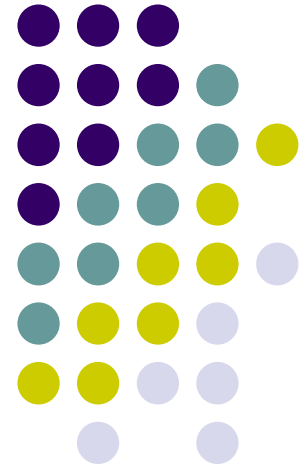
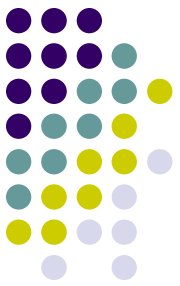


Applied Scaling & Classification Techniques in Political Science

A compressed sparse matrix





Sparsity problem

In text mining, huge matrices are created based on word frequencies with many cells having zero values

This problem is called **sparsity**

Dense Matrix

1	2	31	2	9	7	34	22	11	5
11	92	4	3	2	2	3	3	2	1
3	9	13	8	21	17	4	2	1	4
8	32	1	2	34	18	7	78	10	7
9	22	3	9	8	71	12	22	17	3
13	21	21	9	2	47	1	81	21	9
21	12	53	12	91	24	81	8	91	2
61	8	33	82	19	87	16	3	1	55
54	4	78	24	18	11	4	2	99	5
13	22	32	42	9	15	9	22	1	21

Sparse Matrix

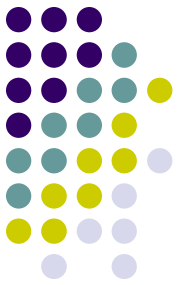
1	.	3	.	9	.	3	.	.	.
11	.	4	2	1
.	.	1	.	.	.	4	.	1	.
8	.	.	.	3	1
.	.	.	9	.	.	1	.	17	.
13	21	.	9	2	47	1	81	21	9
.
.	.	.	.	19	8	16	.	.	55
54	4	.	.	.	11
.	.	2	22	.	21

Sparsity problem



The challenge in this sense is how to store a sparse matrix in a clever way (w/o all those 0s!) so that it occupies less space

Being able of doing it can substantially (and I really mean that!) reduce the time spent by your computer in running any algorithm



Sparsity problem

There are several way of doing it

Quanteda stores the sparse matrix in a clever (and compressed) way! How?

One possibility (just as an illustration): the CSR (the compressed sparse row)

	0	1	2	3
0	a		b	c
1		d		
2			e	f
3				g

Compressed Sparse Row (CSR)

0	3	4	6	7
---	---	---	---	---

Row Pointers

0	2	3	1	2	3	3
---	---	---	---	---	---	---

Column Indices

a	b	c	d	e	f	g
---	---	---	---	---	---	---

Data values



Sparsity problem

For the row pointers: always add a “0” at the beginning,
and then for each row count the number of non-0 values
and keep adding

	0	1	2	3
0	a		b	c
1		d		
2			e	f
3				g

Compressed Sparse Row (CSR)

0	3	4	6	7
---	---	---	---	---

Row Pointers

0	2	3	1	2	3	3
---	---	---	---	---	---	---

Column Indices

a	b	c	d	e	f	g
---	---	---	---	---	---	---

Data values

Sparsity problem



However note that not all the packages in R allows you to work with compressed sparse matrices (and this could be a problem) – for example `randomForest`