

Optimizing Winning Strategies in Regular Infinite Games

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RWTHAACHEN

A Quotation of 50 Years Ago

Alonzo Church

at the “Summer Institute of Symbolic Logic”

Cornell University, 1957:

“Given a requirement which a circuit is to satisfy, we may suppose the requirement expressed in some suitable logistic system which is an extension of restricted recursive arithmetic. The *synthesis problem* is then to find recursion equivalences representing a circuit that satisfies the given requirement (or alternatively, to determine that there is no such circuit).”



Alonzo Church (1903-1995)

APPLICATION OF RECURSIVE ARITHMETIC TO THE PROBLEM OF CIRCUIT SYNTHESIS

Alonzo Church

RESTRICTED RECURSIVE ARITHMETIC

Primitive symbols are individual (i.e., numerical) variables x, y, z, t, \dots , singular functional constants i_1, i_2, \dots, i_μ , the individual constant 0, the accent ' as a notation for successor (of a number), the notation () for application of a singular function to its argument, connectives of the propositional calculus, and brackets [].

Axioms are all tautologous wffs. Rules are modus ponens; substitution for individual variables; mathematical induction,

from $P \supset S_a^a P$ and $S_0^a P$ to infer P ;

and any one of several alternative recursion schemata or sets of recursion schemata.

$$\chi_1(x_1 + M + 1, 0, \dots, 0, 0) \equiv \text{falsehood}$$

.....

$$\chi_N(x_1 + M + 1, M, \dots, M, g) \equiv \text{falsehood}$$

$$\chi_1(x_1 + M + 1, x_2 + M + 1, \dots, 0, 0) \equiv \text{falsehood}$$

.....

$$\chi_1(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, 0) \equiv \text{falsehood}$$

$$\chi_2(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, 0) \equiv \text{falsehood}$$

.....

$$\chi_N(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, 0) \equiv \text{falsehood}$$

$$\chi_1(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, 1) \equiv \text{falsehood}$$

.....

$$\chi_N(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, g) \equiv \text{falsehood}$$

$$\chi_1(0, 0, \dots, 0, t + g + 1) \equiv \chi_1(0, 0, \dots, 0, t + g) \vee$$

$$Q_{100\dots 0}[\chi_1(0, 0, \dots, 0, t), \dots, \chi_N(0, 0, \dots, 0, t), \chi_1(0, 0, \dots, 1, t), \dots, \chi_N(M+1, M+1, \dots, M+1, t)]$$

$$\chi_2(0, 0, \dots, 0, t + g + 1) \equiv \chi_2(0, 0, \dots, 0, t + g) \vee$$

$$\bar{\chi}_1(0, 0, \dots, 0, t + g) Q_{200\dots 0}[\chi_1(0, 0, \dots, 0, t), \dots,$$

$$\chi_N(0, 0, \dots, 0, t), \chi_1(0, 0, \dots, 1, t), \dots, \dots,$$

$$\chi_N(M + 1, M + 1, \dots, M + 1, t)]$$

.....

$$\chi_N(M, M, \dots, M, t + g + 1) \equiv \chi_N(M, M, \dots, M, t + g) \vee$$

$$\bar{\chi}_1(M, M, \dots, M, t + g) \bar{\chi}_2(M, M, \dots, M, t + g) \dots$$

$$\bar{\chi}_{N-1}(M, M, \dots, M, t + g) Q_{NM\dots M}[\chi_1(0, 0, \dots,$$

$$0, t), \dots, \chi_N(0, 0, \dots, 0, t), \chi_1(0, 0, \dots, 1, t),$$

$$\dots, \dots, \chi_N(2M + 1, 2M + 1, \dots, 2M + 1, t)]$$

$$\chi_1(x_1 + M + 1, 0, \dots, 0, t + g + 1) \equiv \chi_1(x_1 + M + 1, 0,$$

$$\dots, 0, t + g) \vee Q_{10\dots 0}[\chi_1(x_1, 0, \dots, 0, t), \dots, \chi_N(x_1, 0, \dots, 0, t), \chi_1(x_1, 0, \dots, 1, t), \dots, \dots, \chi_N(x_1 + 2M + 2, M + 1, \dots, M + 1, t)]$$

$$\chi_2(x_1 + M + 1, 0, \dots, 0, t + g + 1) \equiv \chi_2(x_1 + M + 1, 0, \dots, 0,$$

$$t + g) \vee \bar{\chi}_1(x_1 + M + 1, 0, \dots, 0, t + g) Q_{20\dots 0}[\chi_1(x_1, 0, \dots, 0, t), \dots, \chi_N(x_1, 0, \dots, 0, t), \chi_1(x_1, 0, \dots, 1, t), \dots, \dots, \chi_N(x_1 + 2M + 2, M + 1, \dots, M + 1, t)]$$

.....

$$\chi_1(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g + 1) \equiv$$

$$\chi_1(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g) \vee$$

$$Q_1[\chi_1(x_1, x_2, \dots, x_m, t), \dots, \chi_N(x_1, x_2,$$

$$\dots, x_m, t), \chi_1(x_1, x_2, \dots, x_m + 1, t), \dots,$$

$$\dots, \chi_N(x_1 + 2M + 2, x_2 + 2M + 2, \dots,$$

$$x_m + 2M + 2, t)]$$

$$\chi_2(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g + 1) \equiv$$

$$\chi_2(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g) \vee$$

$$\bar{\chi}_1(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g) Q_2[\chi_1(x_1,$$

$$x_2, \dots, x_m, t), \dots, \chi_N(x_1, x_2, \dots, x_m, t),$$

$$\chi_1(x_1, x_2, \dots, x_m + 1, t), \dots, \dots,$$

$$\chi_N(x_1 + 2M + 2, x_2 + 2M + 2, \dots, x_m + 2M + 2, t)]$$

$$\dots$$

$$\chi_N(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g + 1) \equiv$$

$$\chi_N(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g) \vee$$

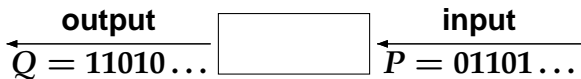
$$\bar{\chi}_1(x_1 + M + 1, x_2 + M + 1, \dots, x_m + M + 1, t + g) \bar{\chi}_2(x_1 + M + 1,$$

$$x_2 + M + 1, \dots, x_m + M + 1, t + g) \dots \bar{\chi}_{N-1}(x_1 + M + 1, x_2 + M + 1,$$

$$\dots, x_m + M + 1, t + g) Q_N[\chi_1(x_1, x_2, \dots, x_m, t), \dots,$$

$$\chi_N(x_1, x_2, \dots, x_m, t), \chi_1(x_1, x_2, \dots, x_m + 1, t), \dots,$$

Game-Theoretic View



For $t = 0, 1, 2, \dots$: Input player (1) supplies bit $P(t)$,
output player (2) responds by bit $Q(t)$

Bitstreams correspond to subsets of \mathbb{N} .

Use variables X, Y for subsets of \mathbb{N} .

Requirement $\varphi(X, Y)$ is considered as winning condition in
an infinite two-person game.

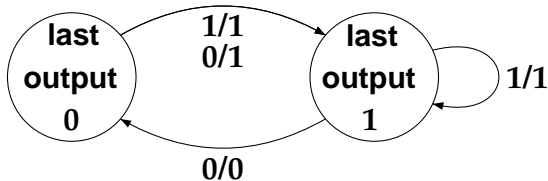
Play $\begin{pmatrix} P(0) \\ Q(0) \end{pmatrix} \begin{pmatrix} P(1) \\ Q(1) \end{pmatrix} \begin{pmatrix} P(2) \\ Q(2) \end{pmatrix} \dots$ is won by 2 if $(\mathbb{N}, \dots) \models \varphi(P, Q)$

Example

$\varphi(X, Y)$:

- $\forall t (X(t) \rightarrow Y(t))$
- $\neg \exists t (\neg Y(t) \wedge \neg Y(t'))$
- $(\exists^\omega t \neg X(t) \rightarrow \exists^\omega t \neg Y(t))$

Solution:



This is a finite-state strategy (realized by a Mealy automaton).

Plan

1. The origin: Church's Problem (done)
2. Muller games
3. Solving Muller games
4. Memory-optimal controllers
5. Optimal solutions for liveness requirements
6. Outlook

Muller Games

Approach for Solution of Church's Problem

1. Translation of formula φ into Muller automaton
2. Conversion of Muller automaton into a Muller game graph
3. Transformation of Muller game into parity game
4. Solution of parity game

Steps 1 and 2 go from logic to automata (and games).

Steps 3 and 4 show how to solve “regular infinite games”.

Muller Automata

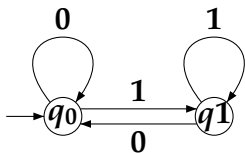
are finite automata $\mathcal{A} = (S, \Sigma, s_0, \delta, \mathcal{F})$ accepting ω -sequences.

Acceptance component: Family $\mathcal{F} = \{F_1, \dots, F_k\}$ of state-sets.

\mathcal{A} accepts $\alpha \Leftrightarrow$ the states occurring infinitely often in the run ρ of \mathcal{A} on α form some set F_i

short: $\text{Inf}(\rho) \in \mathcal{F}$

Example



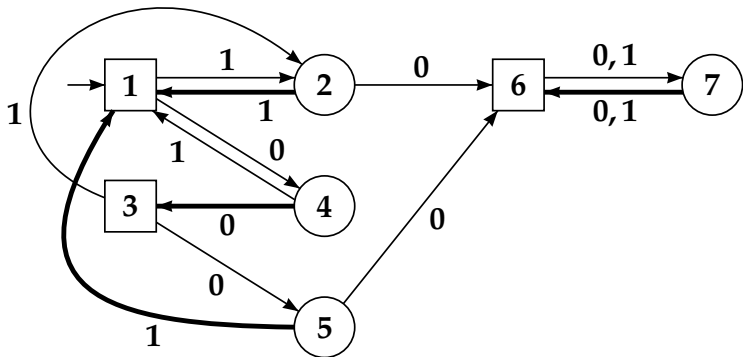
- with $\mathcal{F} = \{\{q_1\}\}$ accepts $(0 + 1)^*1^\omega$
- with $\mathcal{F} = \{\{q_1\}, \{q_0, q_1\}\}$ accepts $(0^*1)^\omega$

We dissolve a transition with $\binom{0}{1}$ into two transitions, marking that Player 1 picks 0 and Player 2 picks 1.

We obtain a “game graph”.

Initial Example

$$\varphi(X, Y): \forall t (X(t) \rightarrow Y(t)) \wedge \neg \exists t (\neg Y(t) \wedge \neg Y(t')) \\ \wedge (\exists^{\omega} t \neg X(t) \rightarrow \exists^{\omega} t \neg Y(t))$$



where $\mathcal{F} = \{\{1, 2, 3, 4\}, \{1, 2, 3, 4, 5\}, \{1, 3, 4, 5\}\}$

Game Graphs

A **game graph** has the form $G = (Q, Q_1, E)$ where $Q_1 \subseteq Q$ and $E \subseteq Q \times Q$ is the transition relation satisfying

$$\forall q \in Q : qE \neq \emptyset \quad (\text{i.e. } \forall q \exists q' : (q, q') \in E)$$

We set $Q_2 := Q \setminus Q_1$

A play is a sequence $\rho = r_0 r_1 r_2 \dots$ with $(r_i, r_{i+1}) \in E$

Intuitively, a token is moved from vertex to vertex via edges,
Player 1 / 2 deciding on the vertices of Q_1 / Q_2

Winning Conditions (Requirements)

in this talk:

- Logical winning condition (e.g. written in MSO)
- Muller condition:
for play ρ : $\text{Inf}(\rho) \in \mathcal{F}$
- Weak Muller condition
for play ρ : $\text{Occ}(\rho) \in \mathcal{F}$

Comparison with Church's Problem

1. Church's Problem uses a trivial graph (over $Q_1 = \{0, 1\}$ and $Q_2 = \{0', 1'\}$) and an MSO winning condition.
2. Model of reactive system: finite game graph and logical winning condition
3. Muller game: Finite game graph and Muller winning condition

Cases 1 and 2 reduce to case 3:

φ is equivalent to Muller automaton $\mathcal{A}_\varphi = (S, Q, s_0, \delta, \mathcal{F})$

Now take game graph over $Q \times S$ with Muller condition referring to second component.

Strategies

A **strategy** for player 2 from q is a function $f : Q^+ \rightarrow Q$, specifying for any play prefix $q_0 \dots q_k$ with $q_0 = q$ and $q_k \in Q_2$ some vertex $r \in Q$ with $(q_k, r) \in E$

A strategy f for player 0 from q is called **winning strategy for player 0 from q** if any play from q which is played according to f is won by player 0 (according to the winning condition).

In the analogous way, one introduces strategies and winning strategies for player 1.

We say: Player 2 wins from q if s/he has a winning strategy from q

Winning Regions

For a game $\Gamma = (G, \varphi)$ with $G = (Q, Q_1, E)$, the **winning regions of players 1 and 2** are the sets

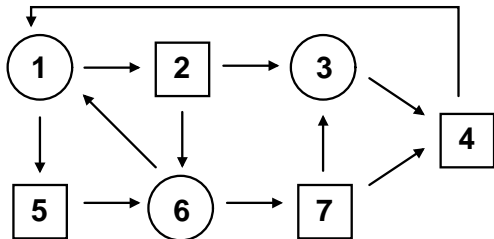
$$W_1 := \{q \in Q \mid \text{player 1 wins from } q\}$$

$$W_2 := \{q \in Q \mid \text{player 2 wins from } q\}$$

Remark: Each vertex q belongs at most to W_1 or W_2 .

An Example

Example:



Winning condition for player 2: Vertex 3 should be reached.

Weak Muller game: Use $\mathcal{F} = \{F \mid 3 \in F\}$

$W_1 = \{1, 2, 4, 5, 6, 7\}$ $W_2 = \{3\}$

Determinacy

In general, the winning regions W_0, W_1 of players 1 and 2 satisfy $W_1 \cap W_2 = \emptyset$

A game is called **determined** if from each vertex either of the two players has a winning strategy.

Remark:

1. There are (exotic) games which are not determined.
2. In descriptive set theory one investigates which abstract winning conditions define determined games.
3. All games in this talk determined.
(They are “Borel games”.)

Church's Problem Reformulated

Given a game $\Gamma = (G, \varphi)$, $G = (Q, Q_1, E)$

1. Decide for each $q \in Q$ whether $q \in W_2$ (i.e. whether player 2 wins from q)
2. In this case:
Construct a suitable winning strategy from q (in the form of an automaton, or program)
3. Optimize the construction of the winning strategy (e.g., time complexity) or optimize parameters of the winning strategy (e.g., size of memory).

Solving a game means to provide algorithms for 1. and 2.

Special Strategies

If Q is finite, then a strategy is a word function $f : Q^+ \rightarrow Q$

There are three basic types of strategies:

1. computable (recursive),
2. finite-state (computable by a Mealy automaton)
3. positional (memoryless, value given by current vertex alone)

Other types: pushdown strategy, counter strategy etc.

Büchi-Landweber Theorem

Finite Muller games are determined, one can compute the winning regions of the two players, and one can compute respective finite-state winning strategies.

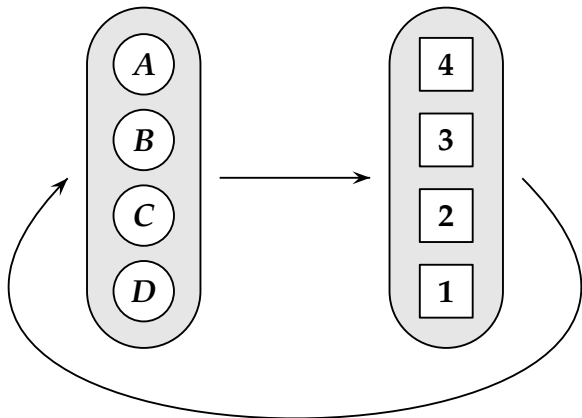
Construction of winning strategies is controller synthesis.

Finite-state controller synthesis is possible in automated manner for MSO- (or LTL-) specifications.

Solving Muller Games

An Interesting Muller Game (DJW-Game)

due to Dziembowski, Jurdziński, Walukiewicz (1997)



Number of letters chosen infinitely often should coincide with the highest number chosen infinitely often.

Latest Appearance Record

Visited letter	LAR
<i>A</i>	<i>ABCD</i>
<i>C</i>	<i>CAB<u>D</u></i>
<i>C</i>	<i><u>C</u>ABD</i>
<i>D</i>	<i>DCAB<u>B</u></i>
<i>B</i>	<i>BDCA<u>A</u></i>
<i>D</i>	<i>D<u>B</u>CA</i>
<i>C</i>	<i>CD<u>B</u>A</i>
<i>D</i>	<i>DC<u>B</u>A</i>
<i>D</i>	<i><u>D</u>CB<u>A</u></i>

Underlined position: “hit”

Example Scenario

Assume the states C and D are repeated infinitely often.

Then:

- the states A and B eventually arrive at the last two positions and are not touched any more; so finally underlinings appear at most on positions 1 and 2
- position 2 is underlined again and again; if only position 1 is underlined from some point onwards, only the same letter would be chosen from there onwards (and not two states C and D as assumed)

Solution of the DJW-Game

LAR-strategy for player 0:

During play, update and use the LAR as follows:

- shift the current letter vertex to the front
underline the position from where the current letter was taken
- move to the number vertex given by underlined position

These are the two items performed by the strategy:

- update of memory
- choice of next step (“output”)

Result: “Finite-state winning strategy” with $n! \cdot n$ states for a game graph with $2n$ vertices

Proof Strategy

Given a Muller game over G , the transition structure of the strategy automata can be constructed from $G = (Q, Q_1, E)$ alone:

- Memory space: $\text{LAR}(Q)$ (LAR's over Q)
- Memory-update during play $\rho \in Q^\omega$ according to LAR-update rule
- Missing item: Output function

Core of Proof

For $\rho \in Q^\omega$ consider induced $\rho' \in \text{LAR}(Q)$

$h :=$ maximal hit occurring infinitely often in ρ'

$R :=$ (eventually fixed) set up to this hit position h

Then: $\text{Inf}(\rho) = R$

Reformulate winning condition using

$c : \text{LAR}(Q) \rightarrow \{1, \dots, 2 \cdot |Q|\}$

$c(\{q_{i_1}, \dots, \underline{q_{i_h}}, \dots, q_{i_n}\}) = 2h$ if $\{q_{i_1}, \dots, q_{i_h}\} \in \mathcal{F}$, else $2h - 1$

Then: $\text{Inf}(\rho) \in \mathcal{F}$ iff $\max(\text{Inf}(c(\rho'))) is even$

This is the “parity condition”

On Parity Games

Emerson-Jutla and Mostowski (1991):

Parity games are determined (even over infinite game graphs), and on the winning region W_i Player i has a positional (!) winning strategy.

Proof by induction over the number of colors

Core of construction of winning strategy: Reachability analysis

Weak Muller Games

Winning condition: $\text{Occ}(\rho) \in \mathcal{F}$

A strategy automaton needs only to remember which states have been visited.

Use “Appearance record” AR rather than LAR.

Introduce weak parity games, with winning condition

“the highest color of a visited vertex is even”

Memory states of strategy automata are sets of vertices rather than lists of vertices.

Looking Back

1. Translation of formula φ into Muller automaton
2. Conversion of Muller automaton into a Muller game graph
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Current Developments

- **Generalizations of the game model:**
Infinite-state, concurrent, stochastic, timed, weighted, distributed, multi-player games
- **Closer analysis (this talk)**
 1. Memory-optimal controllers
 2. Optimal solutions for liveness requirements
- **Other issues:**
Definability of controllers
Generalizing winning strategies

Memory-optimal Controllers

Memory Reduction

Fact: For a Muller game with n states one can construct winning strategies with $n! * n$ states, and $n!$ is also a lower bound.

But: There are two sources of memory:

- construction of Muller game arena
- construction of finite-state controller

Problem 1: How are these two steps related?

Problem 2: Understand the space of strategies

Three Approaches to Memory Reduction

- Reduce memory for given strategy f
Use standard procedure as in DFA minimization
- View the game graph as an automaton and reduce it first
(Holtmann, Löding (Aachen))
- Search the space of all (winning) strategies to find one
with minimal-memory implementation
(open problem, hint by Büchi-Landweber)

Holtmann-Löding Method

General plan:

- Given a (weak) Muller game over Q ,
- transform it into a (weak) parity game over $S \times Q$,
- Forgetting about the partition (Q_1, Q_2) we obtain an automaton with state-set S and input alphabet Q that accepts (with the (weak) parity condition) precisely the winning plays for Player 2.
- Main step: Mimimize / Reduce the size of this automaton in a way that a (weak) parity game over some $S_0 \times Q$ can be extracted.
- Use S_0 as memory space for winning strategy.

Main Technical Points

- Define $(s, q) \sim (s', q)$ iff from s with initial vertex q and from s' with initial vertex q the same plays are accepted.
- Define $s \equiv s'$ iff for all q we have $(s, q) \sim (s', q)$
- Then \equiv -classes can serve as new states.
- Use tests for $(s, q) \sim (s', q)$ (from ω -automata theory)

Result: There are games with $c \cdot n$ vertices where the game graph reduction yields an exponential gain over the standard strategy minimization.

On the other hand, the approach misses some potentials of minimization and is not a complete method.

Optimal Solutions for Liveness Requirements

Optimality in Request-Response Games

Game arena $G = (V, V_0, E)$

Subsets $Rqu_1, \dots, Rqu_k \subseteq V$: “Requests”

Subsets $Rsp_1, \dots, Rsp_k \subseteq V$: “Responses”

RR-condition:

$$\bigwedge_{i=1}^k \forall s (Rqu_i(s) \rightarrow \exists t (s < t \wedge Rsp_i(t)))$$

LTL:

$$\bigwedge_{i=1}^k \mathbf{G}(Rqu_i \rightarrow \mathbf{XF} Rsp_i)$$

Standard Solution of RR-Games

- It suffices to keep a memory for the set of "open requests"
Memory size: 2^k for k conditions
- Reduction to Büchi games
- Result: Winning strategy which ensures bounded waiting time between request and response
(Bound $B := k \cdot |V|$).

Problem: Use finer measure than maximum of waiting times

Measuring Quality of Solution

Penalty function associates to i -th moment of waiting a penalty

- **Linear Penalty model:**

For each moment of waiting (for each RR-condition)
pay 1 unit

- **Quadratic Penalty model:**

For the i -th moment of waiting pay i units

- **More general, use strictly growing unbounded penalty function**

Activation of i -th condition in a play ϱ is a visit to Rqu_i such that all previous visits to Rqu_i are already matched by an Rsp_i -visit.

Values of Plays and Strategies

For a given penalty function define:

- $w_\varrho(n) =$ sum of penalties in $\varrho(0) \dots \varrho(n)$ divided by number of activations

”average penalty sum per activation”

- $w(\varrho) = \limsup_{n \rightarrow \infty} w_\varrho(n)$

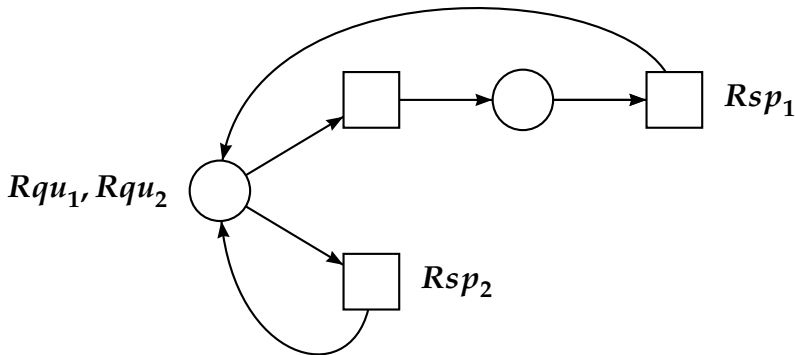
Given a strategy σ for controller and a strategy τ for adversary

- $\varrho(\sigma, \tau) :=$ the play induced by σ and τ
- $w(\sigma) := \sup_{\tau} w(\varrho(\sigma, \tau))$

Call σ optimal if there is no other strategy with smaller value.

On the Linear Penalty

For the linear penalty model, a finite-state optimal strategy does not exist in general:



Theorem

(with F. Horn and N. Wallmeier)

For any strictly increasing unbounded penalty function one can decide whether a RR-game is won by controller and in this case one can compute a finite-state optimal winning strategy.

Proof ingredients:

- **It suffices to consider strategies with value $\leq M$ (induced by bounded waiting time of standard solution).**
- **Conversely: For strategies with value $\leq M$ one can assume bounded waiting time.**
- **Reduction to mean-payoff games (Zwick-Paterson)**

Building a Mean-Payoff Game

From a game graph $G = (V, E)$ with k conditions
proceed to a game graph over $V \times \mathbb{N}^k$

State format: (v, n_1, \dots, n_k)

$n_i =$

current waiting time for i -th condition since last activation

Derived mean-payoff game:

For each edge $e = (u, \bar{m}) \rightarrow (v, \bar{n})$

introduce edge weight

$w(e) = n_1 + \dots + n_k$ (sum of current penalties)

Boundedness Lemma

Let σ be a winning strategy of value $\leq M$

Then one can construct a winning strategy σ' with bounded waiting times such that $w(\sigma') \leq w(\sigma)$.

Consequence:

In the mean-payoff game, it suffices to consider waiting time vectors in a domain $[0, B]^k$ rather than \mathbb{N}^k .

So we obtain a finite MPG which can be solved.

Intuition for Boundedness Lemma

Example scenarium:

Consider a winning strategy σ of value $\leq M$ which allows unbounded waiting times just for the last RR-condition.

States: (v, \bar{m}, m) with $v \in V, \bar{m} \in [0, B]^{k-1}, m \geq 0$

In a play with unbounded waiting times for the last condition, pick a “critical segment” $(v, \bar{m}, m), \dots, (v, \bar{m}, m')$ where each position has a penalty $\geq M$.

In σ' , this play segment is skipped. This decreases

- **the waiting time for the last component**
- **the value of the strategy**
(each deleted step has \geq average penalty)

Outlook

Broader View on Transformations

A strategy defines an operator $T : \{0,1\}^\omega \rightarrow \{0,1\}^\omega$

T is continuous if $T(P)(t)$ depends only on a finite segment of P .

The MSO-condition $\varphi(X, Y)$:

$$(\exists t X(t) \leftrightarrow \forall t Y(t)) \wedge (\forall t Y(t) \vee \forall t \neg Y(t))$$

is solvable only by the non-continuous operator T_0 with

$$T_0(\emptyset) = 0^\omega \text{ and } T_0(P) = 1^\omega \text{ for } P \neq \emptyset$$

On Continuity

Landweber, Hosch (1971): It is decidable whether a MSO winning condition can be solved by a finite-state strategy with bounded delay.

Example 1: Division of a sequence by two

$$T^-: P(0)P(1) \dots \mapsto P(0)P(2)P(4) \dots$$

T^- is continuous, with linearly increasing delay.

Example 2: Doubling a sequence

$$T^+: P(0)P(1) \dots \mapsto P(0)P(0)P(1)P(1) \dots$$

T^+ is bit-by-bit-computable with unbounded memory, or with a sequential machine.

Conclusion

- **Church's Problem is far from closed.**
- **A current challenge is to shift the investigation from decision problems to various optimization problems.**
- **Another challenge is to investigate the synthesis of generalized automata (e.g., sequential machines)**

General Sources

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