

1422 **A APPENDIX**1423 **A.1 Array Index Notation Grammar**

1424 The full syntax of array index notation can be found in Figure 32.

```

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1426 <array_stmt> ::= <access> '=' <expr>
1427 <access>     ::= <tensor> {<index>}
1428 <index>       ::= <index_var> [ <index_slice> ]
1429 <index_slice> ::= '(' <lo> ':' <hi> [ ':' <st> ] ')'
1430 <expr>         ::= <literal> | <access> | <call_expr> | <reduce_expr>
1431           | <binary_expr> | <unary_expr> | '(' <expr> ')'
1432           | <func> '(' <expr> {,} <expr> ')' '
1433 <call_expr>   ::= <func> '(' <expr> {,} <expr> ')'
1434 <reduce_expr> ::= <func> <expr>
1435           <index_var>
1436 <binary_expr> ::= <expr> <op> <expr>
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```

1438 Fig. 32. The syntax of array index notation. Expressions within braces may be repeated any number of times.  
 1439  $\langle \text{func} \rangle$  and  $\langle \text{op} \rangle$  both represent arbitrary (user-defined or predefined) functions and are implemented in the  
 1440 same way; they differ only in how they are invoked.

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1443 **A.2 PyData/Sparse API**

1444 An example of performing the xor operation on two sparse tensors using PyData/Sparse is found  
 1445 below.

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1447 1 import numpy
1448 2 import sparse
1449 3
1450 4 # Create some tensors.
1451 5 dim = 1000
1452 6 A = sparse.random((dim, dim, dim))
1453 7 B = sparse.random((dim, dim, dim))
1454 8 # Perform the XOR computation.
1455 9 C = numpy.logical_xor(A, B)
1456

```

1456 An example performing the GCD operation can be found below:

```

1457 1 import numpy
1458 2 import
1459 3
1460 4 def gcd(x, y):
1461 5     return ... # Compute the GCD between x and y.
1462 6 # Register the gcd function as a ufunc.
1463 7 gcd = np.frompyfunc(gcd, 2, 1)
1464 8
1465 9 # Create some tensors.
1466 10 dim = 1000
1467 11 A = sparse.random((dim, dim, dim))
1468 12 B = sparse.random((dim, dim, dim))
1469 13 # Perform the XOR computation.
1470 14 C = gcd(A, B)

```

1471 While this code is simpler than the code to use our sparse array compiler, users do not have  
1472 control over many factors, such as the formats of the tensors, and are restricted to the predefined  
1473 set of NumPy functions.

### 1474 1475 A.3 Iteration Lattice Construction Algorithm

1476 As described in Section 6, the presented iteration lattice construction algorithm (Algorithm 1)  
1477 supports only array index notation expressions that do not contain repeat tensors. Fig. 18 illustrates  
1478 an example of when iteration sub-spaces do not overlap when the index notation contains a repeated  
1479 tensor. This example motivates our implementation of a filtered Cartesian Product.

1480 We include the full algorithm that does support repeated tensors in Algorithm 2.

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**Algorithm 2** Full iteration lattice construction algorithm
 

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1520
1521 procedure CONSTRUCTLATTICE (FunctionAlgebra A, FunctionArguments args)
1522   // Preprocessing steps
1523   Algebra A = DEMORGAN(A)                                     ▷ Apply De Morgan's Law
1524   Algebra A = AUGMENT(A, args)                                 ▷ Augmentation pass
1525   return BUILDLATTICE(A)
1526 end procedure

1527
1528 // let  $\mathcal{L}$  represent an iteration lattice and  $p$  represent an iteration lattice point
1529 procedure BUILDLATTICE (Algebra A)
1530   if A is Tensor(t) then                                         ▷ Segment Rule
1531     return  $\mathcal{L}(p(\{t\}, \text{producer=true}))$ 
1532   else if A is  $\sim$ Tensor(t) then                                     ▷ Complement Rule
1533      $p_o = p(\{t, \mathbb{U}\}, \text{producer=false})$ 
1534      $p_p = p(\{\mathbb{U}\}, \text{producer=true})$ 
1535     return  $\mathcal{L}(\{p_o, p_p\})$ 
1536   else if A is (left  $\cap$  right) then                                ▷ Intersection Rule
1537      $\mathcal{L}_l, \mathcal{L}_r = \text{BUILDLATTICE}(\text{left}), \text{BUILDLATTICE}(\text{right})$ 
1538     cp = FILTEREDCARTESIANPRODUCT( $\mathcal{L}_l.\text{points}()$ ,  $\mathcal{L}_r.\text{points}()$ )
1539     mergedPoints = { $p(\{p_l + p_r\}, \text{producer}=p_l.\text{producer} \wedge p_r.\text{producer}) : \forall (p_l, p_r) \in cp$ }
1540     mergedPoints = REMOVEDUPLICATES(mergedPoints, ommitterPrecedence)
1541     return  $\mathcal{L}(\text{mergedPoints})$ 
1542   else if A is (left  $\cup$  right) then                                ▷ Union Rule
1543      $\mathcal{L}_l, \mathcal{L}_r = \text{BUILDLATTICE}(\text{left}), \text{BUILDLATTICE}(\text{right})$ 
1544     cp = FILTEREDCARTESIANPRODUCT( $\mathcal{L}_l.\text{points}()$ ,  $\mathcal{L}_r.\text{points}()$ )
1545     mergedPoints = { $p(\{p_l + p_r\}, \text{producer}=p_l.\text{producer} \vee p_r.\text{producer}) : \forall (p_l, p_r) \in cp$ }
1546     mergedPoints = mergedPoints +  $\mathcal{L}_l.\text{points}() + \mathcal{L}_r.\text{points}()$ 
1547     mergedPoints = REMOVEDUPLICATES(mergedPoints, producerPrecedence)
1548     return  $\mathcal{L}(\text{mergedPoints})$ 
1549 end procedure

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1551 procedure FILTEREDCARTESIANPRODUCT (LatticePoints left, LatticePoints right)
1552    $p_{l,\text{root}}, p_{r,\text{root}} = \text{left.root}, \text{right.root}$ 
1553   for ( $p_l$  in left)  $\times$  ( $p_r$  in right) do overlap = true
1554     for tensor in  $p_l$  do
1555       if (tensor in  $p_{r,\text{root}}$ )  $\wedge$  (tensor not in  $p_l$ ) then overlap = false
1556     end for
1557     for tensor in  $p_r$  do
1558       if (tensor in  $p_{l,\text{root}}$ )  $\wedge$  (tensor not in  $p_r$ ) then overlap = false
1559     end for
1560     if overlap then cp +=  $\{(p_l, p_r)\}$ 
1561   end for
1562   return cp
1563 end procedure

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1569 **A.4 Medical Imaging Edge Detection**

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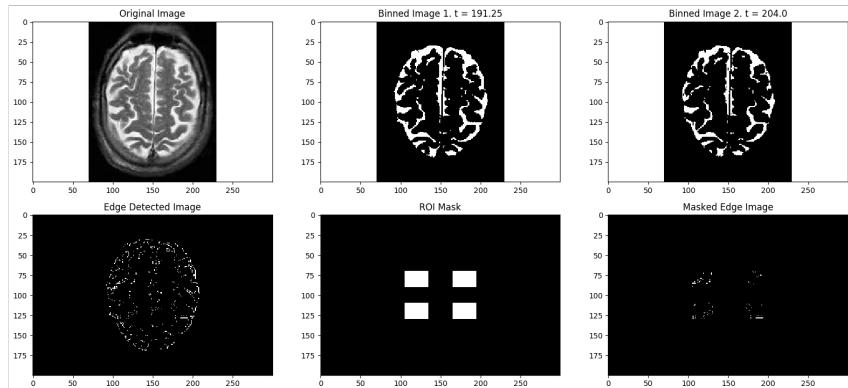
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1585 Fig. 33. Example MRI image, thresholding, ROI mask, and output

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