

Formal Verification of COO to CSR Sparse Matrix Conversion

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We describe a machine-checked correctness proof of a C program that converts a coordinate-form (COO) sparse matrix to a compressed-sparse-row (CSR) matrix. The classic algorithm (sort the COO entries in lexicographic order by row,column; fill in the CSR arrays left to right) is concise but has rather intricate invariants. We illustrate a bottom-up methodology for deriving the invariants from the program.

1 Introduction

We will describe the machine-checked correctness proof of a C program that converts a Coordinate-form sparse matrix (COO) into a Compressed Sparse Row matrix (CSR):

```
struct csr_matrix *coo_to_csr_matrix(struct coo_matrix *p) { . . . }
```

The program itself is given in Listing 9; it implements an algorithm (presumably) known for many decades¹ This paper is meant as a tutorial on the methodology for approaching the specification and proof of numerical algorithms involving both data structures and approximations (keeping the data-structure reasoning separate from the approximation reasoning), and a demonstration of a technique for deriving loop invariants from the properties they must satisfy in a Hoare logic proof.

Sparse matrix-vector multiplication is a fundamental operation in numerical methods, and takes time proportional to the number of nonzero entries in the matrix—which can be much smaller than the size of the corresponding dense matrix. Depending on the structure of sparsity in the matrix, and on whether the multiplication is of the form Ax or $x^T A$, different sparse representations may be appropriate. For the very common case of unstructured sparsity and multiplication Ax , the Compressed Sparse Row (CSR) representation is useful [3, §4.3.1].

The CSR format stores the elements of a sparse $m \times n$ matrix A using three one-dimensional arrays: a floating-point array `val` that stores the nonzero elements of A , an integer array `col_ind` that stores the column indices of the elements, and an integer array `row_ptr` that stores the locations in the array `col_ind` that start a row in A . Figure 1 shows an example.

But before a CSR matrix is used for multiplications, it must be *built*. One does not first build the dense matrix and from it extract the sparse matrix, as that would be quite inefficient. In a typical application scientific/engineering application that gives rise to a sparse matrix, one first translates the problem into a set of triples (row,column,value), that is, a *coordinate form sparse matrix* (COO matrix).

A COO matrix has dimensions `rows×cols`; there are n coordinate triples `row_ind[k]`, `col_ind[k]`, `val[k]` for $0 \leq k < n$. Each of those arrays has size `maxn` $\geq n$ to allow for additional entries in the future.

^{*}This paper accompanies my keynote lecture “Foundational End-to-end Verification of Numerical Programs” at VSS 2025, the International Workshop on Verification of Scientific Software; and covers one of the results described in that talk.

¹Barret *et al.* [3] describe CSR (which they called “compressed row storage”) but do not mention COO nor the method of constructing CSR matrices—however, if CSR matrices were in use then there must have been a method of constructing them.

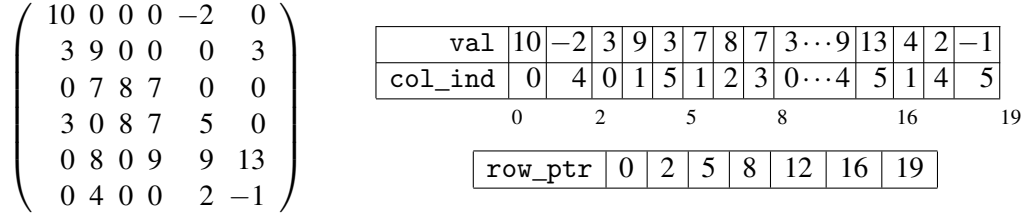


Figure 1: An example of the three arrays (val, col_ind, row_ptr) used to store a matrix compressed sparse row (CSR) format. From Barret *et al.* [3], adjusted for 0-based array indexing.

row_ind	0	0	0	1	1	1	2	2	2	...	4	4	5	5	5
col_ind	0	0	4	0	1	5	1	2	3	5	...	4	5	1	4
val	7	3	-2	3	9	3	7	8	7	3	...	9	13	4	2

Figure 2: A coordinate-form representation of the matrix from Figure 1

A COO matrix may have more than one entry at the same (row,col). If entry k is (i, j, x) and entry k' is (i, j, y) , the matrix this represents has value $x + y$ at position (i, j) —or $x + y$ plus other entries $(i, j, _)$. We call those *duplicates*. For example, the COO in Figure 2 represents the same matrix as the CSR in Figure 1; there is a duplicate $7 + 3$ at position $(0, 0)$. Although the entries in Figure 2 are sorted, generally the entries in a COO matrix can be arranged in any order.

Duplicate entries arise naturally in scientific problems. For example, in a finite-element analysis of a mesh such as Figure 5, each interior vertex adjacent to d regions (elements) contributes d values to the list of coordinate tuples.

COO matrices are not very efficient for performing matrix multiplication (though perhaps better than dense matrices). Their primary purpose is as an intermediate representation for building another form of sparse representation, such as CSR.

The algorithm for converting a COO to a CSR matrix is well known: First, sort the tuples by the lexicographic order of row then column. Then process the sorted tuples in order, adding duplicates together as they are seen; each of the three arrays of the CSR representation can be filled in left-to-right.

Listing 9 shows the program. I have never run this program, I have only proved it correct. But unlike Knuth,² I can be confident that it works, because (almost 50 years later) we have foundationally sound machine-checked program logics.

The Coq proofs corresponding to this draft of the paper may be found at https://github.com/VeriNum/iterative_methods/tree/8de4d78c78f92f280581253af49c2309fb95b2b2c/sparse

²“Beware of bugs in the above code; I have only proved it correct, not tried it.” Donald E. Knuth, Notes on the van Emde Boas construction of priority queues, March 1977. <https://staff.fnwi.uva.nl/p.vanemdeboas/knuthnote.pdf>

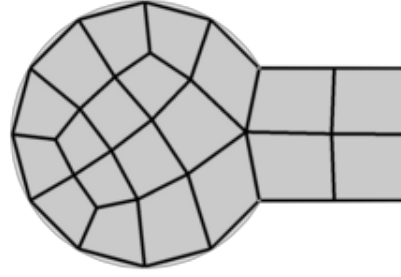
```
struct csr_matrix {
    double *val;
    unsigned *col_ind, *row_ptr;
    unsigned rows, cols;
};
```

Listing 3: CSR struct in C.

```
struct coo_matrix {
    unsigned *row_ind, *col_ind;
    double *val;
    unsigned n, maxn, rows, cols;
};
```

Listing 4: COO struct in C.

Figure 5: Mesh arising from an irregular finite-element problem. From https://en.wikipedia.org/wiki/Mesh_generation



2 The specification

In many application domains, we can prove that a program computes exactly the right answer. However, numerical algorithms generally compute *approximations*, with inaccuracies arising from discretization and from floating-point round-off. Thus the specification of a numerical program will typically bound the distance between the computed answer and the true solution to some mathematical equation. In COO-to-CSR the approximation comes from floating-point round-off when summing duplicate elements.

Reasoning about data structures in C programs (or other low-level imperative programming languages) is difficult enough without having to simultaneously reason about the composition of approximations. So we separate the reasoning into: first, the C program *exactly* implements a low-level specification; and second, that low-level spec approximates the solution to a mathematical equation within specified bounds.

We will use the Verifiable Software Toolchain to specify and prove correctness of the C program. In VST program logic, called *Verifiable C*, we distinguish between mathematical objects and the data structures that represent them. Verifiable C is embedded in the Coq logic, and we can describe mathematical objects directly using Coq types and values. For example, we consider “COO matrix” be a mathematical value whose Coq type is,

```
Record coo_matrix (t: type) := {
  coo_rows:  $\mathbb{Z}$ ;
  coo_cols:  $\mathbb{Z}$ ;
  coo_entries: list ( $\mathbb{Z} * \mathbb{Z} * \text{ftype } t$ )
}.
```

Here, `type` means “floating-point format,” such as IEEE double precision; for generality, our matrix types and our algorithms are parameterized over formats. If `t` is a type, then `ftype t` is the Coq [Type](#) of values in that floating-point format [1].

Therefore, a `coo_matrix` is a tuple (R, C, E) where the matrix is supposed to have dimension $R \times C$ and E is a list of 3-tuples.

A program might represent COO matrices with an array of 3-element records, or Fortran-like with three separate arrays. Either way, we can describe the relation between the mathematical value of type `coo_matrix(Tdouble)` and our data-structure layout. In Verifiable C such a relation is a *memory predicate* or `mpred`. So, for example, in our specification we have,

Definition `coo_rep` (sh: share) (coo: coo_matrix Tdouble) (p: val) : mpred := . . .

The permission-share `sh` tells whether the data structure has read permission or write permission (etc.) and henceforth we will ignore or omit permission shares. (They can be quite useful when describing shared-memory parallel programs, but that is not the focus here.) We can read `coo_rep` as saying, the COO matrix `coo` is laid out in memory as a pointer data structure rooted at address `p`.

Similarly, we can describe the mathematical type of CSR matrices:

```
Record csr_matrix {t: type} := {
  csr_cols: ℤ;
  csr_vals: list (ftype t);
  csr_col_ind: list ℤ;
  csr_row_ptr: list ℤ;
  csr_rows: ℤ := Zlength (csr_row_ptr) - 1
}.
```

That is, a CSR matrix is a 4-tuple (cols,vals,col_ind,row_ptr) where cols is the number of columns in the matrix, vals is a sequence of floating-point values in format t , col_ind is a sequence of integers, and row_ptr is another sequence of integers; and where the number of rows is one less than the length of the row_ptr sequence. See Figure 1 for an example.

One can imagine various data structures with which a C program could represent this structure in memory, but having chosen such a data structure, we can specify it with an mpred relation:

Definition `csr_rep` (sh: share) (csr: csr_matrix Tdouble) (p: val) : mpred := . . .

A COO matrix represents a mathematical matrix, and we can state this with a mathematical relation:

Definition `coo_to_matrix` {t} (coo: coo_matrix t) (m: matrix t) : Prop := . . .

Definition `csr_to_matrix` {t} (csr: csr_matrix t) (m: matrix t) : Prop := . . .

So, for example, the relation `csr_to_matrix` holds between the two mathematical objects depicted in Figure 1; and the relation `coo_to_matrix` holds between the COO shown in Figure 2 and the matrix in Figure 1.

Having defined all these relations, we can now specify what it would mean for the C function `coo_to_csr_matrix` to be correct. Our program takes as input the address p where a COO matrix is stored, and returns the address at which a corresponding CSR matrix is stored.

One might think the specification of this program is as follows:

- Let coo be a COO sparse matrix representation,
- that is laid out in memory at address p (i.e., `coo_rep coo p`);
- let M be the matrix that \bar{M} represents (i.e., `coo_to_matrix coo M`);
- then there exists a CSR sparse matrix csr ,
- that represents matrix M ,
- and that is laid out in memory at address q (i.e., `csr_rep csr q`);
- and q is returned.

This is almost right. The problem is the use of the definite article, “let M be *the* matrix.” A floating-point COO matrix does not represent a unique mathematical matrix, because of the floating-point addition needed when duplicate entries are combined. The duplicate entries may be added together in any order; and while addition in the reals is associative, addition in the floats is not. The possible values of the resulting matrix-entry will all be similar to each other, but not exactly the same. That is, `coo_to_matrix` is a relation but it is not a function. We can say, “let M be *a* matrix that \bar{M} represents.”

To define the relation between \bar{M} and M , we first need a relation defining the floating point sum, *in any order and any tree of associativity*, of a set of floating-point values:

```

Inductive sum_any {t}: list (ftype t) → ftype t → Prop :=
| Sum_Any_0: sum_any nil (Zconst t 0)
| Sum_Any_1: ∀ x, sum_any [x] x
| Sum_Any_split: ∀ al bl a b, sum_any al a → sum_any bl b →
    sum_any (al++bl) (BPLUS a b)
| Sum_Any_perm: ∀ al bl s, Permutation al bl → sum_any al s → sum_any bl s.

```

That is, let t be a floating-point format, so the Coq type `ftype t` contains floating-point values in that format. The relation `sum_any v s`, where the t argument is implicit, says that the list-of-floats v relates to float s as follows:

- If v is the empty list then $s = 0$.
- If v is the singleton list containing x then $s = x$.
- If v is the concatenation of lists al and bl , such that al sums (in any order) to a and bl to b , then v relates to the floating-point sum of a and b , written as `BPLUS a b`.
- If al is a permutation of bl , and al relates to s , then bl also relates to s .

By this definition, `sum_any` relates al to any floating-point value that relates from combining the elements of s , once each, in an arbitrary tree of additions. There are theorems in numerical analysis that bound the distance between any such s and the real number that one would obtain by adding in arbitrary-precision arithmetic without rounding. Such theorems will be useful later, but we don't need them now; we can first specify correctness of COO-to-CSR conversion using `sum_any`, then show that this specification implies the desired accuracy bound.

We can therefore write the definition of `coo_to_matrix`—that a COO matrix `coo` represents a mathematical matrix m :

```

Definition coo_to_matrix {t: type} (coo: coo_matrix t) (m: matrix t) : Prop :=
  coo_rows coo = matrix_rows m ∧
  matrix_cols m (coo_cols coo) ∧
  ∀ i, 0 ≤ i < coo_rows coo →
  ∀ j, 0 ≤ j < coo_cols coo →
    sum_any (map snd (filter (coord_eqb (i,j) oo fst) (coo_entries coo)))
      (matrix_index m (Z.to_nat i) (Z.to_nat j))).

```

Listing 6: `coo_to_matrix`

At index (i, j) we filter all the entries of the COO matrix whose coordinates are equal to (i, j) , take their values (with `map snd`), and use `sum_any` to relate that to m_{ij} .

Listing 7 gives the “funspec” (VST function specification) for the `coo_to_csr_matrix` function. We will summarize here what it means; for more explanation, see [4] or [2]. The lines of the funspec are as follows:

DECLARE: The C name of the function is `coo_to_csr_matrix`.

WITH quantifies over the mathematical variables that will be used in the precondition (and perhaps postcondition): the COO matrix `coo` and address p .

PRE begins the precondition of the function, which takes a C parameter whose C type is “pointer to struct `csr_matrix`.”

PROP: The precondition has three parts, this first of which contains mathematical propositions that must hold of the **WITH**-bound variables; in this case, that the COO matrix is well-formed (for example, that each entry's row- and column-index are within the bounds of the matrix dimensions; see Listing 8).

```

Definition coo_to_csr_matrix_spec [simplified] :=
  DECLARE _coo_to_csr_matrix
  WITH coo: coo_matrix Tdouble,
  PRE [ tptr (Tstruct _coo_matrix noattr) ]
    PROP(coo_matrix_wellformed coo)
    PARAMS( p )
    SEP (coo_rep coo p)
  POST [ tptr (Tstruct _csr_matrix noattr) ]
    EX coo': coo_matrix Tdouble, EX csr: csr_matrix Tdouble,
    EX m: matrix Tdouble, EX q: val,
    PROP(coo_matrix_equiv coo coo';
      coo_to_matrix coo m; csr_to_matrix csr m)
    RETURN( q )
    SEP (coo_rep coo' p; csr_rep csr q).

```

Listing 7: Specification (simplified) of the C function. The full specification describes how this function has access to the memory allocator (malloc/free), needed to allocate space for the new CSR matrix.

```

Definition coo_matrix_wellformed {t} (coo: coo_matrix t) :=
  (0 ≤ coo_rows coo ∧ 0 ≤ coo_cols coo)
  ∧ Forall (fun e ⇒ 0 ≤ fst (fst e) < coo_rows coo ∧ 0 ≤ snd (fst e) < coo_cols coo)
    (coo_entries coo).

```

Listing 8: Wellformedness of COO matrices

PARAM: The value of the C-language parameter is p . We distinguish between mathematical values bound in the logic (such as p) from C-language identifiers that stand for C variables (such as $_p$). In this case, the C variable $_p$ *contains* the value p upon entry to the function.

SEP: This clause contains spatial conjuncts in separation logic, that is, it describes data structures in memory. In this case, that there is a representation of the coo matrix at address p .

POST: The postcondition starts by giving the C type of the return value of the function; in this case, pointer to `struct csr_matrix`.

EX: This postcondition existentially quantifies four mathematical quantities: a COO matrix coo' , a CSR matrix csr , a mathematical matrix m , and an address q .

PROP: Three propositions are asserted about how the variables are related: coo' is equivalent to coo , coo represents the matrix m , and csr also represents m . The postcondition doesn't say so explicitly, but in a typical implementation the difference between coo and coo' is that the entries (coordinate-tuples) of coo' are sorted in order.

RETURN: address q is the value returned by this function.

SEP: In memory when the function returns are the modified COO matrix coo' (at the same address p) and a CSR matrix at newly allocated memory address q .

2.1 The minimum you need to inspect

When one inspects a machine-checked proof, it is not so important to check the proof itself—the machine has done that—but it is critical to check the theorem-statement. For if it's the wrong theorem, it will not

```

1  unsigned coo_count (struct coo_matrix *p) {
2      unsigned i, n = p->n;
3      if (n==0) return 0;
4      unsigned count=1;
5      for (i=1; i<n; i++)
6          if (p->row_ind[i-1]!=p->row_ind[i] || p->col_ind[i-1]!=p->col_ind[i])
7              count++;
8      return count;
9  }
10
11 struct csr_matrix *coo_to_csr_matrix(struct coo_matrix *p) {
12     struct csr_matrix *q;
13     unsigned count, i, r,c, ri, ci, cols, k, l, rows;
14     unsigned *col_ind, *row_ptr, *prow_ind, *pcol_ind;
15     double x, *val, *pval;
16     unsigned n = p->n;
17     coo_quicksort(p, 0, n);
18     k = coo_count(p);
19     rows = p->rows; prow_ind=p->row_ind; pcol_ind=p->col_ind;
20     pval = p->val;
21     q = surely_malloc(sizeof(struct csr_matrix));
22     val = surely_malloc(k * sizeof(double));
23     col_ind = surely_malloc(k * sizeof(unsigned));
24     row_ptr = surely_malloc ((rows+1) * sizeof(unsigned));
25     r=-1;
26     c=0; /* this line is unnecessary but simplifies the proof */
27     l=0; /* partial_csr_0 */
28     for (i=0; i<n; i++) {
29         ri = prow_ind[i]; ci = pcol_ind[i]; x = pval[i];
30         if (ri==r)
31             if (ci==c)
32                 val[l-1] += x; /* partial_CSR_duplicate */
33             else { c=ci; col_ind[l]=ci; val[l]=x; l++;} /* partial_CSR_newcol */
34         else {
35             while (r+1<=ri) row_ptr[++r]=l; /* partial_CSR_skiprow */
36             c=ci; col_ind[l]=ci; val[l]=x; l++; /* partial_CSR_newrow */
37         }
38     }
39     cols = p->cols;
40     while (r+1<=rows) row_ptr[++r]=l; /* partial_CSR_lastrows */
41     q->val = val; q->col_ind = col_ind; q->row_ptr = row_ptr;
42     q->rows = rows; q->cols = cols;
43     return q; /* partial_CSR_properties */
44 }

```

Listing 9: COO-to-CSR conversion

serve the intended purpose. Furthermore, every definition that’s referenced (even indirectly) from the theorem-statement is part of this “trusted base.” So let us review what goes into the theorem we have stated.

- `coo_to_csr_matrix_spec` (Listing 7), 14 lines.
- `coo_to_matrix` (Listing 6), 7 lines.
- `coo_rep` (not shown), 15 lines.
- `coo_matrix_equiv` (not shown), 3 lines.

Although `csr_to_matrix` and `csr_rep` are mentioned in `coo_to_csr_matrix_spec`, it is not strictly necessary to inspect them. That’s because we have elsewhere [5] proved the theorem that our C-language sparse matrix-vector multiply function correctly multiplies a CSR matrix by a vector to produce the correct result that the mathematical matrix M would produce (based on the same definitions of `csr_to_matrix` and `csr_rep`).³ Therefore, you can treat these definitions as an abstract data type; no matter what they are, you will get the intended result when multiplying by the CSR matrix that our function produces.

In the remainder of this paper I will present dozens of lines of definitions and lemma-statements. All of those should be treated as part of the *proof* of the main theorem, stated above; and this proof has been checked by the Coq kernel.

2.2 An alternate specification

Some computational scientists, since they cannot normally achieve *formal verification of correctness and accuracy* such as we do here, rely upon other means of validation and verification. One of those is *regression testing*; that is, testing whether a change to the program has introduced a new bug by comparing the output of the changed program to the output of the original program, or examining the output of specific test cases. For such testing, it is helpful if the program has the property of *bit-for-bit reproducibility*. The COO-to-CSR program as I have specified it above does not have that property, because it permits any order of summation of duplicate elements, and in floating point those values may be very slightly different.

One can write a stronger specification, that gives bit-for-bit reproducibility, by insisting that the duplicate elements be added in left-to-right order of their appearance in the original (unsorted) COO entry list. Then one can easily show that the stronger specification entails the specification I’ve given above. The program I have verified in this paper does not have that property.

A separate issue is that of *reassembly*. Suppose one converts a COO matrix to CSR, and then one has a new (i.e., modified) COO matrix with the same sparsity structure but different values. Then the structure of the CSR matrix will be unchanged, and it should be possible to update just the `val` array of the CSR matrix, slightly more efficiently than building a CSR matrix from scratch. To specify this program, and at the same time achieve bit-for-bit reproducibility, we would divide the `coo_to_csr_matrix` into two parts:

1. Build the structure of the CSR matrix (`row_ptr` and `col_ind`)
2. Fill in the values.

Such a program would be similar in many ways to the program I verify here, and would satisfy very similar loop invariants. However, we leave this for future work.

³The definition of `csr_rep` in the LAProof paper [5] encompasses what, in this paper, we have separated into `csr_to_matrix` and `csr_rep`.

3 The low-level proof

It is useful to stratify a proof to separate the low-level details of C programs and data structures from high-level concerns about algorithms. Often this is done by stating an algorithm as a pure functional program in Coq, proving that the C program refines the algorithm, then proving that the algorithm is correct. Here we will take a different approach: we will prove that the C program is correct provided a certain relation exists with certain properties; then we will give a model of the relation.

Let us examine the program (Listing 9). First (at line 17) the COO entries are sorted in place. Then the `coo_count` function scans the entries in order, counting how many *distinct* (row,column) coordinate-pairs are in the list. This allows (at lines 22–23) the allocation of CSR arrays of exactly the right size.⁴

The main loop begins at line 28, traversing all the entries of the COO matrix in order. In examining each entry (r_i, c_i, x_i) we need to check whether it is a duplicate of some previous entry (r, c, y) , for which purpose the local variables record the previous entry's r and c . The previous y need not be remembered because it has already been added into the appropriate spot of the `val` array.

We use unsigned integers for row and column numbers, in part because that allows a greater range of indexes, i.e., bigger matrices.

To ensure that the first entry is not treated as a duplicate, (at line 25) we initialize r to one less than the smallest possible row number; that is, -1 . Since we are using unsigned (i.e., modulo 2^w) arithmetic, this is really $2^w - 1$, so we must be careful. The program *is* careful, but the version before I debugged it (by proving it correct) was not as careful. For example, while $(r+1 \leq r_i)$ is correct at line 35, but while $(r < r_i)$ would be wrong. One can learn this (as I did) from the fact that VST's proof system, since it is sound, won't permit a proof of an incorrect program.

In the traversal, the variable `l` tracks the number of *distinct* coordinate-pairs seen so far; this indicates the spot in `val` array (and in the `col_ind` array) to be filled in next.

It is straightforward to do a Hoare-logic forward proof in VST to reach line 28. But then we will need a loop invariant. In fact, there are three loops (at lines 28, 35, and 39); we will derive all of their invariants together. At any point during the execution of the loop(s), the program has built a partial CSR matrix that represents, more or less, the first i entries of of the (sorted) COO matrix. This (partial) CSR matrix is represented in the arrays `val`, `col_ind`, and `row_ptr`. Figure 10 illustrates such a configuration.

Before we try to find a model (a definition) for this relation, let us examine the properties such a relation must have. At certain points in the program, the values of variables i and r or the contents of `rowptr`, `colind`, `val` are changed. These points are labeled in Listing 9 with comments of the form `/* partial_CSR ... */`.

At each such point, we can examine the VST assertion that characterizes the current program state, along with the assertion required by the next iteration of the loop, to derive an axiom that the `partial_CSR` relation must satisfy.

Hypothesis 1. *There is a relation `partial_CSR i r coo rowptr colind val` such that the arrays `rowptr`, `colind`, `val` represent the COO matrix entries up to the i th entry and the r th row of the matrix; and furthermore `partial_CSR` satisfies all the lemma-statements of Listing 11 and Listing 12.*

Since the COO entries are sorted in order, first by row and then by column, one might think that i determines r and that it is unnecessary to include r as a parameter of the relation. But not quite. The matrix might have some all-zero rows; the purpose of the loops at lines 35 and 39 is to skip over those

⁴`surely_malloc` calls `malloc` but then, if `malloc` returns `NULL` indicating it failed to allocate memory, `surely_malloc` aborts the program.

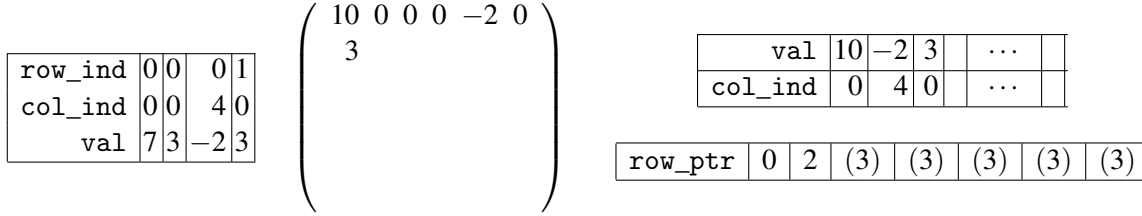


Figure 10: A partially completed CSR representation corresponding to the first four entries in the COO matrix of Figure 2. The (3) values in row_ptr are not actually stored.

rows while recording information about them into the row_ptr array. So the relation must take note of both i and r .

Even though we don't yet know the definition of this relation, we can derive the axioms it must satisfy by attempting a Hoare-logic proof of the program. For the main loop, we choose an invariant (Listing 13) that includes

`partial_CSR i r coo' ROWPTR COLIND VAL`

where i is the value of the loop iteration variable, r is the value local variable r , coo' is the COO matrix that results from the quicksort called at line 17 (of Listing 9), and the remaining variables are the current contents of the arrays row_ptr, col_ind, and val. I have previously proved the correctness of a quicksort implementation,⁵ so I can use that function-specification in the current proof to establish that the entries of coo' are sorted by the relation coord_le.

Consider, then, the situation at line 28 just before the loop starts. We can assume the COO matrix is well formed, that coo' is sorted, and the number of distinct elements in the $coo_entries$ of coo' is representable as an unsigned integer, that is, $\leq \text{Int.max_unsigned}$. We have $r = -1$, $i = 0$, and all three CSR arrays are completely uninitialized. We can summarize this with a lemma called `partial_CSR_0` (see Listing 11) which concludes,

```
partial_CSR 0 (-1) coo
  (Zrepeat Vundef (coo_rows coo + 1))    (* ROWPTR *)
  (Zrepeat Vundef (cd coo))                (* COLIND *)
  (Zrepeat Vundef (cd coo)).                (* VAL *)
```

That is, `partial_CSR` relates the first 0 entries of the COO matrix to a partial CSR matrix filled up to `rows=-1` such that the `ROWPTR` array is a sequence of $(rows+1)$ undefined values, the `COLIND` array is a sequence of undefined values of length `cd coo`, and the `VAL` array is a sequence of undefined (i.e., uninitialized) values of length `cd coo`; where `cd coo` (“count distinct”) is the number of distinct coordinate-pairs in the entry list of `coo`. We will need this many slots to be available for processing all the entries.

Consider another example: at line 33 we have determined that the entry (r_i, c_i, x) is not a duplicate, and we must create a new column in the current row. This corresponds to the lemma named `partial_CSR_newcol` (in Listing 11) which concludes,

```
partial_CSR i r coo ROWPTR COLIND VAL →
partial_CSR (i + 1) r coo ROWPTR
  (upd_Znth (cd_upto i coo) COLIND (Vint (Int.repr c)))
  (upd_Znth (cd_upto i coo) VAL (Vfloat x)).
```

⁵<https://github.com/cverified/cbench-vst/tree/master/qsort>, May 2019

Definition `cd {t} (coo: coo_matrix t) :=`
`count_distinct (coo_entries coo).`

Definition `cd_upto {t} i (coo: coo_matrix t) :=`
`count_distinct (sublist 0 i (coo_entries coo)).`

Lemma `partial_CSR_0:`

`∀ (coo: coo_matrix Tdouble),`
`coo_matrix_wellformed coo →`
`sorted coord_le (coo_entries coo) →`
`cd coo ≤ Int.max_unsigned →`
`partial_CSR 0 (-1) coo`
`(Zrepeat Vundef (coo_rows coo + 1))`
`(Zrepeat Vundef (cd coo)) (Zrepeat Vundef (cd coo)).`

Lemma `partial_CSR_duplicate:`

`∀ h r coo (f: ftype Tdouble) ROWPTR COLIND VAL,`
`0 < h < Zlength (coo_entries coo) →`
`fst (Znth (h-1) (coo_entries coo)) = fst (Znth h (coo_entries coo)) →`
`r = fst (fst (Znth (h-1) (coo_entries coo))) →`
`Znth (cd_upto h coo - 1) VAL = Vfloat f →`
`partial_CSR h r coo ROWPTR COLIND VAL →`
`partial_CSR (h+1) r coo ROWPTR COLIND`
`(upd_Znth (cd_upto h coo - 1) VAL`
`(Vfloat (Float.add f (snd (Znth h (coo_entries coo)))))).`

Lemma `partial_CSR_newcol:`

`∀ i r c x coo ROWPTR COLIND VAL,`
`0 < i < Zlength (coo_entries coo) →`
`Znth i (coo_entries coo) = (r, c, x) →`
`r = fst (fst (Znth (i-1) (coo_entries coo))) →`
`c <> snd (fst (Znth (i-1) (coo_entries coo))) →`
`partial_CSR i r coo ROWPTR COLIND VAL →`
`partial_CSR (i + 1) r coo ROWPTR`
`(upd_Znth (cd_upto i coo) COLIND (Vint (Int.repr c)))`
`(upd_Znth (cd_upto i coo) VAL (Vfloat x)).`

Lemma `partial_CSR_skiprow:`

`∀ i r coo ROWPTR COLIND VAL,`
`0 ≤ i < Zlength (coo_entries coo) →`
`r ≤ fst (fst (Znth i (coo_entries coo))) →`
`partial_CSR i (r-1) coo ROWPTR COLIND VAL →`
`partial_CSR i r coo (upd_Znth r ROWPTR (Vint (Int.repr (cd_upto i coo))))`
`COLIND VAL.`

Listing 11: Axioms for `partial_CSR`

Lemma `partial_CSR_newrow`:

```

  ∀ i r c x coo ROWPTR COLIND VAL,
  0 ≤ i < Zlength (coo_entries coo) →
  Znth i (coo_entries coo) = (r,c,x) →
  (i <> 0 → fst (fst (Znth (i - 1) (coo_entries coo))) <> r) →
  partial_CSR i r coo ROWPTR COLIND VAL →
  partial_CSR (i + 1) r coo ROWPTR
  (upd_Znth (cd_upto i coo) COLIND (Vint (Int.repr c)))
  (upd_Znth (cd_upto i coo) VAL (Vfloat x)).

```

Lemma `partial_CSR_lastrows`:

```

  ∀ r coo ROWPTR COLIND VAL,
  r ≤ coo_rows coo →
  partial_CSR (Zlength (coo_entries coo)) (r-1) coo ROWPTR COLIND VAL →
  partial_CSR (Zlength (coo_entries coo)) r coo
  (upd_Znth r ROWPTR (Vint (Int.repr (cd coo)))) COLIND VAL.

```

Lemma `partial_CSR_properties`:

```

  ∀ coo ROWPTR COLIND VAL,
  partial_CSR (Zlength (coo_entries coo)) (coo_rows coo) coo ROWPTR COLIND VAL →
  ∃ (m: matrix Tdouble) (csr: csr_matrix Tdouble),
    csr_to_matrix csr m ∧ coo_to_matrix coo m
    ∧ coo_rows coo = matrix_rows m
    ∧ coo_cols coo = csr_cols csr
    ∧ map Vfloat (csr_vals csr) = VAL
    ∧ Zlength (csr_col_ind csr) = cd coo
    ∧ map Vint (map Int.repr (csr_col_ind csr)) = COLIND
    ∧ map Vint (map Int.repr (csr_row_ptr csr)) = ROWPTR
    ∧ Zlength (csr_vals csr) = cd coo.

```

Listing 12: More axioms for `partial_CSR`

This says that if we increase i to $i + 1$, and update the k th element of `COLIND` and `ROW` to c and x respectively (where k is the number of distinct elements in the first i entries of `coo`), then the `partial_CSR` entry still holds (provided the other premises of `partial_CSR_newcol` are satisfied). `Int.repr` forces an mathematical integer into a modulo- 2^w machine integer, and `Vint` injects that into the type of `C` values; `Vfloat` injects an IEEE double-precision floating-point number into the `C` value type.

Listing 13 shows the invariant for the main loop. At line 33 of Listing 9 we must reestablish that invariant; the lemma statement `partial_CSR_newcol` is derived by noticing what it will take to reestablish this loop invariant. The other lemma statements are derived analogously, as we reestablish invariants at the places in the program annotated with comments of the form `/* partial_CSR... */`.

Finally, once we reach the end of the loops (at line 42), the relation `partial_CSR` with $i = n$ and $r = \text{rows}$, (which is to say, a completed CSR matrix) should have properties strong enough to derive the desired postcondition of the function; namely, those stated in `partial_CSR_properties` (Listing 12).

Assuming the existence of a `partial_CSR` relation satisfying these axioms allows the `C` program to be proved correct without too much fuss. Now, if we can demonstrate that such a relation exists, then the main theorem will be proved.

```

EX i:ℤ, EX l:ℤ, EX r:ℤ, EX c:ℤ,
EX ROWPTR: list val, EX COLIND: list val, EX VAL: list val,
  PROP(0 ≤ l ≤ k; l ≤ i ≤ n; -1 ≤ r < coo_rows coo'; 0 ≤ c ≤ coo_cols coo';
    partial_CSR i r coo' ROWPTR COLIND VAL;
    l = count_distinct (sublist 0 i (coo_entries coo'));
    l = 0 → r = -1;
    i < 0 → r = (fst (fst (Znth (i-1) (coo_entries coo'))))%Z ∧ c = snd (fst (Znth
      (i-1) (coo_entries coo'))))
  LOCAL (temp_l (Vint (Int.repr l));
    temp_r (Vint (Int.repr r)); temp_c (Vint (Int.repr c));
    temp_row_ptr rowptr_p; temp_col_ind colind_p; temp_val val_p;
    temp_q q; temp_pval vp; temp_pcol_ind cp; temp_prow_ind rp;
    temp_rows (Vint (Int.repr (coo_rows coo')));
    temp_n (Vint (Int.repr n)); temp_p p)
  SEP(FRZL FR1;
    data_at Ews (tarray tuint (coo_rows coo' + 1)) ROWPTR rowptr_p;
    data_at Ews (tarray tuint k) COLIND colind_p;
    data_at Ews (tarray tdouble k) VAL val_p;
    data_at Ews (Tstruct_csr_matrix noattr) q;
    data_at sh t_coo
      (rp, (cp, (vp,
        (Vint (Int.repr (Zlength (coo_entries coo'))),
        (Vint (Int.repr maxn),
        (Vint (Int.repr (coo_rows coo')), Vint (Int.repr (coo_cols coo')))))))) p;
    data_at sh (tarray tuint maxn)
      (map (fun e : ℤ * ℤ * ftype Tdouble ⇒ Vint (Int.repr (fst (fst e))))
        (coo_entries coo') ++ Zrepeat Vundef (maxn - Zlength (coo_entries coo')))
      rp;
    data_at sh (tarray tuint maxn)
      (map (fun e : ℤ * ℤ * ftype Tdouble ⇒ Vint (Int.repr (snd (fst e))))
        (coo_entries coo') ++ Zrepeat Vundef (maxn - Zlength (coo_entries coo')))
      cp;
    data_at sh (tarray tdouble maxn)
      (map (fun e : ℤ * ℤ * float ⇒ Vfloat (snd e)) (coo_entries coo') ++
        Zrepeat Vundef (maxn - Zlength (coo_entries coo'))) vp).

```

Listing 13: Main loop invariant, in VST notation. The EX clauses bind existentially quantified variables. The PROP clauses state propositions about those variables. The LOCAL clauses give the values of C variables. The SEP clauses give the separating conjuncts for data structures in memory. FRZL FR1 is essentially a “frame” of data that are untouched by (and irrelevant to) the loop.

4 Proving the partial_CSR relation

Now we need to prove that a relation exists that satisfies the lemmas in Listing 11 and Listing 12.

```

Derfinition partial_CSR {t: type} (i: ℤ) (r: ℤ) (coo: coo_matrix t)
  (ROWPTR COLIND VAL: list C.val) : Prop :=
  (* fill in a definition here *)

```

In defining this relation, the concept of “COO matrix up to entry i ” will be useful:

```
Definition coo_upto (i:  $\mathbb{Z}$ ) {t} (coo: coo_matrix t) : coo_matrix t :=
{| coo_rows := coo_rows coo;
  coo_cols := coo_cols coo;
  coo_entries := sublist 0 i (coo_entries coo)
|}.
```

We say a `coo_matrix` is well-formed if, in every coordinate-tuple (r_i, c_i, x) we have $0 \leq r_i < \text{coo_rows}$ and $0 \leq c_i < \text{coo_cols}$. Clearly, if `coo` is well-formed, then `coo_upto i coo` is well-formed.

Figure 10 suggests that a partial COO matrix (such as `coo_upto 4 coo`) should relate to a partial CSR matrix (such as the one shown in the figure). But a partial CSR matrix is not well-formed, because the `row_ptr` array is not filled in. However, the partial CSR matrix in that figure can be trivially completed by filling in all the empty slots with 3, i.e., the number of elements in the `val` array.

Therefore the heart of the `partial_CSR` relation should be a relation between a *complete* COO matrix (namely, `coo_upto i coo`) and a *complete* CSR matrix. We call this relation `coo_csr`:

```
Definition coo_csr {t} (coo: coo_matrix t) (csr: csr_matrix t) : Prop := . . .
```

The most natural *semantic* relation is that both of these represent the same matrix:

```
(* wrong *)  $\exists m$ : matrix t, coo_to_matrix coo m  $\wedge$  csr_to_matrix csr m.
```

However, it turns out that this relation is not strong enough to support the induction. (At least, I don’t have evidence that it is strong enough.) The reason is that either matrix might contain explicit zero values, and the COO matrix might have an $(i, j, 0)$ where the CSR matrix has no entry, or vice versa. Therefore, we need a more structural assurance that if the COO matrix has one or more entries at (i, j) then the CSR matrix has a corresponding entry, and vice versa.

The relation is therefore defined as,

```
Inductive coo_csr {t} (coo: coo_matrix t) (csr: csr_matrix t) : Prop :=
  build_coo_csr:  $\forall$ 
    (coo_csr_rows: coo_rows coo = csr_rows csr)
    (coo_csr_cols: coo_cols coo = csr_cols csr)
    (coo_csr_vals: Zlength (csr_vals csr) = count_distinct (coo_entries coo))
    (coo_csr_entries: entries_correspond coo csr)
    (coo_csr_zeros: no_extra_zeros coo csr),
    coo_csr coo csr.
```

That is, the COO and CSR matrix must have the same number of columns; the length of the CSR `val` array must be the number of distinct (i, j) coordinates of the COO; for every entry in the COO matrix there must be a corresponding element of the CSR matrix (`entries_correspond`); and for every element in the CSR matrix there must be a corresponding entry in the COO matrix (`no_extra_zeros`). Those latter relations are defined as,

```
Definition entries_correspond {t} (coo: coo_matrix t) (csr: csr_matrix t) :=
 $\forall h$ ,
 $0 \leq h < \text{Zlength (coo\_entries coo)} \rightarrow$ 
let '(r,c) := fst (Znth h (coo_entries coo)) in
let k := cd_upto (h+1) coo - 1 in
  Znth r (csr_row_ptr csr)  $\leq k < \text{Znth (r+1) (csr\_row\_ptr csr)} \wedge$ 
  Znth k (csr_col_ind csr) = c  $\wedge$ 
  sum_any (map snd (filter (coord_eqb (r,c) oo fst) (coo_entries coo)))
```

```

Inductive csr_matrix_wellformed {t} (csr: csr_matrix t) : Prop :=
  build_csr_matrix_wellformed:
  ∀ (CSR_wf_rows: 0 ≤ csr_rows csr)
    (CSR_wf_cols: 0 ≤ csr_cols csr)
    (CSR_wf_vals: Zlength (csr_vals csr) = Zlength (csr_col_ind csr))
    (CSR_wf_vals': Zlength (csr_vals csr) = Znth (csr_rows csr) (csr_row_ptr csr))
    (CSR_wf_sorted: list_solver.sorted Z.le (0 :: csr_row_ptr csr ++ [Int.max_unsigned]))
    (CSR_wf_rowsorted: ∀ r, 0 ≤ r < csr_rows csr →
      sorted Z.lt
        (-1 :: sublist (Znth r (csr_row_ptr csr)) (Znth (r+1) (csr_row_ptr csr))
          (csr_col_ind csr)
          ++ [csr_cols csr])),
  csr_matrix_wellformed csr.

Inductive partial_CSR (h: ℤ) (r: ℤ) (coo: coo_matrix Tdouble)
  (rowptr: list val) (colind: list val) (val: list val) : Prop :=
  build_partial_CSR: ∀
    (partial_CSR_coo: coo_matrix_wellformed coo)
    (partial_CSR_coo_sorted: sorted coord_le (coo_entries coo))
    (partial_CSR_i: 0 ≤ h ≤ Zlength (coo_entries coo))
    (partial_CSR_r: -1 ≤ r ≤ coo_rows coo)
    (partial_CSR_r': Forall (fun e ⇒ fst (fst e) ≤ r) (coo_entries (coo_upto h coo)))
    (partial_CSR_r'': Forall (fun e ⇒ fst (fst e) ≥ r)
      (sublist h (Zlength (coo_entries coo)) (coo_entries coo)))
    (csr: csr_matrix Tdouble)
    (partial_CSR_wf: csr_matrix_wellformed csr)
    (partial_CSR_coo_csr: coo_csr (coo_upto h coo) csr)
    (partial_CSR_val: sublist 0 (Zlength (csr_vals csr)) val = map Vfloat (csr_vals csr))
    (partial_CSR_colind: sublist 0 (Zlength (csr_col_ind csr)) colind =
      map (Vint oo Int.repr) (csr_col_ind csr))
    (partial_CSR_rowptr: sublist 0 (r+1) rowptr =
      map (Vint oo Int.repr) (sublist 0 (r+1) (csr_row_ptr csr)))
    (partial_CSR_val': Zlength val = count_distinct (coo_entries coo))
    (partial_CSR_colind': Zlength colind = count_distinct (coo_entries coo))
    (partial_CSR_rowptr': Zlength rowptr = coo_rows coo + 1)
    (partial_CSR_dbound: count_distinct (coo_entries coo) ≤ Int.max_unsigned),
  partial_CSR h r coo rowptr colind val.

```

Listing 14: Wellformedness of CSR matrices and definition of partial_CSR

(Znth k (csr_vals csr)).

Definition no_extra_zeros {t} (coo: coo_matrix t) (csr: csr_matrix t) :=
 ∀ r k, 0 ≤ r < coo_rows coo →
 Znth r (csr_row_ptr csr) ≤ k < Znth (r+1) (csr_row_ptr csr) →
 let c := Znth k (csr_col_ind csr) in In (r,c) (map fst (coo_entries coo)).

Having established this coo_csr relation of a complete COO matrix to a complete CSR matrix, we can use it in relating partial matrices, as shown in Listing 14.

With this definition, proofs of all the partial_CSR lemmas proceed in a straightforward (though tedious) manner.

5 Conclusion

The program is not long, but its invariants are surprisingly intricate. The definitions, properties, and lemmas for `coo_csr` and `partial_CSR` took 1571 lines of Coq, and based on those the VST proof took 412 lines. In the process of developing the proof I found and fixed five bugs in the program—four off-by-one errors and one in tricky situation (discussed above) resulting from initializing the unsigned integer variable `r` to -1 (modulo 2 to the wordsize).

This proved-correct module is now ready to serve its intended purpose: as a composable verified component in any verified program—such as finite-element PDE solution—that requires construction of Compressed Sparse Row matrices.

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