

# ABSTRACT FORWARD PROPAGATE

## KILDALL'S ALGORITHM

AbstractForwardPropagate( $\Gamma_c, P$ )

```
S := {l0};  
 $\hat{\mu}(l_0) := \alpha(P)$ ;  
 $\hat{\mu}(l) := \perp$ , for  $l \in L \setminus \{l_0\}$ ;  
while S  $\neq \emptyset$  do{  
    l := Choose S;  
    S := S  $\setminus$  {l};  
    foreach (l, c, l')  $\in T$  do{  
        F :=  $\hat{sp}(\hat{\mu}(l), c)$ ;  
        if  $\neg(F \leq \hat{\mu}(l'))$  then{  
             $\hat{\mu}(l') := \hat{\mu}(l') \sqcup F$ ;  
            S := S  $\cup$  {l'};  
        }  
    }  
}
```

- Does this algorithm actually calculate the abstract JOP?
- What are the conditions under which it is guaranteed to terminate?



# ABSTRACT FORWARD PROPAGATE

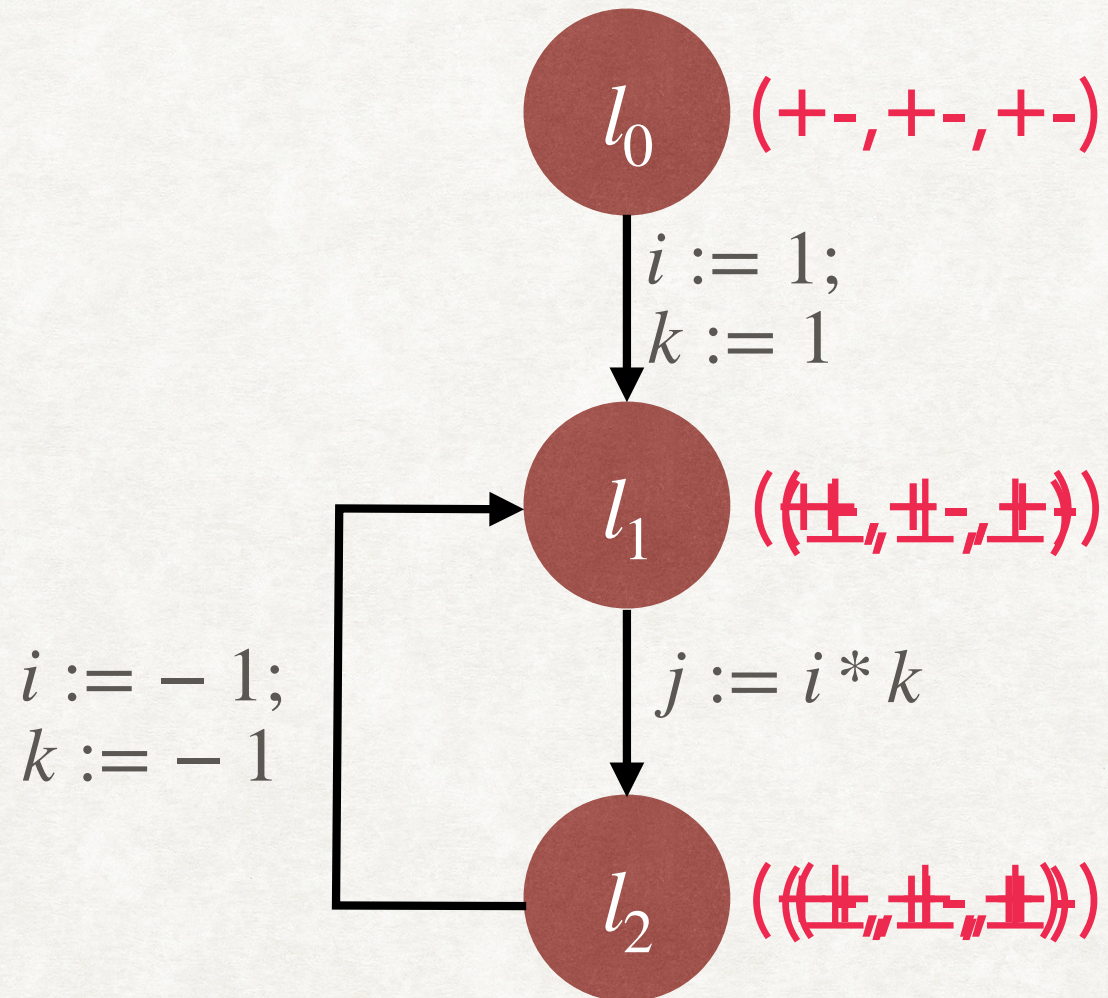
## KILDALL'S ALGORITHM

AbstractForwardPropagate( $\Gamma_c, P$ )

```
S := {l0};  
 $\hat{\mu}_K(l_0) := \alpha(P)$ ;  
 $\hat{\mu}_K(l) := \perp$ , for  $l \in L \setminus \{l_0\}$ ;  
while S  $\neq \emptyset$  do{  
    l := Choose S;  
    S := S  $\setminus$  {l};  
    foreach (l, c, l')  $\in T$  do{  
        F :=  $\hat{f}_c(\hat{\mu}_K(l))$ ;  
        if  $\neg(F \leq \hat{\mu}_K(l'))$  then{  
             $\hat{\mu}_K(l') := \hat{\mu}_K(l') \sqcup F$ ;  
            S := S  $\cup$  {l'};  
        }  
    }  
}
```

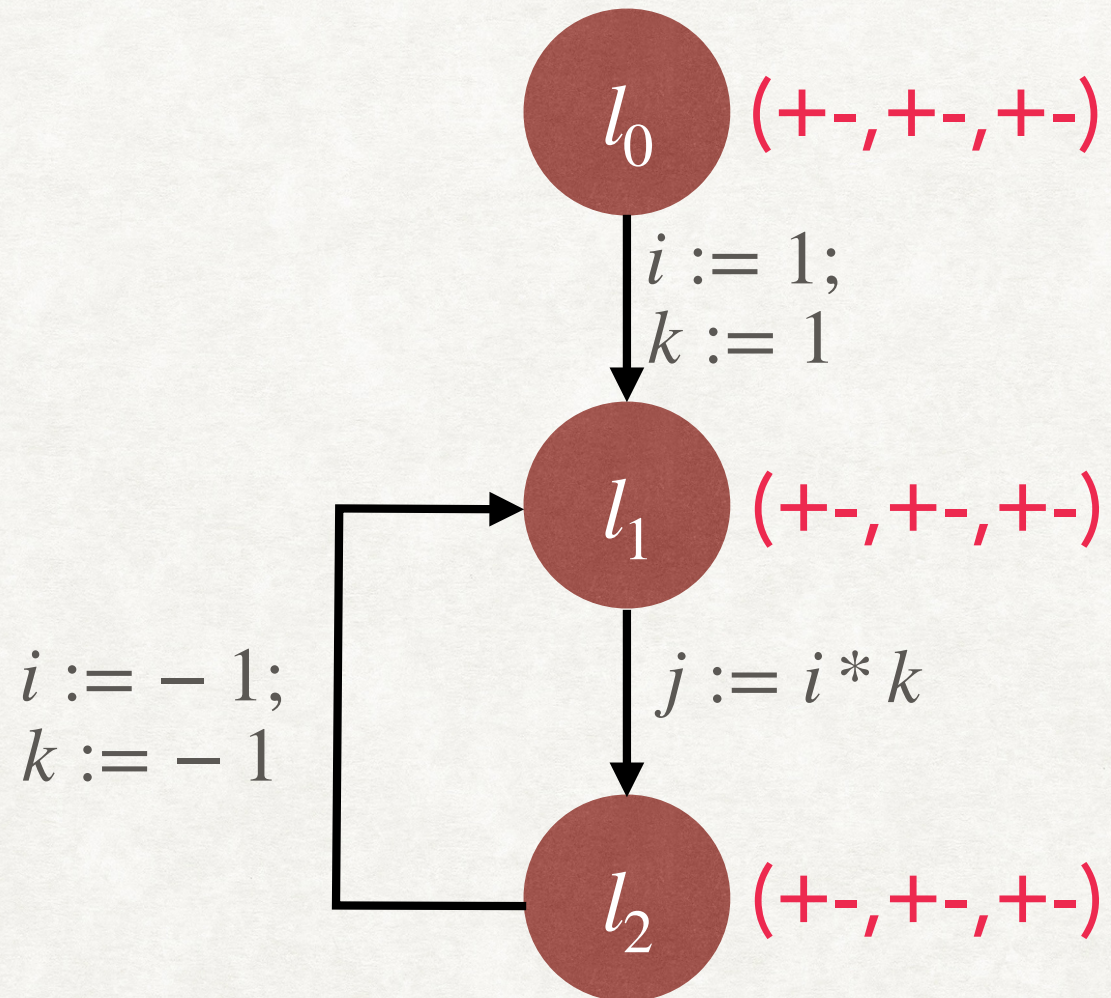


## EXAMPLE - KILDALL'S ALGORITHM



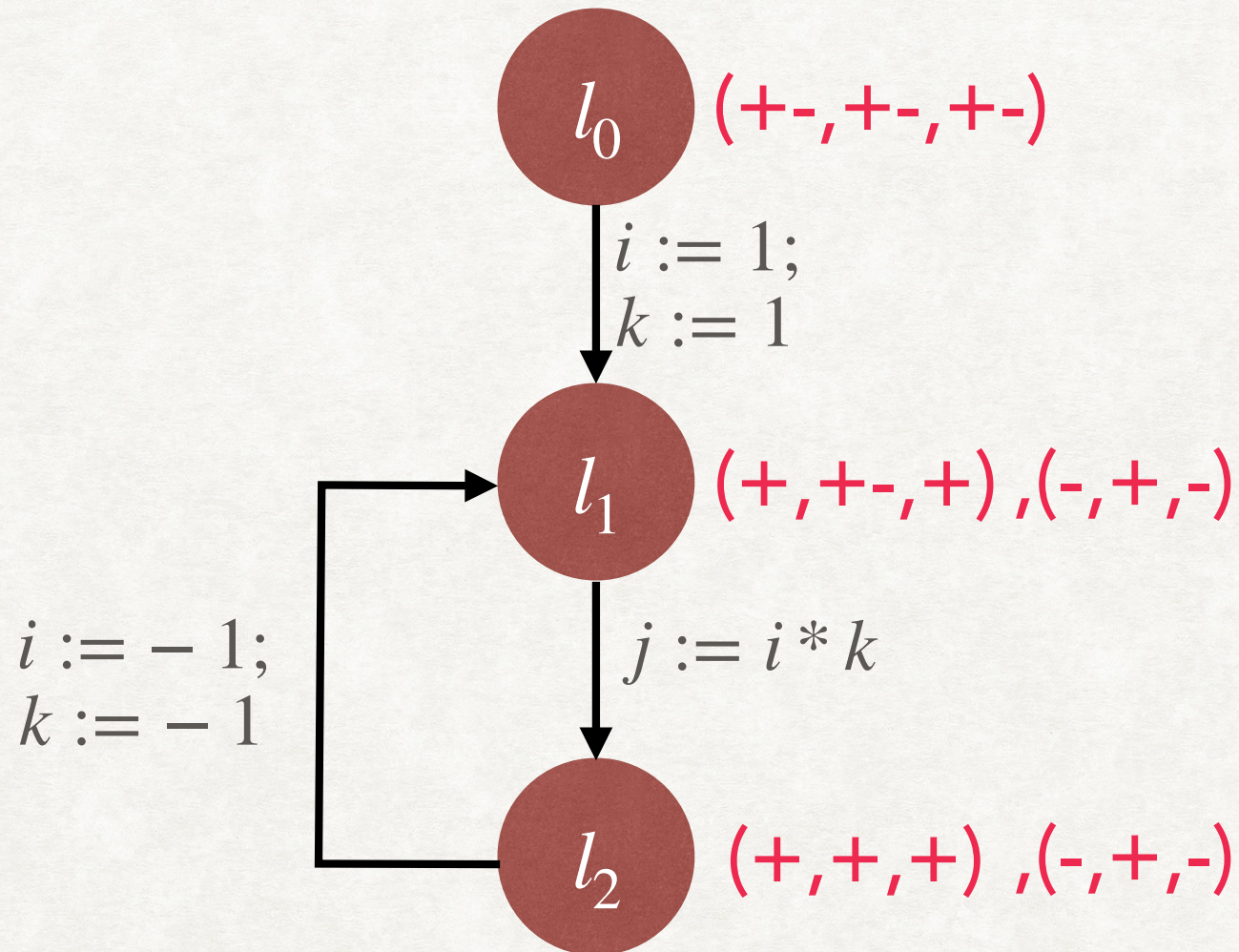


## EXAMPLE - KILDALL'S ALGORITHM



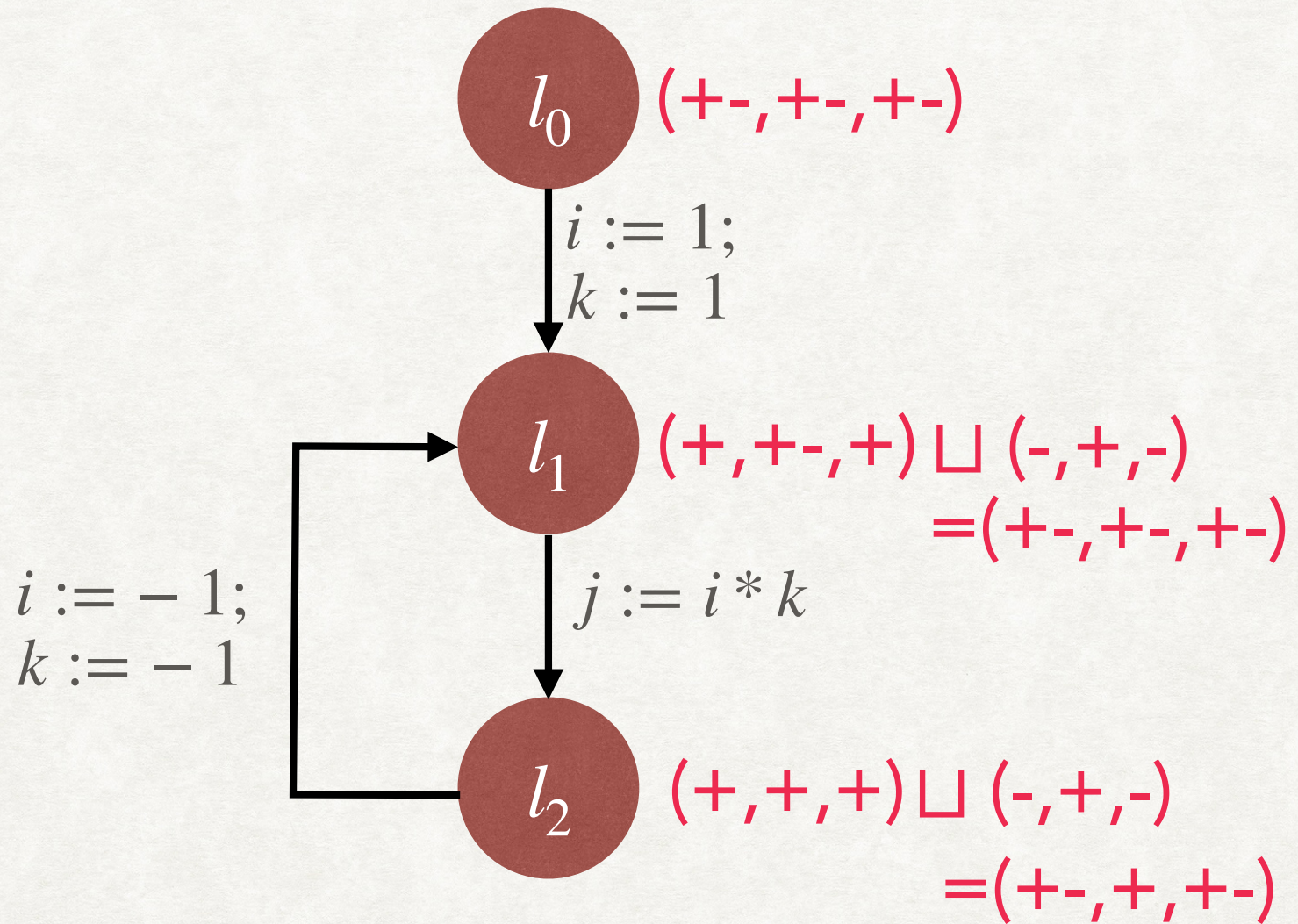


## EXAMPLE - ABSTRACT JOP



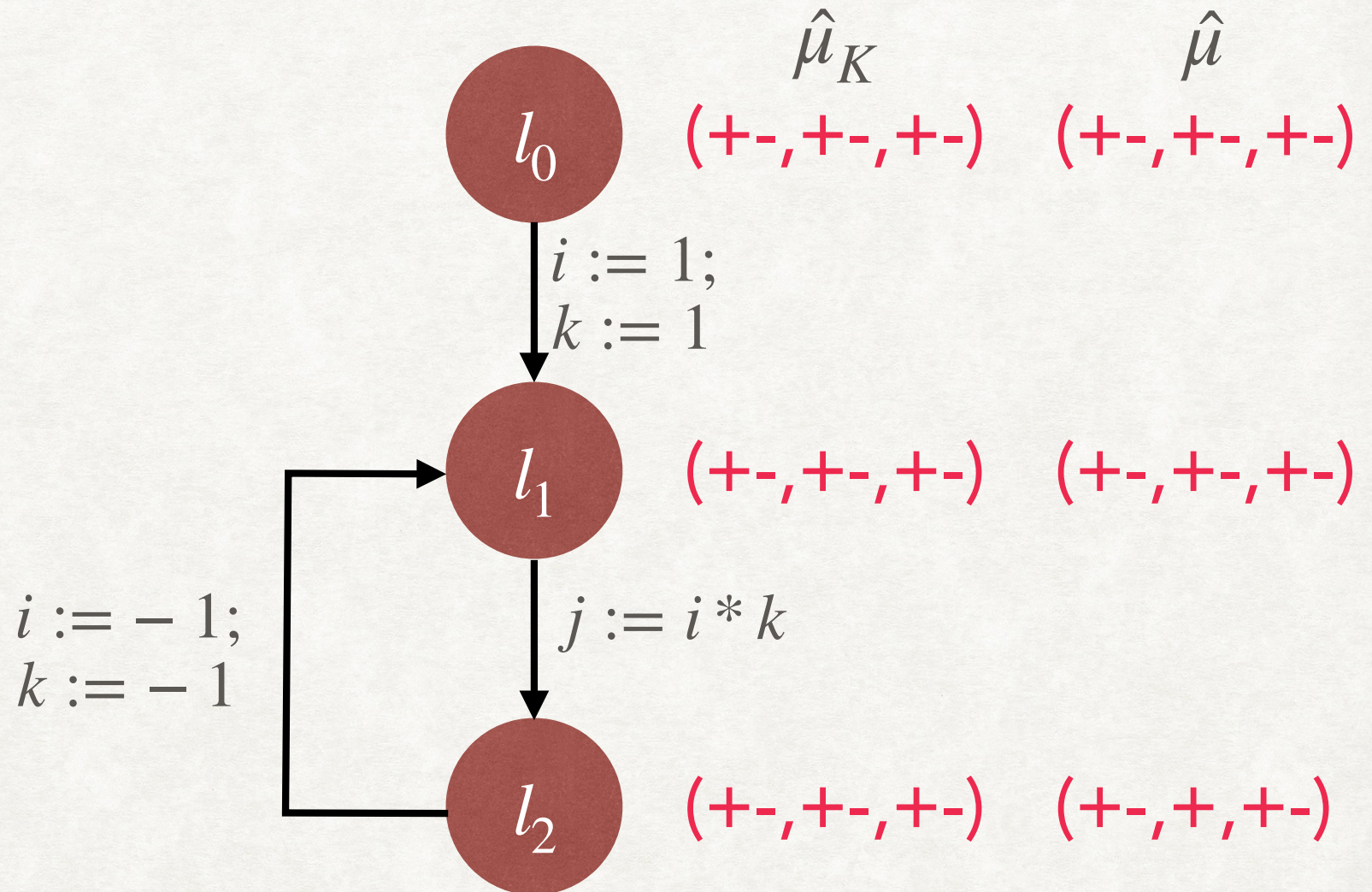


## EXAMPLE - ABSTRACT JOP





# EXAMPLE - KILDALL VS ABSTRACT JOP



$\hat{\mu}_K \neq \hat{\mu}$  : This is because Kildall's Algorithm applies join eagerly  
 We will prove that  $\hat{\mu}_K \geq \hat{\mu}$



# PROPERTIES OF KILDALL'S ALGORITHM

1. The values computed using Kildall's algorithm are an over-approximation of the abstract JOP, if the underlying AI framework is monotonic.
2. In general, Kildall's algorithm computes the least solution to a system of equations.
3. If the abstract domain satisfies the ascending chain condition, then Kildall's algorithm is guaranteed to terminate.

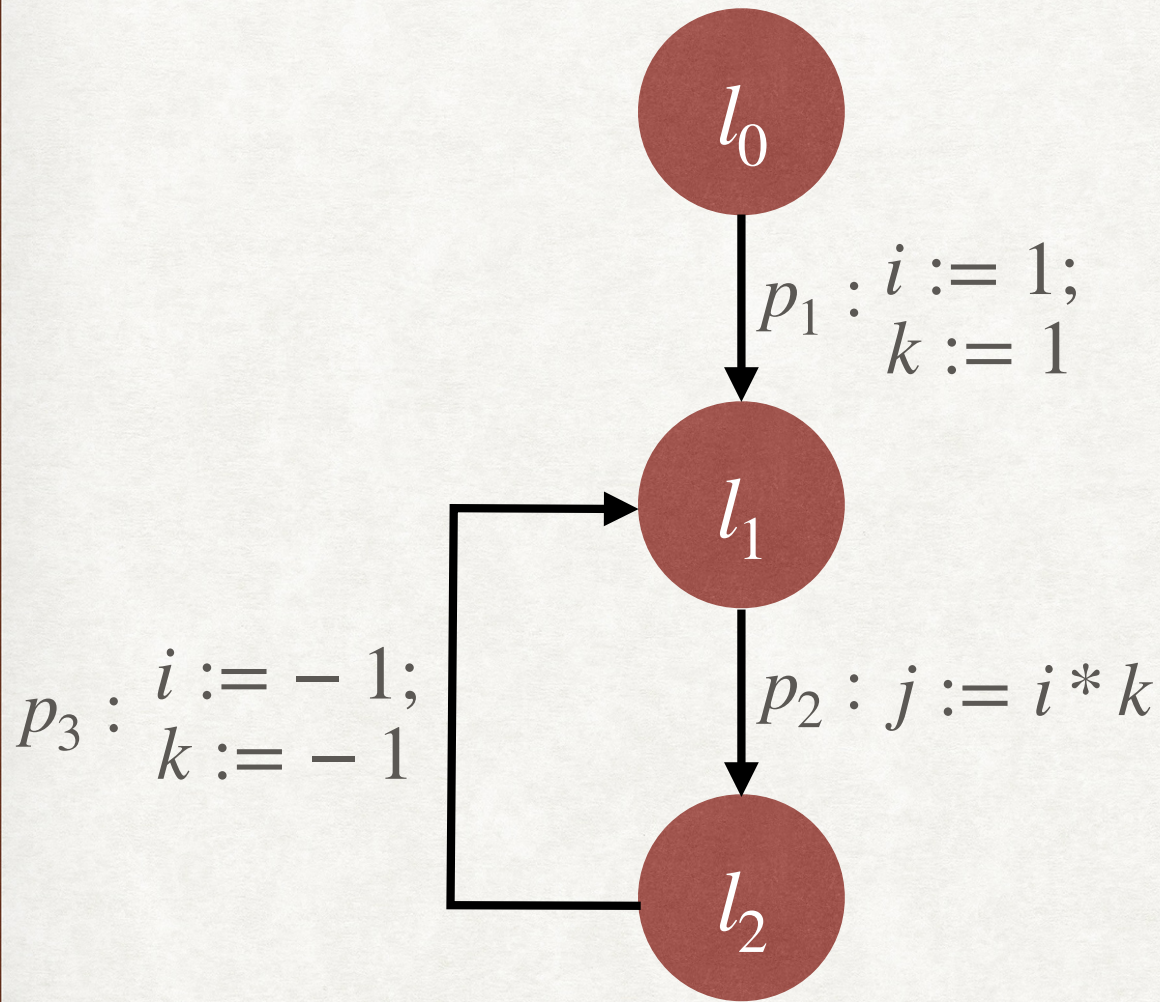


# DATAFLOW EQUATIONS

- Program  $\Gamma_c = (V, L, l_0, l_e, T)$  induces a system of data flow equations:
  - $X_{l_0} = d_0$
  - For all other locations  $l \in L \setminus \{l_0\}$ ,  $X_l = \bigsqcup_{(l', c, l) \in T} \hat{f}_c(X_{l'})$
- For collecting semantics, replace  $d_0$  with  $c_0$ ,  $\sqcup$  with  $\cup$  and  $\hat{f}_c$  with  $f_c$ .



## EXAMPLE - DATAFLOW EQUATIONS



$$X_{l_0} = d_0$$

$$X_{l_1} = \hat{f}_{p_1}(X_{l_0}) \sqcup \hat{f}_{p_3}(X_{l_2})$$

$$X_{l_2} = \hat{f}_{p_2}(X_{l_1})$$



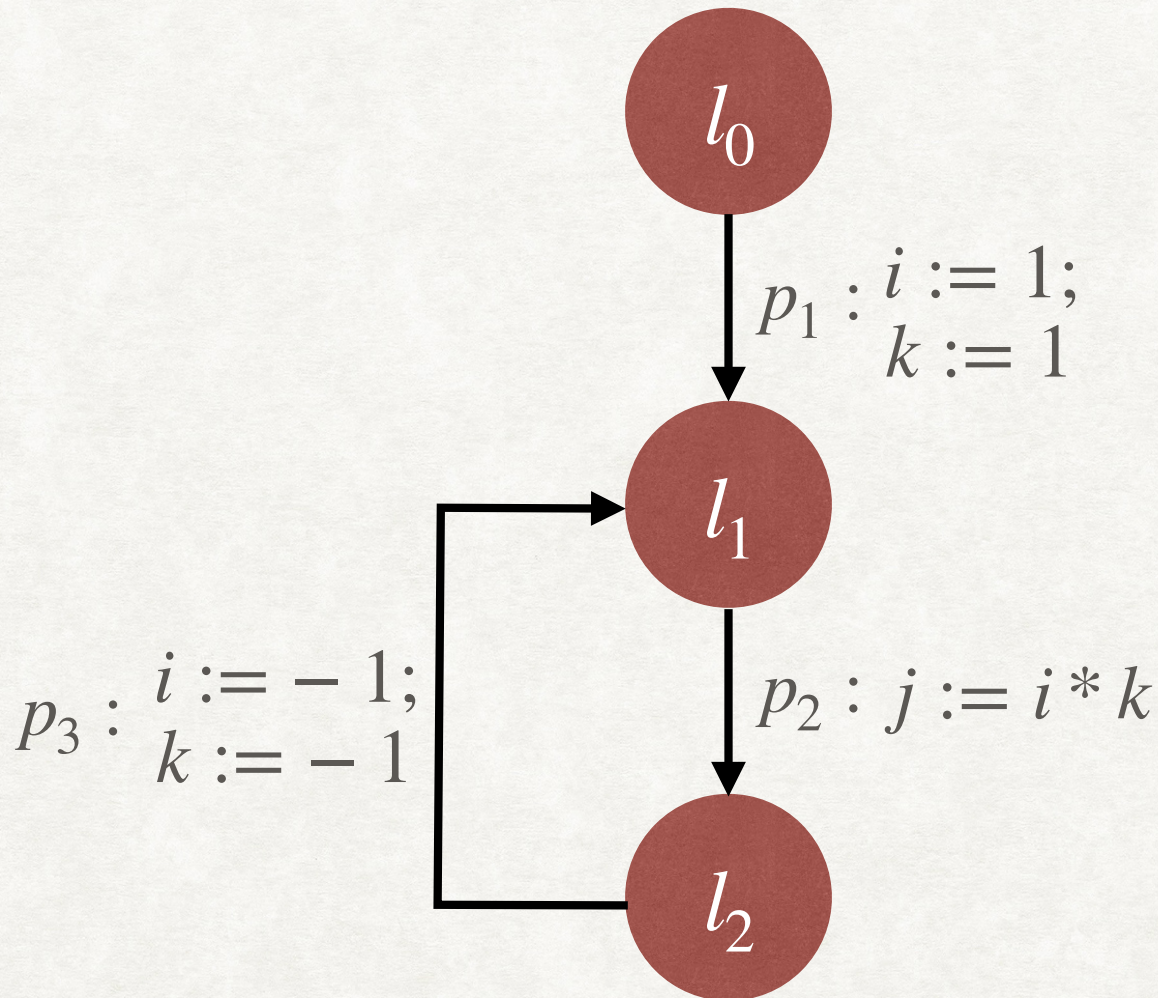
# DATAFLOW EQUATIONS AS FUNCTION

- Consider the 'vectorised' lattice  $(\bar{D}, \bar{\leq})$ , where  $\bar{D} = D^{|L|}$ .
  - $\bar{d} \bar{\leq} \bar{d}' \Leftrightarrow \forall l \in L. \bar{d}(l) \leq \bar{d}'(l)$
  - **Homework:** Prove that if  $(D, \leq)$  is a complete lattice, then  $(\bar{D}, \bar{\leq})$  is also a complete lattice.
- We can view the data flow equations as a function  $\bar{f}: \bar{D} \rightarrow \bar{D}$ :
  - $(\bar{f}(\bar{d}))(l_0) = d_0$
  - $(\bar{f}(\bar{d}))(l) = \bigsqcup_{(l', c, l) \in T} \hat{f}_c(\bar{d}(l'))$



# DATAFLOW EQUATIONS AS FUNCTION

## EXAMPLE



Notice that a  
fixpoint of  $\bar{f}$  is a  
solution to the  
dataflow equations

$$\bar{f}(d_{l_0}, d_{l_1}, d_{l_2}) = (d_0, \hat{f}_{p_1}(d_{l_0}) \sqcup \hat{f}_{p_3}(d_{l_2}), \hat{f}_{p_2}(d_{l_1}))$$



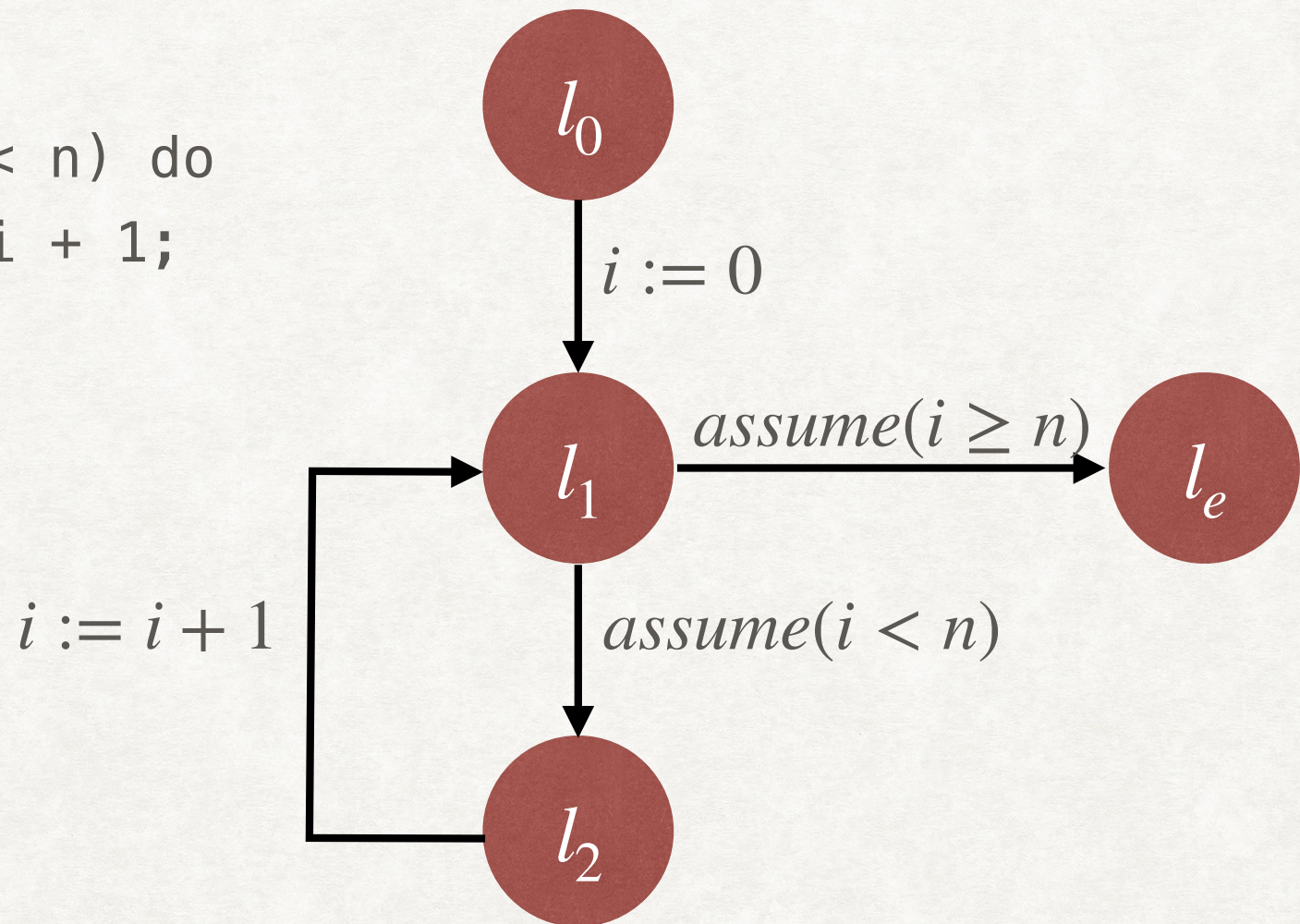
# DATAFLOW EQUATIONS AS FUNCTION

- If every abstract transfer function  $\hat{f} : D \rightarrow D$  is monotonic, then the function  $\bar{f} : \bar{D} \rightarrow \bar{D}$  is also monotonic.
  - **Homework:** Prove this.
- We have a monotonic function  $\bar{f}$  on a complete lattice  $\bar{D}$ . Hence, we can apply Knaster-Tarski fixpoint theorem.
- The least fixpoint  $lfp(\bar{f})$  exists, and is in fact the least solution to the dataflow equations.
- We will show that Kildall's algorithm actually computes  $lfp(\bar{f})$ .
- Note that we can also use the sequence  $\perp, \bar{f}(\perp), \bar{f}^2(\perp), \dots$  to compute  $lfp(\bar{f})$ .
  - This method is also called Kleene Iteration.



## LFP INTRODUCES THE LEAST OVER APPROXIMATION: EXAMPLE

```
i := 0;  
while(i < n) do  
  i := i + 1;
```



$(+ -, +, +, +)$  is a solution to the data flow equations,  
And  $(+ -, + -, + -, + -)$  is also another solution



# ABSTRACT JOP $\leq lfp(\bar{f})$ FOR MONOTONIC AI FRAMEWORK

## PROOF

- Given AI  $(D, \leq, \alpha, \gamma, \hat{F}_D)$ , if all functions in  $\hat{F}_D$  are monotonic, then Abstract JOP  $\leq lfp(\bar{f})$ .

**Proof:** Abstract JOP  $\hat{\mu}(l) = \bigsqcup_{\pi \in \Pi_l} \hat{f}_\pi(d_0)$

Let  $lfp(\bar{f}) = \bar{d}$ . We have to show that  $\forall l \in L. \hat{\mu}(l) \leq \bar{d}(l)$ .

We will show that for all locations  $l$ , all paths  $\pi \in \Pi_l$ ,  $\hat{f}_\pi(d_0) \leq \bar{d}(l)$ .

This proves the required result. Why?

We will use induction on length of the paths.

**Base Case:** Paths  $\pi$  of length 0 are empty and end at  $l_0$ . Hence,  $\hat{f}_\pi(d_0) = d_0$ .

Since  $\bar{f}(\bar{d}) = \bar{d}$  and  $(\bar{f}(\bar{d}))(l_0) = d_0$ , we have  $\bar{d}(l_0) = d_0$ .

Thus,  $\hat{f}_\pi(d_0) \leq \bar{d}(l_0)$

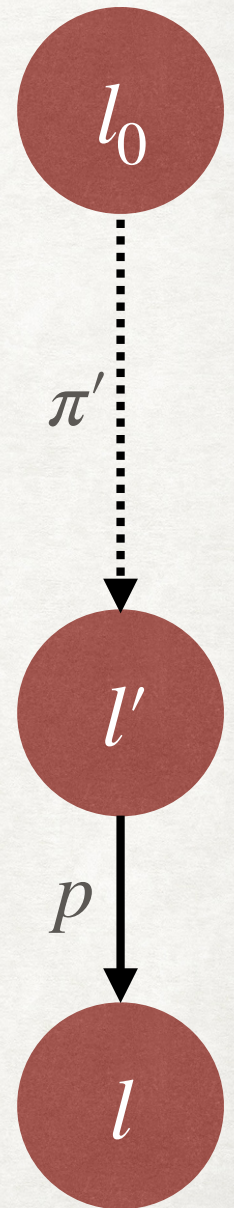


# ABSTRACT JOP $\leq lfp(\bar{f})$ FOR MONOTONIC AI FRAMEWORK

## PROOF

**Inductive Case:** Assume that the claim holds for all paths of length  $n$ .

Consider a path  $\pi$  of length  $n + 1$  ending at location  $l$ .





# ABSTRACT JOP $\leq lfp(\bar{f})$ FOR MONOTONIC AI FRAMEWORK

## PROOF

**Inductive Case:** Assume that the claim holds for all paths of length  $n$ .

Consider a path  $\pi$  of length  $n + 1$  ending at location  $l$ .

Let  $\pi'$  be the prefix of the path of length  $n$ , ending at location  $l'$ .

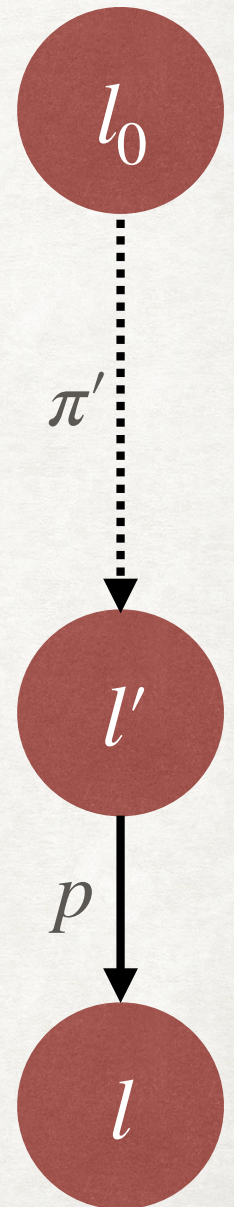
By Inductive Hypothesis,  $\hat{f}_{\pi'}(d_0) \leq \bar{d}(l')$ .

Since  $\hat{f}_p$  is monotonic,  $\hat{f}_p(\hat{f}_{\pi'}(d_0)) \leq \hat{f}_p(\bar{d}(l'))$ .

Hence,  $\hat{f}_{\pi}(d_0) \leq \hat{f}_p(\bar{d}(l'))$ .

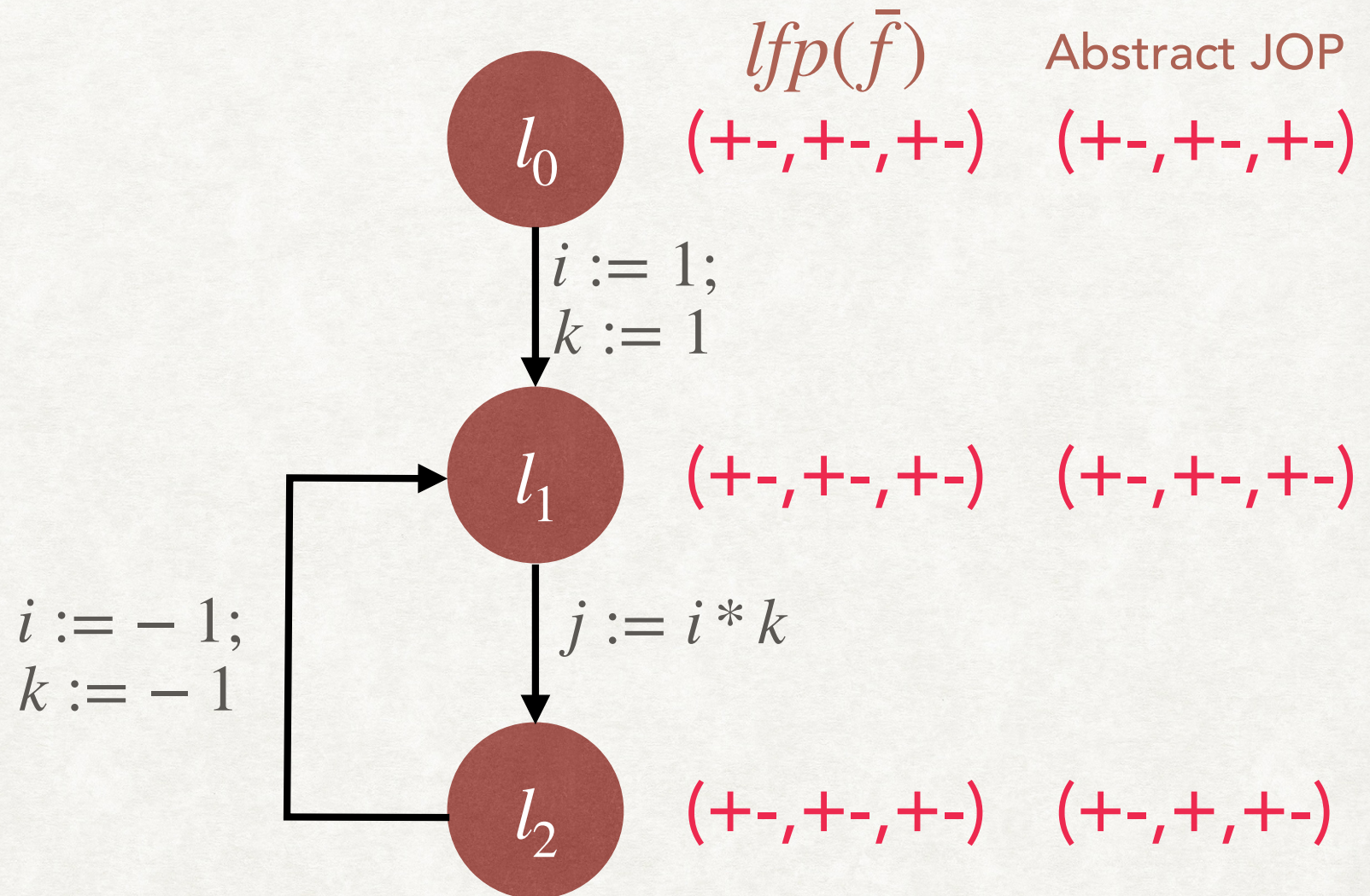
Now  $\bar{f}(\bar{d}) = \bar{d}$ . Hence,  $\bar{d}(l) = \bigsqcup_{(l',p,l) \in T} \hat{f}_p(\bar{d}(l'))$ .

Hence,  $\hat{f}_p(\bar{d}(l')) \leq \bar{d}(l)$ . Thus,  $\hat{f}_{\pi}(d_0) \leq \bar{d}(l)$ .





# EXAMPLE - LFP VS ABSTRACT JOP



$$\bar{f}(d_{l_0}, d_{l_1}, d_{l_2}) = (d_0, \hat{f}_{p_1}(d_{l_0}) \sqcup \hat{f}_{p_3}(d_{l_2}), \hat{f}_{p_2}(d_{l_1}))$$



# DISTRIBUTIVE AND INFINITELY DISTRIBUTIVE FUNCTIONS

- Given two posets  $(D_1, \leq_1)$  and  $(D_2, \leq_2)$ , function  $f: D_1 \rightarrow D_2$  is called distributive if for  $x, y \in D_1$  such that  $x \sqcup_1 y$  exists, then  $f(x) \sqcup_2 f(y)$  also exists, and  $f(x \sqcup_1 y) = f(x) \sqcup_2 f(y)$ .
- Given two posets  $(D_1, \leq_1)$  and  $(D_2, \leq_2)$ , function  $f: D_1 \rightarrow D_2$  is called infinitely distributive if for all  $X \subseteq D_1$  such that  $\sqcup_1 X$  exists, then  $\sqcup_2 f(X)$  also exists, and  $\sqcup_2 f(X) = f(\sqcup_1 X)$ .
- **Exercise:** If  $f$  is distributive, then  $f$  is also monotonic.