



# CodePeer: Re-Engineering Abstract Interpretation

*for Precise, Scalable Whole-  
Program Verification*

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**Flight Software Workshop  
JHU/Applied Physics Lab  
October 2015**

## Where we are going

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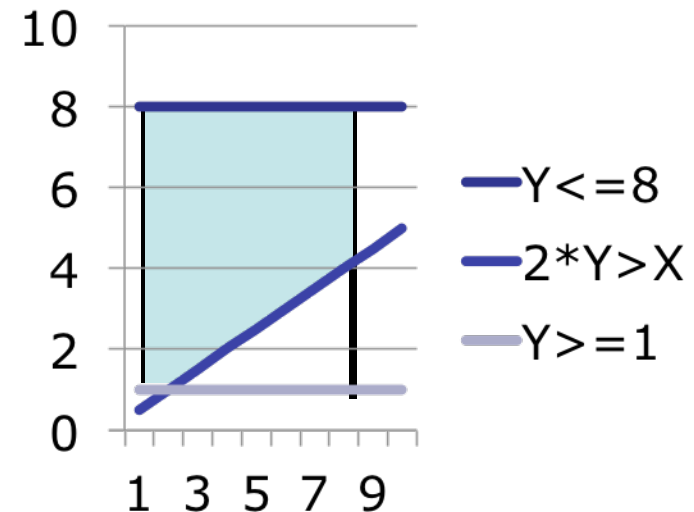
- **Background**
- **Abstract Interpretation and alternatives**
- **Re-engineering with value-number approach in CodePeer Static Analyzer**
- **Inferring pre- and post- conditions in CodePeer**
- **Additional Roles for Abstract-Interpretation-Based Static Analysis**



## What is Abstract Interpretation?

- Approximates the set of possible states of all variables at each point in a program, to allow proofs for safety, security, or correctness.
- Iterates until a fixed-point, then checks for violations.
- *Constructs* the set of possible values, rather than *searching* through them (handles large ranges).
- Represents relationships, e.g.  $2*Y > X$ , using, e.g. polygons/polyhedrons

**X in 1..8, Y in 1..8,  $2*Y > X$ :**



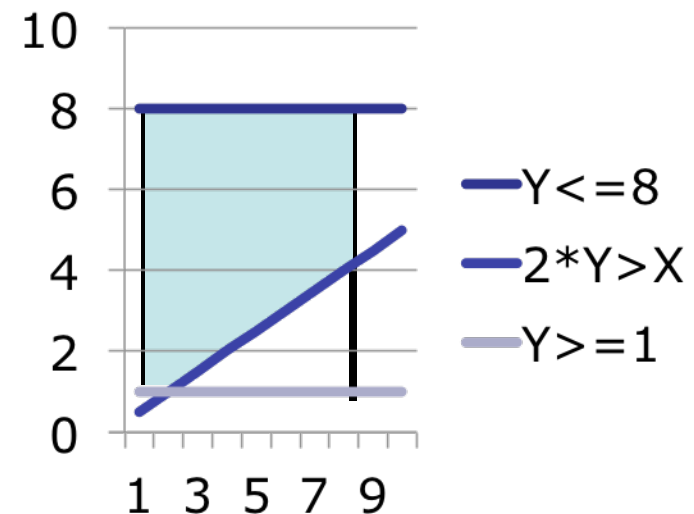
## Alternative Program Proof Techniques

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- **Model Checking (e.g. SPIN model checker)**
  - Searches through state space for states that violate desirable properties
  - State explosion is a challenge; may limit loop iterations
  - Symbolic Model Checking can help
- **Formal Proof (e.g. SPARK 2014 toolset)**
  - Constructs a series of Verification Conditions (VCs) that represent desired safety, security, or correctness properties that should hold at various points in the program
  - Use SMT Solver or equivalent to prove each Verification Condition
  - Use timeout to determine VC cannot be proved
  - Typically relies on programmer to provide pre/postconditions, loop invariants, etc.

## What is the problem with “classic” Abstract Interpretation?

- **Polyhedral representation of relationships between variables is fundamentally limiting (e.g.  $Y > B - X*Z/A$ )**
- **Many approaches exist (courtesy of Wikipedia):**
  - congruence relations on integers
  - convex polyhedra (high computational costs)
  - “octagons”
  - difference-bound matrices
  - linear equalities
- **Other issues:**
  - Initial value set for inputs may require exploring all paths that reach procedure
  - May require a driver or harness to provide realistic input values



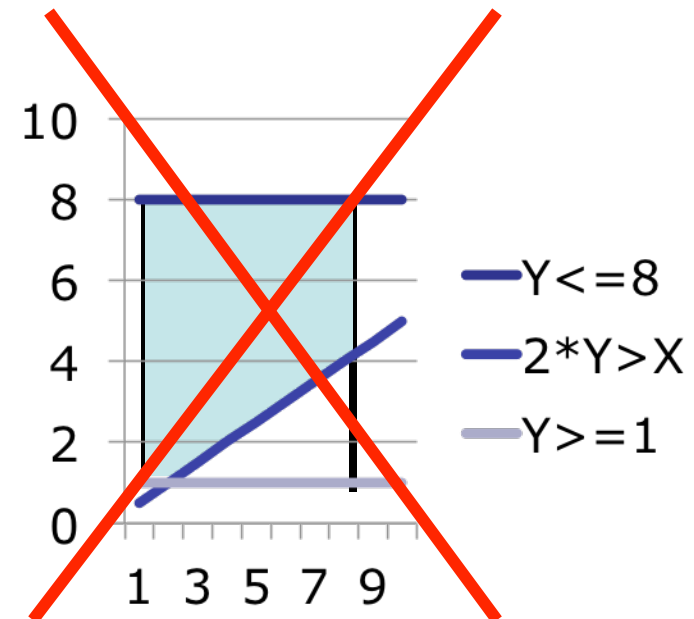
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## Can we Re-Engineer Abstract Interpretation to solve this?

### Basic Trick used in CodePeer Static Analyzer:

- Trick first learned in 1982 in optimizing Ada compiler
- Use "Value Numbers" to represent value of computing *interesting* expressions (e.g. " $Y * 2 - X$ ")
- Associate Value Sets (Vsets) with Value Numbers (VNs)
- Unlike variables, value numbers don't *change* in value over time
  - *but what we know about them does*
- Value set "*shrinks*" when we do a conditional jump
  - e.g. **if**  $Y * 2 > X$  **then** ...
    - in **then** part we know " $Y * 2 - X$ " in  $1 .. +inf$
    - in **else** part we know " $Y * 2 - X$ " in  $-inf .. 0$
- Also *shrinks* when we do a check or an assertion
  - e.g. **assert**  $X + Y > Z \rightarrow "X + Y - Z"$  in  $1 .. +inf$

## How do Value Numbers simplify value-set determination?

- Only need to represent simple sets of integers, floats, or addresses for each value number (no polyhedrons!)
- Relationships between VNs are represented in value-number definition table (aka *computation table*)
- All variables/expressions with same value share a VN
- Each *basic block* and *edge* of Control Flow Graph (CFG) has its own *map* from VN to Value Set
- When one VN's Vset shrinks, we can efficiently *propagate* it to all related VNs in same map

Typical VN=>Vset Map:

$VN1 \Rightarrow \{1..4\}$

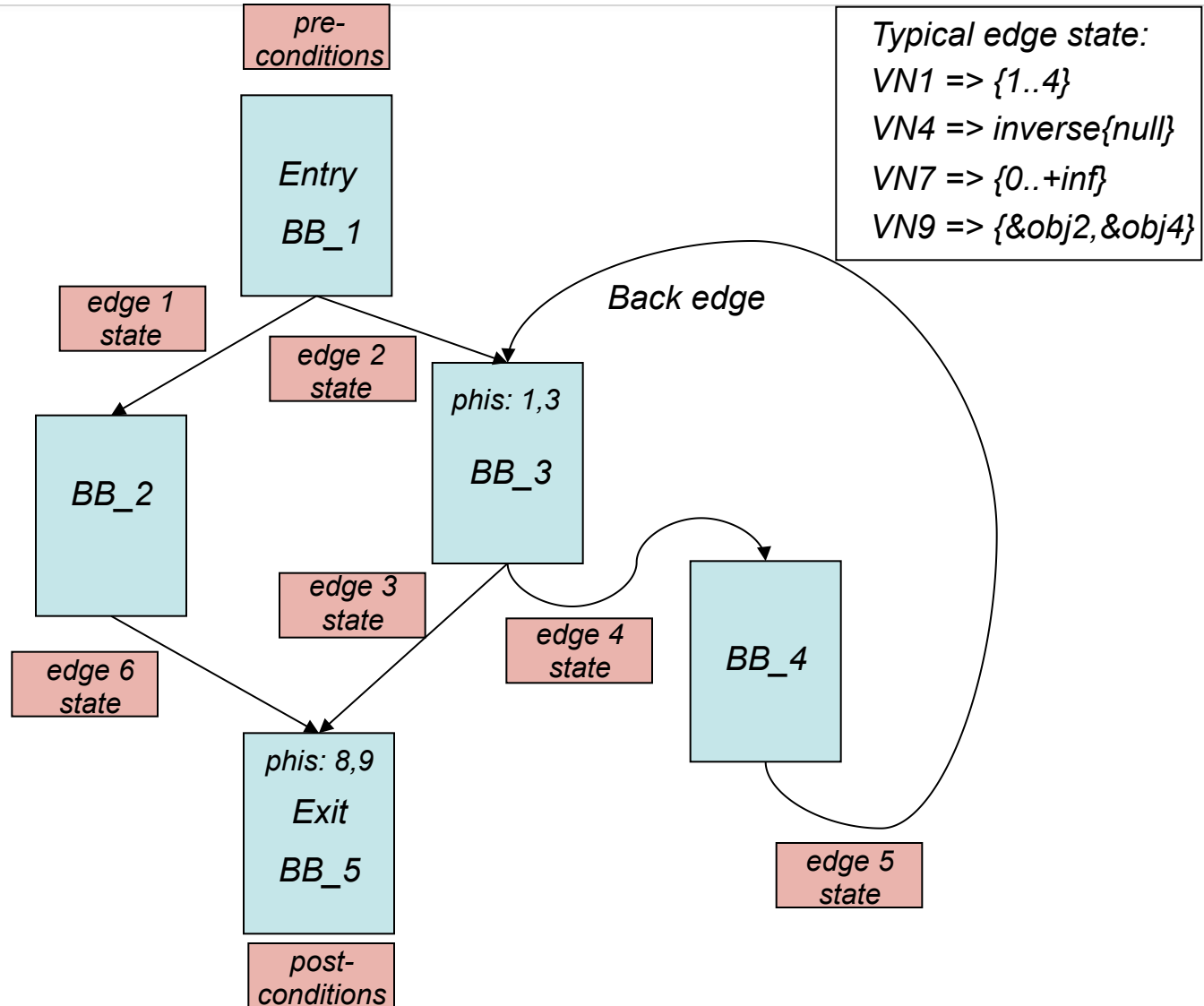
$VN4 \Rightarrow \text{inverse}\{\text{null}\}$

$VN7 \Rightarrow \{0..+\text{inf}\}$

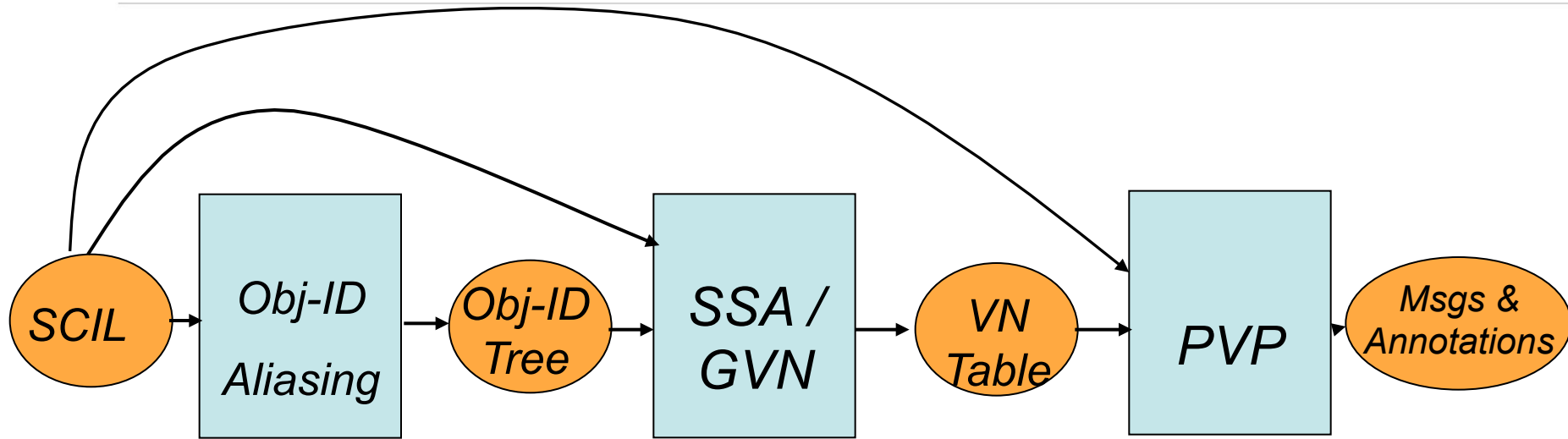
$VN9 \Rightarrow \{\&\text{obj}2, \&\text{obj}4\}$



# Typical Control Flow Graph



## Overall 3-phase Structure of CodePeer Static Analyzer



### *Kinds of Annotations Inferred:*

- SSA { • *Inputs (Live-On-Entry)*
- Obj-ID { • *Outputs (DMODs – Direct Modifications)*  
• *New Objects (Escape Analysis),*
- PVP { • *Preconditions, Postconditions, Presumptions*

## Static Single Assignment/Global Value Numbering (SSA/GVN) phase

*Dominator  
Tree & Loop  
Identification*

*Obj-ID  
Liveness  
(Live-on-  
entry id'ed)*

*Alias  
Identification*

*SSA Phi  
Placement*

*Global Value  
Numbering*

- *Create a “Control Flow Graph” of basic blocks and find loops, etc.*
- *Assign a unique “value number” to every fetch of a variable and every computation*
- *Use “phi” value numbers at join points to represent alternative values*
  - *E.g. if  $X > Y$  then  $Max := X$  else  $Max := Y$  end if;  $Max == ? \Rightarrow PhiVN(X, Y)$*
- *“Kappa” node introduced to represent value of potential alias after assignment*
  - *E.g.  $A[I] := 3$ ;  $A[J] := 5$ ;  $A[I] == ? \Rightarrow KappaVN(I == J? 5: 3)$*

## Goals of Possible Value Propagation (PVP) Phase

- **Compute Possible Value Set for every Value Number in every Basic Block**
  - Map of Value number to Value set
- **Check for failures of run-time checks, user assertions, and preconditions of called routines**
  - Initially assume checks will pass and thereby infer Pre/Postconditions
  - Iterate until a fixed point
  - Make final pass to report checks that still fail

*Typical VN=>Vset Map:*  
VN1 => {1..4}  
VN4 => inverse{null}  
VN7 => {0..+inf}  
VN9 => {&obj2,&obj4}

## PVP Iterative Cycle

- **Works one basic-block (BB) at a time.**
- **Initializes the “checks” table for each BB**
  - Summary of all “implicit” and “explicit” checks performed in BB for each VN
- **Applies checks of BB and then propagates changes in VN value sets until they stabilize**
  - Propagate “down” to constituent VNs, then “up” to composite VNs
  - e.g.  $VN1 = VN2 * VN3$ 
    - Shrink VN1, propagate “down” to VN2 and VN3;
    - Shrink VN3, propagate “up” to VN1.
- **Computes VN state for each BB and for each outgoing edge**
  - Saves edge states for later iterations
  - Saves exit-block state for pre/postconditions

## Inferring Preconditions in PVP phase

- **Each Input** (parameter or global) is given an “Input VN” to represent its (unknown) *initial* value
- **Vset for Input VN is full possible range of type**
  - **e.g.** Inp\_VN1 in  $-2^{31} .. +2^{31}-1$
- **As we apply checks and assertions Vset for Input VN may shrink, eliminating the “bad” values.**
- **VNs corresponding to combinations of Inputs and Literals** (e.g.  $\text{Inp\_VN1} - \text{Inp\_VN2} * 2$ ) **might also undergo checks/assertions and might shrink** (directly or indirectly)
- **At exit block, Vset of Input VN or combination thereof represents the “good” values** (those that survived the checks and assertions), **i.e. a precondition**

**e.g.**  $\text{Inp1} - \text{Inp2} * 2$  in  $1 .. +\text{inf}$ ;       $\text{Inp2}$  in  $-\text{inf} .. -1 \mid +1 .. +\text{inf}$

→ **Preconditions:**  $\text{Inp1} > \text{Inp2} * 2$ ;       $\text{Inp2} \neq 0$

## Inferring Postconditions in PVP phase

Same principle applies for Postconditions ...

- **In Exit Block, Vset associated with a VN that represents the final value of some Output** (parameter, global, or function result) **or combination thereof, represents possible values upon completion, i.e. a *postcondition***

- ***Example:***

```

proc Incr(X : in out Integer) is
  X := X + 1
end proc Incr

```

*initial value of X is Inp\_VN1:*

*Inp\_VN1 in  $-2^{31} .. +2^{31}-1$*

*final value of X is VN2 = Inp\_VN1 + 1:*

*VN2 in  $-2^{31}+1 .. +2^{31}$*

*check that X + 1 doesn't overflow:*

*VN2 in  $-2^{31}+1 .. +2^{31}-1$*

*propagates to:*

*Inp\_VN1 in  $-2^{31} .. +2^{31}-2$*

→ *precondition:  $X'Initial \leq 2^{31}-2$*

→ *postcondition:  $X'Final \text{ in } -2^{31}+1 .. +2^{31}-1; X'Final = X'Initial+1$*

## CodePeer Screen shot showing Inferred Pre/Postconditions

```

P/P  -- Subp: fsw_example
      --
      -- Preconditions:
      --   N >= 1
      --
      -- Postconditions:
      --   A = One-of{1, 101, N - 1}
      --   A in (0..121 | 789..231-2)
      --   B = One-of{2, 102, N}
      --   B in (1..122 | 790..231-1)
      --   B = A + 1
      --
      -- Test Vectors:
      --   N: {123..456}, {457..789}, {1..122 | 790..231-1}
      --   A: {1}, {101}
      --   B: {2}, {102}
      --
2     procedure Fsw_Example (N : Natural;
3                               A, B : out Natural) is
4     begin
5         case N is
6             when 123 .. 456 =>
7                 A := 1;
8                 B := 2;
9             when 457 .. 789 =>
10                A := 101;
11                B := 102;
12            when others =>
13                A := N - 1;
14                B := N;
15        end case;
16        pragma Assert (B - A = 1);
17    end Fsw_Example;

```



## Additional Roles for CodePeer Static Analyzer

- **Formal Prover for “Low-Hanging fruit”**
  - Verification conditions generated for every run-time check and every assertion, precondition, etc. in SPARK code
  - SMT Solvers can be slow on some verification conditions
  - CodePeer can quickly prove about 98% of checks cannot fail
  - Use CodePeer as pre-filter, so SMT solvers only need to deal with remaining 2% of verification conditions
- **Verifier for Model-Based Code Generation**
  - AdaCore QGen tool translates Simulink model into source code, either MISRA C or SPARK subset of Ada
  - Can use CodePeer to check for possible run-time errors in generated SPARK code
  - If CodePeer gives clean bill of health on SPARK code, it applies to generated MISRA C code as well

## Summary

### **Abstract Interpretation is useful for whole-program verification**

- **Advantages:**
  - Can handle large-range types
  - Avoids state-space explosion
  - Needs no user-provided pre/postconditions or loop invariants
- **Problems with classic approach:**
  - Limited ability to represent relationships between variables
  - May require top-down walk of all paths to provide value sets for inputs
  - May require driver or harness to provide realistic inputs
- **Re-engineered value-number based approach in CodePeer:**
  - Can represent arbitrary relationships between variables
  - Uses efficient mechanism to propagate information between value numbers
  - Can infer both numeric and symbolic pre/postconditions so no need for drivers/harnesses or top-down walk
- **Other Roles:** Filter Low-hanging fruit; Validate generated code for model