

# Week 11A + 11B Lecture Notes: Lattices, Fixed Points, Sign Analysis, and Liveness

These two topics are really the same “big idea” in different outfits:

**We want to compute a fact about every program point** (after each instruction), even with branches/loops, **by solving a system of equations**.

Because branches/loops create circular dependencies, we solve the equations by finding a **least fixed point** of a **monotonic function** over a **complete lattice**.

## 0) The Big Picture (what are we trying to achieve?)

Compilers (and static analyzers) often need *guaranteed-safe approximations* of runtime facts:

- **Sign analysis**: “Could `x` be negative here?” (e.g., is `sqrt(x)` safe?)
- **Liveness analysis**: “Will this variable’s current value be used again?”  
(needed for **register allocation** and **dead code elimination**)

Both are examples of **dataflow analysis**:

- pick an *abstract domain* (a lattice),
- define equations for each program point,
- solve them as a **fixed point**.

## 1) Warm-up: Why plain “type inference” isn’t enough (Sign as Type?)

### Example idea

You might think: “let’s refine `int` into `posint/negint/zero` and infer signs like types.”

But the slides point out two killer limitations of type inference in this setting:

1. **Flow-insensitive**: it tries to assign *one* “type” to a variable for the whole program.
2. **Path-insensitive**: it can’t distinguish different control-flow paths (esp. loops/branches).

So instead, we switch from “types” to **abstract interpretation**:

we compute a *sign fact per program point*, and we merge paths conservatively.

## 2) Lattice crash course (the math toolbox)

### 2.1 Partial order ( $\sqsubseteq$ )

Think “is at least as informative as”.

If  $a \sqsubseteq b$ , read: **a is more precise (or equal precision) than b**.

### 2.2 Join ( $\sqcup$ ) and meet ( $\sqcap$ )

- **Join**  $x \sqcup y$ : least upper bound (LUB) = “merge information conservatively”
- **Meet**  $x \sqcap y$ : greatest lower bound (GLB) = “common information”

Analogy:

- Join is like combining witness statements: you only keep what’s safe to claim.
- Meet is like taking intersection of guarantees.

### 2.3 Complete lattice

A lattice is **complete** if *every subset* has a join and meet.

This matters because fixed point theorems rely on it.

### 2.4 Common lattices used in these slides

- **Powerset lattice**  $(P(A), \sqsubseteq)$   $\leftarrow$  used for liveness (sets of variables)
- **Product lattice**  $L_1 \times \dots \times L_n$   $\leftarrow$  tuple of facts
- **Map lattice**  $A \rightarrow L$   $\leftarrow$  mapping variables to abstract values (used for sign states)

## 3) Sign Analysis (Week 11A)

### 3.1 What is the abstract domain?

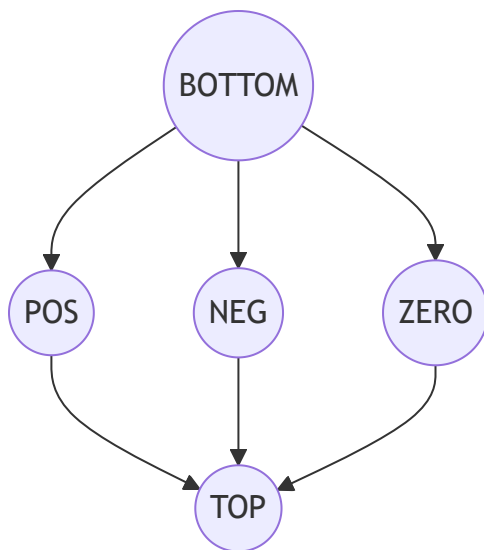
We don't track exact integers, we track **signs**:

- $+$  : any positive integer
- $-$  : any negative integer
- $0$  : exactly zero
- $\top$  : could be  $+$ ,  $-$ , or  $0$  (unknown / most general)
- $\perp$  : no information / impossible / empty (least element)

Interpretation: each abstract value represents a *set* of concrete values.

### 3.2 The sign lattice

A typical shape:



Key intuition:

- Moving **up** = “less precise, more possible values”
- Moving **down** = “more precise”

### 3.3 Abstract operations (++ , --)

To analyze  $x = x + 1$ , we need an abstract version of  $+ 1$ .

The slides define tables like:

- $s1(x) ++ +$  means “current sign of  $x$  plus a positive constant”

- similarly `-- +` for subtracting a positive constant

These tables are conservative:

if you can't be sure, you return `⊤`.

## 3.4 Abstract states

A program state becomes a **map**:

- `State = Var → VarSign`

Example:

- `[x → ⊤, y → 0, t → +]`

This is a **map lattice**: one map is “ $\leq$ ” another if it's  $\leq$  at every variable.

## 3.5 Turning a program into equations

For each labeled instruction `i`, make a state `s_i`.

Each `s_i` is defined from predecessor states + the instruction's transfer rule.

- For straight-line code, this is easy: `s2 = transfer(s1)`.
- For branches/loops, you must merge:  
`s_join = join(predecessor states)`  
 which is typically a pointwise  $\sqcup$  (least upper bound).

## 3.6 Why fixed points appear (loops)

In loops, you get circular equations like:

- `s4` depends on `s9`
- `s9` depends on `s4`

So you can't “solve once”; you must iterate until stable.

# 4) Monotonic functions + Fixed points (Week 11A core engine)

## 4.1 Monotonic function

A function  $f$  is **monotonic** if:

- whenever  $x \sqsubseteq y$ , then  $f(x) \sqsubseteq f(y)$ .

This matters because monotonicity guarantees that iterating from  $\perp$  moves “upward” in a controlled way and eventually stabilizes.

## 4.2 Fixed Point Theorem (why iteration works)

For a complete lattice of finite height:

- every monotonic  $f$  has a **unique least fixed point**  $\text{lfp}(f)$
- and it can be obtained by iterating from  $\perp$  :  
 $\perp, f(\perp), f(f(\perp)), \dots$  until no change

## 4.3 Naive fixed point algorithm (intuition)

1. start with the least element ( $\perp$  / “no info everywhere”)
2. apply  $f$
3. repeat until stable

You can think of it as:

keep propagating facts through the program until nothing new is learned.

## 4.4 How sign analysis fits this pattern

The slides’ “objective” is:

- turn the whole equation system  $(s_0, s_1, \dots, s_n)$  into **one big function**  
 $f((s_0, \dots, s_n)) = (new\_s_0, \dots, new\_s_n)$
- then compute  $\text{lfp}(f)$  by iteration.

## 5) Liveness Analysis (Week 11B)

### 5.1 What does “live” mean?

A variable  $v$  is **live at program point  $i$**  if:

along **some** path from  $i$  to the future, the value currently held in  $v$  might be used before  $v$  is overwritten.

So liveness is a **may** property: “may be needed”.

### 5.2 Abstract domain = sets of variables

At each point  $i$ , liveness state is a set:

- $s_i \subseteq V$  where  $V$  is the set of variables

Domain is  $P(V)$  which forms a **powerset lattice**:

- $\perp = \{\}$  (no variables live)
- $\top = V$  (everything live)
- ordering by subset  $\subseteq$

### 5.3 Backward analysis

Unlike sign analysis (forward), liveness is **backward**:

- current liveness depends on what successors will need.

So at point  $i$ :

- first combine successor facts (join),
- then apply the instruction's effect.

### 5.4 Join for liveness (may analysis)

Because it's “may be live”, the join is typically **union**:

- if a variable is live on *either* successor path, it is live here.

## 5.5 Transfer rules (the core intuition)

For an assignment:

- `t <- expr`
  - `t` becomes **not live** immediately *before* the assignment (it's overwritten)
  - variables used in `expr` become **live** (their values are needed)

So conceptually:

- $IN = (OUT - \{t\}) \cup \text{uses}(\text{expr})$

Other instructions:

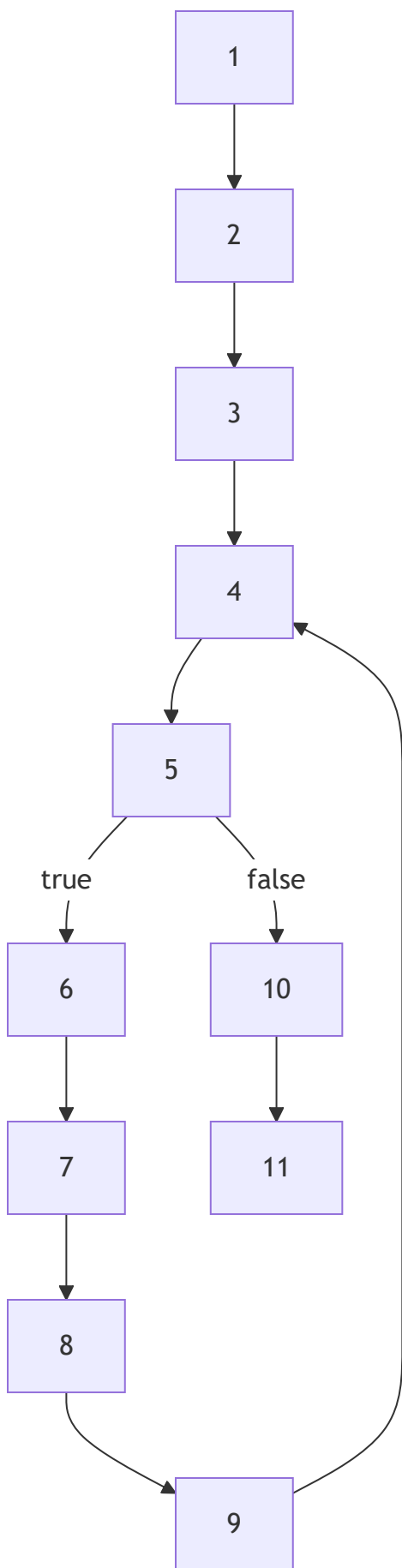
- conditional branch uses the condition variable  $\rightarrow$  it becomes live
- return uses the returned variable(s)

## 6) Worked Example (from the slides): PA1 liveness

Program (labels 1..11):

```
1: x <- input
2: y <- 0
3: s <- 0
4: b <- y < x
5: ifn b goto 10
6: y <- y + 1
7: t <- s
8: s <- s + y
9: goto 4
10: rret <- s
11: ret
```

CFG sketch:








The slides construct a single monotonic function  $f_1$  over the tuple  $(s_{11}, s_{10}, \dots, s_1)$  and then apply fixed point iteration.

Final liveness sets shown in the slides (interpretation: “live before that line”):

- $s_1 = \{\text{input}\}$
- $s_2 = \{x\}$
- $s_3 = \{y, x\}$
- $s_4 = \{y, x, s\}$
- $s_5 = \{y, x, s, b\}$
- $s_6 = \{y, x, s\}$
- $s_7 = \{y, x, s\}$
- $s_8 = \{y, x, s\}$
- $s_9 = \{y, x, s\}$
- $s_{10} = \{s\}$
- $s_{11} = \{\}$

Quick sanity checks (how to “feel” the result):

- At 11: `ret` , nothing is needed anymore  $\rightarrow \{\}$  
- Just before 10: `rret <- s` , we must have `s` available  $\rightarrow \{s\}$  
- Inside the loop (around 4–9), `x` , `y` , `s` keep being needed for comparisons/updates  $\rightarrow \{x, y, s\}$  

## 7) Forward vs Backward, May vs Must (Week 11B wrap-up)

### Forward vs backward

- **Sign analysis**: forward (depends on predecessors)
- **Liveness**: backward (depends on successors)

### May vs must (with powerset domains)

- **May analysis**: over-approximation, join with **union** ( $\cup$ ) (liveness)
- **Must analysis**: under-approximation, join with **intersection** ( $\cap$ )

The slides present **deadness analysis** as the dual of liveness:

- instead of “may be live”, we track “must be dead”
- join becomes meet-like (intersection-style), and the transfer rules flip accordingly.

## 8) Takeaway summary

- Pick a lattice that represents the *kind* of information you want.
  - signs: 5-point sign lattice, states are maps  $\text{Var} \rightarrow \text{VarSign}$
  - liveness: powerset lattice  $\mathcal{P}(V)$
- Define transfer equations at each instruction.
- Merges/loops make it recursive.
- Turn the whole system into a monotonic function  $f$ .
- Compute  $\text{lfp}(f)$  by iterating from the least element until stable.

That’s the core pattern you’ll keep seeing in compiler analyses.