## Static Analysis and Dataflow Analysis

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But wait? Isn't that impossible?

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It's a classic paradox!

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Striking the right balance is key to a useful analysis

## Approximation

Modeled program behaviors


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Consider some behaviors possible when they are not.

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Ignore some behaviors that are possible.

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- Dynamic Analysis
- Analyzed $\subseteq$ Feasible
- As \# tests $\uparrow$, Analyzed $\rightarrow$ Feasible



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- As \# tests $\uparrow$, Analyzed $\rightarrow$ Feasible
- Static Analysis
- Feasible $\subseteq$ Analyzed
- As infeasible paths $\downarrow$, Analyzed $\rightarrow$ Feasible
- The two areas complement each other
- Static analysis can help generate useful tests
- Dynamic analysis can help identify infeasibility


## Abstract Interpretation

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$$
\begin{aligned}
& \text { concrete }(x)=5 \mapsto \operatorname{abstract}(x)=+ \\
& \text { concrete(y) = -3 } \mapsto \text { abstract }(\mathrm{y})=- \\
& \text { concrete(z) }=0 \mapsto \operatorname{abstract}(z)=0
\end{aligned}
$$

Combines sets of the concrete domain

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- ...
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This type of approximation is called abstract interpretation.

## Abstract Interpretation


6) print (sum)
7) print(i)

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## 6) print (sum) 7) print(i)

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Does the process ever end?

## Abstract Interpretation



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- Abstract state can only move up lattice at a statement



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Why does this specific example terminate?


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- The transfer function
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- But in theory a lattice need not be finite!
(ranges/intervals, linear constraints, ...)


## Abstract Interpretation

- Guarantee termination by carefully choosing
- The abstract domain
- The transfer function
- For basic analyses, use a monotone framework
- But in theory a lattice need not be finite!
- Widening operators can still make it feasible (e.g., heuristically raise to T )


## Abstract Interpretation

- What properties should a good abstraction have?


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\mathbf{x}_{\mathrm{A}} \prod \mathrm{x}_{\mathrm{B}}=?
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Will it always terminate?

## Dataflow Analysis

- Note: need to model program state before and after each statement
- Proper ordering \& a work list algorithm improves the efficiency


## Worklist Algorithms

```
work = nodes()
state(n) = \perp\forall n G nodes()
while work = Ø:
    unit = take(work)
    old = state(unit)
    before = Пstate(p)
        | \in preds(unit)
    new = transfer(before, unit)
    if old # after:
        work = work U succs(unit)
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```

work: | 1 | 2 | 3 | 4 |
| :--- | :--- | :--- | :--- |

state:

(3 $\mapsto \perp)$
( $4 \mapsto \perp$ )

\}

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unit $=1$

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work: | 2 | 3 | 4 |
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## Worklist Algorithms



work: | 2 | 3 | 4 |
| :--- | :--- | :--- |

state: $\left\{\begin{array}{l}(\boxed{1} \mapsto \perp) \\ (\boxed{2} \mapsto \perp)\end{array}\right.$
$(3 \mapsto \perp)$
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\begin{tabular}{ll} 
unit \(=\) & 1 \\
old \(=\) & \(\perp\) \\
new & \(\underset{i}{\text { sum } \rightarrow 0}\) \\
\(=\) &
\end{tabular}
    if old # after:
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\left\{\begin{array}{ll}
\left(\sqrt{1} \mapsto \operatorname{sum} \rightarrow 0_{i \rightarrow+}\right. & (\sqrt{3} \mapsto \perp) \\
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\end{array}\right\}
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work: $\quad$| $-\cdots$ | 4 |
| :--- | :--- | :--- | :--- |

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4,3 were added back to the list
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- Control flow
- Many different paths are summarized together
- Abstraction
- Deliberately throwing away information
- Granularity of program state affects correlations across variables


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For one path $p: f_{p}(\perp)=f_{n}\left(f_{n-1}\left(\ldots f_{1}\left(f_{0}(\perp)\right)\right)\right)$

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For all paths $\mathrm{p}: \quad \Pi \mathrm{pfp}(\perp)$

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- When $f()$ is distributive, MFP=MOP

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\mathrm{f}(\mathrm{x} \sqcap \mathrm{y} \sqcap \mathrm{z})=\mathrm{f}(\mathrm{x}) \sqcap \mathrm{f}(\mathrm{y}) \sqcap \mathrm{f}(\mathrm{z})
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- This applies to an important class of problems called bitvector frameworks.


## Bitvector Frameworks

- When the property concerns subsets of a finite set, the abstract domain \& lattice are easy:
- Concrete: D = \{a, b, c, d, ... $\}$
- Abstract: $\wp(D)=\{\{ \},\{a\},\{b\}, \ldots,\{a, b\},\{a, c\}, \ldots\}$
- Lattice: Defined by subset relation:



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- Abstract: $\wp(D)=\{\{ \},\{a\},\{b\}, \ldots,\{a, b\},\{a, c\}, \ldots\}$
- Lattice: Defined by subset relation: What would the meet operator be?



## Bitvector Frameworks

- Why is this convenient?
- Hint: bitvector frameworks


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- Why is this convenient?
- Hint: bitvector frameworks
$-X=\{a, b\}, Y=\{c, d\} \rightarrow X \sqcup Y=\{a, b\} \cup\{c, d\}=\{a, b, c, d\}$
- We can implement the abstract state using efficient bitvectors!


## Effect of Approximation

- If approximation yields imprecise results, why do we do it?


## Recap: Dataflow Analysis

Analyze complex behavior with approximation:

- Abstract domain: e.g. $\{-, 0,+\} \cup\{T, \perp\}$
- Transfer functions: -++ $\rightarrow$ T
- Bounded domain lattice height:
- Concern for false + \& -



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Implementation:

- Computing using work lists

- Speeding up by sorting CFG nodes


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Let's see an example

## File Policy Analysis

Goal: Identify potential misuses of open/closed files

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What should our design actually be?

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- Transfer functions?
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What should our design actually be?

- Abstract domain?
- Transfer functions?
- Lattice?
[DEMO]


## Flow Insensitive Analysis

- Saw flow sensitive analysis
- Modeling state at each statement is expensive
- Scales to functions and small components
- Usually not beyond 1000s of lines without care


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- Saw flow sensitive analysis
- Modeling state at each statement is expensive
- Scales to functions and small components
- Usually not beyond 1000s of lines without care
- Flow insensitive analyses aggregate into a global state
- Better scalability
- Less precision
- "Does this function modify global variable X?"


## Context Sensitive Analyses

- Program behavior may be dependent on the call stack / calling context.
- "If bar() is called by foo(), then it is exception free."
- Can enable more precise interprocedural analyses


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Can you imagine how to solve this? What problems might arise?

## Context Sensitivity

- Recall that we can extract a call graph
- Just as you are doing in your first project!

```
def a():
    b()
    *"
        b()
def b():
```



The behavior of c() could be affected by each "..."

Modeling them can make analysis more precise.

## Context Sensitivity

- Simplest Approach
- Add edges between call sites \& targets
- Perform data flow on this larger graph

```
def main():
    \(x=7\)
    \(r=p(x)\)
    \(x=r\)
    \(z=p(x+10)\)
```

def $p(a):$
if a < 9:
$y=0$
else:
$y=1$

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\text { def } & \text { main }(): \\
x & =7 \\
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$$

main()

$$
\begin{aligned}
& x=7 \\
& \text { call } p(x)
\end{aligned}
$$

$$
\operatorname{def} p(a):
$$

$$
\text { if } a<9:
$$

$$
y=0
$$

```
r = return p(x)
x = r
call p(x+10)
```



$$
z=\text { return } p(x+10)
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$r=7$

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main()
$x=7$
call $p(x)$

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- How could we address it?


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- Make a copy of the function at each call site


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- Make a copy of the function at each call site
- What problems arise?
- What other strategies can we use?


## Context Sensitivity

- Solution 3: Make a Copy
- Make one copy of each function per call site


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- Make one copy of each function per call site

1) def main():
2) $a()$
3) $a()$
4) $\operatorname{def} a():$
5) $b()$
6) $\operatorname{def} b():$
7) pass

## Context Sensitivity

- Solution 3: Make a Copy
- Make one copy of each function per call site
return a()
call a()
b()\#\#5
pass

```
```

    a()##3
    ```
```

    a()##3
    ```
```

    a()##3
    call b()

```
```

call b()

```
```

call b()

```
```


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- Solution 3: Make a Copy
- Make one copy of each function per call site

1) def main():
$\begin{array}{ll}\text { 2) } & \text { a() } \\ 3) & a()\end{array}$
2) $\operatorname{def} a():$
3) $b()$
4) $\operatorname{def} b():$
5) pass


So far, so good
b()\#\#5
pass

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- Solution 3: Make a Copy
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3) $b()$
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5) pass

How can we improve it?

## Context Sensitivity

Generalized:

- Make a bounded number of copies


## Context Sensitivity

Generalized:

- Make a bounded number of copies
- Choose a key/feature that determines which copy to use
- Bounded calling context/call stack (call site sensitivity)
- Allocation sites of objects (object sensitivity)


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- Modify the treatment of calls slightly:

On foo(in) with context C :

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If (foo, C) doesn't have a summary, process foo(in) in C and save the result to $S$.

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If the summary $S$ already approximates foo(in), use $S$

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Otherwise, process foo(in) in $C$ and update $S$ with (in $\Pi$ S.in).

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On foo(in) with context C:
If (foo, C) doesn't have a summary, process foo(in) in C and save the result to $S$.
If the summary $S$ already approximates foo(in), use $S$
Otherwise, process foo(in) in $C$ and update $S$ with (in $\Pi$ S.in). If the result changes, reprocess all callers of (foo, C)

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- Set of dataflow facts D is finite
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- Domain and range of transfer functions is $\mathscr{P}(D)$
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- Set of dataflow facts $D$ is finite
- Transfer functions are distributive $[f(x \sqcap y)=f(x) \sqcap f(y)]$
- Domain and range of transfer functions is $\mathscr{P}(\mathrm{D})$
- Lattice ordering is set containment
(Interprocedural Finite Distributive Subsets)


## Context Sensitivity - IFDS

- Consider an undefined variable analysis...

$$
\begin{aligned}
& a=7 \\
& b=a \\
& c=d
\end{aligned}
$$

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x=r \\
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string: ()$_{2}$

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string: ()$_{1} \quad$ unreachable

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string: $\left.\left(I_{1}\right)_{2}\right)_{2}$

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- A fact $f$ holds before a node if $f$ is CFL-Reachable in a language of matched parentheses


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- Does constant propagation fit our definition of IFDS?


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- Does constant propagation fit our definition of IFDS?
- Can you think of ways that it could be made to fit into IFDS?


## Dataflow Configurations

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## Dataflow Configurations

Can be configured in many ways:

- Forward / Backward (e.g. reaching vs liveness)
- May / Must ( $\cup$ vs $\cap$ in lattice when paths П)
- Sensitivity \{Path? Flow? Context?\}

The configuration is ultimately driven by the property/problem of interest

## Static Analysis

- We've already seen a few static analyses:
- Call graph construction
- Points-to graph construction (What are MAY/MUST?)
- Static slicing


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- We've already seen a few static analyses:
- Call graph construction
- Points-to graph construction (What are MAY/MUST?)
- Static slicing
- The choices for approximation are why these analyses are imprecise.


## Other (Traditionally) Static Approaches

- Type based analyses
- Bounded state exploration
- Symbolic execution
- Model checking

Many of these have been integrated into dynamic analyses, as we shall see over the semester.

## Static Analysis Summary

- Considers all possible executions


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- Considers all possible executions
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- Can answer queries like:
- Must my program always ...?
- May my program ever ...?
- Dataflow analysis is one common form of static analysis

