

Sparse Matrices

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Outline

Coordinate format (COO)

Compressed Sparse Row (CSR)

Compressed Sparse Column (CSC)

Matrix times vector

$$\text{COO } y = Ax$$

$$\text{CSR, CSC } y = Ax$$

$$y = A^T x$$

$$\text{CSR } y = A^T x$$

Example 1.

$$A = \begin{bmatrix} 0 & 1 & 0 & 2 & 0 \\ 3 & 0 & 0 & 0 & 0 \\ 0 & 0 & 4 & 0 & 0 \\ 5 & 0 & 6 & 7 & 0 \\ 0 & 0 & 0 & 0 & 8 \end{bmatrix}$$

Store nonzeros only.

Representations

- Coordinate or triplet format, COO
- Compressed Sparse Row (CSR), also called Compressed Row Storage (CRS)
- Compressed Sparse Column (CSC), also called Compressed Column Storage (CCS)

Coordinate format (COO)

Store nonzero a_{ij} as the triplet (i, j, a_{ij}) .

Example 2.

$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \left[\begin{array}{ccccc} & 1 & & & \\ 3 & & & 2 & \\ & & 4 & & \\ 5 & & 6 & 7 & \\ & & & & 8 \end{array} \right] \end{matrix}$$

row_index	1	1	2	3	4	4	4	5
col_index	2	4	1	3	1	3	4	5
value	1	2	3	4	5	6	7	8

array	stores
row_index	row indexes
col_index	column indices
value	nonzeros

Compressed Sparse Row (CSR)

From COO, compress the row indices as a `row_pointer` array.

Example 3.

row_index	1	1	2	3	4	4	4	5	
offset	1	2	3	4	5	6	7	8	9
row_pointer	1	3	4	5	8	9			

Assume A is $m \times n$. Denote the number of nonzeros by `nnz`.

- `row_pointer`($i + 1$) – `row_pointer`(i) is number of nonzeros in row i
- `nnz` = `row_pointer`($m + 1$) – `row_pointer`(1) is total number of nonzeros.

Example 3. cont.

$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \left[\begin{array}{ccccc} & 1 & & & 2 \\ 3 & & & & \\ & & 4 & & \\ 5 & & 6 & 7 & \\ & & & & 8 \end{array} \right] \end{matrix}$$

CSR

row_pointer

1	3	4	5	8	9
---	---	---	---	---	---

col_index
value

2	4	1	3	1	3	4	5
1	2	3	4	5	6	7	8

Compressed Sparse Column (CSC)

Compress column indices into a `col_pointer` array.

Example 4.

$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{matrix} & \begin{bmatrix} & 1 & & & \\ & & 1 & & 2 \\ 3 & & & & \\ & & & 4 & \\ 5 & & 6 & 7 & \\ & & & & 8 \end{bmatrix} \end{matrix}$$

CSC

`col_pointer`

1	3	4	6	8	9
---	---	---	---	---	---

offset

1 2 3 4 5 6 7 8 9

`row_index`

2	4	1	3	4	1	4	5
---	---	---	---	---	---	---	---

value

3	5	1	4	6	2	7	8
---	---	---	---	---	---	---	---

Matrix times vector

Consider $y = Ax$, where A is $m \times n$.

$$y_i = \sum_{j=1}^n a_{ij}x_j, \quad i = 1 : m.$$

In the algorithms that follow, assume that initially the entries of y are set to 0.

Algorithm 1 ($y = Ax$).

for $i = 1 : m$

 for $j = 1 : n$

$y(i) += A(i, j) * x(j)$

COO CSR CSC $y = Ax$ $y = A^T x$

COO $y = Ax$

If A is in COO format, $y = Ax$ can be computed as

Algorithm 2 (COO $y = Ax$).

for $k = 1 : \text{nnz}$

$i = \text{row_index}(k)$

$j = \text{col_index}(k)$

$y(i) += \text{value}(k) * x(j)$

COO CSR CSC $y = Ax$ $y = A^T x$

CSR, CSC $y = Ax$

Algorithm 3 (CSR $y = Ax$).

```
for  $i = 1 : m$ 
    for  $k = \text{row\_pointer}(i) : \text{row\_pointer}(i + 1) - 1$ 
         $j = \text{col\_index}(k)$ 
         $y(i) += \text{value}(k) * x(j)$ 
```

Algorithm 4 (CSC $y = Ax$).

```
for  $j = 1 : n$ 
    for  $k = \text{col\_pointer}(j) : \text{col\_pointer}(j + 1) - 1$ 
         $i = \text{row\_index}(k)$ 
         $y(i) += \text{value}(k) * x(j)$ 
```

$$y = A^T x$$

Consider $y = A^T x$, where A is $m \times n$; A^T is $n \times m$.

$$y_i = \sum_{j=1}^m a_{ji} x_j, \quad i = 1 : n$$

Algorithm 5 ($y = A^T x$).

for $i = 1 : n$

 for $j = 1 : m$

$$y(i) += A(j, i) * x(j)$$

Inefficient if we process by rows. Swap the loops.

Algorithm 6 ($y = A^T x$).

for $j = 1 : m$

 for $i = 1 : n$

$$y(i) += A(j, i) * x(j)$$

Algorithm 7 (CSR $y = A^T x$).

for $j = 1 : m$

 for $k = \text{row_pointer}(j) : \text{row_pointer}(j + 1) - 1$

$i = \text{col_index}(k)$

$y(i) += \text{value}(k) * x(j)$