

Time complexity analysis of + between two SparseVector

Consider two `SparseVector`'s. The first has nnz_1 non-zero values, and the second has nnz_2 non-zero values. The two vectors have respectively attributes `rowidx1`, `nzval1` and `rowidx2`, `nzval2`. We recall the pseudo-code of the algorithm :

```

1: newRowIdx ← new empty array of size  $nnz_1 + nnz_2$ 
2: newNzVal ← new empty array of size  $nnz_1 + nnz_2$ 
3: newNnz ← 0
4: Two pointers  $i \leftarrow 0, j \leftarrow 0$ 
5: while  $i \leq nnz_1$  and  $j \leq nnz_2$  do
6:   if rowidx1[ $i$ ] < rowidx2[ $j$ ] then
7:     newRowIdx[newNnz] ← rowidx1[ $i$ ]
8:     newNzVal[newNnz] ← nzval1[ $i$ ]
9:      $i \leftarrow i + 1$ 
10:    newNnz ← newNnz + 1
11:   else if rowidx1[ $i$ ] > rowidx2[ $j$ ] then
12:     newRowIdx[newNnz] ← rowidx2[ $j$ ]
13:     newNzVal[newNnz] ← nzval2[ $j$ ]
14:      $j \leftarrow j + 1$ 
15:     newNnz ← newNnz + 1
16:   else
17:     newRowIdx[newNnz] ← rowidx1[ $i$ ]
18:     newNzVal[newNnz] ← nzval1[ $i$ ] + nzval2[ $j$ ]
19:      $i \leftarrow i + 1$ 
20:      $j \leftarrow j + 1$ 
21:     newNnz ← newNnz + 1
22:   end if
23: end while
24: while  $i < nnz_1$  do
25:   newRowIdx[newNnz] ← rowidx1[ $i$ ]
26:   newNzVal[newNnz] ← nzval1[ $i$ ]
27:    $i \leftarrow i + 1$ 
28:   newNnz ← newNnz + 1
29: end while
30: while  $j < nnz_2$  do
31:   newRowIdx[newNnz] ← rowidx2[ $j$ ]
32:   newNzVal[newNnz] ← nzval2[ $j$ ]
33:    $j \leftarrow j + 1$ 
34:   newNnz ← newNnz + 1
35: end while
36: finalRowIdx ← new empty array of size newNnz
37: finalNzVal ← new empty array of size newNnz
38: Copy newNnz first elements of newRowIdx to finalRowIdx
39: Copy newNnz first elements of newNzVal to finalNzVal
40: Free newRowIdx and newNzVal
41: Initialize new SparseVector resVector with newNnz, finalRowIdx, finalNzVal
42: Free finalRowIdx and finalNzVal
    return resVector

```

From line 4 to line 23, the loop consists in two pointers i and j going from 0 to maximum either nnz_1 or nnz_2 . In any case, thanks to the assumption made in the statement, one execution inside the loop is done in $\mathcal{O}(1)$. The worst case is when the two pointers go respectively to nnz_1 and nnz_2 . So the total execution asymptotic time of this loop is $\mathcal{O}(nnz_1 + nnz_2)$. From line 24 to 29, the worst case is when the pointer i is still at 0. Also thanks to the assumption made in the statement, the total execution time of this loop is $\mathcal{O}(nnz_1)$. The exact same analysis can be made for the loop going from line 30 to 35, it has time complexity $\mathcal{O}(nnz_2)$. Finally, lines 38 and 39 are made in $\mathcal{O}(\text{newNnz})$ since copying m elements from an array to another is made in $\mathcal{O}(m)$. The same observation can be made for the initialization of the new `SparseVector` object in line 41. Thus the total execution time is made in

$\mathcal{O}(nnz_1 + nnz_2) + \mathcal{O}(nnz_1) + \mathcal{O}(nnz_2) + \mathcal{O}(\text{newNnz})$. Since $nnz_1 \leq nnz_1 + nnz_2$, $nnz_2 \leq nnz_1 + nnz_2$ and $\text{newNnz} \leq nnz_1 + nnz_2$, the total execution time is $\mathcal{O}(nnz_1 + nnz_2)$.

Space complexity analysis of + between two SparseVector

Memory is allocated in lines 1, 2, 36, 37 and 41 of the algorithm written above. In lines 1 and 2, $\mathcal{O}(nnz_1 + nnz_2)$ space is allocated. In lines 36, 37 and 41, $\mathcal{O}(\text{newNnz})$ space is allocated. Thus the total space allocated is $\mathcal{O}(nnz_1 + nnz_2) + \mathcal{O}(\text{newNnz})$. Since $\text{newNnz} \leq nnz_1 + nnz_2$, the total space allocated is $\mathcal{O}(nnz_1 + nnz_2)$. Note that in this analysis we do not consider when $\mathcal{O}(1)$ space is allocated such as in line 4.

Time complexity analysis of * between a SparseMatrix and a Vector

Consider the $*$ operator between a `SparseMatrix` with attributes $m, n, M, \text{rowidx}, \text{nzval}$, and a `Vector` of size n .

```

1: Initialize new zero Vector res of size m
2: for i ∈ {1, ..., n} do
3:   for j ∈ {1, ..., M} do
4:     k ← Mi + j
5:     if rowidx[k] = -1 then
6:       Break
7:     end if
8:     res[rowidx[k]] ← res[rowidx[k]] + nzval[k] × (i-th element of v)
9:   end for
10: end for

```

Note that in this analysis we consider that `rowidx` is a $n \times M$ matrix. The initialization in line 1 is made in $\mathcal{O}(m)$ time complexity since copying an array of size s in another is made in $\mathcal{O}(s)$. Then, from line 2 to 10, the two loops consist in browsing `rowidx` line by line, stopping each time that a -1 is reached. Thanks to the assumption made in the statement, each execution after having checked the value of `rowidx` is made in $\mathcal{O}(1)$. Suppose that $nnz \geq n$, then the time complexity from line 2 is $\mathcal{O}(nnz)$ since all non-zero values of a line is placed before the first -1 in `rowidx`. Now, suppose that $n > nnz$, then the algorithm still checks the values of `rowidx` at each line (which is also made in $\mathcal{O}(1)$). In this case, the time complexity is thus $\mathcal{O}(n)$. Thus the total time complexity is $\mathcal{O}(m) + \mathcal{O}(\max(n, nnz))$. If $m \geq \max(n, nnz)$, then the time complexity is $\mathcal{O}(m)$, otherwise it is $\mathcal{O}(\max(n, nnz))$. We can summarize this by saying that the time complexity is $\mathcal{O}(\max(m, n, nnz))$.

Space complexity analysis of * between a SparseMatrix and a Vector

The only space allocation is made in line 1. It allocates $\mathcal{O}(m)$ space to store the solution. Note that $\mathcal{O}(m) = \mathcal{O}(\max(m, n, nnz))$. Indeed if $m = \max(m, n, nnz)$, then it is direct that $\mathcal{O}(m) = \mathcal{O}(\max(m, n, nnz))$, otherwise $\mathcal{O}(m) = \mathcal{O}(\max(m, n, nnz))$ since $m < \max(m, n, nnz)$.

Description of the memory states during the call of = of SparseVector

First, note that the variables `mSize` and `nnz` of `this` always point to the same place in memory, but their value might change. Concerning the variables `rowidx` and `nzval`, there are two cases. Either the value of `nnz` of `this` is the same as the value of `nnz` of `otherVector`, or it is not the case. In the first case, the places in memory to which `rowidx` and `nzval` of `this` point do not change, although new information from `otherVector` is copied in these places. In the second case (`nnz` are not the same), the places to which `rowidx` and `nzval` point are freed. New places in memory are then allocated, and `rowidx` and `nzval` now point to these new places. Again, information from `otherVector` is copied in these new places.

In the case of `rowidx`, an illustration is given in Figure 1.

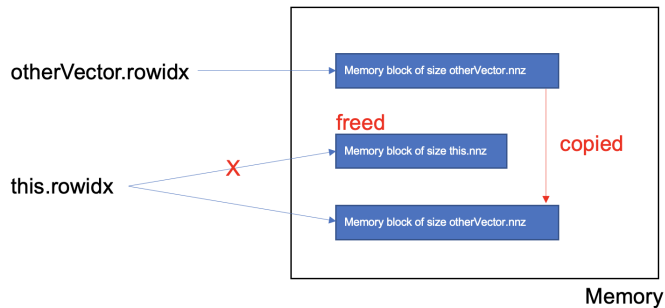


Figure 1: `=` operator memory modifications if `this.nnz` and `otherVector.nnz` have not identical values