

A Nominal Approach to Probabilistic Separation Logic

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$X \leftarrow \text{flip } 1/2;$

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X and Y are independent random variables

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$$(\mathcal{F}, \mu) \bullet (\mathcal{G}, \nu) \models P * Q \quad \text{if} \quad \begin{array}{l} (\mathcal{F}, \mu) \models P \\ (\mathcal{G}, \nu) \models Q \end{array}$$

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\mathcal{F}, \mathcal{G} are σ -algebras,
 μ, ν are probability measures

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↑
independent combination
("disjoint union for spaces")

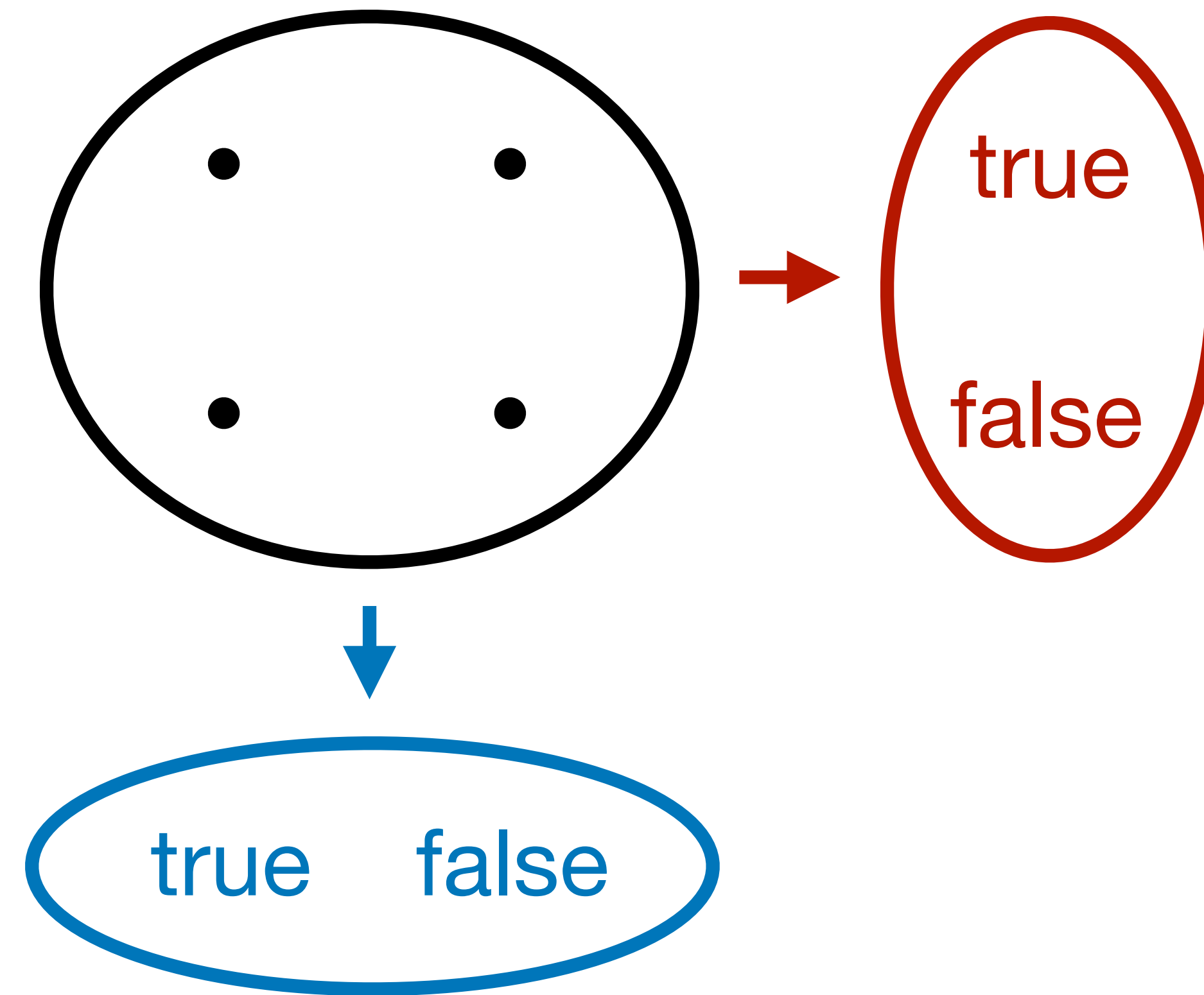
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This paper: a nominal answer

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Set[≪]

"absolutely continuous sets"

This paper: a nominal answer

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Lilac's independent
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A more abstract view: resources to atomic sheaves

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R

symmetric monoidal

"resources"

A more abstract view: resources to atomic sheaves

R

S

symmetric monoidal, atomic, ...

"resource shapes"

A more abstract view: resources to atomic sheaves

$$|\cdot| : \mathbf{R} \xrightarrow{\text{monoidal}} \mathbf{S}$$

"forget the contents of the resource"

A more abstract view: resources to atomic sheaves

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In $\text{Sh}_{\text{atomic}}(\mathbf{S})$,

$\text{Res} = \varinjlim_{r:\text{Core}(\mathbf{R})} \mathbf{S}(-, |r|)$ is a sheaf of resources

$\text{Res} \otimes \text{Res}$ is a sheaf of separated resources

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Lemma C.23

A more abstract view: atomic sheaves to G-sets

- Under suitable conditions, one can find an object s_∞ that produces an equivalence $i : \mathrm{Sh}_{\mathrm{atomic}}(\mathbf{S}) \simeq \mathrm{Aut}(s_\infty)\text{-sets}$.

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- This equivalence gives a correspondence

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in

$\mathrm{Sh}_{\mathrm{atomic}}(\mathbf{S})$

\sim

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 EMS_d Set_d^{\ll}

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- Across this equivalence,

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discrete independent combination comes
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Theorem 3.21

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Lilac's independent combination comes
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Theorem 4.24

See the paper for...

- Precise definitions
- Separation logic details
- Constructing suitable s_∞ s
- Properties of **EMS**_{std} (monoidal, atomic, subcanonical)

Summary

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- Lilac's independent combination can be explained in terms of the familiar product of probability spaces

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Probabilistic Programming Semantics for Name Generation

MARCIN SABOK, McGill University, Canada
SAM STATON, University of Oxford, United Kingdom
DARIO STEIN, University of Oxford, United Kingdom
MICHAEL WOLMAN, McGill University, Canada

Probability Sheaves and the Giry Monad*

Alex Simpson

Equivalence and Conditional Independence in Atomic Sheaf Logic

Alex Simpson*

Thanks!

EMS

\approx

Set^{\ll}

independence via
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Lilac's independent
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The folklore: separation logic in **Sch**

- **Sch** = $\text{Sh}_{\text{atomic}}(\mathbf{FinInj}^{\text{op}})$

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↑
"heap shapes"

The folklore: separation logic in **Sch**

- **Sch** = $\text{Sh}_{\text{atomic}}(\mathbf{FinInj}^{\text{op}})$
- $(\mathbf{FinInj}^{\text{op}}, +, \emptyset)$ is a monoidal category.
- Yields a monoidal structure (\otimes, I) on **Sch**, by Day convolution.

The folklore: separation logic in Sch

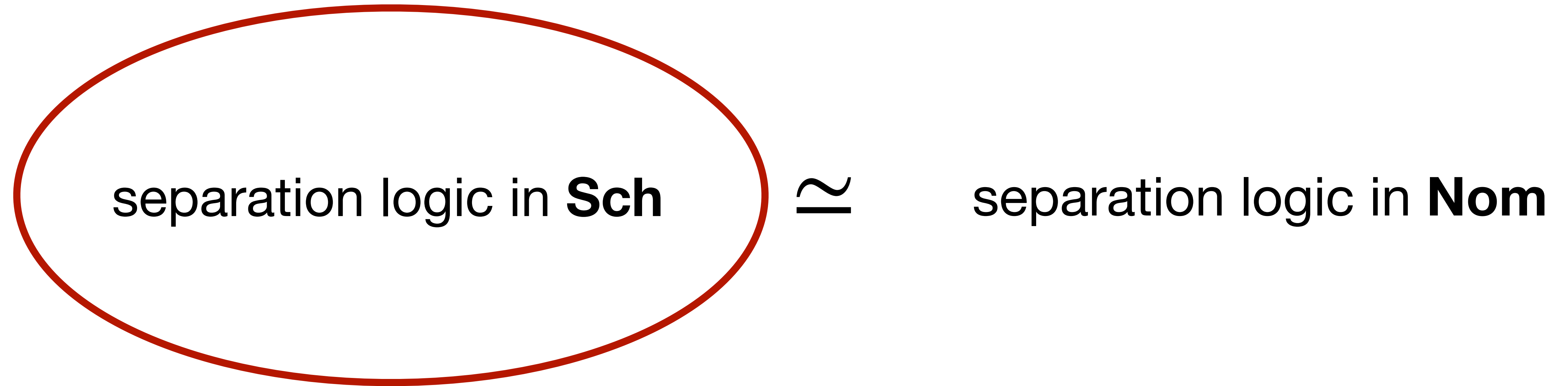
- There is a sheaf of heaps $\mathbb{H}(L) = L \rightarrow_{\text{fin}} \mathbb{Z}$.
- The convolution $\mathbb{H} \otimes \mathbb{H}$ is a sheaf of separated heaps:

$$(\mathbb{H} \otimes \mathbb{H})(L) = \{ (h, h') \mid \text{dom}(h) \cap \text{dom}(h') = \emptyset \}$$

- These form the basic ingredients of separation logic.

The folklore: separation logic in **Sch**

- Our results are probabilistic analogs of the following fact:


$$\text{separation logic in } \mathbf{Sch} \approx \text{separation logic in } \mathbf{Nom}$$

The folklore: separation logic in **Nom**

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\approx

separation logic in **Nom**

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The folklore: separation logic in **Nom**

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- Heaps: $\overline{H} = \mathbb{N} \rightarrow_{\text{fin}} \mathbb{Z}$
- Separated heaps:

$$\overline{H}_{\text{sep}} = \{ (h, h') \in \overline{H} \times \overline{H} \mid \text{dom}(h) \cap \text{dom}(h') = \emptyset \}$$

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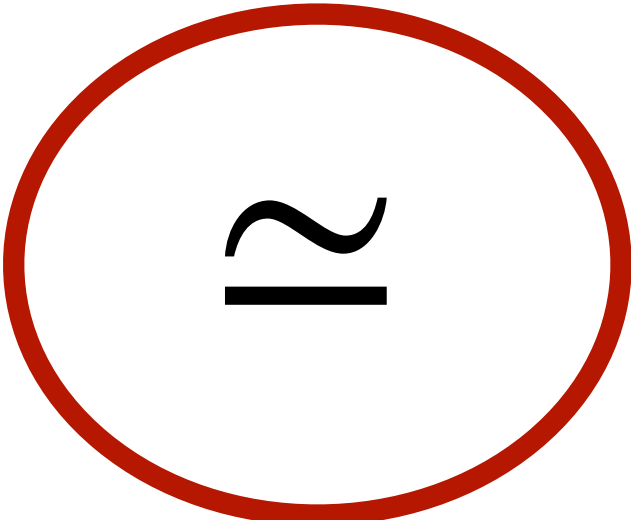
separation logic in **Sch**

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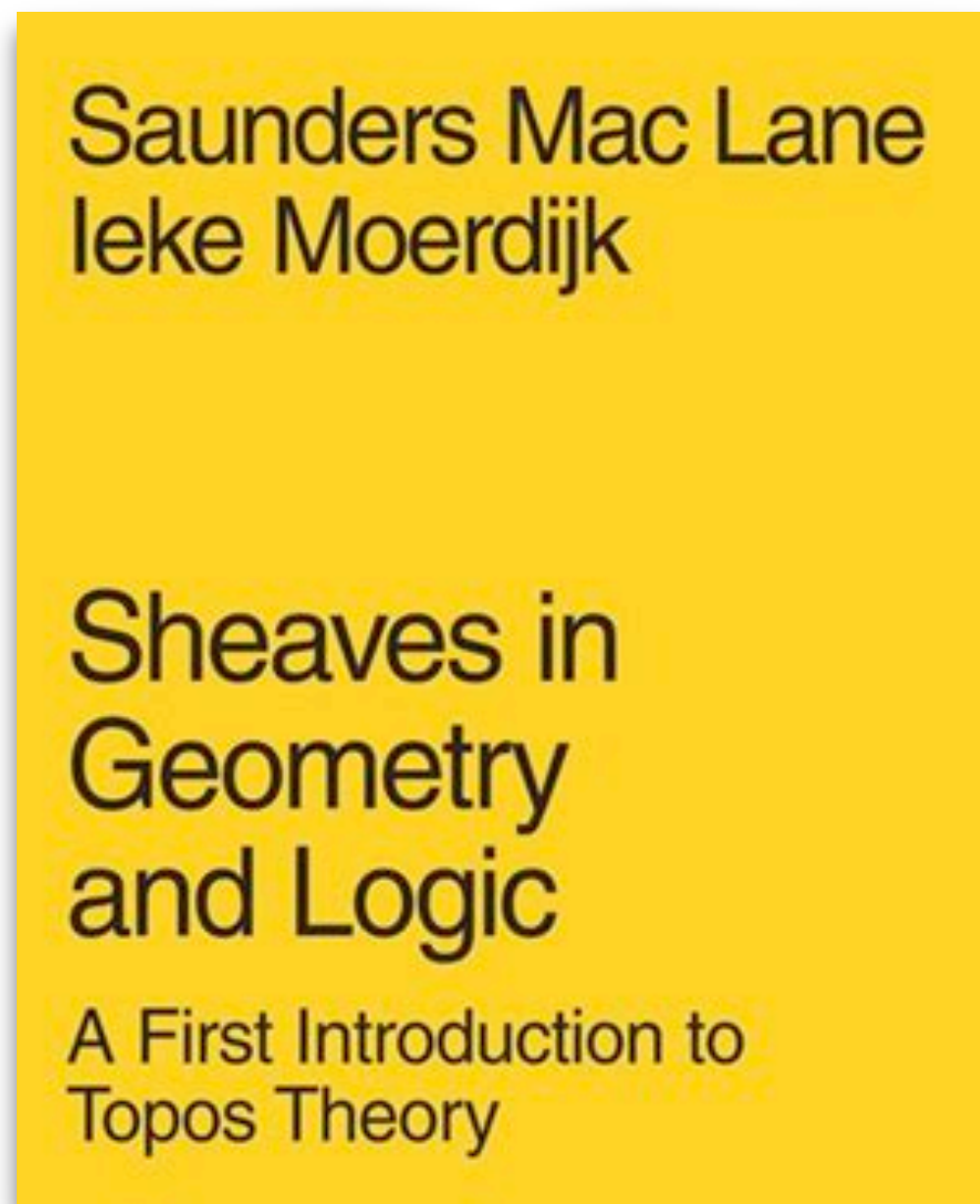
separation logic in **Nom**

The folklore: the equivalence

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separation logic in **Sch**  separation logic in **Nom**

The folklore: the equivalence



, Theorem III.9.2: **Sch** \simeq **Nom**.

The folklore: the equivalence

- Across this equivalence,

Sch

Nom

\mathbb{H}

corresponds to

$\overline{\mathbb{H}}$

$\mathbb{H} \otimes \mathbb{H}$

corresponds to

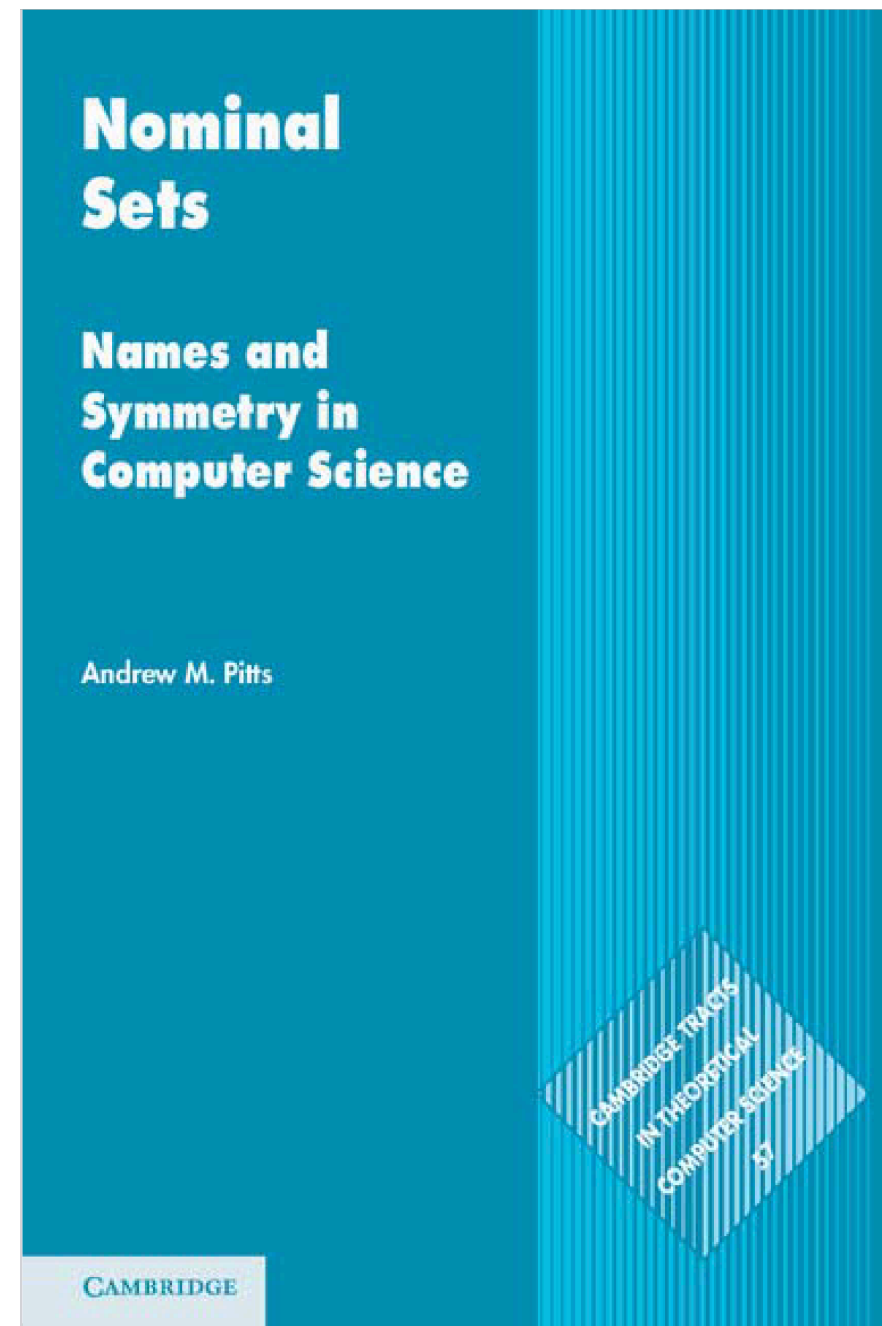
$\overline{\mathbb{H}}_{\text{sep}}$

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, Lemma 1.14 (Homogeneity):

$$\begin{array}{ccc} N & \xrightarrow{\pi} & N \\ \uparrow & & \uparrow \\ A & \xrightarrow{f} & B \end{array}$$

Our probabilistic analog: the discrete case

- Key lemma:

$$\begin{array}{ccc} \mathbb{N} & \overset{\pi}{\dashrightarrow} & \mathbb{N} \\ \uparrow & & \uparrow \\ A & \xrightarrow{f} & B \end{array}$$

becomes

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- Proof roughly boils down to: any two nonnegligible measurable subsets of $[0, 1]$ are measurably isomorphic.

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- Proof requires some heavy-duty measure theory.