

Preprocessing

Document 1: To err is human, but to really foul things up you need a computer

Document 2: Computer science is no more about computers than astronomy is about telescopes

Document 3: A computer once beat me at chess, but it was no match for me at kick boxing



Stem nouns to singular

Document 1: To err is human, but to really foul **thing** up you need a computer

Document 2: Computer science is no more about **computer** than astronomy is about **telescope**

Document 3: A computer once beat me at chess, but it was no match for me at kick boxing



Stoplist: a about at but for is it me than thing to was you

Document 1: ~~To~~ err ~~is~~ human, ~~but-to~~ really foul **thing** up ~~you~~ need ~~a~~ computer

Document 2: Computer science ~~is~~ no more ~~about~~ **computer** ~~than~~ astronomy ~~is-about~~ **telescope**

Document 3: ~~A~~ computer once beat ~~me-at~~ chess, ~~but-it-was~~ no match ~~for-me-at~~ kick boxing



Remove terms that occur once

Document 1: ~~To~~ err ~~is~~ **human**, ~~but-to~~ really foul **thing** up ~~you~~ need ~~a~~ computer

Document 2: Computer **science** ~~is~~ no ~~more~~ ~~about~~ **computer** ~~than~~ **astronomy** ~~is-about~~ **telescope**

Document 3: ~~A~~ computer ~~once-beat-me-at~~ chess, ~~but-it-was~~ no ~~match~~ ~~for-me-at~~ kick boxing

Bag of Words: Document by Term Matrix

Document 1: ~~To~~ ~~err~~ ~~is~~ ~~human~~, ~~but~~ ~~to~~ ~~really~~ ~~foul~~ ~~thing~~ ~~up~~ ~~you~~ ~~need~~ ~~a~~ computer

Document 2: Computer science ~~is~~ ~~no~~ ~~more~~ ~~about~~ computer ~~than~~ ~~astronomy~~ ~~is~~ ~~about~~ ~~telescope~~

Document 3: ~~A~~ computer ~~once~~ ~~beat~~ ~~me~~ ~~at~~ ~~chess~~, ~~but~~ ~~it~~ ~~was~~ ~~no~~ ~~match~~ ~~for~~ ~~me~~ ~~at~~ ~~kick~~ ~~boxing~~

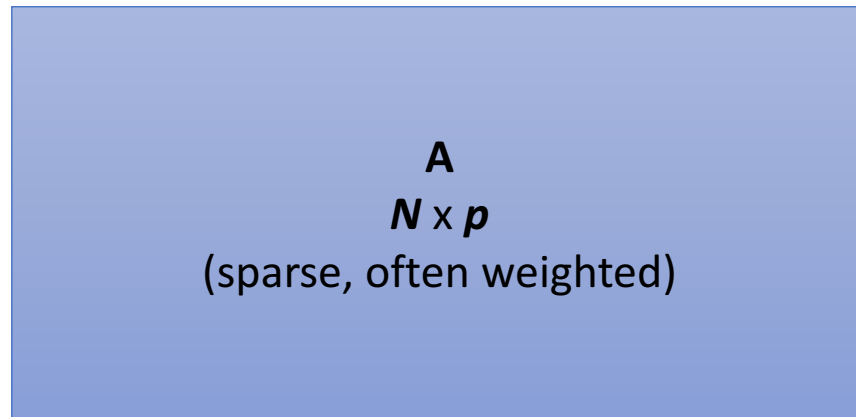


	computer	no
Document 1	1	0
Document 2	2	1
Document 3	1	1

Bag of Words: Matrix Factorization

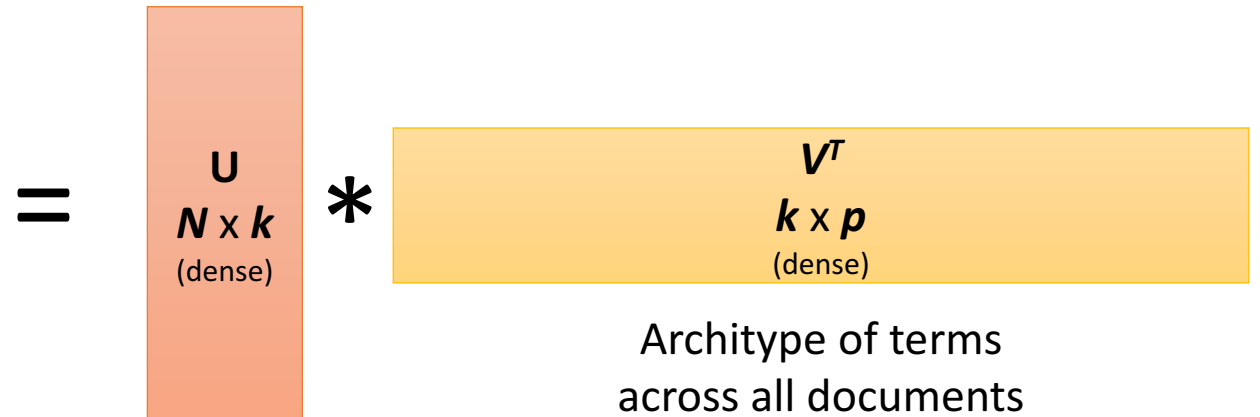
Select k features to represent the corpus

$$A = U \Sigma V^T$$



Term by Document Matrix
 N terms
 p documents

Architypes of documents
across all terms

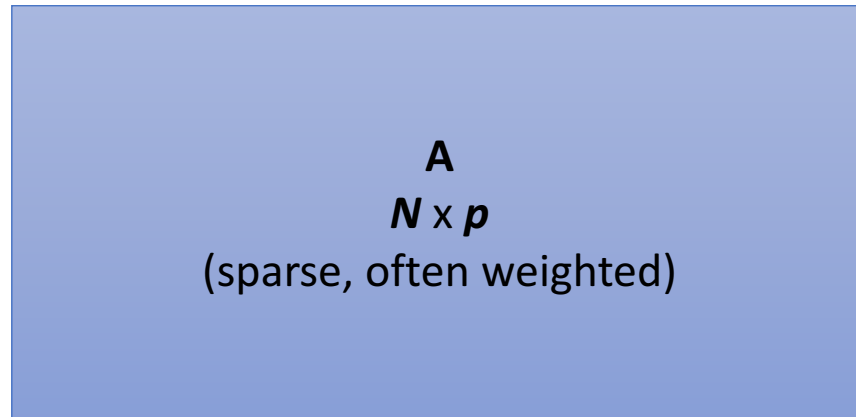


Architype of terms
across all documents

Bag of Words: Matrix Factorization

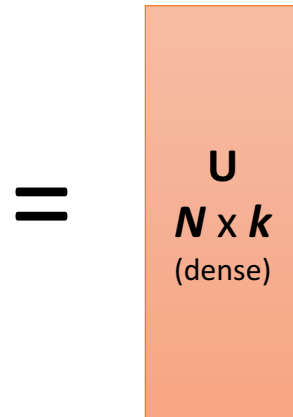
Select k features to represent the corpus

$$A = U \Sigma V^T$$



Term by Document Matrix
 N terms
 p documents

Archetypes of documents
across all terms

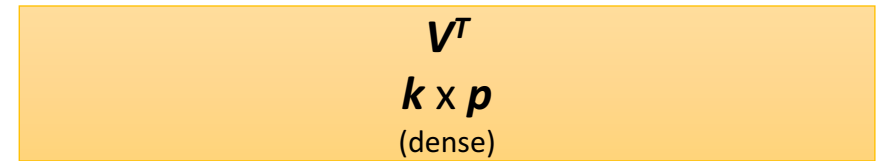


Best for analyzing the relationship between
topics in the documents and each term, i.e.
topics composed of similar terms

=

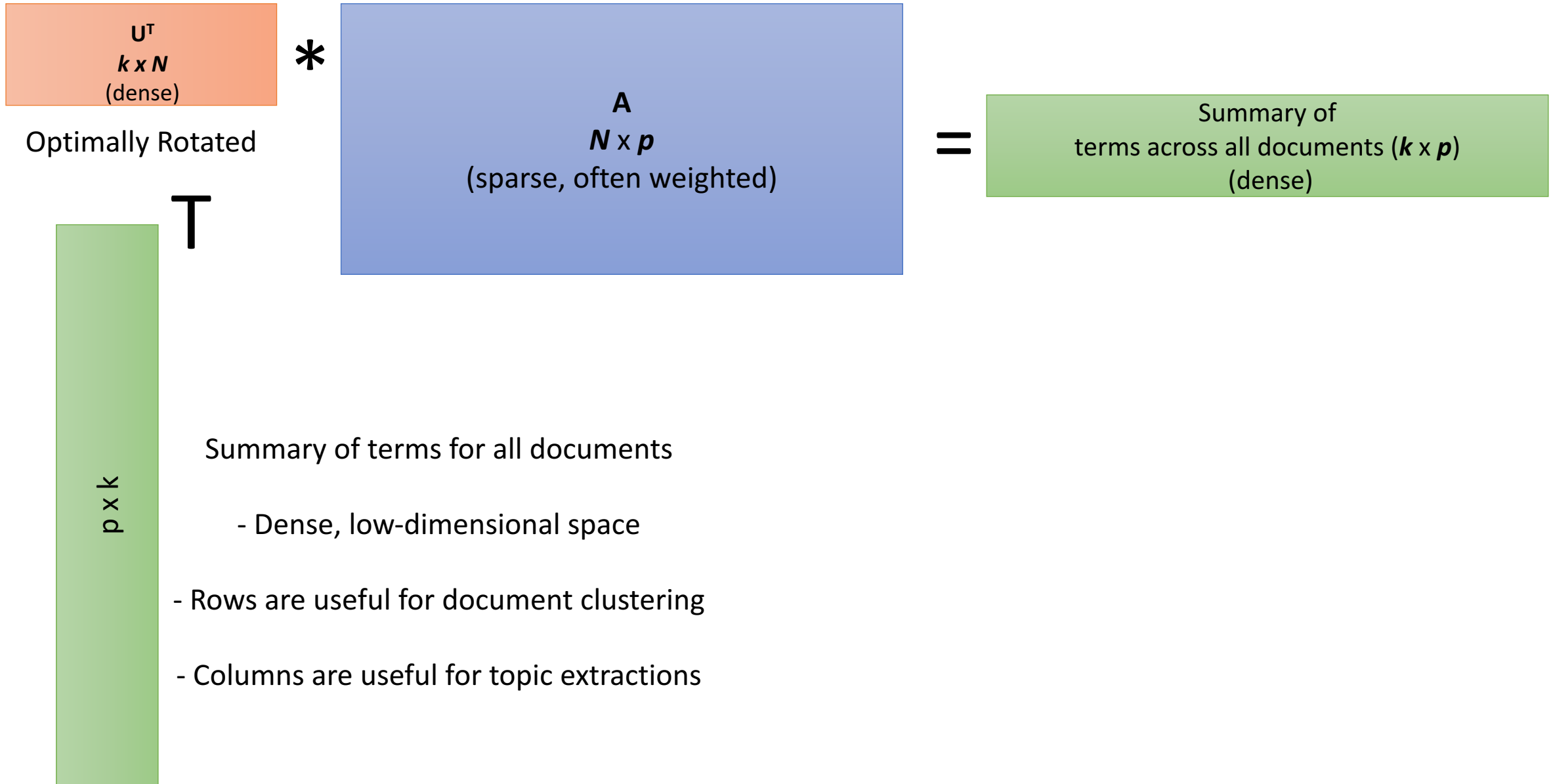
*

Best for analyzing the relationship
between topics in the documents and
each document, i.e. document clusters



Archetype of terms
across all documents

Bag of Words: Matrix Factorization: document clustering and topic extraction in SAS Text Miner



Content-Sensitive: Matrix Factorization: Term Embedding: GloVe

	Term 1	Term 2	Term 3	Term 4	Term 5	...
Term1	90	2	0	1	0	...
Term 2	2	56	1	6	0	...
Term 3	0	1	78	0	1	...
Term 4	1	6	0	24	0	...
⋮	⋮	⋮	⋮	⋮	⋮	⋮

**Sparse, wide, fixed-length vectors that record
term co-occurrence**

**Matrix
Factorization**

log bilinear
weighted least squares

	Factor 1	Factor 2	Factor 3
Term 1	1.304	0.582	0.892
Term 2	0.897	0.843	0.885
Term 3	0.745	1.129	1.002
Term 4	0.921	0.962	0.714
⋮	⋮	⋮	⋮

**Dense, fixed-length vectors for each
term in the corpus**

Each row vector
represents a term
("distributed
representation")

The output of a hidden layer of a neural network is used to embed terms into a fixed-length vector space from a simple encoding

