

# CSC410

# Data Flow Analyses

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# Properties of Transfer Functions

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**Monotonicity:**  $\forall x, y \in \mathbf{L} : x \sqsubseteq y \implies f(x) \sqsubseteq f(y).$

Transfer functions in a dataflow analysis must be monotone!

**Distributivity:**  $\forall x, y \in \mathbf{L} : f(x \sqcap y) = f(x) \sqcap f(y).$

But not necessarily distributive!

# Formally

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For a path  $\pi = init \dots l$

Transfer Function for location  $l$

$$f_{\pi} = f_{init} \circ \dots \circ f_l$$

Set of all paths to  $l$

Initial information at “init”

$$MOP_{\circ}(l) = \bigcap_{\pi \in Path(l)} f_{\pi}(I).$$

$$MOP_{\bullet}(l) = f_l(MOP_{\circ}(l)).$$

# Good News

Instead, compute the *maximal fixed point solution* (MFP).

in the meet lattice

Consider the set of constraints below:

$$MFP_{\circ}(l) = \begin{cases} \iota & l = init \\ \bigcap_{(l',l) \in flow} MFP_{\bullet}(l') & \text{otherwise} \end{cases}$$

$$MFP_{\bullet}(l) = f_l(MFP_{\circ}(l))$$

# The Coincidence

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If transfer functions are **monotone**:

$$MOP_{\circ}(l) \supseteq MFP_{\circ}(l) \qquad MOP_{\bullet}(l) \supseteq MFP_{\bullet}(l)$$

Less Precise!

The fixpoint solution **over-approximates** the result!

If transfer functions are **distributive**:

$$MOP_{\circ}(l) = MFP_{\circ}(l) \qquad MOP_{\bullet}(l) = MFP_{\bullet}(l)$$

The two solutions **coincide**!

# Proof of their relation

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If transfer functions are **monotone**:

$$MOP_{\circ}(l) \sqsupseteq MFP_{\circ}(l) \qquad MOP_{\bullet}(l) \sqsupseteq MFP_{\bullet}(l)$$

(proof on the board)

**Monotonicity:**  $\forall x, y \in \mathbf{L} : x \sqsubseteq y \implies f(x) \sqsubseteq f(y).$

**alternatively:**  $\forall x, y \in \mathbf{L} : f(x \sqcap y) \sqsubseteq f(x) \sqcap f(y).$

(exercise for you: prove the two definitions are equivalent)