# An Introduction to Tensor Tiling in MLIR

Kunwar Grover Mahesh Ravishanker

# Why This Tutorial?

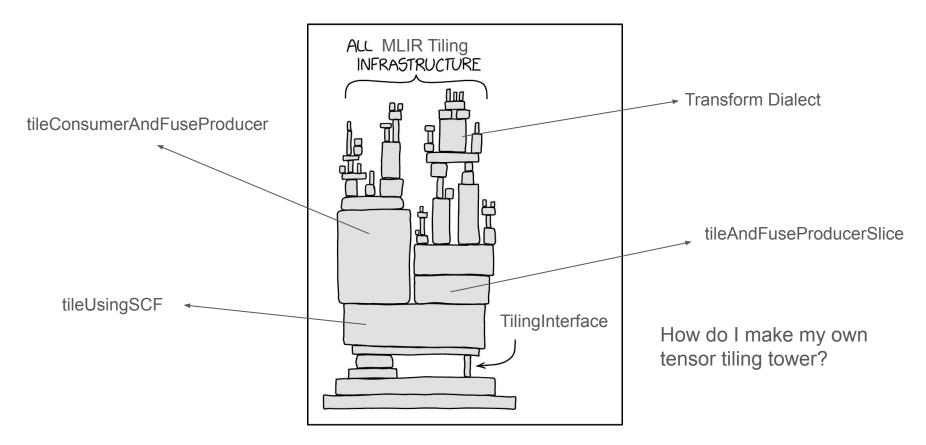
#### People frequently ask me:

- How do I build my own tensor compiler using upstream dialects?
- Why isn't there a simple pass to tile tensor operations?

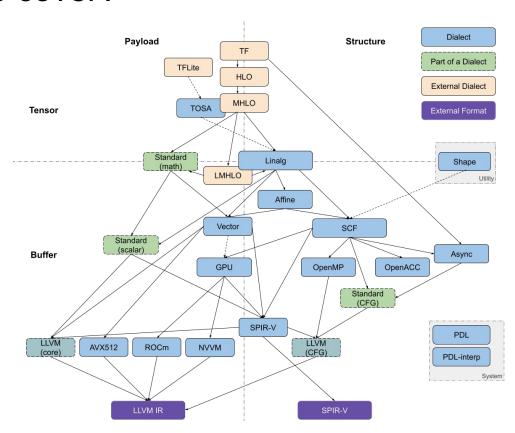


160k+ LoC (C++)

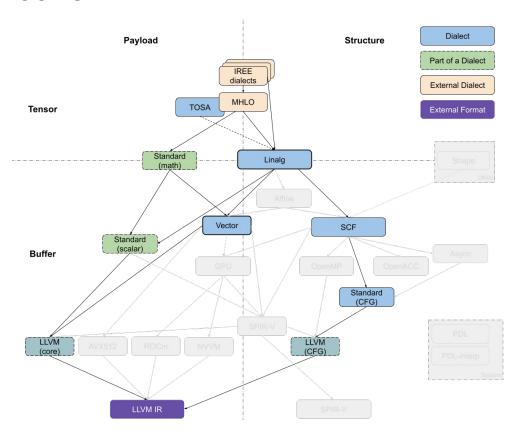
# A 100 foot view of Tensor Tiling in MLIR



#### What will we cover?

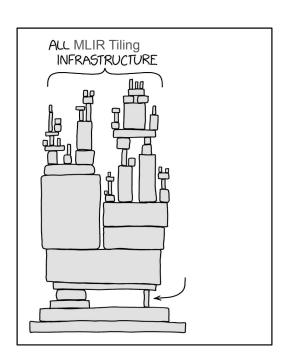


#### What will we cover?



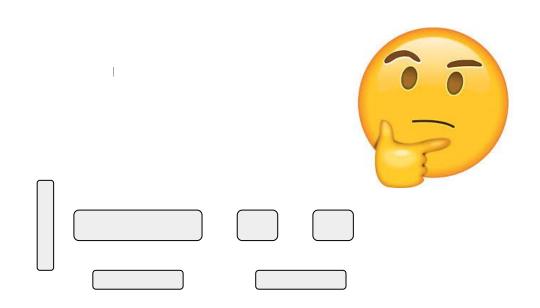
- 1. Observe
- 2. Understand
- 3. Build
- 4. Extend

- 1. Observe
- 2. Understand
- 3. Build
- 4. Extend

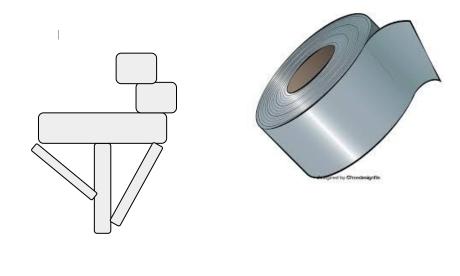




- 1. Observe
- 2. Understand
- 3. Build
- 4. Extend

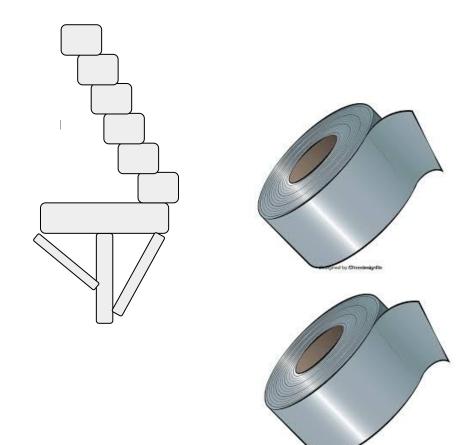


- 1. Observe
- 2. Understand
- 3. Build
- 4. Extend

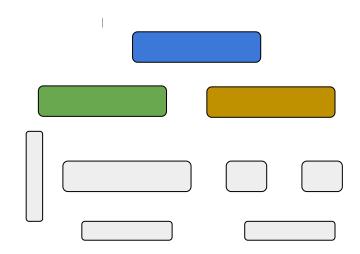


~400 LoC

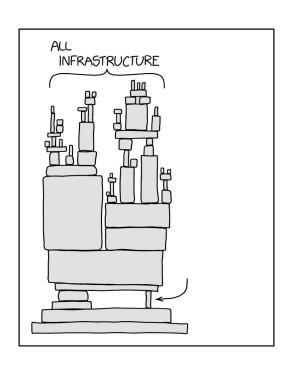
- 1. Observe
- 2. Understand
- 3. Build
- 4. Extend



- 1. Observe
- 2. Understand
- 3. Build
- 4. Extend
- 5. Advanced



# Observe



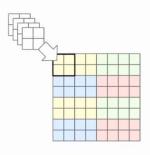






#### **Fusion into Loops**

```
linalg.generic
scf.forall (%i, %j))in (2, 4) {
    linalg.generic {
        indexing_maps = ...,
        iterator_types = ...,
    } ins(memref<4x2xf32>, memref<4x2xf32>, f32)
    outs(memref<4x2xf32>) {
        ...
     }
}
```



transform.structured.fuse\_into\_containing\_op
%structured into %loop

Similar to compute\_at as long as the loop has been materialized.

Google Research

#### Tutorial: Controllable Transformations in MLIR

Alex Zinenko

## Replicating Halide For Convolutions

```
%init = linalg.broadcast ins(%bias) ...
%conv = linalg.conv2d ins(%filter, %input) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...
```

Source IR: Conv2d + ReLU

**Objective**: Fuse Broadcast, Conv and ReLU, and target the Conv to a good tile size for the hardware

```
for n
 for v
   for x
     for c
       conv[n, y, x, c] = bias[c]
for n
 for v
   for x
     for c
       for rz
         for rv
           for rx
             conv[n, y, x, c] += filter[rx, rz, ry, c] * input[n, y+rz, x+ry, rx]
for n
 for v
   for x
     for c
       relu[n, y, x, c] = max(0, conv[n, y, x, c])
```

#### Replicating Halide For Convolutions

```
%init = linalg.broadcast ins(%bias) ...
%conv = linalg.conv2d ins(%filter, %input) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...
```

```
for n
    for y
    for x
    for c
        conv[n, y, x, c] = bias[c]
        for rz
            for ry
            for rx
                  conv[n, y, x, c] += filter[rx, rz, ry, c] * input[n, y+rz, x+ry, rx]
                  relu[n, y, x, c] = max(0, conv[n, y, x, c])
```

Source IR: Conv2d + ReLU

**Objective**: Fuse Broadcast, Conv and ReLU, and target the Conv to a good tile size for the hardware

#### **Transform Dialect**

Transform Dialect: Describe transformations on operations

#### **Transform Dialect**

Transform Dialect: Describe transformations on operations

```
transform.sequence {
    ^bb0(%bias, %conv, %relu):
    ...
    %tiled_conv, %forall2 =
        transform.structured.fuse_into_containing_op %conv into %forall
    %tiled_conv, %forall2 =
        transform.structured.fuse_into_containing_op %bias into %forall
}
```

# Replicating Halide For Convolutions: Reduction Tiling

```
^bb0(%bias, %conv, %relu):

...

%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../5, .../64) {
    %filter_tile = tensor.extract_slice %filter ...
    %input_tile = tensor.extract_slice %input ...
    %bias_tile = tensor.extract_slice %bias ...

%init_tile = linalg.broadcast ins(%bias_tile)
    %conv_tile = linalg.conv2d ...
    %relu_tile = linalg.elementwise ...

*scf.forall.yield* %relu_tile
}

^bb0(%bias, %conv, %relu):

...

*red_fill, %red_conv, %combining, %forloops =
transform.structured.tile_reduction_using_for %conv3
    // n x y c rz ry rx
    tile_sizes [0, 0, 0, 0, 1, 1, 1]
}

}

**scf.forall.yield* %relu_tile
}
```

# Replicating Halide For Convolutions: Reduction Tiling

```
^bb0(%bias, %conv, %relu):
                                                                  %red fill, %red conv, %combining, %forloops =
%relu =
                                                                  transform.structured.tile reduction using for %conv3
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter tile = tensor.extract slice %filter ...
                                                                                                    // n x y c rz ry rx
 %input_tile = tensor.extract_slice %input ...
                                                                                           tile sizes [0, 0, 0, 0, 1, 1, 1]
 %bias_tile = tensor.extract_slice %bias ...
 %init tile = linalg.broadcast ins(%bias tile)
 %conv tile =
 scf.for %rz ... {
   scf.for %ry ... {
     scf.for %rx ... {
       %filter_subtile = tensor.extract_slice %filter_tile ...
       %input subtile = tensor.extract slice %input tile ...
       %conv subtile = linalg.conv2d ...
       scf.vield %conv subtile
 %relu tile = linalg.elementwise ...
 "scf.forall.yield" %relu tile
```

#### Replicating Halide For Convolutions: Loop Structure

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter tile = tensor.extract slice %filter ...
 %input_tile = tensor.extract_slice %input ...
 %bias_tile = tensor.extract_slice %bias ...
 %init tile = linalg.broadcast ins(%bias tile)
 %conv tile =
 scf.for %rz ... {
   scf.for %ry ... {
      scf.for %rx ... {
       %filter subtile = tensor.extract slice %filter tile ...
       %input subtile = tensor.extract slice %input tile ...
       %conv subtile = linalg.conv2d ...
       scf.vield %conv subtile
 %relu tile = linalg.elementwise ...
  "scf.forall.yield" %relu tile
```

#### Replicating Halide For Convolutions: Vectorization

```
transform.sequence {
                                                                 ^bb0(%bias, %conv, %relu):
                                                                   transform.structured.vectorize children and apply patterns %func
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter tile = tensor.extract slice %filter ...
 %input_tile = tensor.extract_slice %input ...
 %bias_tile = tensor.extract_slice %bias ...
 %init tile = linalg.broadcast ins(%bias tile)
 %conv tile =
 scf.for %rz ... {
   scf.for %ry ... {
     scf.for %rx ... {
       %filter_subtile = tensor.extract_slice %filter_tile ...
       %input subtile = tensor.extract slice %input tile ...
       %conv subtile = linalg.conv2d ...
       scf.vield %conv subtile
 %relu tile = linalg.elementwise ...
  "scf.forall.yield" %relu tile
```

#### Replicating Halide For Convolutions: Vectorization

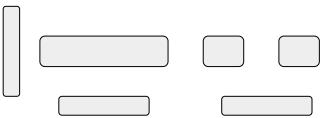
```
^bb0(%bias, %conv, %relu):
                                                                    . . .
                                                                   transform.structured.vectorize children and apply patterns %func
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %bias tile = vector.transfer read %bias ...
 %init_tile = vector.broadcast %bias_tile
 %conv tile =
 scf.for %rz ... {
   scf.for %ry ... {
     scf.for %rx ... {
       %filter subtile = vector.transfer_read %filter ...
       %input_subtile = vector.transfer_read %input ...
       %conv mul = arith.mulf ... : vector<...>
       %conv_sum = arith.addf ... : vector<...>
       scf.yield %conv sm
 %relu tile = arith.maxnumf %conv tile ... : vector<...>
  "scf.forall.yield" %relu_tile
```

#### Replicating Halide For Convolutions: Bufferization

```
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
    %bias_tile = vector.transfer_read %bias ...
    %init_tile = vector.broadcast %bias_tile
    %conv_tile =
    scf.for %rz ... {
        scf.for %rx ... {
            %filter_subtile = vector.transfer_read %filter ...
            %input_subtile = vector.transfer_read %input ...
            %conv_mul = arith.mulf ... : vector<...>
            %conv_sum = arith.addf ... : vector<...>
            scf.yield %conv_sm
        }
    }
    relu_tile = arith.maxnumf %conv_tile ... : vector<...>
    vector.transfer_write %relu_tile, %relu ...
}
```

```
transform.sequence {
^bb0(%bias, %conv, %relu):
  . . .
 transform.bufferization.one shot bufferize %func {
    bufferize function boundaries = true,
   function boundary type conversion = 1 : i32
```

# Understand





#### Main Observations

- Other than tiling, rest of the pipeline is fixed
- Getting the loop structure right is important
- Loop structure is built using Tiling and Fusing

#### Under The Hood: TilingInterface

Tiling an Operation: tileUsingSCF

## Under The Hood: Tiling By Fusion

```
%init = linalg.broadcast ...
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
    %filter_tile = tensor.extract_slice %filter ...
    %input_tile = tensor.extract_slice %input ...
    %init_tile = tensor.extract_slice %init ...

%conv_tile = linalg.conv2d ins(%filter_tile, %input_tile)
    outs(%init_tile)
    %relu_tile = linalg.elementwise ...

"scf.forall.yield" %relu_tile
}
```

```
transform.sequence {
    ^bb0(%bias, %conv, %relu):
    ...
    %tiled_conv, %forall2 =
        transform.structured.fuse_into_containing_op %conv into %forall
    %tiled_conv, %forall2 =
        transform.structured.fuse_into_containing_op %bias into %forall
}
```

Tiling by Fusion:

tileAndFuseProduceSlice tileAndFuseConsumerSlice

## Under The Hood: Tiling By Fusion

```
%init = linalg.broadcast ...
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
    %filter_tile = tensor.extract_slice %filter ...
    %input_tile = tensor.extract_slice %input ...
    %init_tile = tensor.extract_slice %init ...

%conv_tile = linalg.conv2d ins(%filter_tile, %input_tile)
    outs(%init_tile)
    %relu_tile = linalg.elementwise ...

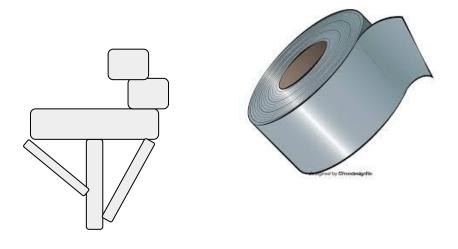
"scf.forall.yield" %relu_tile
}
```

```
transform.sequence {
    ^bb0(%bias, %conv, %relu):
    ...
    %tiled_conv, %forall2 =
        transform.structured.fuse_into_containing_op %conv into %forall
    %tiled_conv, %forall2 =
        transform.structured.fuse_into_containing_op %bias into %forall
}
```

Tiling by Fusion:

tileAndFuseProduceSlice tileAndFuseConsumerSlice

# Build



https://github.com/Groverkss/tinytile

#### TinyTile: Greedy TileAndFuse

#### TinyTile: Greedy TileAndFuse

```
%init = linalg.broadcast ins(%bias) ...
// tile conv using scf.forall:
           n x y c rz rx ry
// tile sizes = [1, 1, 5, 64, None, None, None]
%conv = linalg.conv2d ins(%filter, %input) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...
%init = linalg.broadcast ...
%conv =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter tile = tensor.extract slice %filter ...
 %input_tile = tensor.extract_slice %input ...
 %init_tile = tensor.extract_slice %init ...
 %conv tile = linalg.conv2d ins(%filter tile, %input tile)
                           outs(%init tile) ...
  "scf.forall.yield" %conv tile
%relu = linalg.elementwise ...
```

```
LogicalResult tileAndFuse(TilingInterface op,
                         ArrayRef<int64 t> tileSizes) {
 // Control how to tile the operation.
 scf::SCFTilingOptions tilingOptions;
 tilingOptions.setTileSizes(tileSizes);
 // Tile the operation.
 scf::tileUsingSCF(rewriter, tilingOp, tilingOptions);
```

#### TinyTile: Greedy TileAndFuse

```
LogicalResult tileAndFuse(TilingInterface op
                          ArrayRef<int64 t> tileSizes) {
 // Tile the operation
 while (true) {
   // Get a candidate slice.
   Operation *candidate = ...
   // Fuse producer.
   if (isa<tensor::ExtractSliceOp>(candidate)) {
      scf::tileAndFuseProducerOfSlice(rewriter, candidate, loops);
    // Fuse consumer.
    if (isa<tensor::InsertSliceOp,</pre>
            tensor::ParallelInsertSliceOp>(candidate)) {
     scf::tileAndFuseConsumerOfSlice(rewriter, candidate, loops);
```

#### TinyTile: Greedy TileAndFuse

```
%init = linalg.broadcast ...
%conv =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter_tile = tensor.extract_slice %filter ...
 %input tile = tensor.extract slice %input ...
 %init tile = tensor.extract slice %init ...
 %conv tile = linalg.conv2d ins(%filter tile, %input tile)
                             outs(%init tile) ...
  "scf.forall.yield" %conv tile
%relu = linalg.elementwise ...
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter tile = tensor.extract slice %filter ...
 %input tile = tensor.extract slice %input ...
  %bias_tile = tensor.extract_slice %bias ...
 %init tile = linalg.broadcast ins(%bias tile)
 %conv tile = linalg.conv2d ...
 %relu tile = linalg.elementwise ...
  "scf.forall.vield" %relu tile
```

```
LogicalResult tileAndFuse(TilingInterface op
                          ArrayRef<int64 t> tileSizes) {
 // Tile the operation
  while (true) {
   // Get a candidate slice.
    Operation *candidate = ...
    // Fuse producer.
   if (isa<tensor::ExtractSliceOp>(candidate)) {
      scf::tileAndFuseProducerOfSlice(rewriter, candidate, loops);
    // Fuse consumer.
    if (isa<tensor::InsertSliceOp,</pre>
            tensor::ParallelInsertSliceOp>(candidate)) {
      scf::tileAndFuseConsumerOfSlice(rewriter, candidate, loops);
```

#### TinyTile: Tracking Slices

```
%init = linalg.broadcast ins(%bias) ...
// tile conv using scf.forall:
          n x y c rz rx ry
// tile sizes = [1, 1, 5, 64, None, None, None]
%conv = linalg.conv2d ins(%filter, %input) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...
%init = linalg.broadcast ...
%conv =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter tile = tensor.extract slice %filter ...
 %input tile = tensor.extract slice %input ...
 %init_tile = tensor.extract_slice %init ...
 %conv tile = linalg.conv2d ins(%filter tile, %input tile)
                           outs(%init tile) ...
  "scf.forall.yield" %conv tile
%relu = linalg.elementwise ...
```

#### TinyTile: Tracking Slices

```
%init = linalg.broadcast ins(%bias) ...
// tile conv using scf.forall:
            n x y c rz rx ry
// tile sizes = [1, 1, 5, 64, None, None, None]
%conv = linalg.conv2d ins(%filter, %input) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...
%init = linalg.broadcast ...
%conv =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter tile = tensor.extract slice %filter ...
 %input tile = tensor.extract slice %input ...
 %init_tile = tensor.extract_slice %init ...
 %conv tile = linalg.conv2d ins(%filter tile, %input tile)
                           outs(%init tile) ...
  "scf.forall.yield" %conv tile
%relu = linalg.elementwise ...
```

```
struct SliceListener : public RewriterBase::Listener {
  void notifyOperationInserted(Operation* op,
                               OpBuilder::InsertPoint) override {
    if (isa<tensor::ExtractSliceOp,</pre>
            tensor::InsertSliceOp,
            tensor::ParallelInsertSliceOp>(op)) {
      candidates.push back(op);
  std::deque<Operation*> candidates;
LogicalResult tileAndFuse(TilingInterface op
                          ArrayRef<int64 t> tileSizes) {
  SliceListener listener:
  rewriter.setListener(&listener);
  while (!candidates.empty()) {
   // Get a candidate slice.
    Operation *candidate = listener.candidates.front();
    listener.candidates.pop front()
```

#### TinyTile: Multiple Tiling Levels

```
LogicalResult tileAndFuse(TilingInterface op) {
    ...
    // Control how to tile the operation.
    scf::SCFTilingOptions tilingOptions;

SmallVector<int64_t> tileSizes = getTileSizes(op);
    tilingOptions.setTileSizes(tileSizes);
}
```

#### TinyTile: Multiple Tiling Levels

#### TinyTile: Pass Pipeline

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter_tile = tensor.extract_slice %filter ...
 %input_tile = tensor.extract_slice %input ...
 %bias_tile = tensor.extract_slice %bias ...
 %init tile = linalg.broadcast ins(%bias tile)
 %conv tile =
 scf.for %rz ... {
   scf.for %ry ... {
      scf.for %rx ... {
       %filter subtile = tensor.extract slice %filter tile ...
       %input_subtile = tensor.extract_slice %input_tile ...
       %conv subtile = linalg.conv2d ...
       scf.vield %conv subtile
 %relu tile = linalg.elementwise ...
  "scf.forall.yield" %relu tile
```

```
void createPassPipeline(PassManager &pm) {

   // Parallel tiling using scf.forall
   {
     tutorial::TutorialTileAndFuseOptions options;
     options.tilingLevel = tutorial::TilingLevel::Parallel;
     pm.addPass(tutorial::createTutorialTileAndFuse(options));
   }

   // Reduction tiling using scf.for
   {
     tutorial::TutorialTileAndFuseOptions options;
     options.tilingLevel = tutorial::TilingLevel::Reduction;
     pm.addPass(tutorial::createTutorialTileAndFuse(options));
   }

   ...
}
```

#### TinyTile: Vectorization

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
    %bias_tile = vector.transfer_read %bias ...
    %init_tile = vector.broadcast %bias_tile
    %conv_tile =
    scf.for %rz ... {
        scf.for %ry ... {
            %filter_subtile = vector.transfer_read %filter ...
            %input_subtile = vector.transfer_read %input ...
            %conv_mul = arith.mulf ... : vector<...>
            %conv_sum = arith.addf ... : vector<...>
            scf.yield %conv_sm
        }
    }
}
%relu_tile = arith.maxnumf %conv_tile ... : vector<...>
"scf.forall.yield" %relu_tile
```

```
// Vectorization
{
   pm.addPass(createLinalgGeneralizeNamedOpsPass());
   pm.addPass(tutorial::createTutorialVectorization());
   // Cleanup
   pm.addPass(createCanonicalizerPass());
   pm.addPass(createCSEPass());
   pm.addPass(tensor::createFoldTensorSubsetOpsPass());
}
```

#### TinyTile: Vectorization

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
    %bias_tile = vector.transfer_read %bias ...
    %init_tile = vector.broadcast %bias_tile
    %conv_tile =
    scf.for %rz ... {
        scf.for %ry ... {
            %filter_subtile = vector.transfer_read %filter ...
            %input_subtile = vector.transfer_read %input ...
            %conv_mul = arith.mulf ... : vector<...>
            %conv_sum = arith.addf ... : vector<...>
            scf.yield %conv_sm
        }
    }
}
%relu_tile = arith.maxnumf %conv_tile ... : vector<...>
"scf.forall.yield" %relu_tile
```

```
// Vectorization
  pm.addPass(createLinalgGeneralizeNamedOpsPass());
  pm.addPass(tutorial::createTutorialVectorization());
  // Cleanup
  pm.addPass(createCanonicalizerPass());
  pm.addPass(createCSEPass());
  pm.addPass(tensor::createFoldTensorSubsetOpsPass());
SmallVector<linalg::GenericOp> candidates;
  funcOp.walk([&](linalg::GenericOp op) {
    candidates.push back(op);
 });
  for (linalg::GenericOp candidate : candidates) {
    (void)linalg::vectorize(rewriter, candidate);
```

#### TinyTile: Bufferization

```
// Bufferization
{
  bufferization::OneShotBufferizePassOptions options;
  options.bufferizeFunctionBoundaries = true;
  options.functionBoundaryTypeConversion =
       bufferization::LayoutMapOption::IdentityLayoutMap;
  pm.addPass(bufferization::createOneShotBufferizePass(options));
  pm.addPass(createCanonicalizerPass());
  pm.addPass(createCSEPass());
  pm.addPass(memref::createFoldMemRefAliasOpsPass());
}
```

#### TinyTile: A Matmul/Convolution Compiler!

```
%init = linalg.broadcast ins(%bias) ...
%conv = linalg.conv2d ins(%filter, %input) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...
```

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
 %filter_tile = tensor.extract_slice %filter ...
  %input_tile = tensor.extract_slice %input ...
  %bias_tile = tensor.extract_slice %bias ...
  %init tile = linalg.broadcast ins(%bias tile)
  %conv tile =
  scf.for %rz ... {
   scf.for %ry ... {
      scf.for %rx ... {
        %filter subtile = tensor.extract slice %filter tile ...
        %input_subtile = tensor.extract_slice %input_tile ...
        %conv_subtile = linalg.conv2d ...
        scf.vield %conv subtile
  %relu tile = linalg.elementwise ...
  "scf.forall.yield" %relu tile
```

#### TinyTile: A Matmul/Convolution Compiler!

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
  %filter_tile = tensor.extract_slice %filter ...
  %input_tile = tensor.extract_slice %input ...
  %bias tile = tensor.extract slice %bias ...
  %init tile = linalg.broadcast ins(%bias tile)
  %conv tile =
  scf.for %rz ... {
   scf.for %ry ... {
      scf.for %rx ... {
        %filter subtile = tensor.extract slice %filter tile ...
        %input_subtile = tensor.extract_slice %input_tile ...
        %packed_filter = linalg.pack %filter_subtile ...
        %packed input = linalg.pack %input subtile ...
        %conv_subtile = linalg.conv2d ...
        scf.yield %conv subtile
  %relu tile = linalg.elementwise ...
  "scf.forall.yield" %relu_tile
```

#### TinyTile: A Matmul/Convolution Compiler!

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
  %filter_tile = tensor.extract_slice %filter ...
  %input tile = tensor.extract_slice %input ...
  %bias tile = tensor.extract slice %bias ...
  %init tile = linalg.broadcast ins(%bias tile)
  %conv tile =
  scf.for %rz ... {
   scf.for %ry ... {
      scf.for %rx ... {
        %filter subtile = tensor.extract slice %filter tile ...
        %input subtile = tensor.extract slice %input tile ...
        %packed_filter = linalg.pack %filter_subtile ...
        %packed input = linalg.pack %input subtile ...
        %conv subtile = linalg.conv2d ...
        scf.yield %conv subtile
  %relu tile = linalg.elementwise ...
  %scale tile = tensor.extract slice %scale ...
  %actv tile = linalg.elementwise ...
  "scf.forall.yield" %relu_tile
```

#### TinyTile: Controlling Tiling Scheme

```
func.func @kernel (...) -> ...
attributes { tiling_transform_spec = "__halide" } {
    ...
}

transform.named_sequence @__halide(...) {
    ...
}
```

#### TinyTile: Controlling Tiling Scheme

```
func.func @kernel (...) -> ...
attributes { tiling_transform_spec = "__halide" } {
    ...
}

transform.named_sequence @__halide(...) {
    ...
}
```

```
// Get transform entry point.
auto entryPoint =
    funcOp->getAttrOfType<StringAttr>("transform_tiling_spec");

// Create a dynamic pass pipeline to run.
OpPassManager modulePassManager(ModuleOp::getOperationName());
transform::InterpreterPassOptions options;
options.entryPoint = entryPoint.str();
modulePassManager.addPass(transform::createInterpreterPass(options));

// Run pipeline on the module.
runPipeline(modulePassManager, moduleOp);
```

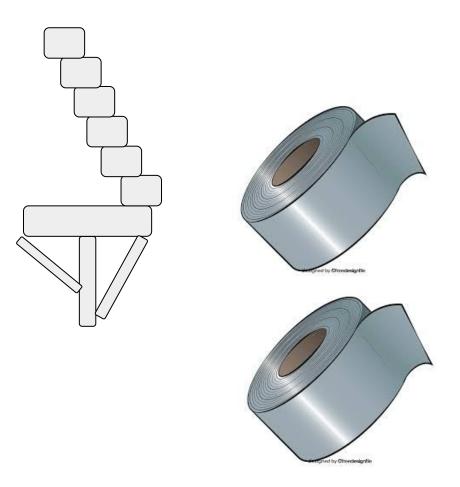
#### TinyTile: Controlling Tiling Scheme

```
func.func @kernel (...) -> ...
attributes { tiling_transform_spec = "__halide" } {
    ...
}

transform.named_sequence @__halide(...) {
    ...
}
```

```
// Get transform entry point.
auto entryPoint =
 funcOp->getAttrOfType<StringAttr>("transform tiling spec");
// Create a dynamic pass pipeline to run.
OpPassManager modulePassManager(ModuleOp::getOperationName());
transform::InterpreterPassOptions options;
options.entryPoint = entryPoint.str();
modulePassManager.addPass(transform::createInterpreterPass(options));
// Run pipeline on the module.
runPipeline(modulePassManager, moduleOp);
// Apply required transform spec.
pm.addPass(tutorial::createTutorialApplyTilingSpec());
// Parallel Tiling
// Reduction Tiling
```

# Extend



#### A New Op?: TilingInterface

```
%deqi = tutorial.dequant %input, %scale
%init = linalg.broadcast ins(%bias) ...
%conv = linalg.conv2d ins(%filter, %deqi) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...
```

### TilingInterface: Tiling

```
%deqi = tutorial.dequant %input, %scale
```

#### TilingInterface: getIterationDomain

#### TilingInterface: getIterationDomain

```
def Tutorial DequantOp : Op<Tutorial Dialect,</pre>
for row in range(M):
                                                                                                "dequant",
  for col in range(N):
    deqi[row][col] = dequant(input[row][col], scale[row][col])
                                                                   [DeclareOpInterfaceMethods<TilingInterface,
Iteration Domain: 0 <= row < M, 0 <= col < N
                                                                     ["getIterationDomain",
                                                                 ]>]>
  int64 t rank = getInputType().getRank();
  SmallVector<OpFoldResult> sizes =
    tensor::getMixedSizes(getInput());
  SmallVector<Range> loopBounds(rank);
  for (auto dim : llvm::seq<int64 t>(rank)) {
    loopBounds[dim].offset = 0;
    loopBounds[dim].size = sizes[dim];
    loopBounds[dim].stride = 1;
  return loopBounds;
```

#### TilingInterface: getLoopIteratorTypes

#### TilingInterface: getLoopIteratorTypes

```
def Tutorial DequantOp : Op<Tutorial Dialect,</pre>
for row in range(M):
                                                                                                  "dequant",
  for col in range(N):
    deqi[row][col] = dequant(input[row][col], scale[row][col])
                                                                     [DeclareOpInterfaceMethods<TilingInterface,
                                                                       ["getIterationDomain",
Both loops are parallel
                                                                        "getLoopIteratorTypes",
                                                                  ]>]>
int64_t rank = getInput().getType().getRank();
return SmallVector<utils::IteratorType>(rank,
                                         utils::IteratorType::parallel);
```

#### TilingInterface: getTiledImplementation



Given: iteration domain

Objective: generate code to compute output tile

]>]>

#### TilingInterface: getTiledImplementation



Given: iteration domain

Objective: generate code to compute output tile

#### TilingInterface: getResultTilePosition



Given: iteration domain

Objective: offset, size of result tiles

#### TilingInterface: getResultTilePosition



Given: iteration domain

Objective: offset, size of result tiles

```
resultOffsets = offsets;
resultSizes = sizes;
```

#### TilingInterface: Producer Fusion

```
%deqi = tutorial.dequant %input, %scale
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
    %filter_tile = tensor.extract_slice %filter ...
    %input_tile = tensor.extract_slice %deqi ...
    %bias_tile = tensor.extract_slice %bias ...

%init_tile = linalg.broadcast ins(%bias_tile)
%conv_tile = linalg.conv2d ...
%relu_tile = linalg.elementwise ...

"scf.forall.yield" %relu_tile
}
```

#### TilingInterface: getIterationDomainFromResultTile



Given: offsets, size of result tiles

Objective: iteration domain

#### TilingInterface: getIterationDomainFromResultTile



Given: offsets, size of result tiles

Objective: iteration domain

iterDomainOffsets = offsets; iterDomainSizes = sizes;

#### TilingInterface: generateResultTileValue



Given: offsets, size of result tiles

Objective: generate code to compute result tile

#### TilingInterface: generateResultTileValue



Given: offsets, size of result tiles

Objective: generate code to compute result tile

#### TilingInterface: Consumer Fusion

#### TilingInterface: getIterationDomainTileFromOperand



Given: offsets, size of an input tile

Objective: iteration domain

#### TilingInterface: getIterationDomainTileFromOperand



Given: offsets, size of an input tile

Objective: iteration domain

```
def Tutorial DequantOp : Op<Tutorial Dialect,</pre>
                               "dequant",
  [DeclareOpInterfaceMethods<TilingInterface,
    ["getIterationDomain",
     "getLoopIteratorTypes",
     "getTiledImplementation",
     "getResultTilePosition"
     // producer fusion
     "getIterationDomainFromResultTile",
     "generateResultTileValue"
     // consumer fusion
     "getIterationDomainTileFromOperandTile"
1>1>
```

iterDomainOffsets = llvm::to\_vector(offsets);
iterDomainSizes = llvm::to\_vector(sizes);

#### TilingInterface: getTiledImplementationFromOperand



Given: offsets, size of an input tile

Objective: generate code to compute result tile

#### TilingInterface: getTiledImplementationFromOperand



Given: offsets, size of an input tile

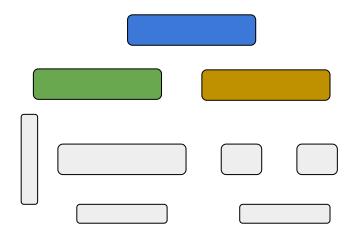
Objective: generate code to compute result tile

```
def Tutorial DequantOp : Op<Tutorial Dialect,</pre>
                               "deauant".
  [DeclareOpInterfaceMethods<TilingInterface,
    ["getIterationDomain",
     "getLoopIteratorTypes",
     "getTiledImplementation",
     "getResultTilePosition"
     // producer fusion
     "getIterationDomainFromResultTile",
     "generateResultTileValue"
     // consumer fusion
     "getIterationDomainTileFromOperandTile"
     "getTiledImplementationFromOperandTile"
1>1>
```

#### TinyTile: Extended

```
%relu =
scf.forall (%n, %x, %y, %c) in (.../1, .../1, .../5, .../64) {
  %filter tile = tensor.extract slice %filter ...
  %input_tile = tensor.extract_slice %input ...
  %scale tile = tensor.extract slice %scale ...
  %bias tile = tensor.extract slice %bias ...
  %init tile = linalg.broadcast ins(%bias tile)
  %conv tile =
  scf.for %rz ... {
   scf.for %ry ... {
      scf.for %rx ... {
        %filter_subtile = tensor.extract_slice %filter_tile ...
        %input subtile = tensor.extract slice %input tile ...
        %scale subtile = tensor.extract slice %scale tile ...
        %packed degi = tutorial.deguant ...
        %packed filter = linalg.pack %filter subtile ...
        %packed_input = linalg.pack %input_subtile ...
        %conv subtile = linalg.conv2d ...
        scf.vield %conv subtile
  %relu_tile = linalg.elementwise ...
 %scale tile = tensor.extract_slice %scale ...
  %actv tile = linalg.elementwise ...
  "scf.forall.yield" %relu tile
```

## Advanced



```
scf.forall (%warp_x, %warp_y) in (2, 2) {
...
} { mapping = [#gpu.warp<x>, #gpu.warp<y>] }
```

```
%init = linalg.broadcast ins(%bias) ...
%conv = linalg.conv2d ins(%filter, %input) outs(%init) ...
%relu = linalg.elementwise ins(%conv, 0) ...

scf::SCFTilingOptions options;
options.setMapping(...)
...
tileUsingSCF(op, options);
```

```
%init = linalg.broadcast ins(%bias) ...
%conv =
scf.forall (%warp_x, %warp_y) in (2, 2) {
    %filter_slice = tensor.extract_slice %filter
    %input_slice = tensor.extract_slice %input
    %init_slice = tensor.extract_slice %init

    %conv_tile = linalg.conv2d ...

    "scf.forall_yield" %conv_tile
} { mapping = [#gpu.warp<x>, #gpu.warp<y>] }
%relu = linalg.elementwise ins(%conv, 0) ...
```

```
scf::SCFTilingOptions options;
options.setMapping(...)
...
tileUsingSCF(op, options);
```

```
%init = linalg.broadcast ins(%bias) ...
%relu =
scf.forall (%warp_x, %warp_y) in (2, 2) {
    %filter_slice = tensor.extract_slice %filter
    %input_slice = tensor.extract_slice %input
    %init_slice = tensor.extract_slice %init

    %conv_tile = linalg.conv2d ...
    %relu_tile = linalg.elementwise ...

    "scf.forall_yield" %conv_tile
} { mapping = [#gpu.warp<x>, #gpu.warp<y>] }
```

#### Applying Greedy Tile And Fuse

```
%relu =
scf.forall (%warp_x, %warp_y) in (2, 2) {
    %filter_slice = tensor.extract_slice %filter
    %input_slice = tensor.extract_slice %input
    %bias_slice = tensor.extract_slice %bias

%init_tile = linalg.broadcast ...
    %conv_tile = linalg.conv2d ...
    %relu_tile = linalg.elementwise ...

"scf.forall_yield" %conv_tile
} { mapping = [#gpu.warp<x>, #gpu.warp<y>] }
```

#### Applying Greedy Tile And Fuse

#### Reduction Tiling

```
enum class ReductionTilingStrategy {
 // [reduction] -> [reduction1, reduction2]
 // -> loop[reduction1] { [reduction2] }
 FullReduction,
 // [reduction] -> [reduction1, parallel2]
 // -> loop[reduction1] { [parallel2] }; merge[reduction1]
 PartialReductionOuterReduction,
 // [reduction] -> [parallel1, reduction2]
 // -> loop[parallel1] { [reduction2] }; merge[parallel1]
 PartialReductionOuterParallel
```

#### Reduction Tiling: Split-K

```
scf::SCFTilingOptions options;
options.setReductionTilingStrategy(
   PartialReductionOuterParallel
);
...
tileUsingSCF(op, options);
```

#### Reduction Tiling: Split-K

```
scf::SCFTilingOptions options;
options.setReductionTilingStrategy(
   PartialReductionOuterParallel
);
...
tileUsingSCF(op, options);
```

## Thank You!

https://github.com/Groverkss/tinytile

