

HW: Latent Semantic Analysis (LSA)

- Given a sparse term-doc matrix
 - E.g., 4 terms and 3 docs

Term	Doc		
	2.3	0.0	4.2
	0.0	1.3	2.2
	3.8	0.0	0.5
	0.0	0.0	0.0

- Each entry is weighted by $TF \times IDF$ score

Row #Tem	Col. # Doc	Nonzero entries
4	3	6
2	2	2 nonzero entries at Col 0
0	2.3	Col 0, Row 0
2	3.8	Col 0, Row 2
1	1	1 nonzero entry at Col 1
1	1.3	Col 1, Row 1
3	3	3 nonzero entries at Col 2
0	4.2	Col 2, Row 0
1	2.2	Col 2, Row 1
2	0.5	Col 2, Row 2

- Perform SVD to obtain corresponding term and doc vectors represented in the latent semantic space
- Evaluate the information retrieval capability of the LSA approach by using varying sizes (e.g., 100, 200, ..,600 etc.) of LSA dimensionality

HW: Latent Semantic Analysis (cont.)

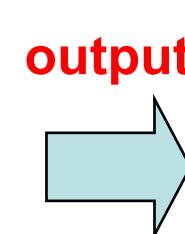
- Example: term-docmatrix

Indexing Term no.	Doc no.	Nonzero entries
51253	2265	218852
77		
508	7.725771	
596	16.213399	
612	13.080868	
709	7.725771	
713	7.725771	
744	7.725771	
1190	7.725771	
1200	16.213399	
1259	7.725771	
.....		

- SVD command (IR_svd.bat)

```
svd -r st -o LSA100 -d 100 Term-Doc-Matrix
```

sparse matrix input prefix of output files No. of reserved eigenvectors name of sparse matrix input



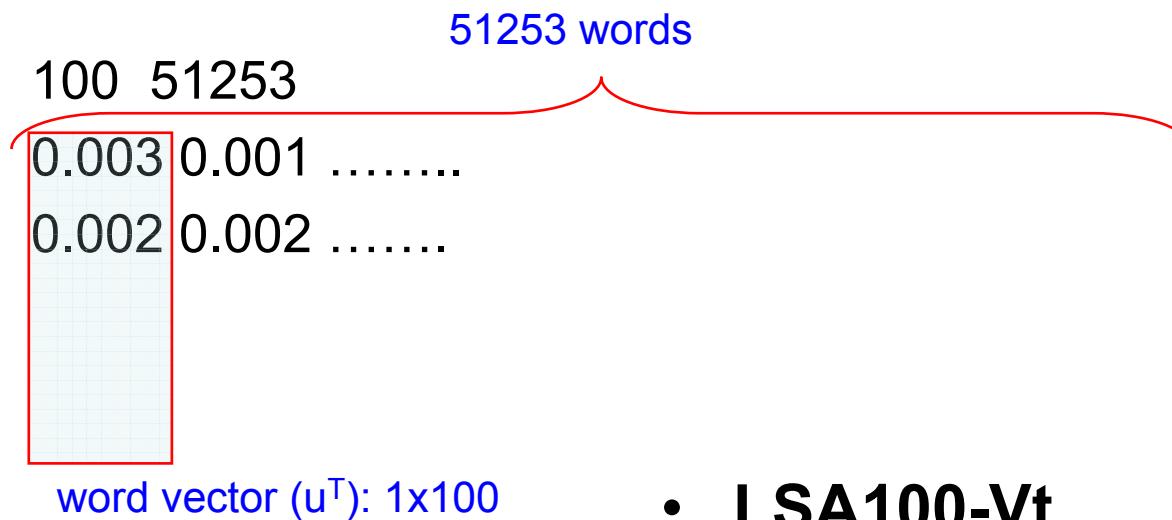
LSA100-Ut

LSA100-S

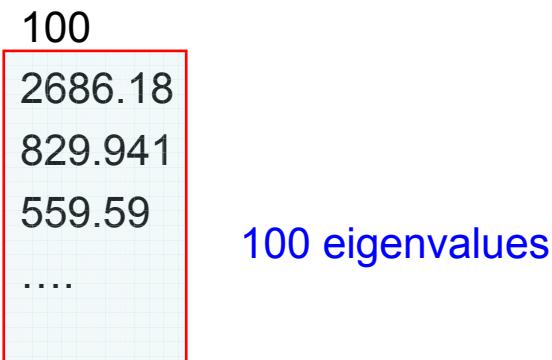
LSA100-Vt

HW: Latent Semantic Analysis (cont.)

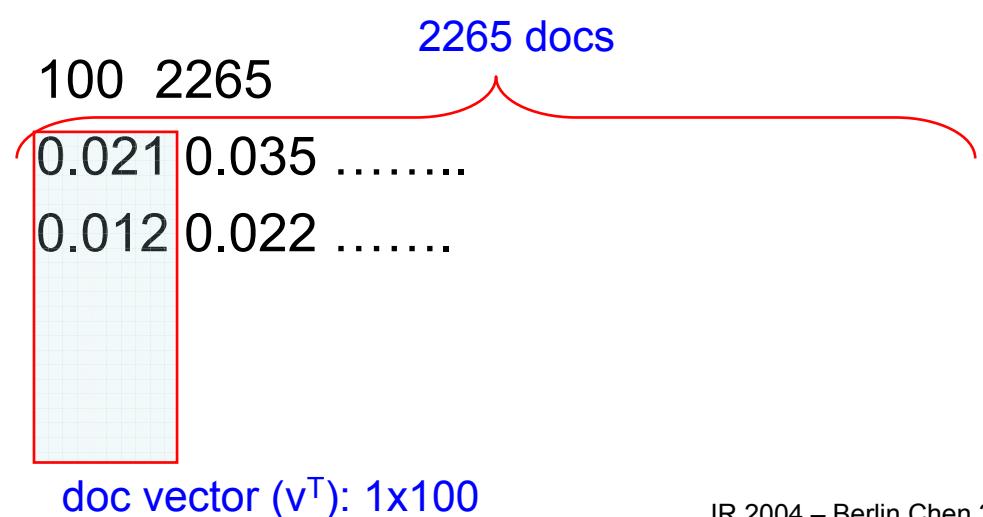
- **LSA100-Ut**



- **LSA100-S**



- **LSA100-Vt**



HW: Latent Semantic Analysis (cont.)

- Fold-in a new $m \times 1$ query vector

$$\hat{q}_{1 \times k} = \begin{pmatrix} q^T \\ 1 \times m \end{pmatrix} U_{m \times k} \Sigma_{k \times k}^{-1}$$

The separate dimensions are differentially weighted

Just like a row of V

Query represented by the weighted sum of its constituent term vectors

TFxIDF weighted beforehand

- Cosine measure between the query and doc vectors in the latent semantic space

$$sim(\hat{q}, \hat{d}) = \text{coincide}(\hat{q}\Sigma, \hat{d}\Sigma) = \frac{\hat{q}\Sigma^T \hat{d}}{\|\hat{q}\Sigma\| \|\hat{d}\Sigma\|}$$

SVDLIBC

- Doug Rohde's SVD C Library version 1.3 is based on the SVDPACKC library
- Download it at <http://tedlab.mit.edu/~dr/>