

## Matrix Decomposition

- Singular Value Decomposition
  - Is a technique that splits a matrix into three components with these properties.

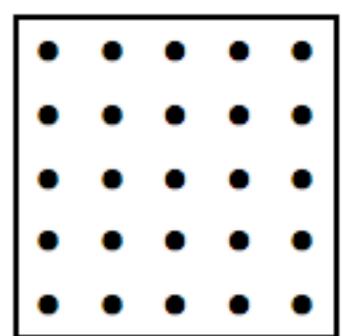
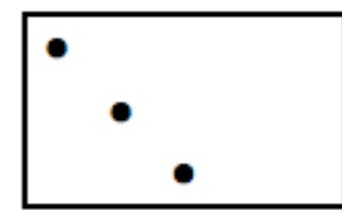
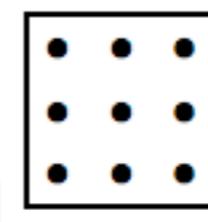
$$C = U \Sigma V^T$$



## Matrix Decomposition

- Singular Value Decomposition
  - SVD reduces the size of your term-document matrix
    - reduces the **dimensionality** or the **rank**
    - you can arbitrarily reduce the dimensionality by putting zeros in the bottom right of sigma
    - this is the mathematically optimal way of reducing dimensions

$$U \Sigma V^T$$



## Matrix Decomposition

- Singular Value Decomposition
  - If the old dimensions were based on **terms**
    - after reducing the rank of the matrix the dimensionality is based on **concepts** or **semantics**
    - a concept is a **linear combination** of terms

$$SVD_{dimension_1} = a * td_{dim_1} + b * td_{dim_2} + c * td_{dim_3} + d * td_{dim_4}$$

$$SVD_{dimension_2} = a' * td_{dim_1} + b' * td_{dim_2} + c' * td_{dim_3} + d' * td_{dim_4}$$

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- Singular Value Decomposition

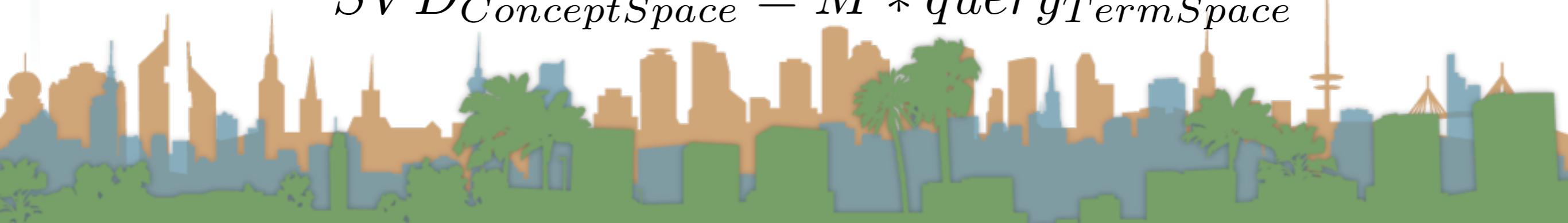
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$$SVD_{ConceptSpace} = M * queryTermSpace$$



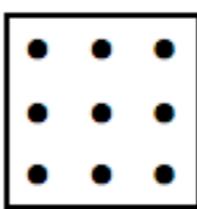
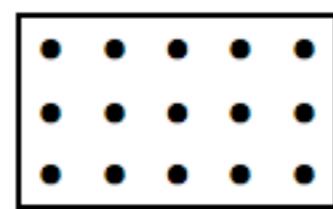
## Matrix Decomposition

- Singular Value Decomposition

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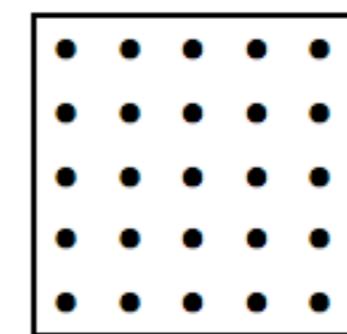
$$C = U \Sigma V^T$$



$$\Sigma$$



$$V^T$$



## Matrix Decomposition

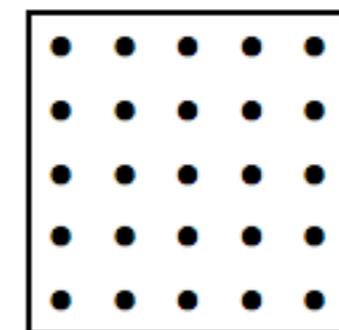
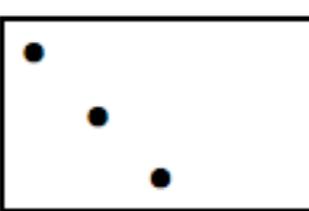
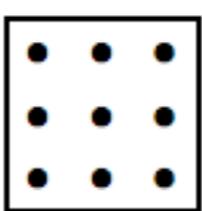
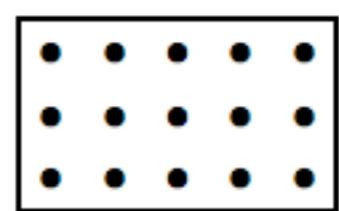
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$$SVD_{ConceptSpace} = M * queryTermSpace$$

$$M = \Sigma_k^{-1} U_k^T$$

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$$SVD_{ConceptSpace} = M * queryTermSpace$$

$$M = \Sigma_k^{-1} U_k^T$$

$$queryConceptSpace = \Sigma_k^{-1} U_k^T queryTermSpace$$



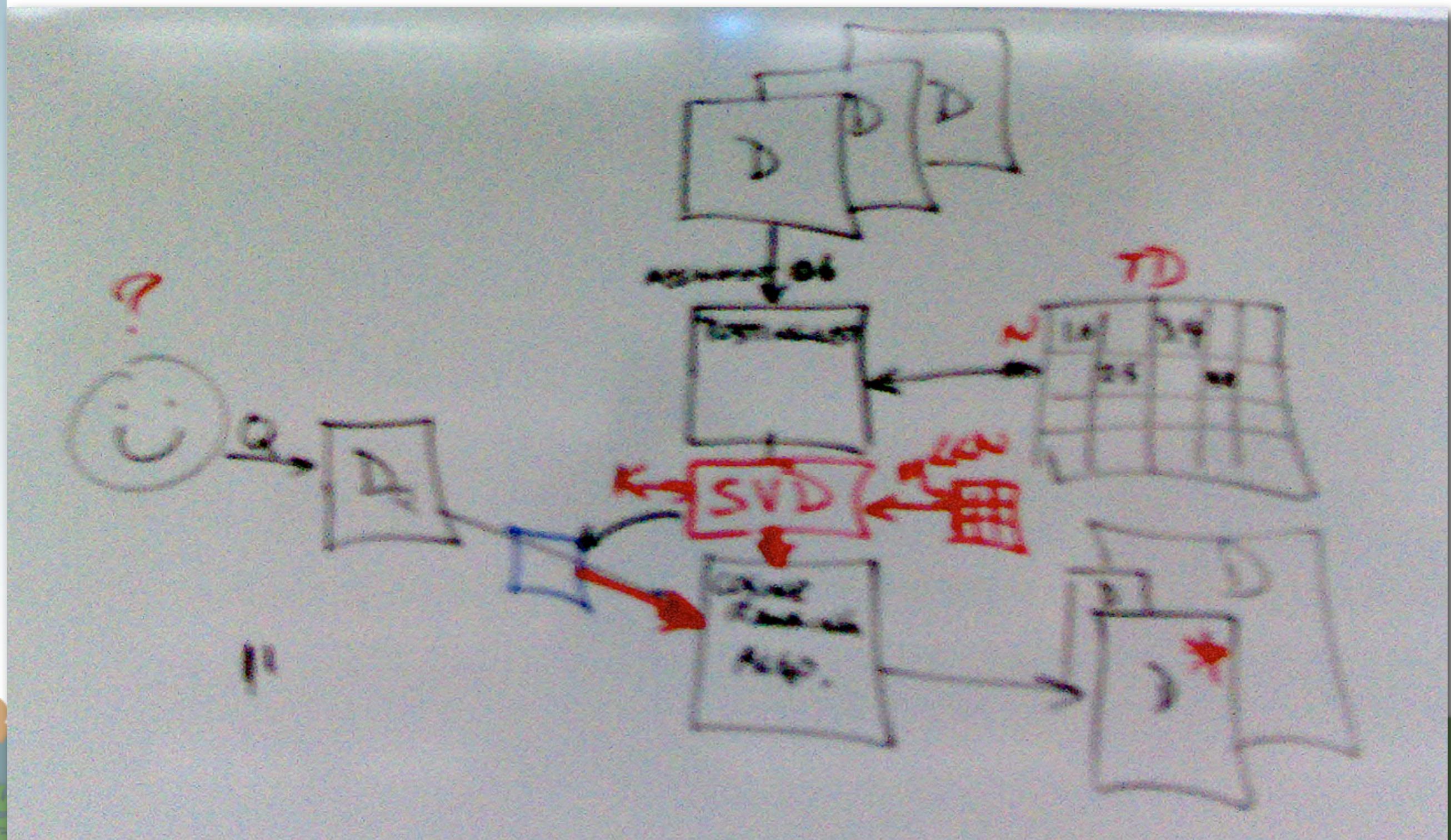
## Matrix Decomposition

- Singular Value Decomposition
  - SVD is an algorithm that gives us  $\Sigma \ U \ V^T$
  - With these quantities we can reduce dimensionality
  - With reduced dimensionality
    - **synonyms are mapped onto the same location**
      - “bat” “chiroptera”
    - **polysemies are mapped onto different locations**
      - “bat” (baseball) vs. “bat” (small furry mammal)



## Matrix Decomposition

- Big Picture



## Matrix Decomposition

- “I am not crazy”
- Netflix



## Matrix Decomposition

- “I am not crazy”
  - Netflix
  - Machine translations
    - Just like “bat” and “chiroptera” map the same
    - “bat” and “murciélagos” can map to the same thing

The math is hard but it's beautiful and powerful

La matemáticas es dura pero es hermosa y de gran alcance

Jene mathematisch ist hart, aber ist und an langer Reichweite schön

That one mathematically is hard, but is beautiful and at long range

