A APPENDIX

A.1 Array Index Notation Grammar

The full syntax of array index notation can be found in Figure 32.

```
 \langle array\_stmt \rangle ::= \langle access \rangle \ '=' \langle expr \rangle 
 \langle access \rangle \qquad ::= \langle tensor \rangle_{\{\langle index \rangle\}} 
 \langle index \rangle \qquad ::= \langle index\_var \rangle [ \langle index\_slice \rangle ] 
 \langle index\_slice \rangle ::= \langle (' \langle lo \rangle \ ':' \langle hi \rangle [ \ ':' \langle st \rangle ] \ ')' 
 \langle expr \rangle \qquad ::= \langle literal \rangle | \langle access \rangle | \langle call\_expr \rangle | \langle reduce\_expr \rangle 
 | \langle binary\_expr \rangle | \langle unary\_expr \rangle | \ '(' \langle expr \rangle \ ')' 
 \langle call\_expr \rangle ::= \langle func \rangle \ \langle (expr) \ \langle (',' \langle expr \rangle) \ \rangle' 
 \langle call\_expr \rangle ::= \langle func \rangle \ \langle (expr) \ \langle (',' \langle expr \rangle) \ \rangle' 
 \langle binary\_expr \rangle ::= \langle (expr) \ \langle (op) \rangle \langle (expr) \rangle
```

Fig. 32. The syntax of array index notation. Expressions within braces may be repeated any number of times. $\langle func \rangle$ and $\langle op \rangle$ both represent arbitrary (user-defined or predefined) functions and are implemented in the same way; they differ only in how they are invoked.

A.2 PyData/Sparse API

An example of performing the xor operation on two sparse tensors using PyData/Sparse is found below.

```
1 import numpy
2 import sparse
3
4 # Create some tensors.
5 dim = 1000
6 A = sparse.random((dim, dim, dim))
7 B = sparse.random((dim, dim, dim))
8 # Perform the XOR computation.
9 C = numpy.logical_xor(A, B)
```

An example performing the GCD operation can be found below:

```
import numpy
import

def gcd(x, y):
    return ... # Compute the GCD between x and y.
    # Register the gcd function as a ufunc.
    gcd = np.frompyfunc(gcd, 2, 1)

# Create some tensors.
    dim = 1000
A = sparse.random((dim, dim, dim))
B = sparse.random((dim, dim, dim))
# Perform the XOR computation.
C = gcd(A, B)
```

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While this code is simpler than the code to use our sparse array compiler, users do not have control over many factors, such as the formats of the tensors, and are restricted to the predefined set of NumPy functions.

A.3 Iteration Lattice Construction Algorithm

As described in Section 6, the presented iteration lattice construction algorithm (Algorithm 1) supports only array index notation expressions that do not contain repeat tensors. Fig. 18 illustrates an example of when iteration sub-spaces do not overlap when the index notation contains a repeated tensor. This example motivates our implementation of a filtered Cartesian Product.

We include the full algorithm that does support repeated tensors in Algorithm 2.

Algorithm 2 Full iteration lattice construction algorithm

```
procedure ConstructLattice (FunctionAlgebra A, FunctionArguments args)
    // Preprocessing steps
    Algebra A = DeMorgan(A)
                                                                                              ▶ Apply De Morgan's Law
    Algebra A = Augment(A, args)
                                                                                                     ▶ Augmentation pass
    return BuildLattice(A)
end procedure
// let \mathcal{L} represent an iteration lattice and p represent an iteration lattice point
procedure BuildLattice (Algebra A)
    if A is Tensor(t) then
                                                                                                           ▶ Segment Rule
        return \mathcal{L}(p(\{t\}, producer=true))
    else if A is ~Tensor(t) then
                                                                                                      ▶ Complement Rule
         p_o = p(\{t, U\}, producer=false)
         p_p = p(\{ \mathbb{U} \}, \text{ producer=true})
         return \mathcal{L}(\{p_o, p_p\})
    else if A is (left \cap right) then
                                                                                                       ▶ Intersection Rule
         \mathcal{L}_l, \mathcal{L}_r = BuildLattice(left), BuildLattice(right)
        cp = FilteredCartesianProduct(\mathcal{L}_l.points(), \mathcal{L}_r.points())
        mergedPoints = { p(\{p_l + p_r\}, \text{producer} = p_l.\text{producer} \land p_l.\text{producer}) : \forall (p_l, p_r) \in \text{cp} \}
        mergedPoints = RemoveDuplicates(mergedPoints, ommitterPrecedence)
         return \mathcal{L}(mergedPoints)
    else if A is (left ∪ right) then
                                                                                                              ▶ Union Rule
         \mathcal{L}_l, \mathcal{L}_r = BuildLattice(left), BuildLattice(right)
        cp = FilteredCartesianProduct(\mathcal{L}_l.points(), \mathcal{L}_r.points())
        mergedPoints = { p(\{p_l + p_r\}, \text{producer} = p_l.\text{producer} \lor p_l.\text{producer}) : \forall (p_l, p_r) \in \text{cp} }
        mergedPoints = mergedPoints + \mathcal{L}_{I}.points() + \mathcal{L}_{r}.points()
        mergedPoints = RemoveDuplicates(mergedPoints, producerPrecedence)
        return \mathcal{L}(mergedPoints)
end procedure
procedure Filtered Cartesian Product (Lattice Points left, Lattice Points right)
    p_{l,\text{root}}, p_{r,\text{root}} = \text{left.root}, \text{right.root}
    for (p_l \text{ in left}) \times (p_r \text{ in right}) do overlap = true
        for tensor in p_1 do
             if (tensor in p_{r,\text{root}}) \land (tensor not in p_l) then overlap = false
        end for
        for tensor in p_r do
             if (tensor in p_{l,root}) \land (tensor not in p_l) then overlap = false
        if overlap then cp += \{(p_l, p_r)\}
    end for
    return cp
end procedure
```

A.4 Medical Imaging Edge Detection

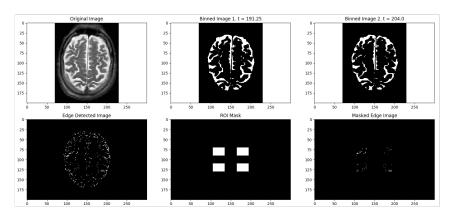


Fig. 33. Example MRI image, thresholding, ROI mask, and output