

# Tensors and low rank tensor decompositions

$S_{ij}$  = score of person  $i$  on test  $j$

	Classics	Math	Music	...
Alice	19	26	17	...
Bob	8	17	9	...
Carol	7	12	7	...
⋮	⋮	⋮	⋮	⋮

rank = 2  
 $S = \vec{x}_{quant} \vec{y}_{quant}^T + \vec{x}_{verb} \vec{y}_{verb}^T$

$S_{ij} \approx \underbrace{\vec{x}_{quant,i}}_{\text{person } i \text{ quant. meas.}} \cdot \underbrace{\vec{y}_{quant,j}}_{\text{test } j \text{ quant. meas.}} + \vec{x}_{verb,i} \cdot \vec{y}_{verb,j}$

	Classics	Math	Music	...
Alice	19	26	17	...
Bob	8	17	9	...
Carol	7	12	7	...
⋮	⋮	⋮	⋮	⋮

=

	Quantitative	Verbal
Alice	4	3
Bob	3	1
Carol	2	1
⋮	⋮	⋮

=

	Classics	Math	Music	...
Quantitative	1	5	2	...
Verbal	5	2	3	...

$\vec{x}_{quant}$        $\vec{y}_{quant}$

$S = \vec{x}_1 \vec{y}_1^T + \vec{x}_2 \vec{y}_2^T$

$19 = 4 \cdot 1 + 3 \cdot 5$

$S = \begin{bmatrix} 1 & 1 \\ 4 & 3 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} -y_1^T \\ -y_2^T \end{bmatrix}$

	Classics	Math	Music	...
Alice	19	26	17	...
Bob	8	17	9	...
Carol	7	12	7	...
⋮	⋮	⋮	⋮	⋮

=

	Quantitative	Verbal
Alice	1	3
Bob	2	1
Carol	1	1
⋮	⋮	⋮

=

	Classics	Math	Music	...
Quantitative	1	5	2	...
Verbal	6	7	5	...

$$S = \vec{X}_{\text{quant}} \otimes \vec{y}_{\text{quant}} \otimes \vec{z}_{\text{quant}} + \vec{X}_{\text{verb}} \otimes \vec{y}_{\text{verb}} \otimes \vec{z}_{\text{verb}}$$

$$\vec{x} \in \mathbb{R}^{n_1}$$

$n_1 = \# \text{ people}$

$$\vec{y} \in \mathbb{R}^{n_2}$$

$n_2 = \# \text{ subjects}$

$$\vec{z} \in \mathbb{R}^{n_3}$$

$n_3 = 2$  (# times of day)

	Quantitative	Verbal
Alice	4	3
Bob	3	1
Carol	2	1
⋮	⋮	⋮

$\vec{X}_{\text{quant},1}$

⊗

	Quantitative	Verbal
Classics	1	5
Math	5	2
Music	2	3
⋮	⋮	⋮

$\vec{y}_{\text{quant},1}$

⊗

	Quantitative	Verbal
Day	1	1
Night	2	1
⋮	⋮	⋮

$\vec{z}_{\text{quant},1}$   
 day ↘  
 night ↙

$S_{ijk} = \text{score of person } i \text{ on test } j \text{ at time } k \in \{1, 2\}$

$n_1 \times n_2 \times n_3$

$$S_{ijk} = \vec{X}_{\text{quant},i} \cdot \vec{y}_{\text{quant},j} \cdot \vec{z}_{\text{quant},k} + \vec{X}_{\text{verb},i} \cdot \vec{y}_{\text{verb},j} \cdot \vec{z}_{\text{verb},k}$$

$H_{i,j,k} (*)$

Cor: If  $\{\vec{X}_{\text{quant}}, \vec{X}_{\text{verb}}\}$ ,  $\{\vec{y}_{\text{quant}}, \vec{y}_{\text{verb}}\}$ ,  $\{\vec{z}_{\text{quant}}, \vec{z}_{\text{verb}}\}$ ,  
 lin indep                      lin indep                      lin indep

then (\*) is unique (up to scaling)

Def. A  $n_1 \times n_2 \times \dots \times n_k$  array  $A \in \mathbb{R}^{n_1 \times n_2 \times \dots \times n_k}$  is called a k-tensor.

vector = 1-tensor

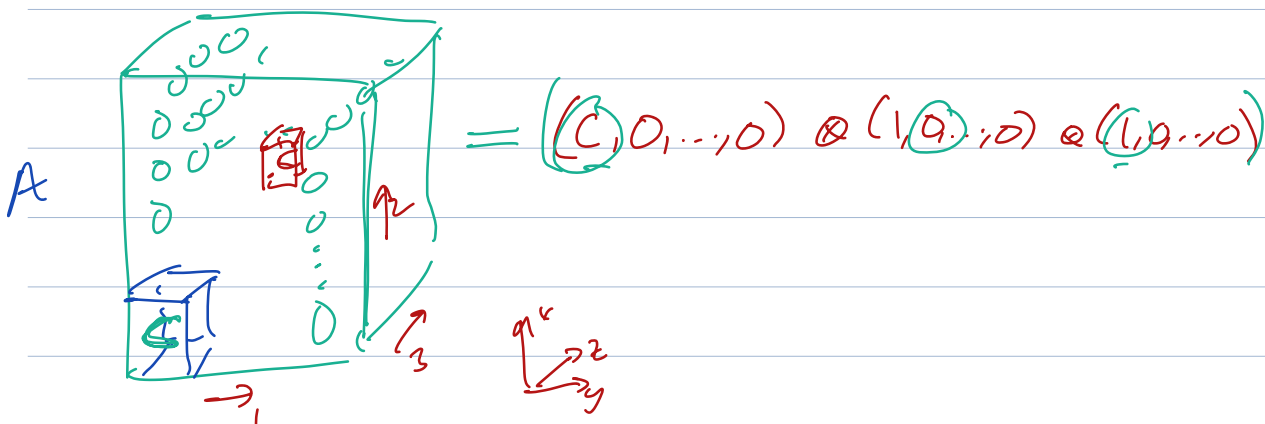
matrix = 2-tensor

Rank: A rank-1 k-tensor is of the form

$$(\vec{u}_1 \otimes \vec{u}_2 \otimes \vec{u}_3 \otimes \dots \otimes \vec{u}_k)_{i_1 i_2 \dots i_k} \\ := (\vec{u}_1)_{i_1} (\vec{u}_2)_{i_2} \dots (\vec{u}_k)_{i_k}$$

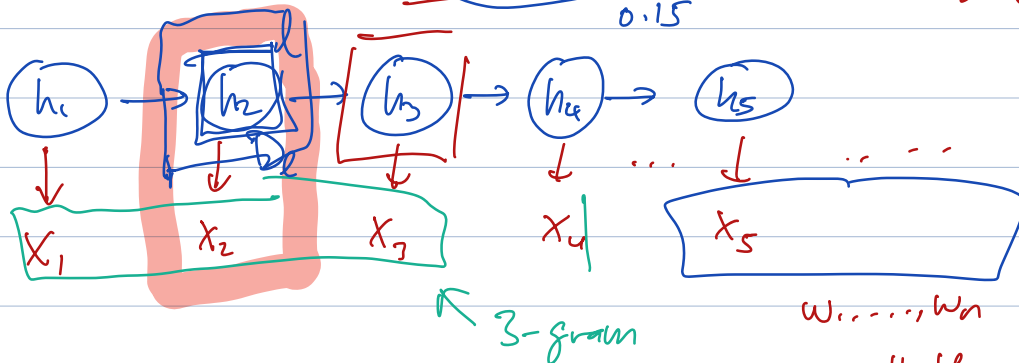
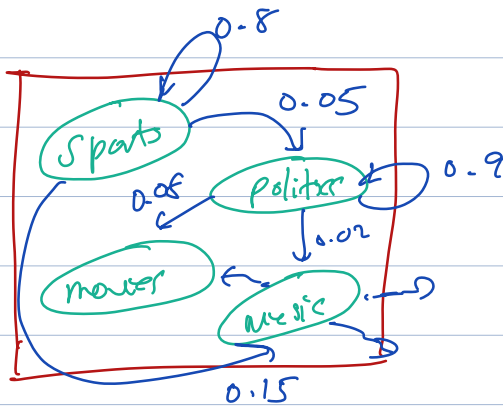
$$\vec{u}_1 \otimes \vec{u}_2 \otimes \vec{u}_3 \otimes \dots \otimes \vec{u}_k \\ \vec{u}_i \in \mathbb{R}^{n_i}, i=1, \dots, k$$

Rank of a k-tensor  $A$  = smallest  $r$  st. we can write  $A$  as a sum of rank-1 tensors



HMM: Topics

$T = \# \text{ topics}$



$$A_{ijk} = P[X_1 x_2 x_3 = w_i w_j w_k]$$

$$= \sum_{l=1}^T P[h_2 = l] \cdot \left. \begin{array}{l} P[X_2 = w_j | h_2 = l] \\ P[X_3 = w_k | h_2 = l] \\ P[X_1 = w_i | h_2 = l] \end{array} \right\}$$

$$A = \sum_{l=1}^T \lambda_l (g_l \otimes x_l \otimes y_l)$$

$$(g_l)_j = P[X_2 = w_j | h_2 = l]$$

$$(x_l)_k = P[X_3 = w_k | h_2 = l] \quad \dots$$







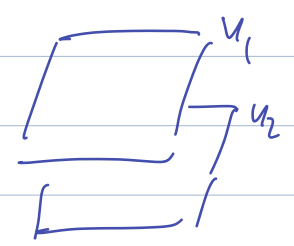
$n \times n \times n$  tensor as input

Alg: Look at random slices of the tensor.

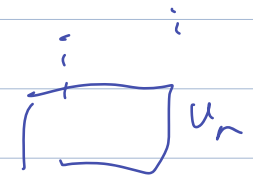
$$T = \sum_{i=1}^r \lambda_i (\vec{x}_i \otimes \vec{y}_i \otimes \vec{z}_i)$$

Pick two random vectors  $\vec{u}, \vec{v}$

(1)  $T_{\vec{u}} = \sum_{i=1}^n u_i T[:, :, i]$



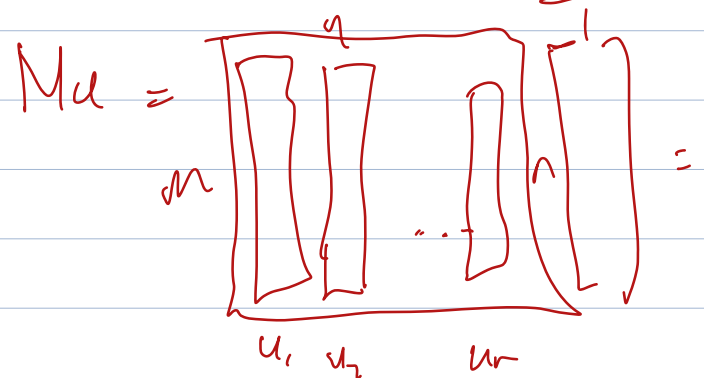
(2)  $T_{\vec{v}} = \sum_{i=1}^n v_i T[:, :, i]$



(3)  $\vec{x}_i =$  eigenvectors of  $T_{\vec{u}} (T_{\vec{u}})^{-1}$

$\vec{y}_i =$  eigenvectors of  $T_{\vec{v}} (T_{\vec{v}})^{-2}$

Note that matrix-vec mult is also  
a slice





$$\boxed{T_{\vec{u}} = X D_{\vec{u}} Y^T \quad T_{\vec{v}} = X D_{\vec{v}} Y^T}$$

"Simultaneous" SVD is unique (!)

$$T_{\vec{u}} (T_{\vec{v}})^{-1} = X D_{\vec{u}} Y^T Y D_{\vec{v}}^{-1} X^{-1}$$

$$= X D_{\vec{u}} D_{\vec{v}}^{-1} X^{-1}$$

←—————

eigenvalues give the  $X$ 's