  appear?





# References



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