



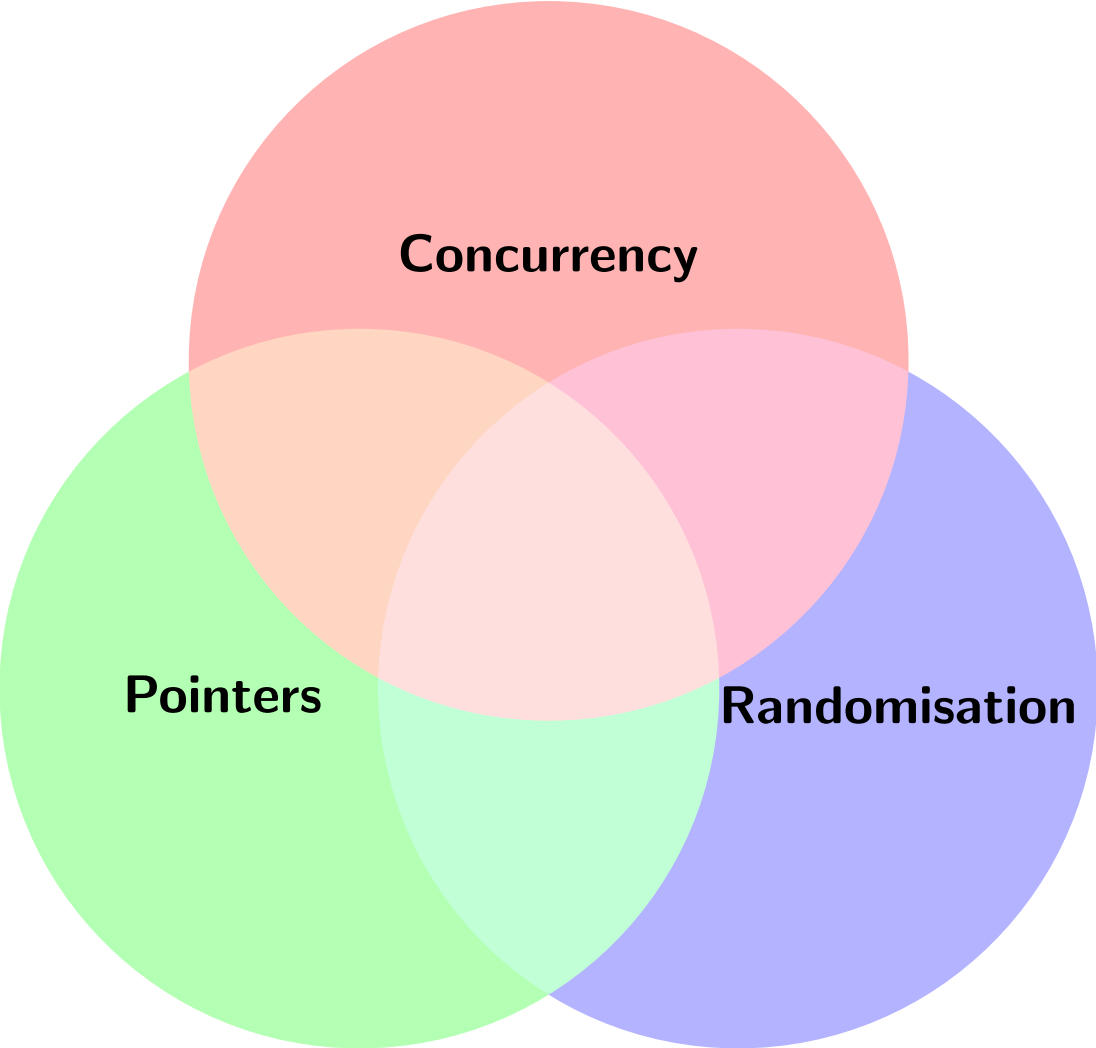
Concurrent Separation Logic with Probabilities

Ira Fesefeldt Joost-Pieter Katoen Thomas Noll

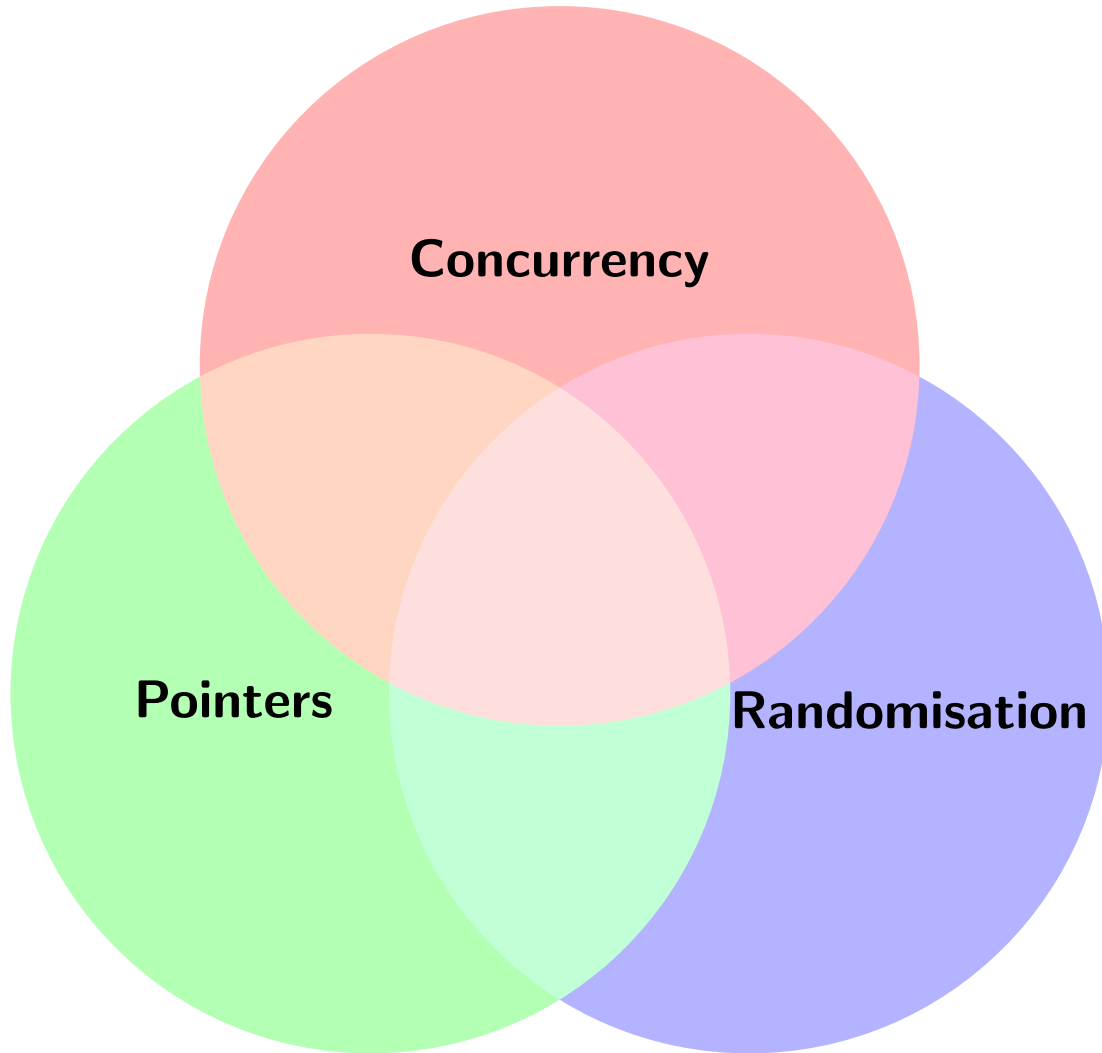
RWTH Aachen University, Germany

OPCT 2023, Bertinoro, Italy; June 26–30, 2023

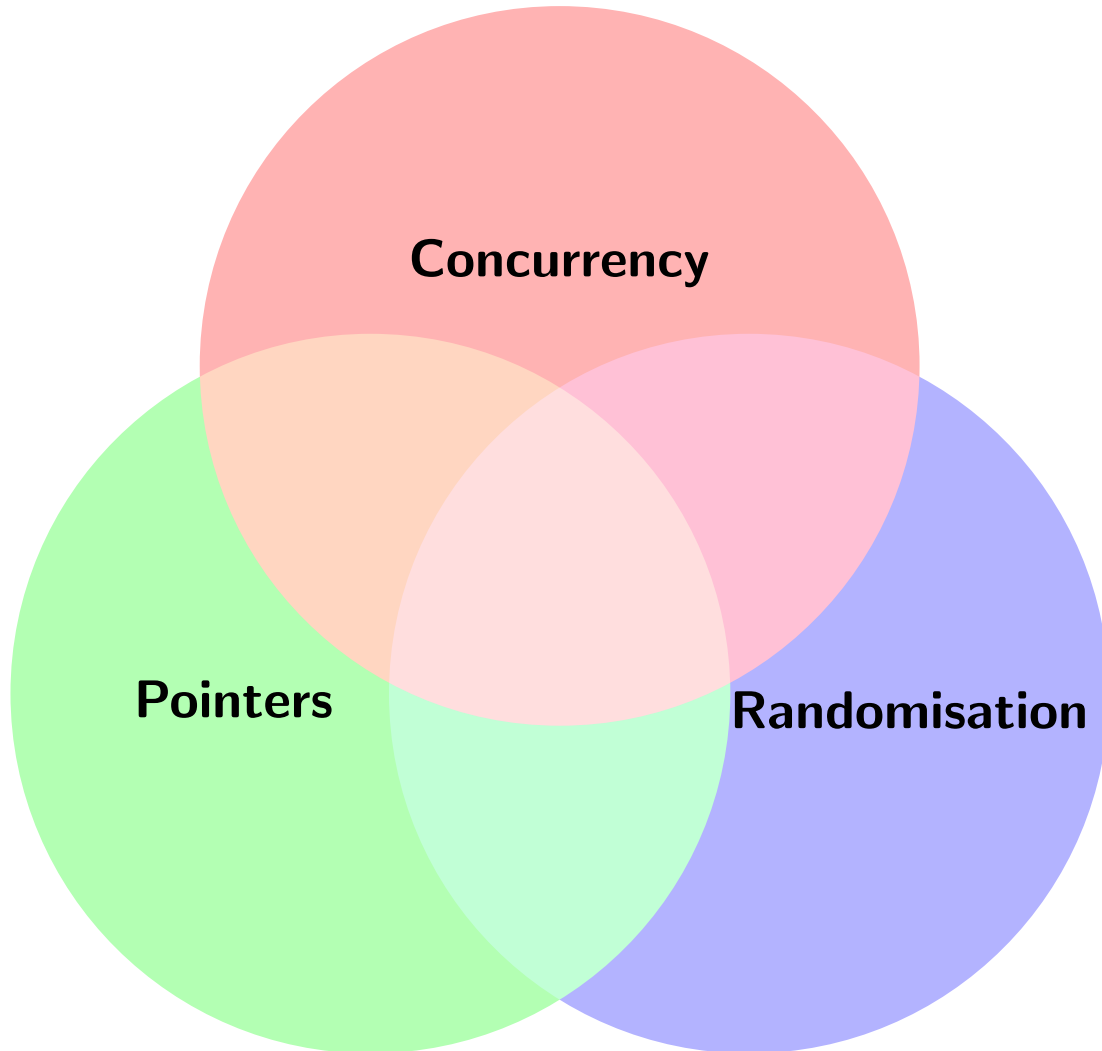
Motivation



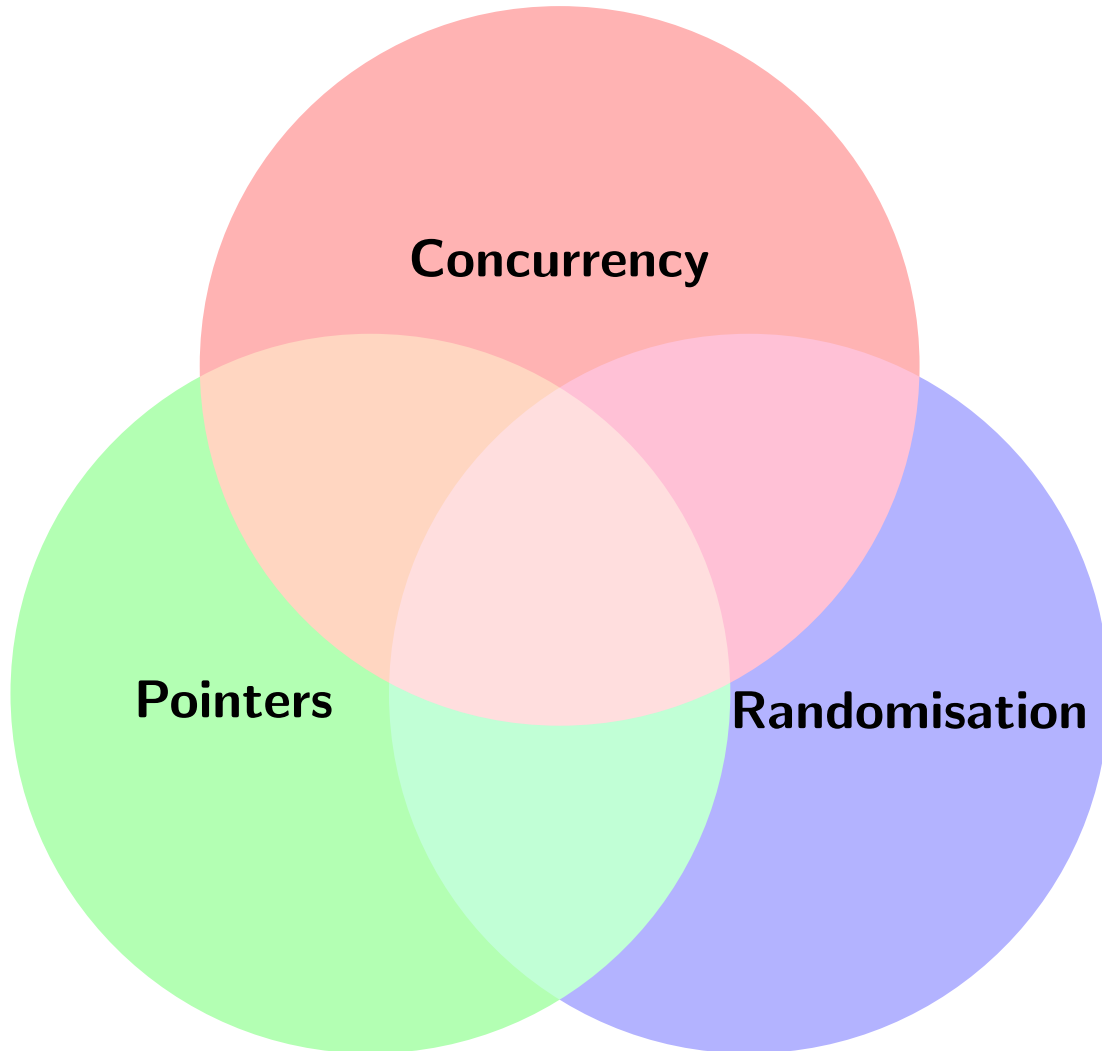
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- Concurrency
 - distributed computing
 - parallel algorithms
 - ...

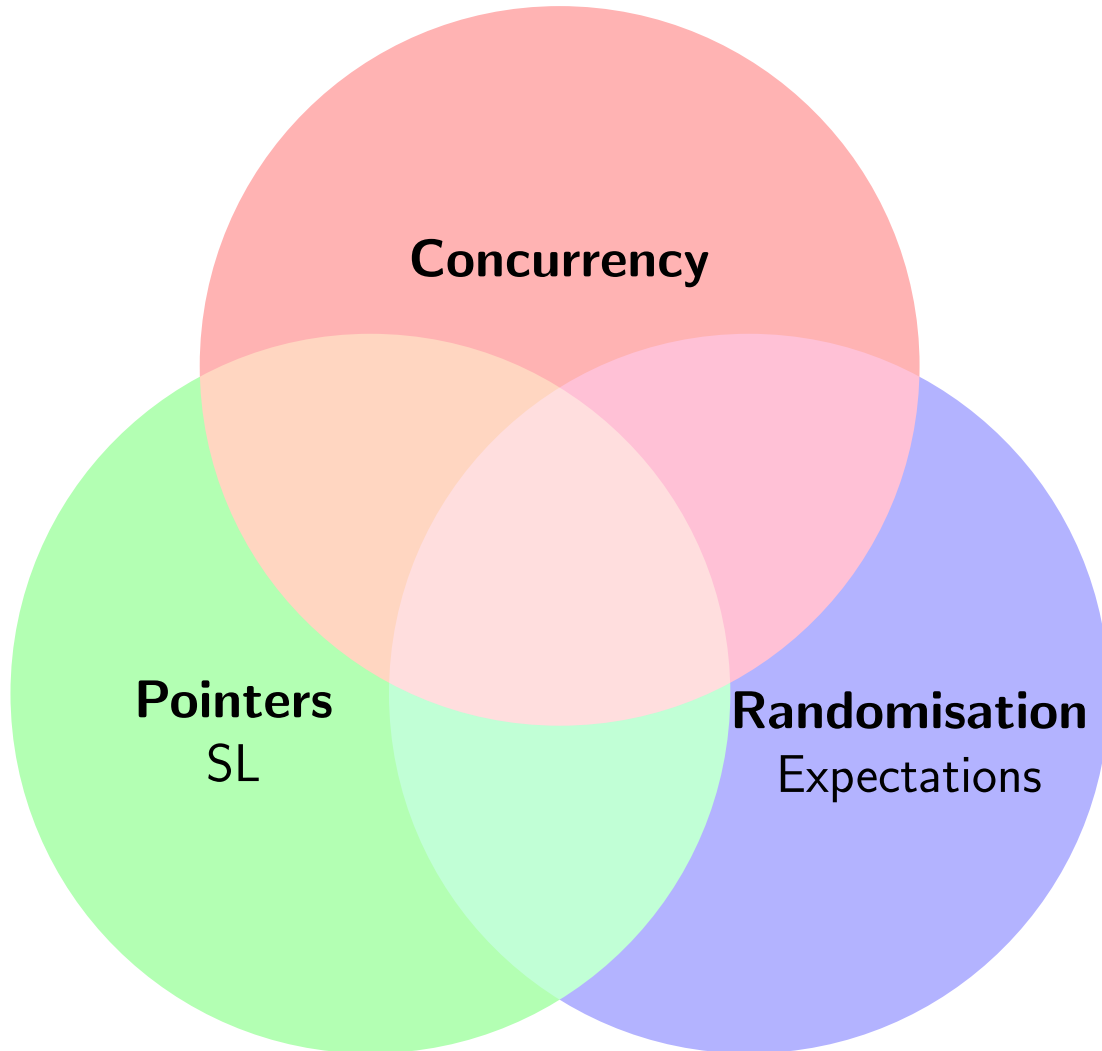


- Concurrency
 - distributed computing
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- Pointers
 - dynamic data structures
 - shared memory
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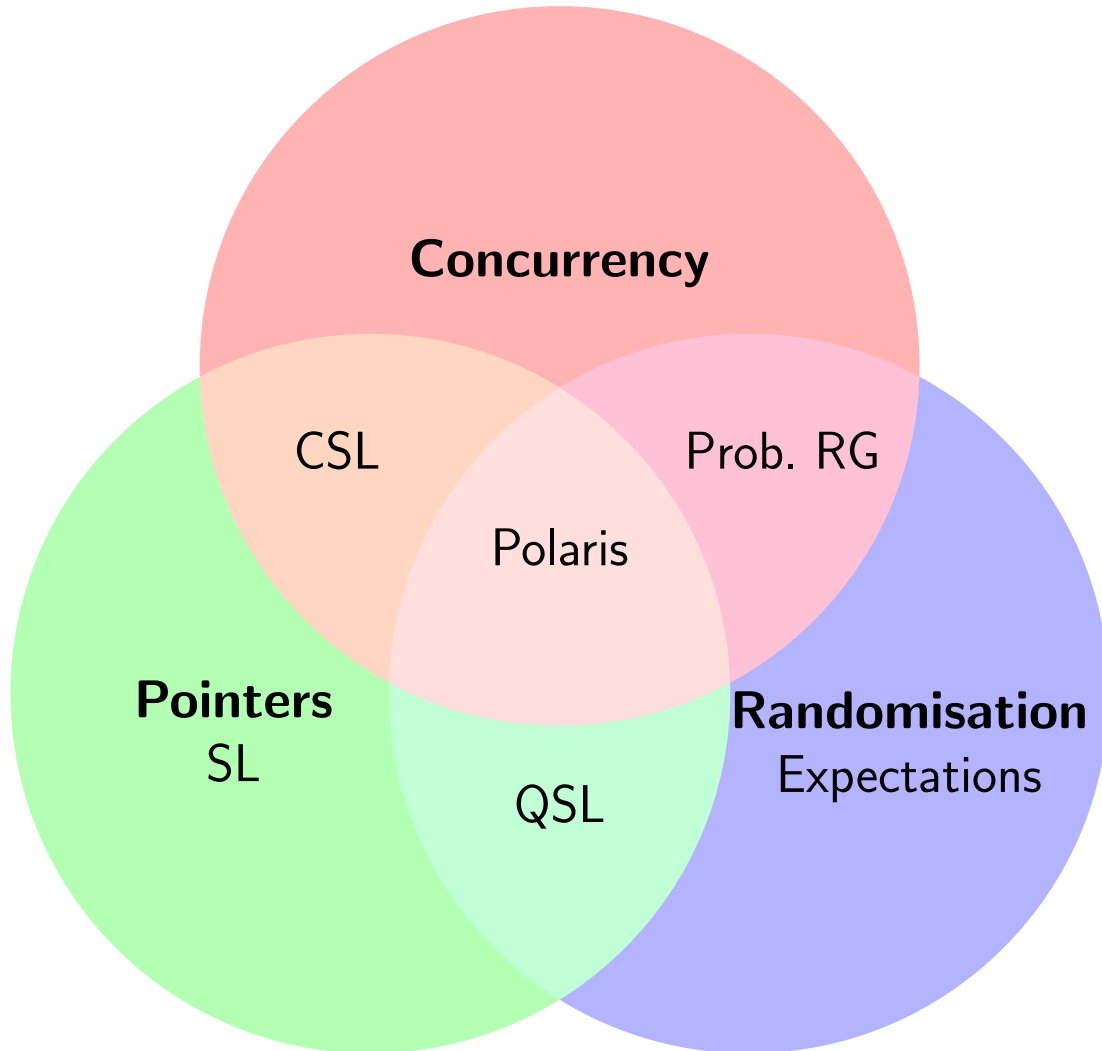
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 - randomised algorithms
 - unreliable hardware
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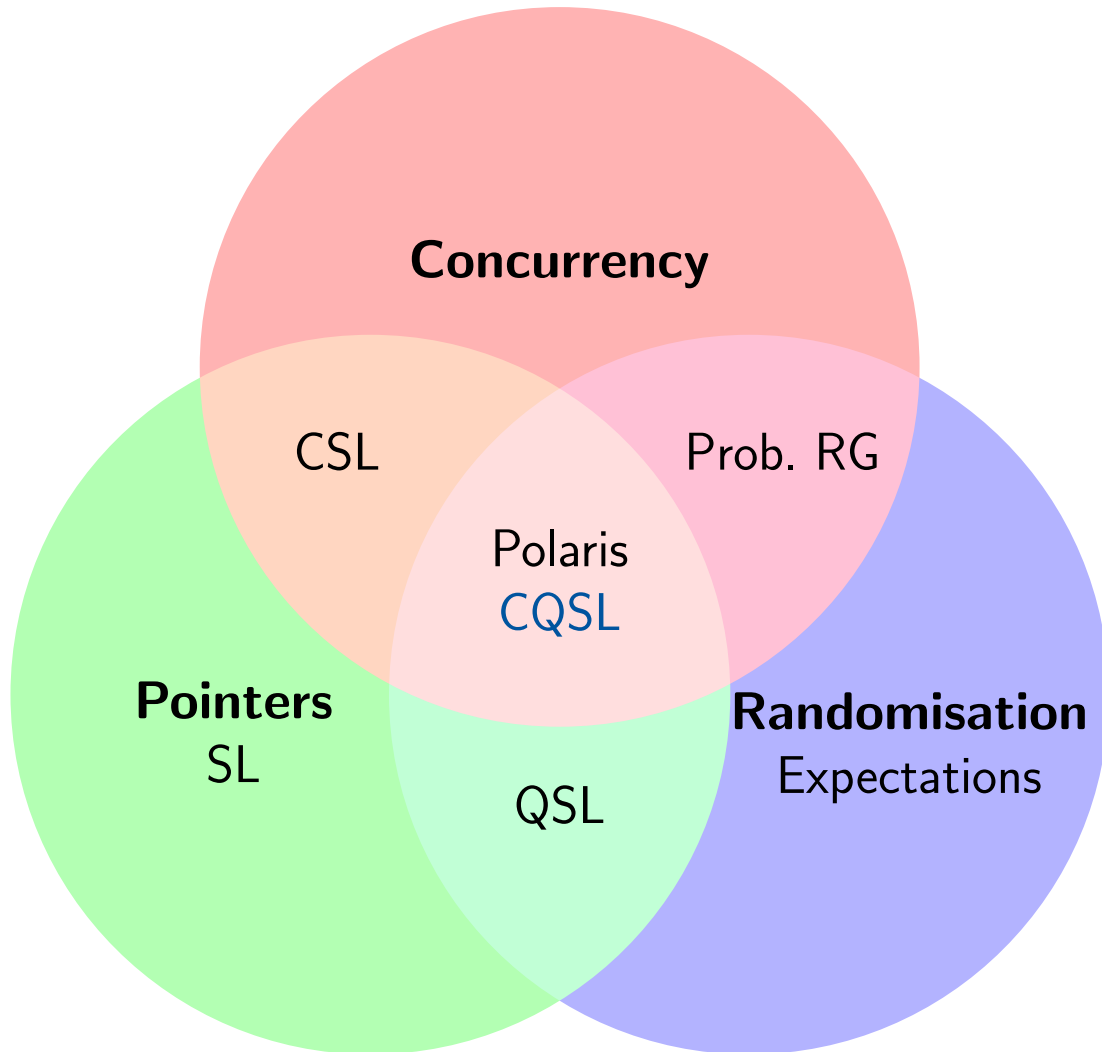
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An Example

```
//  
<r> := -1;  
//  
//  
<r> := 0  
//  
//  
//  
//  
//  
//
```

||

```
//  
y := <r>;  
//  
while(y = -1) { y := <r>}  
//  
//
```

An Example

```
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<r> := -1;  
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<r> := 0  
  
//  
// y = 0
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//  
y := <r>;  
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while (y = -1) { y := <r> }  
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An Example

Given resource invariant $I_R := r \mapsto -1 \vee r \mapsto 0$, we have:

```
// true
<r> := -1;
// true
// true
<r> := 0
// true
// y = 0
||
// true
y := <r>;
// y = -1 ∨ y = 0
while(y = -1) { y := <r> }
// y = 0
```

An Example

Given resource invariant $I_R := r \mapsto -1 \vee r \mapsto 0$, we have:

```
// ???
<r> := -1;
// ???
// ???
{<r> := 0}
  [0.5]
{<r> := 1}
// true
// y = 0
||
// true
y := <r>;
// y = -1 ∨ y = 0
while(y = -1) { y := <r>}
// y = 0
```

Definition

For program C , expectations¹ $X, Y: States \rightarrow [0, 1]$ (with $States := Stacks \times Heaps$):

$X \leq \text{wlp}[[C]](Y)$ iff X lower bounds the expected value of Y after executing C
+ probability to diverge

¹Actually random variables and not expected values.

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probability of Y when $Y: States \rightarrow \{0, 1\}$

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Example

- $0.5 \leq \text{wlp}[[C_{\text{asgn}}]]([r \mapsto 0])$
- $0.5 \leq \text{wlp}[[C_{\text{asgn}}]]([r \mapsto 1])$

$C_{\text{asgn}} : \{ \langle r \rangle := 0 \} [0.5] \{ \langle r \rangle := 1 \}$

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- $0.5 \leq \text{wlp}[[C_{\text{run}}]]([y = 0])$

$$C_{\text{asgn}} : \{ \langle r \rangle := 0 \} [0.5] \{ \langle r \rangle := 1 \}$$

$$C_{\text{run}} : \langle r \rangle := -1; \\ \left\{ \begin{array}{l} \{ \langle r \rangle := 0 \} \\ [0.5] \\ \{ \langle r \rangle := 1 \} \end{array} \right\} \parallel \left\{ \begin{array}{l} y := \langle r \rangle; \\ \text{while}(y = -1) \{ \\ \quad y := \langle r \rangle; \} \end{array} \right.$$

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- $0.5 \leq \text{wlp}[[C_{\text{run}}]]([y = 0])$
- $1 \leq \text{wlp}[[C_{\text{run}}]]([y = 0])$
(probability of 0.5 to not terminate)

$$C_{\text{asgn}} : \{ \langle r \rangle := 0 \} [0.5] \{ \langle r \rangle := 1 \}$$

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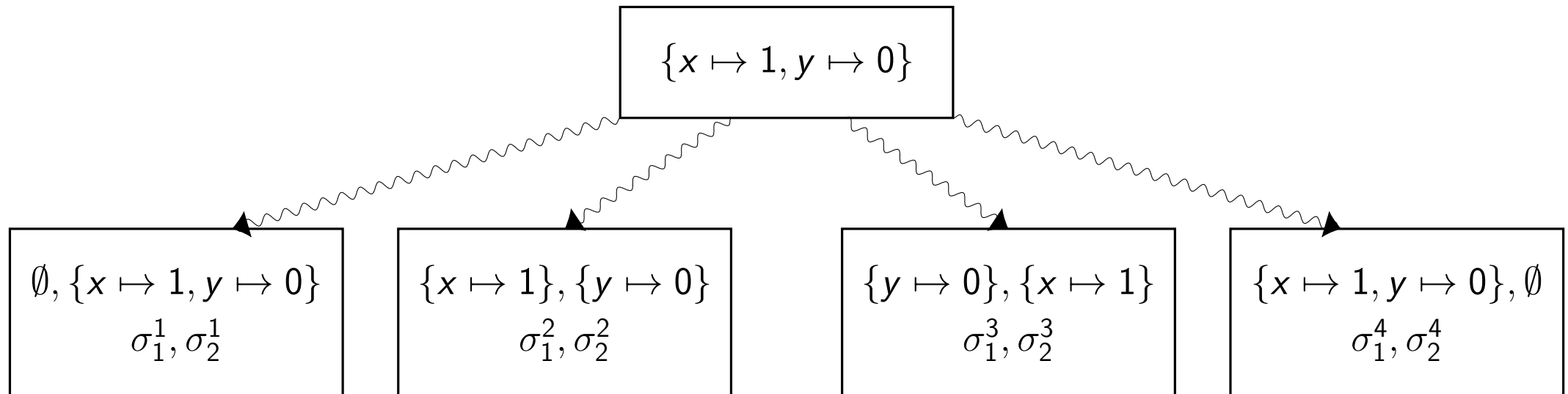
Separating conjunction: $\sigma \models \psi \star \varphi$ iff $\exists \sigma_1, \sigma_2 : \sigma = \sigma_1 \uplus \sigma_2, \sigma_1 \models \psi \wedge \sigma_2 \models \varphi$

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For state $\sigma \in States$ and expectations $X, Y : States \rightarrow [0, 1]$, we define

$$(X \star Y)(\sigma) = \sup \{ X(\sigma_1) \cdot Y(\sigma_2) \mid \sigma = \sigma_1 \uplus \sigma_2 \}$$

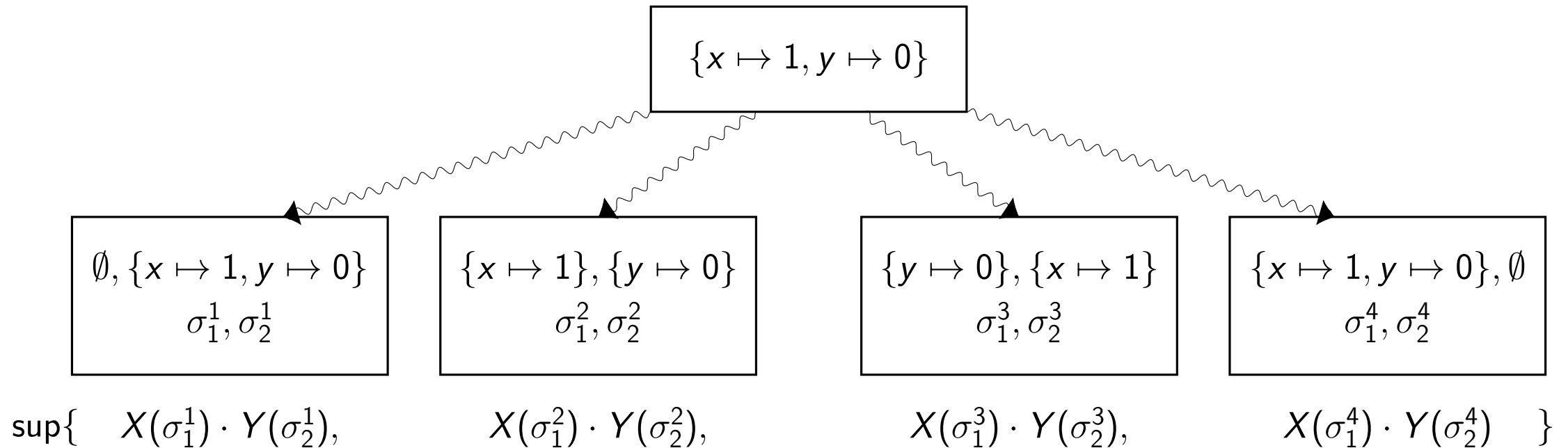


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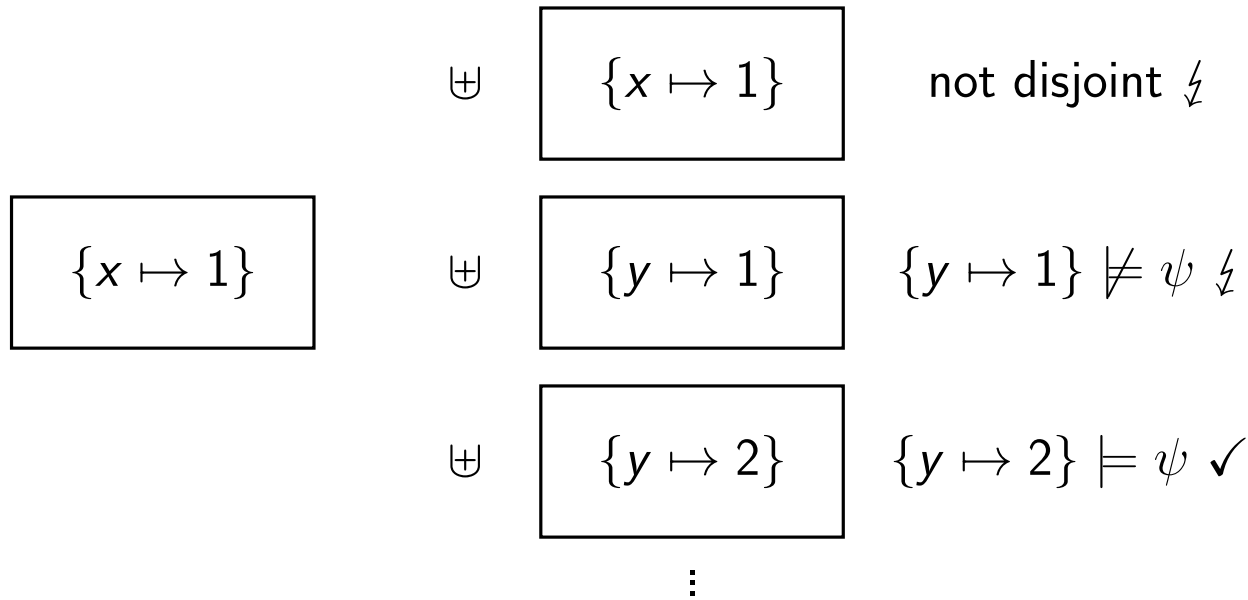
Qualitative magic wand (separating implication): $\sigma \models \psi \multimap \varphi$ iff $\forall \sigma' : \sigma' \models \psi \implies \sigma \uplus \sigma' \models \varphi$

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For state $\sigma \in States$, predicate $\psi : States \rightarrow \{0, 1\}$ and expectation $Y : States \rightarrow [0, 1]$, we define

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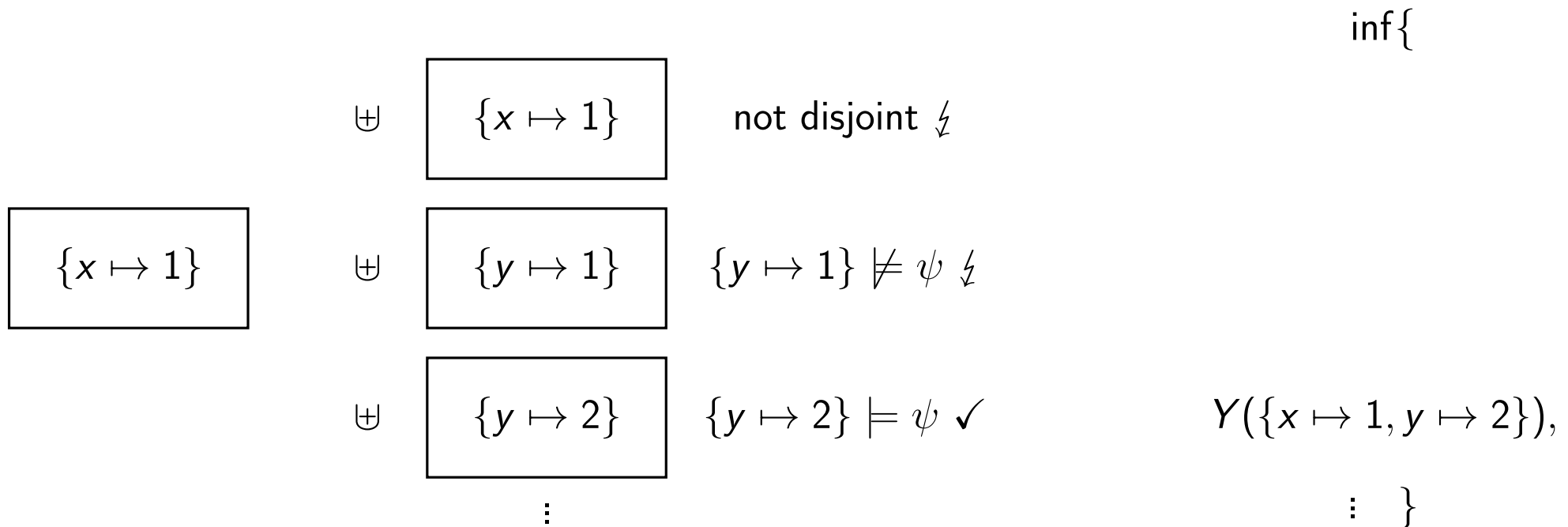


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For program C , expectations $X, Y: States \rightarrow [0, 1]$ and **resource invariant** $\xi: States \rightarrow \{0, 1\}$

$X \leq \text{wrlp}[[C]](Y \mid \xi)$ iff X lower bounds the expected value of Y after executing C
+ probability to diverge
and ξ always preserved by some sub-heap during execution

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$$\frac{X \leq \xi \longrightarrow \text{wlp}[[x := e]](Y \star \xi)}{X \leq \text{wrlp}[[x := e]](Y \mid \xi)} \text{ assign-adv}$$

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$$\frac{X \leq \text{wrlp}[[C]](Y \mid \xi \star \pi)}{X \star \pi \leq \text{wrlp}[[C]](Y \star \pi \mid \xi)} \text{share}$$

Example

Pick $\xi = \max \{ [r \mapsto -1], [r \mapsto 0] \}$:

$$\langle r \rangle := -1$$

$$\{ \langle r \rangle := 0 \} [0.5] \{ \langle r \rangle := 1 \}$$

$$\parallel 1 \star [y = 0] \mid \xi$$

$$y := \langle r \rangle;$$

$$\text{while}(y = -1) \{ y := \langle r \rangle \};$$

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Pick $\xi = \max \{ [r \mapsto -1], [r \mapsto 0] \}$:

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$$\begin{aligned} & // \text{ 0.5 } \mid \xi \\ & \{ \langle r \rangle := 0 \} [0.5] \{ \langle r \rangle := 1 \} \\ & // \text{ 1 } \mid \xi \end{aligned}$$

$$// \text{ 1 } \star [y = 0] \mid \xi$$

$$\begin{aligned} & // \text{ 1 } \mid \xi \\ & y := \langle r \rangle; \\ & // \max \{ [y = -1], [y = 0] \} \mid \xi \\ & \text{while} (y = -1) \{ y := \langle r \rangle \}; \\ & // [y = 0] \mid \xi \end{aligned}$$

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Pick $\xi = \max \{ [r \mapsto -1], [r \mapsto 0] \}$:

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$\parallel 1 \mid \xi$

$\parallel 1 \star [y = 0] \mid \xi$

$\parallel 1 \mid \xi$

$y := \langle r \rangle;$

$\parallel \max \{ [y = -1], [y = 0] \} \mid \xi$

$\text{while}(y = -1) \{ y := \langle r \rangle \};$

$\parallel [y = 0] \mid \xi$

$\implies 0.5 \leq \text{wrlp}[\![C_{run}]\!] ([y = 0] \mid \xi)$

For the empty resource invariant, wrlp and wlp coincide:

Lemma

For the empty-state predicate $[\mathbf{emp}]$, a program C and an expectation $Y : \text{States} \rightarrow [0, 1]$, we have:

$$\text{wrlp}[[C]](Y \mid [\mathbf{emp}]) = \text{wlp}[[C]](Y)$$

For the empty resource invariant, $wrlp$ and wlp coincide:

Lemma

For the empty-state predicate $[\mathbf{emp}]$, a program C and an expectation $Y : \text{States} \rightarrow [0, 1]$, we have:

$$wrlp\llbracket C \rrbracket (Y \mid [\mathbf{emp}]) = wlp\llbracket C \rrbracket (Y)$$

Can be used to compute lower bounds on wlp using $wrlp$:

Example

$$\begin{aligned} & 0.5 \leq wrlp\llbracket C_{run} \rrbracket ([y = 0] \mid \xi) && \text{(previous example)} \\ \implies & 0.5 \star \xi \leq wrlp\llbracket C_{run} \rrbracket ([y = 0] \star \xi \mid [\mathbf{emp}]) && \text{(share rule)} \\ \implies & 0.5 \star \xi \leq wrlp\llbracket C_{run} \rrbracket ([y = 0] \mid [\mathbf{emp}]) && \text{(monotonicity)} \\ \implies & 0.5 \star \xi \leq wlp\llbracket C_{run} \rrbracket ([y = 0]) && \text{(Lemma)} \end{aligned}$$

Conservativity w.r.t. CSL [O'Hearn/Vafeiadis]:

Theorem

If C is non-probabilistic and every trace in C enables certain framing criteria (left out here):

$$\xi \models_{\text{CSL}} \{\psi\} C \{\varphi\} \quad \text{iff} \quad \psi \leq \text{wrlp}[[C]](\varphi \mid \xi)$$

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Superlinearity to eliminate addition:

Lemma

When C is almost surely terminating:

$$\text{wlp}[[C]](X) + \text{wlp}[[C]](Y) \leq \text{wlp}[[C]](X + Y)$$

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Many more rules requiring **preciseness** of ξ :

Definition

ξ is precise iff for $\forall (s, h) \in \text{States}$ with $(s, h) \models \xi$ there exists no $h' \in \text{Heaps}$ such that $h \subsetneq h'$ and $(s, h') \models \xi$.

The End

Conclusion

- CQSL combines CSL and QSL to support
 - concurrency
 - pointers
 - probabilities
- Conservative w.r.t. to both CSL and QSL

Example

$$C: \langle r \rangle := -1;$$
$$\begin{array}{l} \{ \langle r \rangle := 0 \} \\ [0.5] \\ \{ \langle r \rangle := 1 \} \end{array} \parallel \begin{array}{l} y := \langle r \rangle; \\ \text{while}(y = -1) \{ \\ \quad y := \langle r \rangle; \\ \} \end{array};$$

$$0.5 \star \xi \leq \text{wlp}[[C]] ([y = 0])$$

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Open Problems

- Certain rules still require **preciseness** of resource invariant
- **Quantitative** resource invariants
(require fully quantitative version of magic wand)
- Upper bounds for **weakest (resource-safe) liberal preexpectations**
- Extension to weakest **non-liberal** preexpectations

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Proof Automation

- Expectations are **generalised propositions**
- **Quantitative extensions** for dealing with probabilities required

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