

Towards a logical foundation of computational complexity

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Background

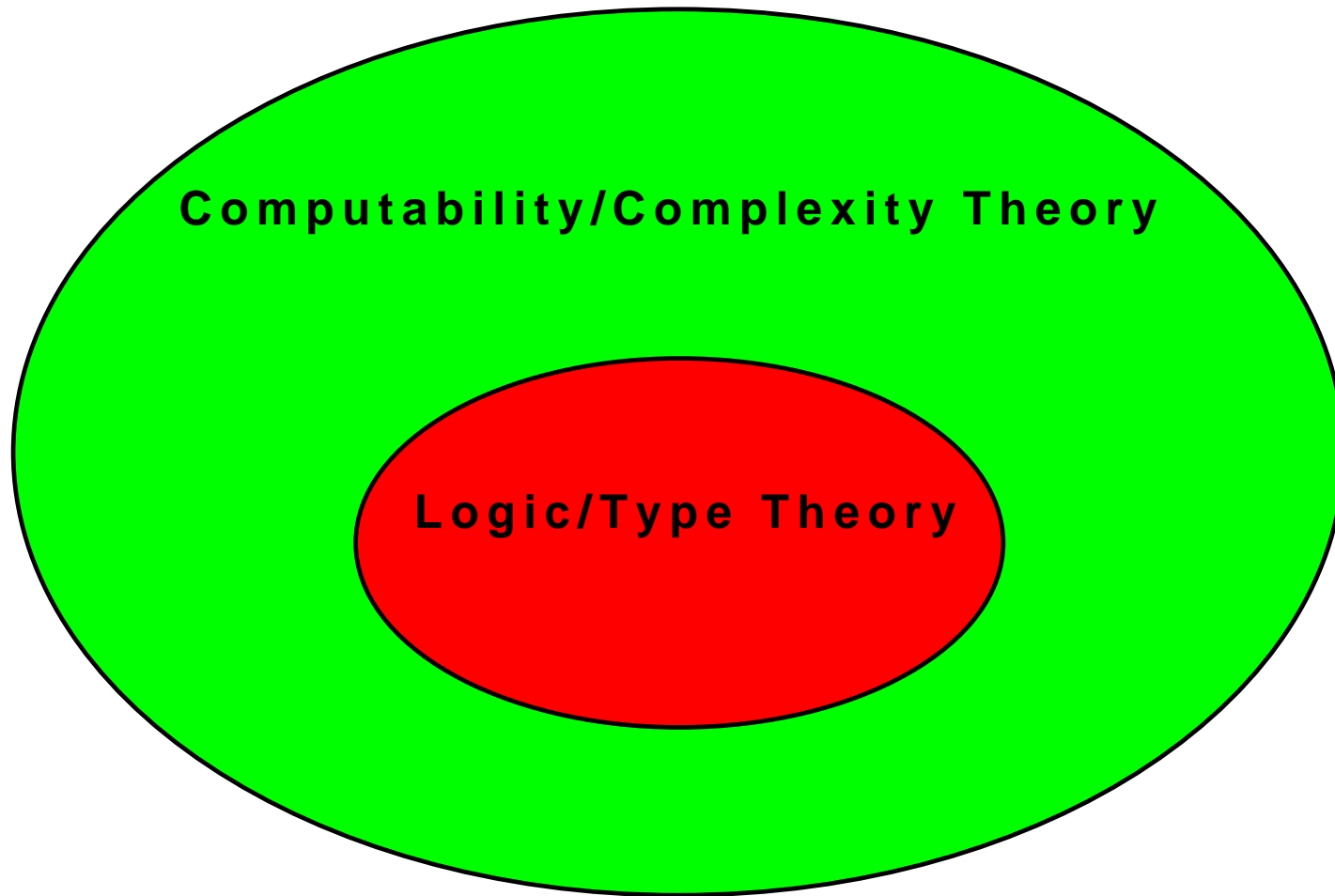
- Logic and type theory. **Proofs = Programs.**
- Complexity issues arise in
 - type checking/inference,
 - verification,
 - **normalization**, etc.
- **Implicit computational complexity:**
Machine-independent, parameter-free characterizations of complexity classes (such as P)

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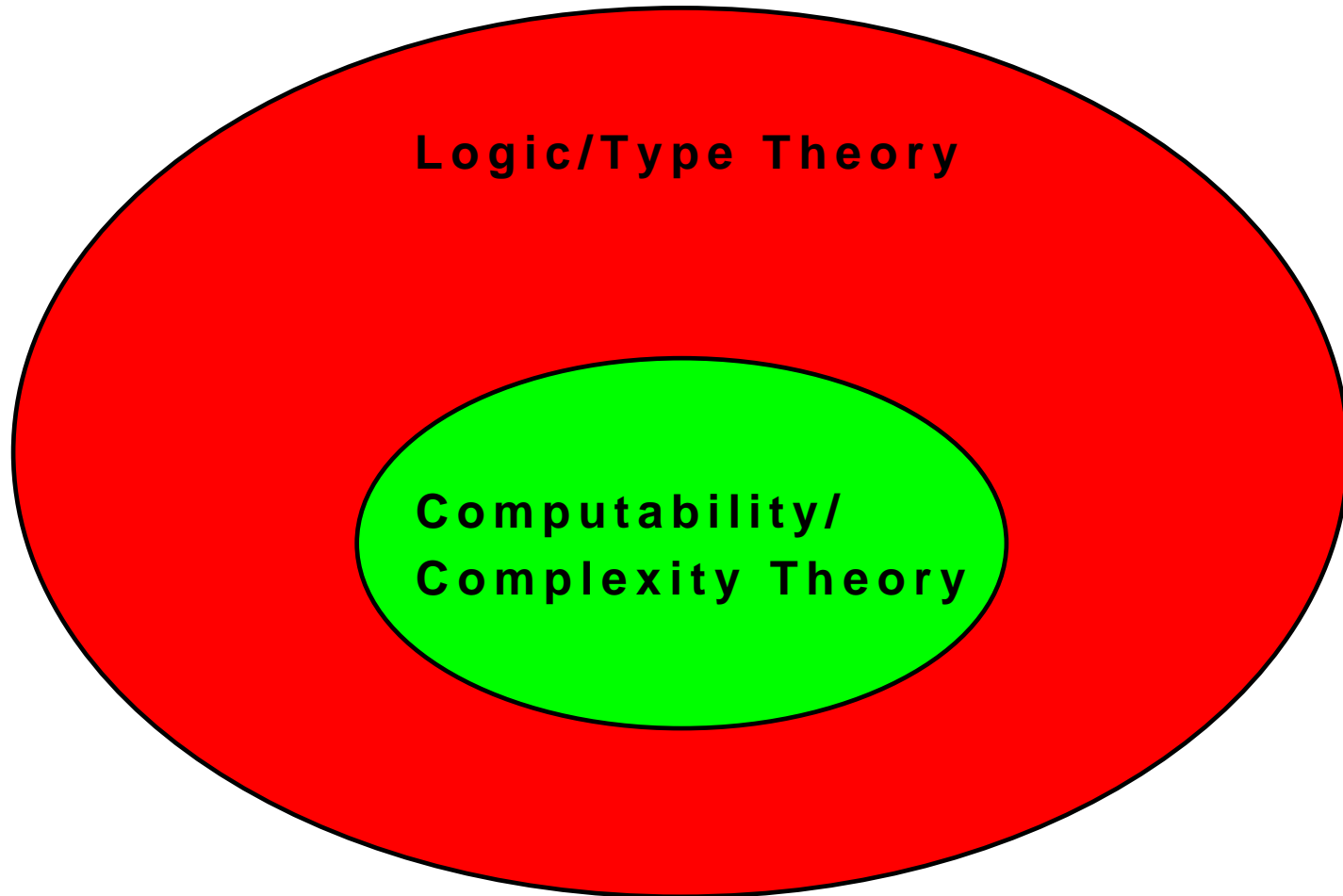
- Gödel's T + restrictions:
 - Safety (Bellantoni-Cook, Leivant, Marion, ...)
 - Linearity at higher order (Bellantoni-Niggl-Schwichtenberg, Hofmann, ...)
 - Cons-free (Jones, Kristiansen, ...)
- Girard's F + restrictions:
 - Light linear logic (Girard, Asperti, Baillot-T., ...)
 - Soft linear logic (Lafont, Gaboardi-Ronchi, Hofmann-Schöpp, ...)
- Complexity of simply typed lambda calculus (Schubert)
- Complexity of fragments of linear logic (Mairson-T.)
- Parallel complexity of proof nets (T.)

Current Status

When talking about complexity of proofs and programs ...



Our Ambition



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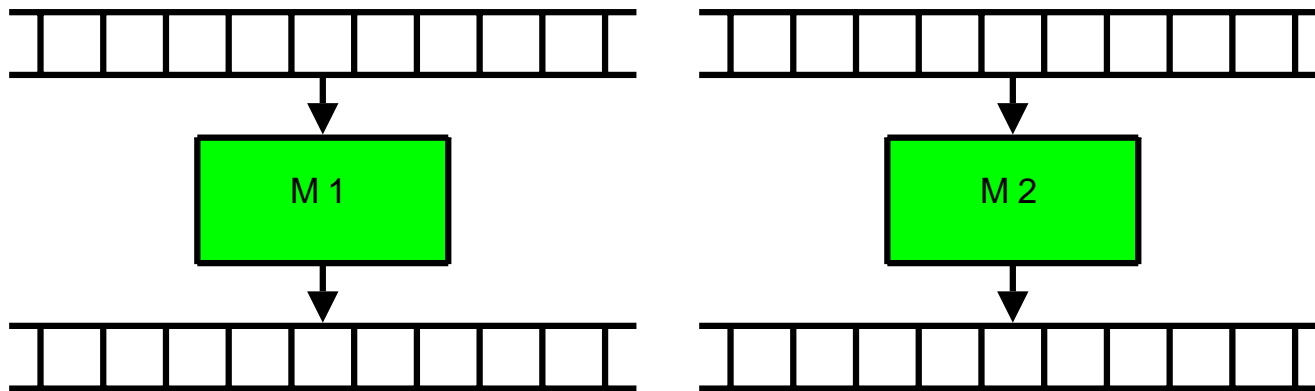
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- Which logic? — [Ludics](#) (Girard 2001).

Outline

1. Time and space sensitive compositions in lambda calculus
2. What is ludics?
3. Data and computation in ludics
4. Arbitrary data sets
5. Language operators and internal completeness
6. Space compression and focalization
7. Conclusion

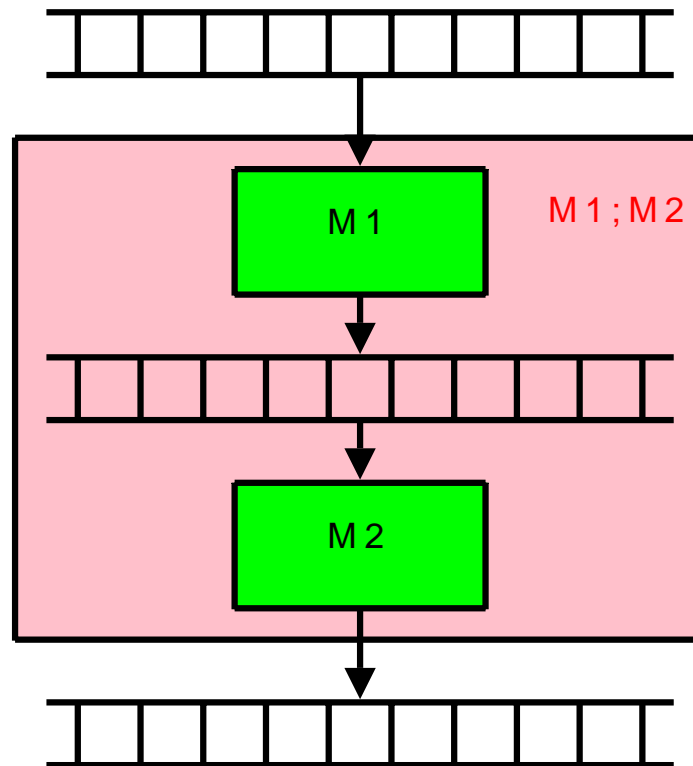
Composition of TMs

How do you compose two TMs?



Composition of TMs

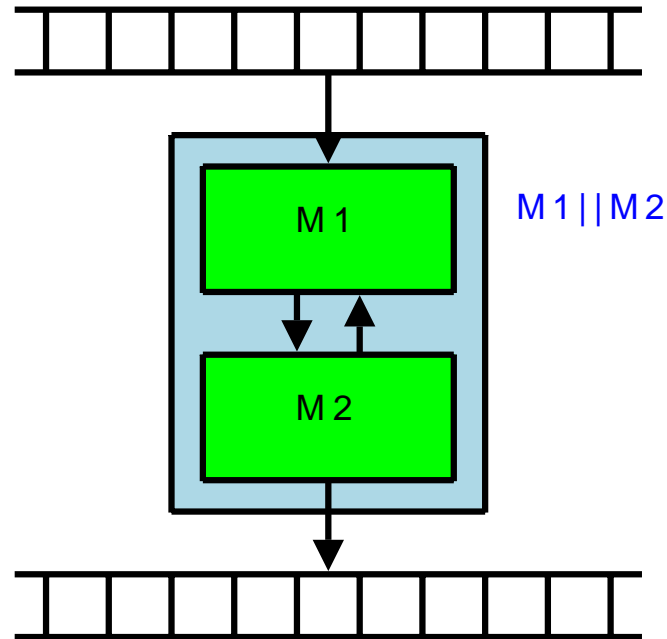
- Sequential composition $M_1; M_2$: first simulate M_1 , then M_2



- Time efficient, but not space efficient.

Composition of TMs

- Interactive composition $M_1 || M_2$: simulate a dialogue between M_1 and M_2



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Evaluations of lambda terms

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 - Composition is time efficient, but not space efficient.
- Is there a space efficient evaluation method?

Krivine's Abstract Machine

- A **pointer** machine working on (graphs of) untyped λ -terms
- Equipped with **environments** ρ (for variables) and **stacks** π (of arguments)

$$(x\rho, \pi) \longrightarrow (\rho(x), \pi) \quad \text{if } x \in \text{Dom}(\rho)$$

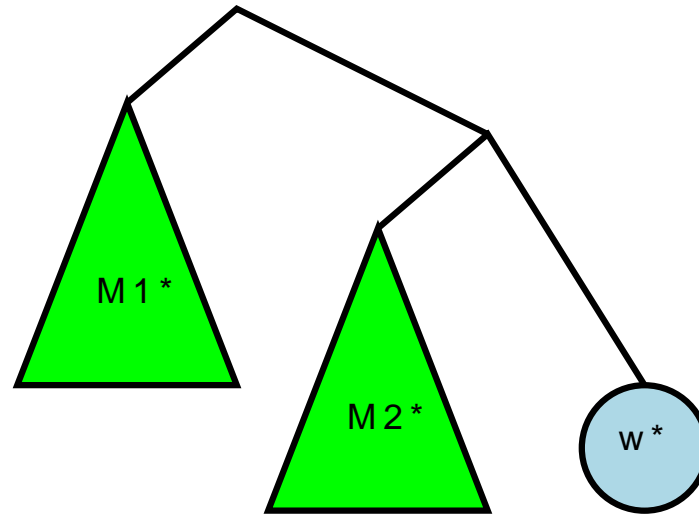
$$((tu)\rho, \pi) \longrightarrow (t\rho, u\rho : \pi)$$

$$((\lambda x.t)\rho, u\rho' : \pi) \longrightarrow (t\rho[x \mapsto u\rho'], \pi)$$

- **Fact:** There is no evaluator that is **significantly** and **uniformly** more space-efficient than (optimized) KAM.

Time-space tradeoff in λ -calculus

- Compose encodings M_1^* , M_2^* of TMs M_1 and M_2 :



- CBV simulates sequential composition $M_1; M_2$
- KAM simulates interactive composition $M_1 || M_2$
- Time-space tradeoff shows up in a different way in lambda calculus.

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What is ludics?

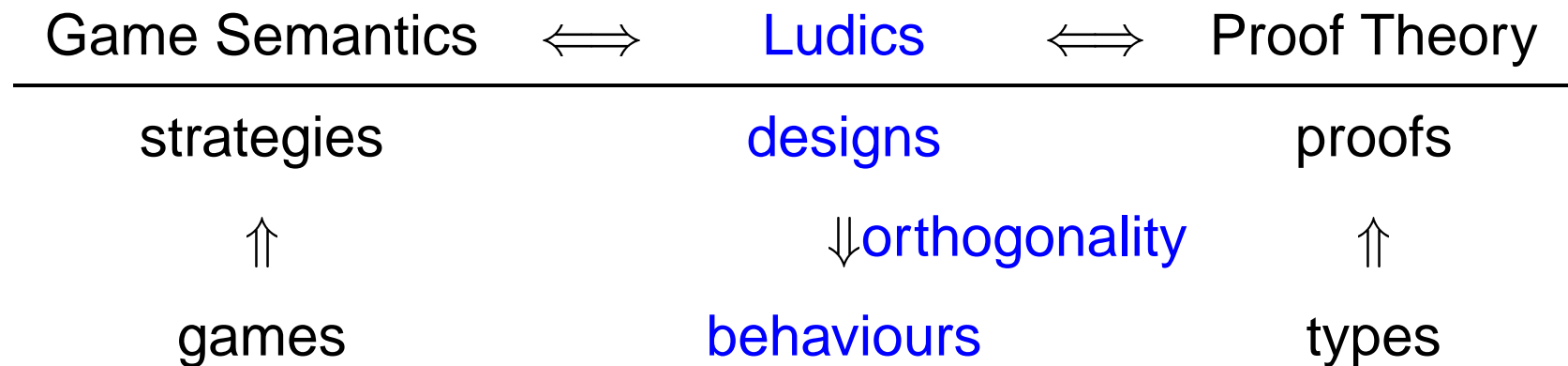
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- **Keywords:** Monism, existentialism, interaction/orthogonality:

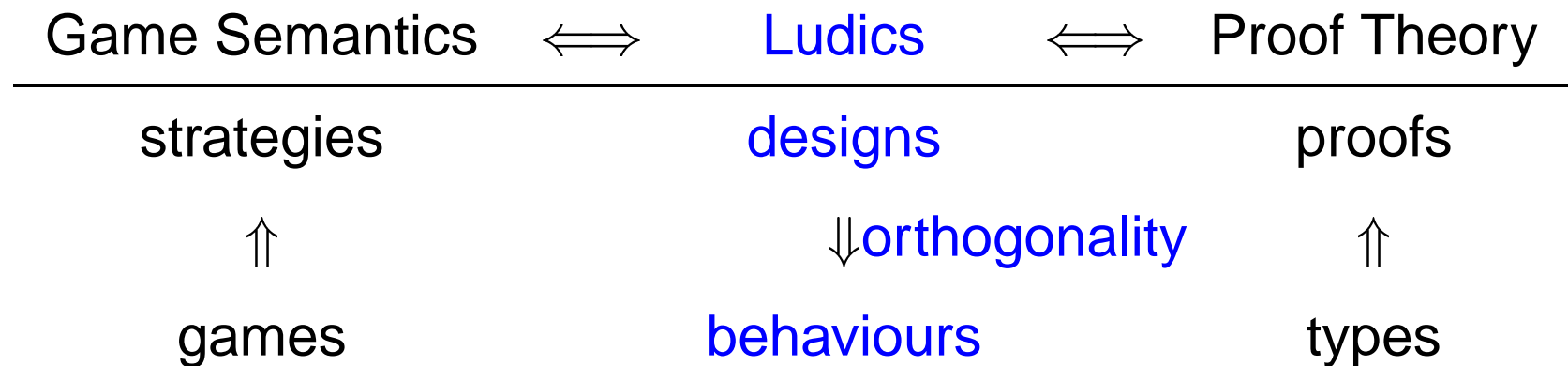
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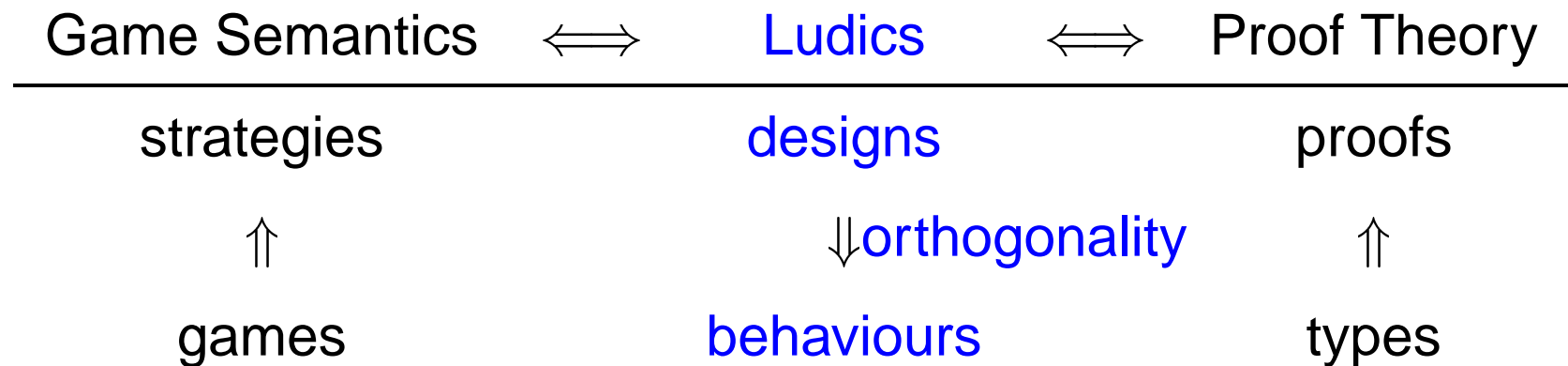
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- **Orthogonality $P \perp N$:** “Players P and N well socialize”
- **Construction of behaviours:** $\{P\}^\perp$, $\mathbf{P}^\perp = \mathbf{N}$, $\mathbf{N}^\perp = \mathbf{P}$
 “Pair (\mathbf{P}, \mathbf{N}) of two player sets **form** a game.”

From C & C to Ludics

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- There is no ontological distinction between M^\bullet and w^\bullet .
 \perp is homogeneous and symmetric.

Computational Ludics

- We introduce a modified version: **computational ludics**.
 - Absolute addresses \implies **Term calculus** with variable binding
 - No care of finiteness \implies Sensitive to **finite generation**
 - Cut-free designs \implies **Cut-ful** ones

Well-behaved frag. of simply typed λ -calculus

- Types: $\tau ::= \iota \mid \tau \rightarrow \tau$
- Positive terms P and negative terms N are defined by:

$$P^\iota ::= (N_0^{\tau_1 \rightarrow \dots \tau_n \rightarrow \iota}) N_1^{\tau_1} \dots N_n^{\tau_n}$$
$$N^{\tau_1 \rightarrow \dots \tau_n \rightarrow \iota} ::= x \mid \lambda x_1^{\tau_1} \dots x_n^{\tau_n}. P^\iota$$

- Reduction: the **arity** n always agrees.

$$(\lambda x_1 \dots x_n. P) N_1 \dots N_n \longrightarrow P[N_1/x_1, \dots, N_n/x_n]$$

Towards ludics

- Designs in ludics:
 - Type-free; arity agreement is ensured in another way.
 - Infinitary (coinductive).
 - Daimon (immediate termination)
 - Additive superimposition: $N_1 + N_2 + N_3 + \dots$
 - Various actions (rather than the single pair $\lambda/@$) given by a signature.
- Signature: $\mathcal{A} = (A, ar)$
 - A is a set of **names**,
 - $ar : A \longrightarrow \mathcal{N}$ gives an **arity** to each name.

Computational designs

- The set of **designs** is coinductively defined by:

P	::=	\boxtimes	Daimon
		Ω	Divergence
		$N_0 \bar{a} \langle N_1, \dots, N_n \rangle$	Proper positive action
N	::=	x	Variable
		$\sum a(\vec{x}_a).P_a$	Proper negative action

- where $ar(a) = n$, $\vec{x}_a = x_1, \dots, x_n$
- $\sum a(\vec{x}_a).P_a$ is built from $\{a(\vec{x}_a).P_a\}_{a \in A}$. Compare it with:

$$P ::= (N_0)N_1 \dots N_n$$

$$N ::= x \mid \lambda x_1 \dots x_n. P$$

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- Reduction rule:

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- Compare it with

$$(\lambda x_1 \dots x_n.P) N_1 \dots N_n \longrightarrow P[N_1/x_1, \dots, N_n/x_n]$$

Orthogonality

- A positive design P is one of the following forms:

$x|\bar{a}\langle N_1, \dots, N_n \rangle$ Head normal form

$(\sum a(\vec{x}_a).P_a) |\bar{a}\langle N_1, \dots, N_n \rangle$ Cut

\boxtimes Daimon

Ω Divergence

- **Fact:** For any **closed** positive design P ,

$P \longrightarrow^* \boxtimes$ or diverges.

- **Orthogonality:** Suppose $fv(P) \subseteq \{x_0\}$ and $fv(N) = \emptyset$.

$P \perp N \iff P[N/x_0] \Downarrow \boxtimes.$

Normalization: the general case

- **Head reduction:** $(\sum a(\vec{x}_a).P_a) |\bar{a}\langle \vec{N}_a \rangle \longrightarrow P_a[\vec{N}_a/\vec{x}]$.
- By **corecursion**, it can be extended to $\llbracket \cdot \rrbracket$:

$$\begin{aligned}\llbracket P \rrbracket &= \mathbf{\times} && \text{if } P \Downarrow \mathbf{\times}; \\ &= x|\bar{a}\langle \llbracket N_1 \rrbracket, \dots, \llbracket N_n \rrbracket \rangle && \text{if } P \Downarrow x|\bar{a}\langle N_1, \dots, N_n \rangle; \\ &= \Omega && \text{if } P \Uparrow;\end{aligned}$$

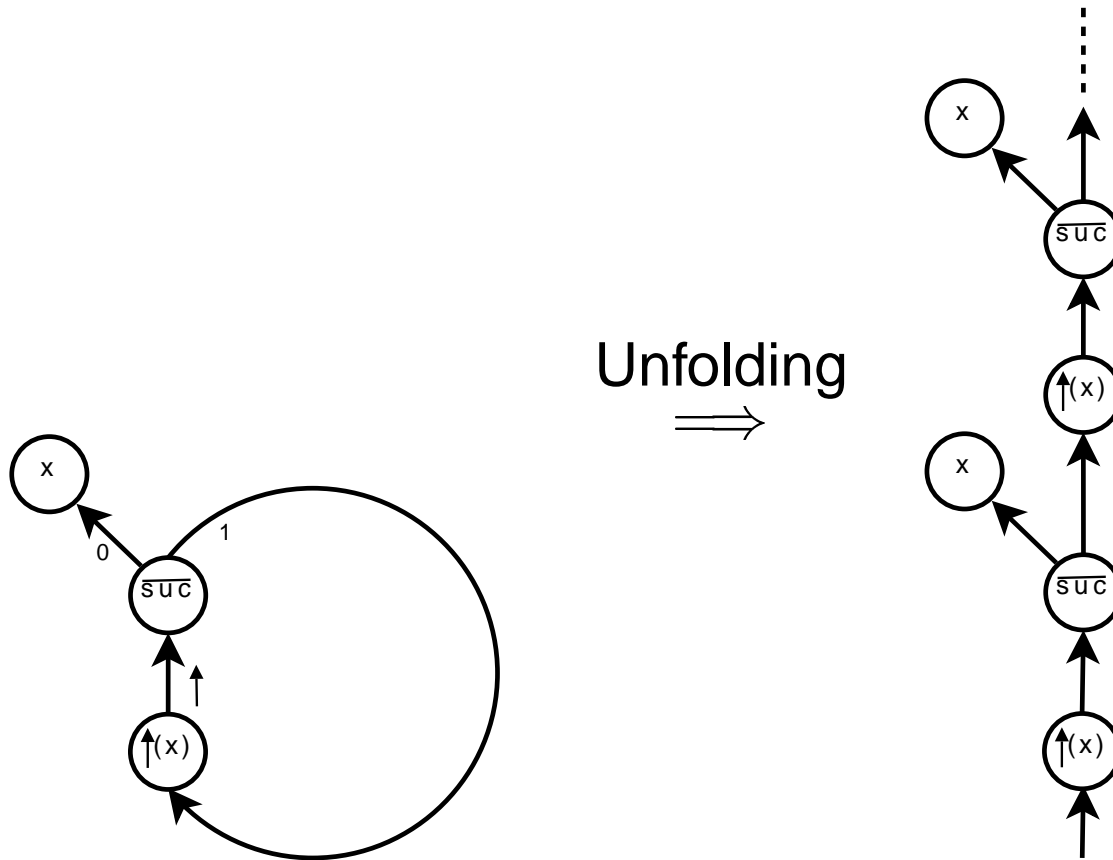
$$\llbracket x \rrbracket = x;$$

$$\llbracket \sum a(\vec{x}_a).P_a \rrbracket = \sum a(\vec{x}_a).\llbracket P_a \rrbracket.$$

- **Non-effective:** it works on infinite designs; renaming and substitution involved.

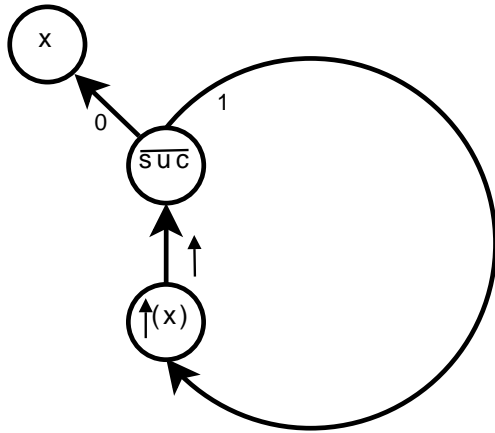
Finite generation

- **Finite generation:** Some **infinite** I-designs can be obtained from a **finite** graph by unfolding:



Finite generation

- Is there any normalization procedure that directly works on graph representations?



- Krivine's abstract machine** can be adapted to do so.

L-designs

- P is **total** if $P \neq \Omega$.
- T is **linear** if for any subterm $N_0|a\langle N_1, \dots, N_n \rangle$, $fv(N_0), \dots, fv(N_n)$ are pairwise disjoint.
- x is **identity** if it occurs in a bracket $N_0|\bar{a}\langle N_1, \dots, x, \dots, N_n \rangle$.
- **L-designs**: total, linear, identity-free designs with finitely many free variables.

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What are data?

- Examples: integers, words, trees, lists, records, etc.
- Data must be:
 - structured (eg. list = head + tail)
 - linearly duplicable (“linear” = “machine-like”)
 - compressable (eg. binary int. \rightarrow hexadecimal int.)
- Fix a unary name $\uparrow \in A$.
- Notation: $\downarrow = \bar{\uparrow}$, $\uparrow \bar{a} \langle \vec{N} \rangle = \uparrow (x).x | \bar{a} \langle \vec{N} \rangle$
- The set of **data designs** is coinductively defined by

$$d ::= \uparrow \bar{a} \langle d, \dots, d \rangle, \quad a \in A.$$

Data: examples

- Natural numbers

$$\begin{aligned}0^\bullet &= \uparrow \overline{\text{zero}} \\ n + 1^\bullet &= \uparrow \overline{\text{suc}\langle n^\bullet \rangle}\end{aligned}$$

- Ordinal omega

$$\omega^\bullet = \uparrow \overline{\text{suc}\langle \omega^\bullet \rangle}.$$

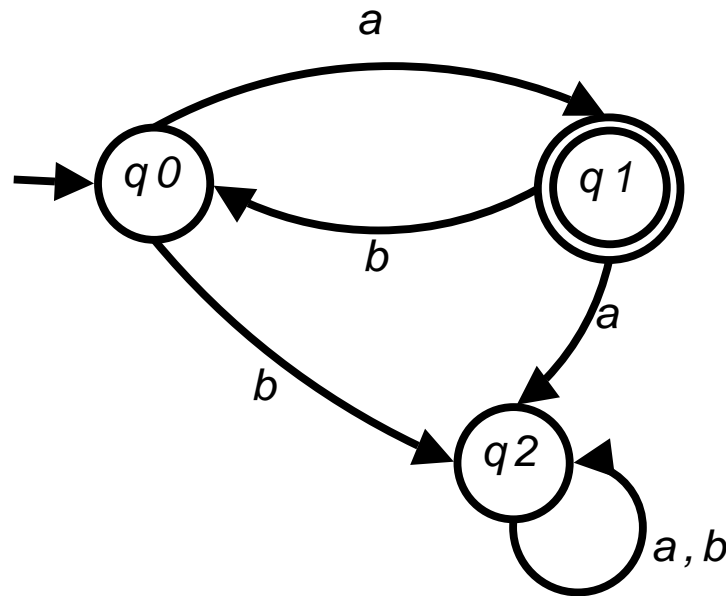
- Words, labelled binary trees, and lists:

$$\begin{aligned}\epsilon^\bullet &= \uparrow \overline{\text{nil}} \\ aba^\bullet &= \uparrow \overline{a}\langle \uparrow \overline{b}\langle \uparrow \overline{a}\langle \uparrow \overline{\text{nil}} \rangle \rangle \rangle \\ \text{node}_a(\text{leaf}_b, \text{leaf}_c)^\bullet &= \uparrow \overline{a}\langle \uparrow \overline{b}, \uparrow \overline{c} \rangle\end{aligned}$$

- Infinite words and trees are also representable.

From DFAs to cut-free l-designs

A DFA accepting $a(ba)^*$:



$$\begin{array}{ll}
 P_0 = x \downarrow \langle N_0 \rangle, & N_0 = a(x).P_1 + b(x).P_2 + \text{nil}.\Omega, \\
 P_1 = x \downarrow \langle N_1 \rangle, & N_1 = a(x).P_2 + b(x).P_0 + \text{nil}.\mathbf{\times}, \\
 P_2 = x \downarrow \langle N_2 \rangle, & N_2 = a(x).P_2 + b(x).P_2 + \text{nil}.\Omega.
 \end{array}$$

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 P_2 = x \mid \downarrow \langle N_2 \rangle, & N_2 = a(x).P_2 + b(x).P_2 + \text{nil}.\Omega.
 \end{array}$$

$$\begin{array}{l}
 P_0[aba^\bullet/x] = P_0[\uparrow(x).x \mid \bar{a} \langle ba^\bullet \rangle / x] \\
 \longrightarrow N_0 \mid \bar{a} \langle ba^\bullet \rangle \\
 \longrightarrow P_1[ba^\bullet/x] \\
 \longrightarrow^* P_0[a^\bullet/x] \\
 \longrightarrow^* P_1[\uparrow(x).x \mid \bar{\text{nil}} / x] \\
 \longrightarrow N_1 \mid \bar{\text{nil}} \\
 \longrightarrow^* \blacklozenge
 \end{array}$$

DFAs = f.g. cut-free l-designs

● **Theorem:** DFA $M \iff$ finitely generated cut-free l-design P :

$$M \text{ accepts } w \iff P \perp w^\bullet,$$

for every $w \in \Sigma^*$.

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- To enrich automata, one equips them with **stacks**.
- To enrich designs, one equips them with **cuts**.

L-designs with cuts

- Successors, Discriminators
- Duplicator $Dup[x]$. For any **finite** data design d ,

$$\llbracket Dup[d] \rrbracket = \uparrow \overline{\text{pair}}(d, d).$$

- Cf. Duplicators in linear λ -calculus (with limited rec.)

$$\begin{aligned} Dup_{\mathbf{B}}(x) &= \text{case } x = \text{true} && \Rightarrow \text{true} \otimes \text{true} \\ & && x = \text{false} && \Rightarrow \text{false} \otimes \text{false} \end{aligned}$$

$$\begin{aligned} Dup_{\mathbf{N}}(x) &= \text{case } x = \text{zero} && \Rightarrow \text{zero} \otimes \text{zero} \\ & && x = \text{suc}(y) && \Rightarrow \text{let } z_1 \otimes z_2 = Dup_{\mathbf{N}}(y) \\ & && && \text{in } \text{suc}(z_1) \otimes \text{suc}(z_2) \end{aligned}$$

- Cut is essential for finite generation.
- Q: Does Dup duplicate ω^\bullet ?

L-designs with cuts

- **Theorem:** TM $M \iff$ finitely generated (**cut-ful**) I-design P :

$$M \text{ accepts } w \iff P \perp w^\bullet.$$

- **Proof.**

(\Rightarrow) Successors, discriminators, duplicators and the **general recursion scheme** are available with cuts.

(\Leftarrow) Krivine's abstract machine works effectively on finite graph representations.

- Finitely generated cut-ful I-designs capture the **r.e. languages**.
- What about **arbitrary data sets**?

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Strong separation for data designs

- **Böhm's theorem** in lambda calculus: Given $t \neq_{\beta\eta} u$, there is $C[]$ such that $C[t] =_{\beta\eta} \lambda xy.x$ and $C[u] =_{\beta\eta} \lambda xy.y$.

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- **Separation theorem** in ludics: for any (atomic) cut-free N, M ,
 $N = M$ iff $P \perp N \iff P \perp M$ for any P

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- **Separation theorem** in ludics: for any (atomic) cut-free N, M ,
 $N = M$ iff $P \perp N \iff P \perp M$ for any P
- **Strong separation for finite data designs:**
for any finite data design d , there is a **counter design** d^c such that for any e ,

$$d = e \quad \text{iff} \quad d^c \perp e.$$

Strong separation for data designs

- **Böhm's theorem** in lambda calculus: Given $t \neq_{\beta\eta} u$, there is $C[]$ such that $C[t] =_{\beta\eta} \lambda xy.x$ and $C[u] =_{\beta\eta} \lambda xy.y$.
- **Separation theorem** in ludics: for any (atomic) cut-free N, M ,
 $N = M$ iff $P \perp N \iff P \perp M$ for any P
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for any finite data design d , there is a **counter design** d^c such that for any e ,
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- How do we separate an arbitrary **set** of data designs?

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- Behaviours are rich enough to capture all sets of finite data.

Behaviours

- Given a set \mathbf{T} of I-designs (atomic, of the same polarity),

$$\mathbf{T}^\perp = \{U : \forall T \in \mathbf{T}. T \perp U\}.$$

- Forms a Galois connection:

$$\mathbf{P} \subseteq \mathbf{N}^\perp \iff \mathbf{N} \subseteq \mathbf{P}^\perp$$

- Behaviour: $\mathbf{T} = \mathbf{T}^{\perp\perp}$.

- Analogue of formulas, types, computability predicates, and languages.

- Fact: Any set of the form $\{P\}^\perp$ is a behaviour.

Behaviours

- Any **set** D of finite data designs ‘forms’ a behaviour $\mathbf{D} \doteq \{D^c\}^\perp$
- Any **r.e. set** $L \subseteq \Sigma^*$ can be expressed as $\{P\}^\perp$ where P is **finitely generated**.
- Any **regular set** $L \subseteq \Sigma^*$ can be expressed as $\{P\}^\perp$ where P is **finitely generated and cut-free**.
- One can apply **logical connectives** to obtain a new behaviour.

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2. What is ludics?
3. Data and computation in ludics
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5. Language operators and internal completeness
6. Space compression and focalization
7. Conclusion

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 - By interaction: $\{P\}^\perp$ for an I-design P .
 - By construction: $a.\mathbf{D} \cup b.\mathbf{E}$

Internal completeness

- Harmony of two approaches is ensured by **internal completeness**:

$$a.\mathbf{D} \cup b.\mathbf{E} \doteq (a.\mathbf{D} \cup b.\mathbf{E})^{\perp\perp}.$$

- The key step when proving **full completeness theorem**:

$$P \in T^\circ \iff P \text{ interprets a proof of } T.$$

- Think of the case

$$\begin{aligned} T &= D \oplus E \\ T^\circ &= (\iota_1.D \cup \iota_2.E)^{\perp\perp} \end{aligned}$$

- Also a key to **DFAs = Regular Expressions**.

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Space compression and Focalization

- **Space compression theorem:** based on compression of data by using more symbols.

$$(0110)_2 \mapsto (12)_4.$$

In terms of data designs,

$$\uparrow \bar{0} \langle \uparrow \bar{1} \langle \uparrow \bar{1} \langle \uparrow \bar{0} \langle \uparrow \bar{\text{nil}} \rangle \rangle \rangle \rangle \mapsto \uparrow \bar{1} \langle \uparrow \bar{2} \langle \uparrow \bar{\text{nil}} \rangle \rangle.$$

- This map can be derived from a general principle of **focalization**:

$$\bar{\alpha} \langle \uparrow \bar{\beta} \langle \mathbf{N} \rangle \rangle \cong \overline{\alpha\beta} \langle \mathbf{N} \rangle.$$

Focalization principle

- In proof search in linear logic, one has to **focus** on a formula in the end sequent:

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- Two connectives of the same polarity can be combined together: $A \otimes (B \oplus C) = \otimes \oplus (A, B, C)$.

Conclusion

- **Ludics:** general setting for logically analyzing computation
 - Supports various data, higher order, concurrency
 - Importance of finite generation and cuts
 - Arbitrary I-designs: arbitrary sets of finite data
 - F.g. I-designs: r.e. languages
 - F.g. cut-free I-designs: regular languages
- “Adding stacks to automata = adding cuts to designs”
- Uses of logical theorems of ludics (separation, linearity, internal completeness, focalization)
 - WIP
 - Internal/full completeness \rightsquigarrow DFA = Regular Expr.
 - Focalization \rightsquigarrow Space compression.