

# Computing Bounds of SSA Values in MLIR

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# Introduction

- Compute LB/UB/EQ of index-typed SSA values or (dynamic) dimension sizes of shaped values (tensor/memref).
- Compare two index-typed SSA values or dimension sizes.
- Op interface driven: [ValueBoundsOpInterface](#)
- Built on top of the MLIR Presburger library.
- Use cases (examples):
  - *Allocation Hoisting*: Compute an upper bound for a dynamic memory allocation size.
  - *Enable Vectorization*: Compute an upper bound of a dynamically-shaped tensor computation.
  - *Subset-based Programming / Bufferization / etc*:  
Prove that two slices/subviews into the same tensor/memref are equivalent/non-overlapping.

# Public API by Example

# Compare Values / Dimensions

```
#include "mlir/Interfaces/ValueBoundsOpInterface.h"

/// Return "true" if "lhs cmp rhs" was proven to hold. Return "false" if the
/// specified relation could not be proven. This could be because the
/// specified relation does in fact not hold or because there is not enough
/// information in the constraint set. In other words, if we do not know for
/// sure, this function returns "false".

static bool ValueBoundsConstraintSet::compare(
    const Variable &lhs, ComparisonOperator cmp, const Variable &rhs);
```

one of:

- index attribute
- index-typed SSA value
- shaped value + dim
- affine map + operands (vars)

# Example: Index-typed Values

```
func.func @test_case(%arg0: index, %arg1: index) {  
    %0 = arith.addi %arg0, %arg1 : index  
    %1 = arith.addi %arg1, %arg0 : index  
    return  
}
```

lhs		rhs	
%0	$\equiv$	%1	→ true
%0	$\geq$	%1	→ true
%0	$<$	%1	→ false
%0	$\geq$	%arg0	→ false

---

```
ValueBoundsConstraintSet::compare(lhs, ValueBoundsConstraintSet::EQ, rhs);
```

# Example: Index-typed Values

```
func.func @scf_for(%lb: index, %ub: index, %s: index) {  
    scf.for %iv = %lb to %ub step %s { }  
    return  
}
```

lhs		rhs	
%iv	$\geq$	%lb	→ true
%iv	<	%ub	→ true

# Example: Dimensions of Shaped Values

```
func.func @scf_for_tensor(%init: tensor<?xf32>) {  
    %r = scf.for %iv = %a to %b step %  
        iter_args(%t = %init) -> tensor<?xf32> {  
            %0 = tensor.insert ... into %t[...]  
            scf.yield %0 : tensor<?xf32>  
    }  
    return  
}
```

lhs	rhs	
$\dim(\%r, 0)$	$=$	$\dim(\%init, 0)$ → true
$\dim(\%r, 0)$	$=$	$\dim(\%t, 0)$ → true

ValueBoundsConstraintSet::compare(

```
{lhs, /*dim=*/0}, ValueBoundsConstraintSet::EQ, {rhs, /*dim=*/0});
```

# Compute Bounds

```
// Compute a bound for the given index-typed value or shape dimension size.  
// The computed bound is stored in `resultMap`. The operands of the bound are  
// stored in `mapOperands`. An operand is either an index-type SSA value  
// or a shaped value and a dimension.  
  
static LogicalResult ValueBoundsConstraintSet::computeBound(  
    AffineMap &resultMap, ValueDimList &mapOperands,  
    presburger::BoundType type, const Variable &value,  
    StopConditionFn stopCondition, bool closedUB = false);
```

no guarantees how tight the bound is

determines which SSA values are allowed to appear in the bound

# API Example

```
func.func @test_case(%arg0: tensor<?xf32>) {  
    %0 = tensor.insert ... into %arg0[...] : tensor<?xf32>  
    %1 = linalg.generic outs(%0 : tensor<?xf32>) ...  
}  
  
affine_map<()>[s0] -> (s0)  
  
AffineMap map;  
[(%arg0, 0)]  
  
ValueDimList operands; // SmallVector<std::pair<Value, std::optional<int64_t>>>  
LogicalResult status = ValueBoundsConstraintSet::computeBound(map, operands, BoundType::EQ, {val, /*dim=*/0},  
/*stopCondition*/[])(Value v, std::optional<int64_t> dim, ...){  
    auto bbArg = dyn_cast<BlockArgument>(v);  
    if (!bbArg)  
        return false;  
    return isa<FunctionOpInterface>(bbArg.getOwner());  
});
```

The diagram shows several annotations with arrows pointing to specific parts of the code:

- An annotation labeled "affine\_map<()>[s0] -> (s0)" points to the declaration of the affine map type.
- An annotation labeled "AffineMap map;" points to the declaration of the map variable.
- An annotation labeled "[(%arg0, 0)]" points to the entry point of the map.
- An annotation labeled "%1" points to the result of the linalg.generic operation.
- An annotation labeled "Only func bbArg in the bound" points to the condition where the block argument is checked.

# API Example

```
func.func @test_case(%arg0: tensor<?xf32>) {  
    %0 = tensor.insert ... into %arg0[...] : tensor<?xf32>  
    %1 = linalg.generic outs(%0 : tensor<?xf32>) ...  
}
```

```
AffineMap map;           failure: could not compute bound  
ValueDimList operands;   vector<std::pair<Value, std::optional<int64_t>>>  
LogicalResult status = ValueBoundsConstraintSet::computeBound(map, operands, BoundType::EQ, {val, /*dim=*/0},  
/*stopCondition*/[](Value v, std::optional<int64_t> dim, ...) { return false; }));
```

No SSA values in the bound  
(constant bound)

# Related: ReifyRankedShapedTypeOpInterface

```
// RUN: mlir-opt -resolve-ranked-shaped-type-result-dims %s

%0 = tensor.insert %f into %arg0[%idx1] : tensor<?xf32>
%1 = tensor.insert %f into %0[%idx2] : tensor<?xf32>
%dim = tensor.dim %1, %c0 : tensor<?xf32>
```

- Fixed-point iteration of rewrite pattern: rewrite tensor/memref op result dim in terms of operands.
- Materializes IR for every operation (in theory, less efficient).
- Op interface driven.

# Related: ReifyRankedShapedTypeOpInterface

```
// RUN: mlir-opt -resolve-ranked-shaped-type-result-dims %s

%0 = tensor.insert %f into %arg0[%idx1] : tensor<?xf32>
%1 = tensor.insert %f into %0[%idx2] : tensor<?xf32>
%dim = tensor.dim %1, %c0 : tensor<?xf32>
→
%dim = tensor.dim %0, %c0 : tensor<?xf32>
```

- Fixed-point iteration of rewrite pattern: rewrite tensor/memref op result dim in terms of operands.
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# Related: ReifyRankedShapedTypeOpInterface

```
// RUN: mlir-opt -resolve-ranked-shaped-type-result-dims %s

%0 = tensor.insert %f into %arg0[%idx1] : tensor<?xf32>
%1 = tensor.insert %f into %0[%idx2] : tensor<?xf32>
%dim = tensor.dim %1, %c0 : tensor<?xf32>
→
%dim = tensor.dim %0, %c0 : tensor<?xf32>
→
%dim = tensor.dim %arg0, %c0 : tensor<?xf32>
```

- Fixed-point iteration of rewrite pattern: rewrite tensor/memref op result dim in terms of operands.
- Materializes IR for every operation (in theory, less efficient).
- Op interface driven.

# ValueBoundsOpInterface

# Example: arith.addi

```
struct AddIOpInterface
: public ValueBoundsOpInterface::ExternalModel<AddIOpInterface, AddIOp> {
void populateBoundsForIndexValue(Operation *op, Value value,
                                ValueBoundsConstraintSet &cstr) const {
    auto addIOp = cast<AddIOp>(op);
    assert(value == addIOp.getResult() && "invalid value");
}
}
```

op result or block argument

get affine expr for SSA value

affine expression, constant or SSA value

comparison operators(==, <, <=, >=, >) are overloaded

# Example: arith.addi

```
struct AddIOpInterface
: public ValueBoundsOpInterface::ExternalModel<AddIOpInterface, AddIOp> {
void populateBoundsForIndexValue(Operation *op, Value value,
                                ValueBoundsConstraintSet &cstr) const {
    auto addIOp = cast<AddIOp>(op);
    assert(value == addIOp.getResult() && "invalid value");

    AffineExpr lhs = cstr.getExpr(addIOp.getLhs()), rhs = cstr.getExpr(addIOp.getLhs());
    cstr.bound(value) == lhs + rhs;
}
```

getExpr has side effects



# Example: tensor.pad

quite a bit simpler than  
ReifyRankedShapedTypeOpInterface

```
struct PadOpInterface
: public ValueBoundsOpInterface::ExternalModel<PadOpInterface, PadOp> {
void populateBoundsForShapedValueDim(Operation *op, Value value, int64_t dim,
                                     ValueBoundsConstraintSet &cstr) const {
    auto padOp = cast<PadOp>(op);
    assert(value == padOp.getResult() && "invalid value");
    AffineExpr srcSize = cstr.getExpr(padOp.getSource(), dim);
    AffineExpr lowPad = cstr.getExpr(padOp.getMixedLowPad()[dim]);
    AffineExpr highPad = cstr.getExpr(padOp.getMixedHighPad()[dim]);
    cstr.bound(value)[dim] == srcSize + lowPad + highPad;
}
};
```

add bound for dim size of shaped value

# Implementation Details

# Constraint Set: FlatLinearConstraints

$$Ax + b = 0$$

$$Ax + b \geq 0$$

one variable per SSA value

coefficients stored as a matrix

- Linear combination of variables
- Multiplication/division of variables is not supported
- Corresponds to the “flattened form” of `AffineExprs`
- Relevant API:
  - project out a variable
  - compute LB/UB of a variable
  - check if constraint set is “empty”

# Example: Constraints

computed bound should have only func args

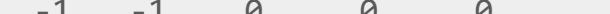
```
// RUN: mlir-opt %s -test-affine-reify-value-bounds="reify-to-func-args"
```

```
func.func @test_case(%arg0: index, %arg1: index, %arg2: index) -> index {  
    %0 = arith.addi %arg0, %arg1 : index          // %0 - %arg0 - %arg1 = 0  
    %1 = arith.addi %0, %arg2 : index            // %1 - %0      - %arg2 = 0  
    %r = "test.reify_bound"(%1) {type = "EQ"} : (index) -> (index)  
    return %r : index  
}
```

replace this op with the computed bound

Constraint set: 5 variables  
 $(\%1 \quad \%0 \quad \%arg2 \quad \%arg0 \quad \%arg1 \quad const)$

1	-1	-1	0	0	0	= 0
0	1	0	-1	-1	0	= 0



$A$        $b$

# Example: Constraints

computed bound should have only func args

// RUN: mlir-opt %s -test-affine-reify-value-bounds="reify-to-func-args"

```
func.func @test_case(%arg0: index, %arg1: index, %arg2: index) -> index {  
    %0 = arith.addi %arg0, %arg1 : index      // %0 - %arg0 - %arg1 = 0  
    %1 = arith.addi %0, %arg2 : index          // %1 - %0 - %arg2 = 0  
    %r = "test.reify_bound"(%1) {type = "EQ"} : (index) -> (index)  
    return %r : index  
}
```

replace this op with the computed bound

project out %0

Constraint set: 4 variables  
(%1    %0    %arg2 %arg0 %arg1 const)  
1    0    -1    -1    -1    0    = 0

A

b

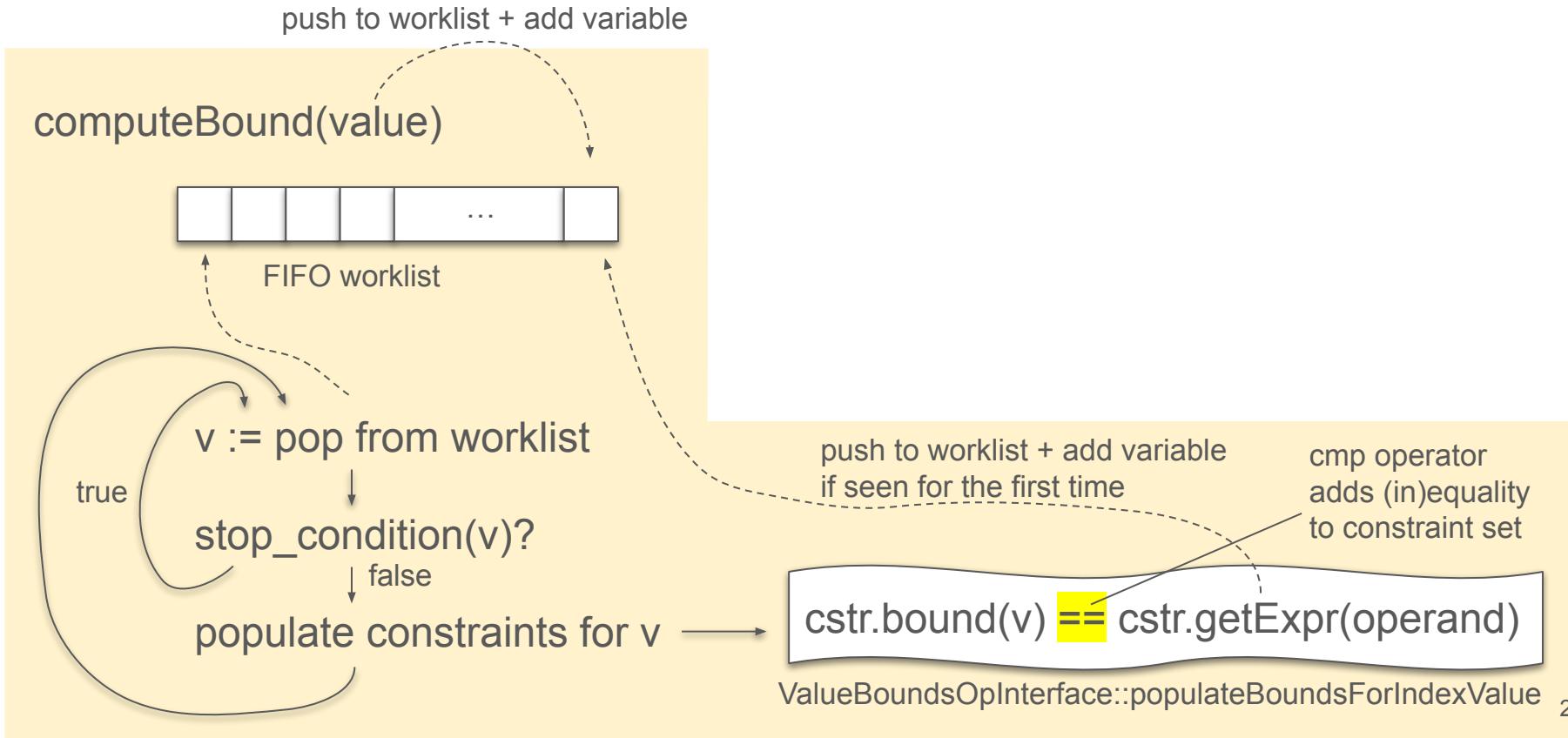
# Example: Constraints

```
#map = affine_map<()>[s0, s1, s2] -> (s0 + s1 + s2)>

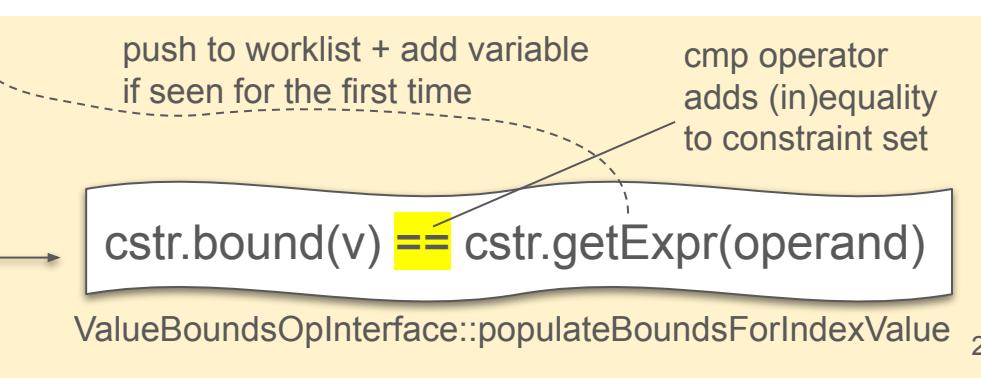
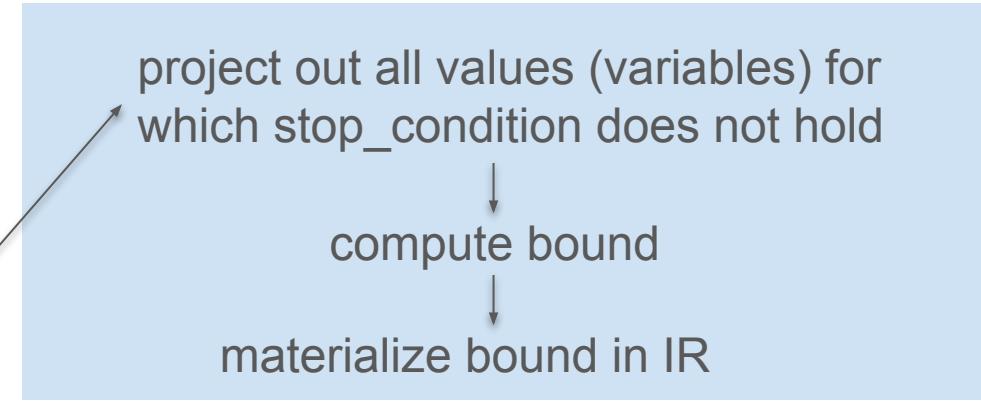
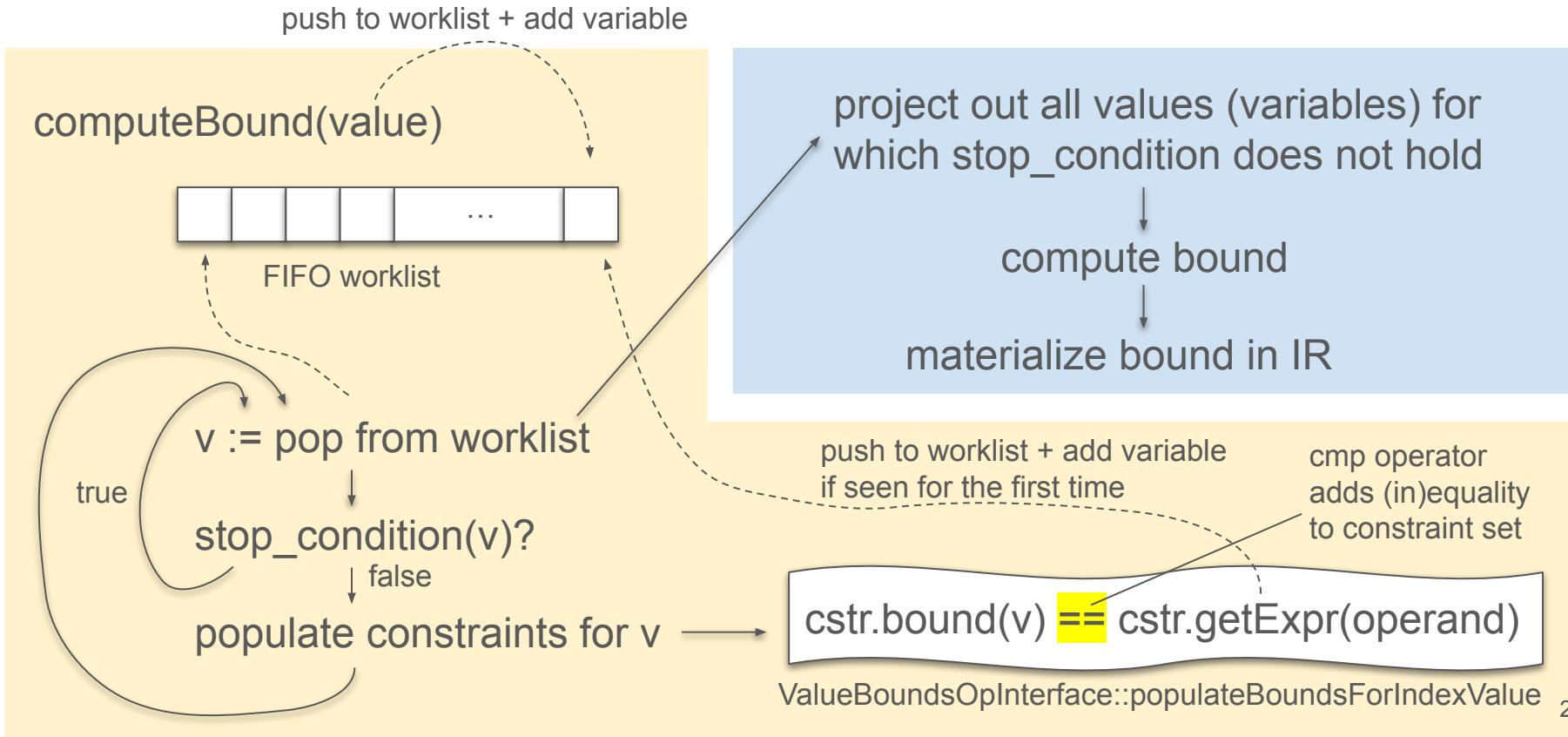
func.func @test_case(%arg0: index, %arg1: index, %arg2: index) -> index {
    %0 = arith.addi %arg0, %arg1 : index
    %1 = arith.addi %0, %arg2 : index
    %r = affine.apply #map()[%arg2, %arg0, %arg1]
    return %r : index
}
```

can also be reified with arith dialect ops

# Computing Bounds: Worklist-Driven IR Analysis



# Computing Bounds: Worklist-Driven IR Analysis



# Overview: Constraints for Various Operations

- `%r = arith.addi %a, %b : index`
  - `%r == %a + %b`
- `%r = arith.constant 5 : index`
  - `%r == 5`
- `%r = memref.subview %m[%offset0, %offset1] [%size0, %size1] [1, 1]`  
`: memref<?x?xf32> to memref<?x?xf32, strided<[?, 1], offset: ?>>`
  - `dim(%r, 0) == %size0`
  - `dim(%r, 1) == %size1`
- Destination-style Op: `%r = out(%t : tensor<?x?xf32>)`
  - `dim(%r, 0) == dim(%t, 0)`
  - `dim(%r, 1) == dim(%t, 1)`
- `%r = affine.max affine_map<()[(s0] -> (s0, 2)>())[%a]`
  - `%r <= 2`
  - `%r <= %a`
- `%r = affine.apply affine_map<()(s0, s1] -> (s0 + s1 mod 8)>())[%a, %b]`
  - `%r == expr(%a) + (expr(%b) mod 8)`

# Constraints for scf.if

```
%r = scf.if %c -> index {  
    scf.yield %a : index  
} else {  
    scf.yield %b : index  
}
```

$\%r \geq \min(\%a, \%b)$

$\%r \leq \max(\%a, \%b)$

cannot be represented in  
the constraint set

# Constraints for scf.if

```
%r = scf.if %c -> index {  
    scf.yield %a : index  
} else {  
    scf.yield %b : index  
}
```

If  $\%a \leq \%b$ :

- $\%r \geq \%a$
- $\%r \leq \%b$

If  $\%b \leq \%a$ :

- $\%r \geq \%b$
- $\%r \leq \%a$

# Constraints for scf.if

```
%r = scf.if %c -> index {  
    scf.yield %a : index  
} else {  
    scf.yield %b : index  
}
```

If %a <= %b:

- %r >= %a
- %r <= %b

```
struct IfOpInterface  
: public ValueBoundsOpInterface::ExternalModel<IfOpInterface, IfOp> {  
  
    void populateBoundsForIndexValue(Operation *op, Value value,  
                                     ValueBoundsConstraintSet &cstr) const {  
        unsigned int resultNum = cast<OpResult>(value).getResultNumber();  
        Value thenValue = ifOp.thenYield().getResults()[resultNum];  
        Value elseValue = ifOp.elseYield().getResults()[resultNum];  
  
        if (cstr.populateAndCompare(thenValue, ComparisonOperator::LE, elseValue)) {  
            cstr.bound(value) >= thenValue;  
            cstr.bound(value) <= elseValue;  
        }  
        ...  
    }  
};
```

If %b <= %a:

- %r >= %b
- %r <= %a



# Constraints for scf.for

```
%r = scf.for %iv = %lb to %ub step %s iter_args(%a = %t) -> tensor<?xf32>
{
    // ...
    scf.yield %0 : tensor<?xf32>
}
```

$\%iv \geq \%lb$

$\%iv < \%ub$

If  $\text{dim}(\%0, 0) == \text{dim}(\%a, 0)$ :

- $\text{dim}(\%r, 0) == \text{dim}(\%t, 0)$
- $\text{dim}(\%a, 0) == \text{dim}(\%t, 0)$

# Comparing Values/Dimensions

Prove that  $\%a \geq \%b$ . Prove by contradiction:

- Populate constraints for  $\%a$  and  $\%b$ .
- Assert that the constraint set is not empty (i.e., has a solution).
- Insert the inverse constraint:  $\%a < \%b$
- If the constraint set is now empty,  $\%a \geq \%b$  holds.

*Stop condition:* Keep traversing IR and populating constraints until the relation can be proven (or until we run out of IR to analyze).

# Limitations and Future Work

# Limitations and Future Work

- Non-flattenable expressions
  - Expressions that cannot be represented as linear combination of variables are not supported.
  - Example: `%r = arith.muli %a, %b : index`
- Performance: Still a lot of room for improvement
  - Stop function-based traversal may traverse more IR than necessary.
  - New constraint set is built for each computed bound (with new IR traversal).
- Cases where `FlatLinearConstraints` computes multiple bounds are not supported.
  - Example: `%r = affine.max affine_map<()[(s0, s1) -> (s0, s1)]>(%a, %b)`
- Unify with `ReifyRankedShapedTypeOpInterface` and/or `InferIntRangeInterface`?

# Questions?

- [ValueBoundsOpInterface](#)
- [ValueBoundsConstraintSet](#)
- [ReifyRankedShapedTypeOpInterface](#)
- [InferIntRangeInterface](#)
- [DestinationStyleOpInterface](#)
- [MLIR Presburger Library](#) (FlatLinearConstraints)
- Compare values/dimensions
- Compute LB/UB/EQ bound of value/dimension
- Worklist-Driven IR Analysis
- Stop condition
- Materialize bound with affine/arith dialect ops
- Branches (scf.if, arith.select)
- Loops (scf.for)
- Non-flattenable expressions (e.g., multiplications)
- Computing bounds for multiple values/dimensions

# Appendix

# API: Convenience Functions

- `static FailureOr<int64_t> ValueBoundsConstraintSet::computeConstantBound(  
presburger::BoundType type, const Variable &var,  
StopConditionFn stopCondition = nullptr, bool closedUB = false);`  
→ Like computeBound, but stop condition is optional.
- `static FailureOr<bool> areEqual(const Variable &var1, const Variable &var2);`  
→ Implemented in terms of compare:
  - var1 == var2: true
  - var1 < var2 or var2 > var1: false
  - otherwise: failure
- `static FailureOr<bool> areEquivalentSlices(MLIRContext *ctx, HyperrectangularSlice slice1,  
HyperrectangularSlice slice2);`
- `static FailureOr<bool> areOverlappingSlices(MLIRContext *ctx, HyperrectangularSlice slice1,  
HyperrectangularSlice slice2);`

# Ex.: Matmul Tiling [4, 4, 4]

```
module attributes {transform.with_named_sequence} {
    transform.named_sequence @_transform_main(%arg1: !transform.any_op {transform.readonly}) {
        %0 = transform.structured.match ops{["linalg.matmul"]} in %arg1 : (!transform.any_op) -> !transform.any_op
        %1, %loops:3 = transform.structured.tile_using_for %0 [4, 4, 4]
        : (!transform.any_op) -> (!transform.any_op, !transform.any_op, !transform.any_op, !transform.any_op)
        transform.yield
    }
}

func.func @tile_linalg_matmul(%arg0: tensor<128x128xf32>, %arg1: tensor<128x128xf32>, %arg2: tensor<128x128xf32>) -> tensor<128x128xf32> {
    %0 = linalg.matmul ins(%arg0, %arg1: tensor<128x128xf32>, tensor<128x128xf32>)
                outs(%arg2: tensor<128x128xf32>) -> tensor<128x128xf32>
    return %0 : tensor<128x128xf32>
}
```

# Ex.: Matmul Tiling [4, 4, 4]

```
func.func @tile_linalg_matmul(%arg0: tensor<128x128xf32>, %arg1: tensor<128x128xf32>, %arg2: tensor<128x128xf32>) -> tensor<128x128xf32> {
    %0 = scf.for %arg3 = %c0 to %c128 step %c4 iter_args(%arg4 = %arg2) -> (tensor<128x128xf32>) {
        %1 = scf.for %arg5 = %c0 to %c128 step %c4 iter_args(%arg6 = %arg4) -> (tensor<128x128xf32>) {
            %2 = scf.for %arg7 = %c0 to %c128 step %c4 iter_args(%arg8 = %arg6) -> (tensor<128x128xf32>) {
                %extracted_slice = tensor.extract_slice %arg0[%arg3, %arg7] [4, 4] [1, 1] : tensor<128x128xf32> to tensor<4x4xf32>
                %extracted_slice_0 = tensor.extract_slice %arg1[%arg7, %arg5] [4, 4] [1, 1] : tensor<128x128xf32> to tensor<4x4xf32>
                %extracted_slice_1 = tensor.extract_slice %arg8[%arg3, %arg5] [4, 4] [1, 1] : tensor<128x128xf32> to tensor<4x4xf32>
                %3 = linalg.matmul ins(%extracted_slice, %extracted_slice_0 : tensor<4x4xf32>, tensor<4x4xf32>
                    outs(%extracted_slice_1 : tensor<4x4xf32>) -> tensor<4x4xf32>
                %inserted_slice = tensor.insert_slice %3 into %arg8[%arg3, %arg5] [4, 4] [1, 1] : tensor<4x4xf32> into tensor<128x128xf32>
                scf.yield %inserted_slice : tensor<128x128xf32>
            }
            scf.yield %2 : tensor<128x128xf32>
        }
        scf.yield %1 : tensor<128x128xf32>
    }
    return %0 : tensor<128x128xf32>
}
```

# Ex.: Matmul Tiling [4, 9, 4]

```
func.func @tile_linalg_matmul(%arg0: tensor<128x128xf32>, %arg1: tensor<128x128xf32>, %arg2: tensor<128x128xf32>) -> tensor<128x128xf32> {  
    %0 = scf.for %arg3 = %c0 to %c128 step %c4 iter_args(%arg4 = %arg2) -> (tensor<128x128xf32>) {  
        %1 = scf.for %arg5 = %c0 to %c128 step %c9 iter_args(%arg6 = %arg4) -> (tensor<128x128xf32>) {  
            %2 = scf.for %arg7 = %c0 to %c128 step %c4 iter_args(%arg8 = %arg6) -> (tensor<128x128xf32>) {  
                %3 = affine.min affine_map<(d0) -> (-d0 + 128, 9)>(%arg5)  
                %extracted_slice = tensor.extract_slice %arg0[%arg3, %arg7] [4, 4] [1, 1] : tensor<128x128xf32> to tensor<4x4xf32>  
                %extracted_slice_0 = tensor.extract_slice %arg1[%arg7, %arg5] [4, %3] [1, 1] : tensor<128x128xf32> to tensor<4x?xf32>  
                %extracted_slice_1 = tensor.extract_slice %arg8[%arg3, %arg5] [4, %3] [1, 1] : tensor<128x128xf32> to tensor<4x?xf32>  
                %4 = linalg.matmul ins(%extracted_slice, %extracted_slice_0 : tensor<4x4xf32>, tensor<4x?xf32>)  
                    outs(%extracted_slice_1 : tensor<4x?xf32>) -> tensor<4x?xf32>  
                %inserted_slice = tensor.insert_slice %4 into %arg8[%arg3, %arg5] [4, %3] [1, 1] : tensor<4x?xf32> into tensor<128x128xf32>  
                scf.yield %inserted_slice : tensor<128x128xf32>  
            }  
            scf.yield %2 : tensor<128x128xf32>  
        }  
        scf.yield %1 : tensor<128x128xf32>  
    }  
    return %0 : tensor<128x128xf32>  
}
```

tile size does not  
divide tensor evenly

# Ex.: Matmul Tiling [4, 9, 4] – Rediscover static information

```
func.func @tile_linalg_matmul(%arg0: tensor<128x128xf32>, %arg1: tensor<128x128xf32>, %arg2: tensor<128x128xf32>) -> tensor<128x128xf32> {  
    %0 = scf.for %arg3 = %c0 to %c128 step %c4 iter_args(%arg4 = %arg2) -> (tensor<128x128xf32>) {  
        %1 = scf.for %arg5 = %c0 to %c128 step %c9 iter_args(%arg6 = %arg4) -> (tensor<128x128xf32>) {  
            %2 = scf.for %arg7 = %c0 to %c128 step %c4 iter_args(%arg8 = %arg6) -> (tensor<128x128xf32>) {  
                %3 = affine.min affine_map<(d0) -> (-d0 + 128, 9)>(%arg5)  
                extracted_slice = tensor.extract_slice %arg0[%arg3, %arg7] [4, 4] [1, 1] : tensor<128x128xf32> to tensor<4x4xf32>  
                extracted_slice_0 = tensor.extract_slice %arg1[%arg7, %arg5] [4, %3] [1, 1] : tensor<128x128xf32> to tensor<4x?xf32>  
                %extracted_slice_1 = tensor.extract_slice %arg8[%arg3, %arg5] [4, %3] [1, 1] : tensor<128x128xf32> to tensor<4x?xf32>  
                %4 = linalg.matmul ins(%extracted_slice, %extracted_slice_0 : tensor<4x4xf32>, tensor<4x?xf32>)  
                    outs(%extracted_slice_1 : tensor<4x?xf32>) -> tensor<4x?xf32>  
                %inserted_slice = tensor.insert_slice %4 into %arg8[%arg3, %arg5] [4, %3] [1, 1] : tensor<4x?xf32> into tensor<128x128xf32>  
                scf.yield %inserted_slice : tensor<128x128xf32>  
            }  
            scf.yield %2 : tensor<128x128xf32>  
        }  
        scf.yield %1 : tensor<128x128xf32>  
    }  
    return %0 : tensor<128x128xf32>  
}
```

compute constant UB for dim(%4, 1)  
→ 10 (open bound)

# Ex.: Matmul Tiling [4, 9, 4] – Rediscover static information

```
func.func @tile_linalg_matmul(%arg0: tensor<128x128xf32>, %arg1: tensor<128x128xf32>, %arg2: tensor<128x128xf32>) -> tensor<128x128xf32> {  
    %0 = scf.for %arg3 = %c0 to %c128 step %c4 iter_args(%arg4 = %arg2) -> (tensor<128x128xf32>) {  
        %1 = scf.for %arg5 = %c0 to %c128 step %c9 iter_args(%arg6 = %arg4) -> (tensor<128x128xf32>) {  
            %2 = scf.for %arg7 = %c0 to %c128 step %c4 iter_args(%arg8 = %arg6) -> (tensor<128x128xf32>) {  
                %3 = affine.min affine_map<(d0) -> (-d0 + 128, 9)>(%arg5)  
                %extracted_slice = tensor.extract_slice %arg0[%arg3, %arg7] [4, 4] [1, 1] : tensor<128x128xf32> to tensor<4x4xf32>  
                %extracted_slice_0 = tensor.extract_slice %arg1[%arg7, %arg5] [4, %3] [1, 1] : tensor<128x128xf32> to tensor<4x?xf32>  
                %extracted_slice_1 = tensor.extract_slice %arg8[%arg3, %arg5] [4, %3] [1, 1] : tensor<128x128xf32> to tensor<4x?xf32>  
                %4 = linalg.matmul ins(%extracted_slice, %extracted_slice_0 : tensor<4x4xf32>, tensor<4x?xf32>)  
                    outs(%extracted_slice_1 : tensor<4x?xf32>)  
                %inserted_slice = tensor.insert_slice %4 into %arg8  
                scf.yield %inserted_slice : tensor<128x128xf32>  
            }  
            scf.yield %2 : tensor<128x128xf32>  
        }  
        scf.yield %1 : tensor<128x128xf32>  
    }  
    return %0 : tensor<128x128xf32>  
}
```

compute constant UB for dim(%4, 1)  
→ 10 (open bound)

Constraint set: 4 variables

%4	%extracted_slice_1	%3	%arg7	const	
1	-1	0	0	0	= 0
0	1	-1	0	0	= 0
0	0	-1	-1	128	>= 0
0	0	-1	0	9	>= 0