

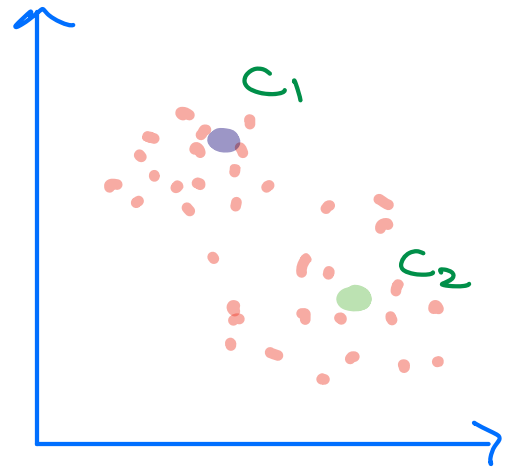
# Agenda

- Matrix Factorization in Clustering
- How do we find  $c_j$
- Non-Negative MF

$$x_i \rightarrow 1-n$$

$$c_j \rightarrow 1-1$$

$$\min_{c_j} \sum_{j=1}^K \sum_{x_i \in c_j} \|x_i - c_j\|^2$$



	1	2	3	4	...	k
1						
2	0	0	0	1	0	0
3						
4	0	1	0	0	0	0
...						
3						

Data point belongs to cluster 4

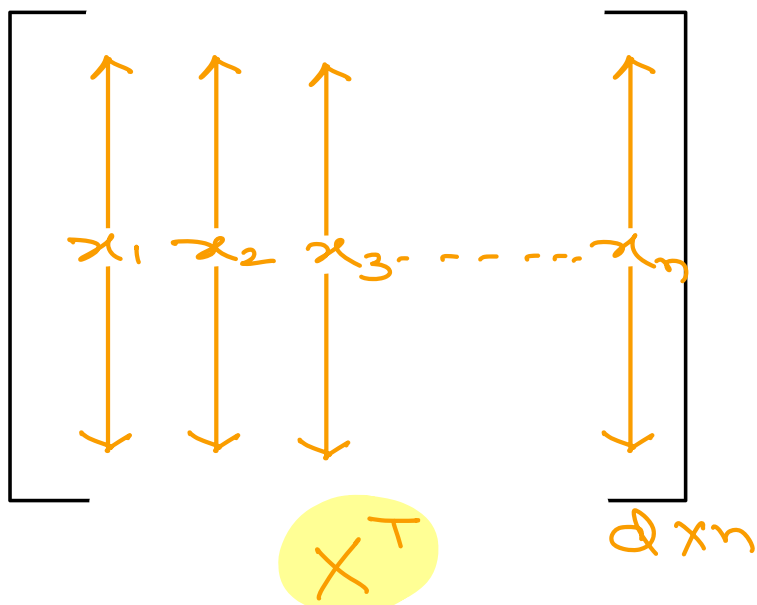
Cluster Assignment Matrix

Represents the cluster  $x_i$ 's belong to

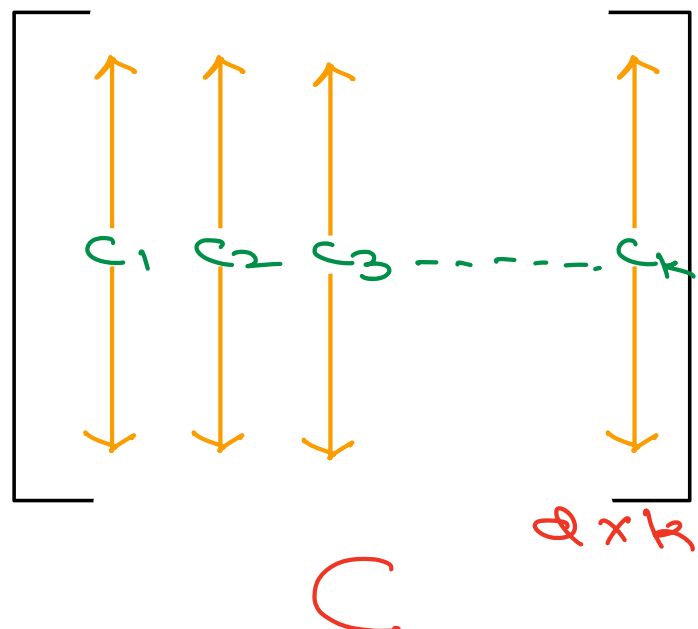
Constraint 1:  $Z_{ij} \in \{0, 1\}$

Constraint 2:  $\sum_{j=1}^K Z_{ij} = 1$

(Hard Clustering)



(data Matrix)



(cluster Co-ordinates)

$$\min_{C_j} \sum_{j=1}^K \sum_{x_i \in c_j} \|x_i - c_j\|^2$$

matrix Form

$$\min_{C_j, Z_{ij}} \sum_{i=1}^n Z_{ij} \|X_i - C_j\|^2$$

$$\min_{C, Z} \|X - ZC^T\|^2$$

$(n \times d)$      $(n \times k)$   $(k \times d)$   
 $\underbrace{\hspace{1.5cm}}_{n \times d}$

# Forbenious Norm

$$||\vec{x}|| \ni \sqrt{x_1^2 + x_2^2 + \dots + x_d^2}$$

Square of  
L2 Norm

$$||\vec{x}||^2 \ni x_1^2 + x_2^2 + \dots + x_d^2$$

$$||\vec{x}||^2 \ni \sum_{i=1}^d x_i^2$$

We can extend this to Matrix  
as well

$$||X||^2 = \sum_{\text{rows}} \sum_{\text{cols}} (x_{\text{row}, \text{col}})^2$$

## Forbenious Norm

$$||X - ZC^T||_F^2$$

$(n \times d)$     $(n \times k)$     $(k \times d)$   
 $\underbrace{\hspace{10em}}_{n \times d}$

$X$

$Z$  and  $C$

s.t.  $\forall Z_{ij} \in \{0, 1\}$

$\sum_{j=1}^k Z_{ij} = 1$

RecSys

Optimize

$$\min_{U_i, I_j} \sum_{i,j} (A_{ij} - U_i I_j)^2$$

s.t.  $A_{ij} \neq \text{Null}$

So in Clustering, we can reformulate the objective as decomposition of matrix  $X$  into  $Z$  and  $C$  where  $Z$  has two constraints

Clustering can be solved with Constrained Matrix Factorization

Soft-Clustering

$$\min_{C, Z} \left\| X - ZC^T \right\|_F^2$$

$(n \times d) \quad (n \times k)(k \times d)$   
 $\quad \quad \quad \underbrace{\hspace{2cm}}_{n \times d}$

$X$

$Z$  and  $C$

s.t. ①  $Z_{ij} \in \{0, 1\} \quad X \rightarrow Z_{ij} \in [0, 1]$

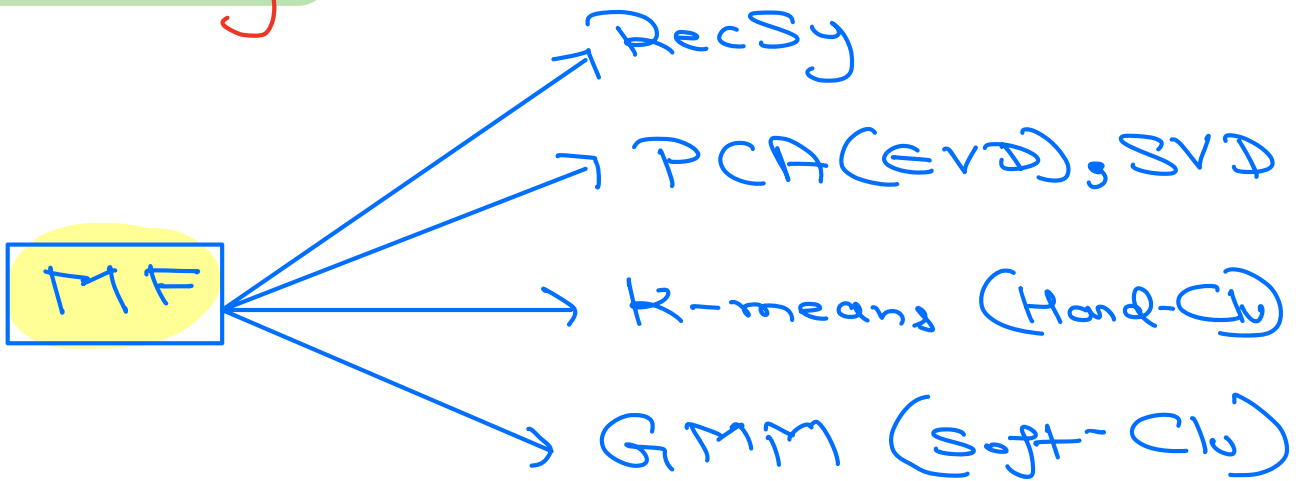
②  $\sum_{j=1}^k Z_{ij} = 1$

$N$

	1	2	3	4	-	-	-	-	k
1									
2									
3									
4	0.1	0.2	0	-	-	-	-	-	0.3
-									
-									
3									

$M = 1$

## Takeaway



How do we find "d"?

RecSys

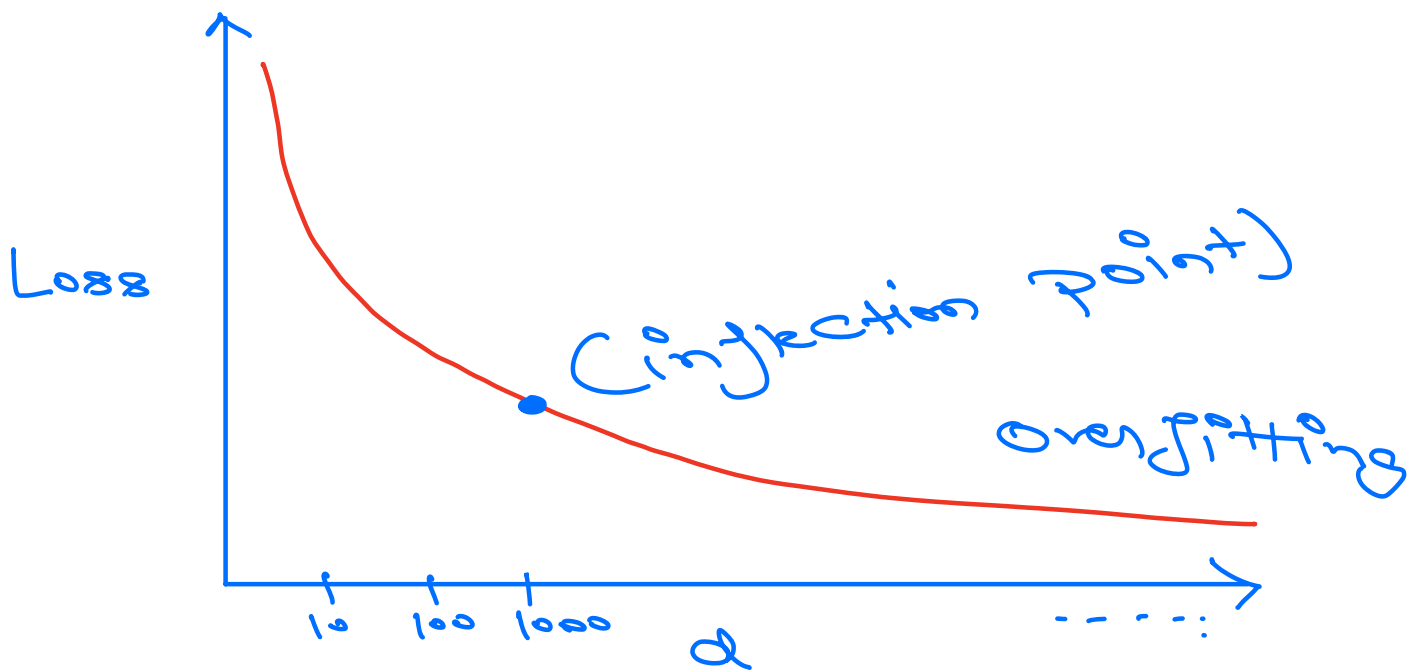
Optimize

$$\min_{U, I} \sum_{i,j} (A_{ij} - U_i I_j)^2$$

s.t.  $A_{ij} \neq \text{Null}$

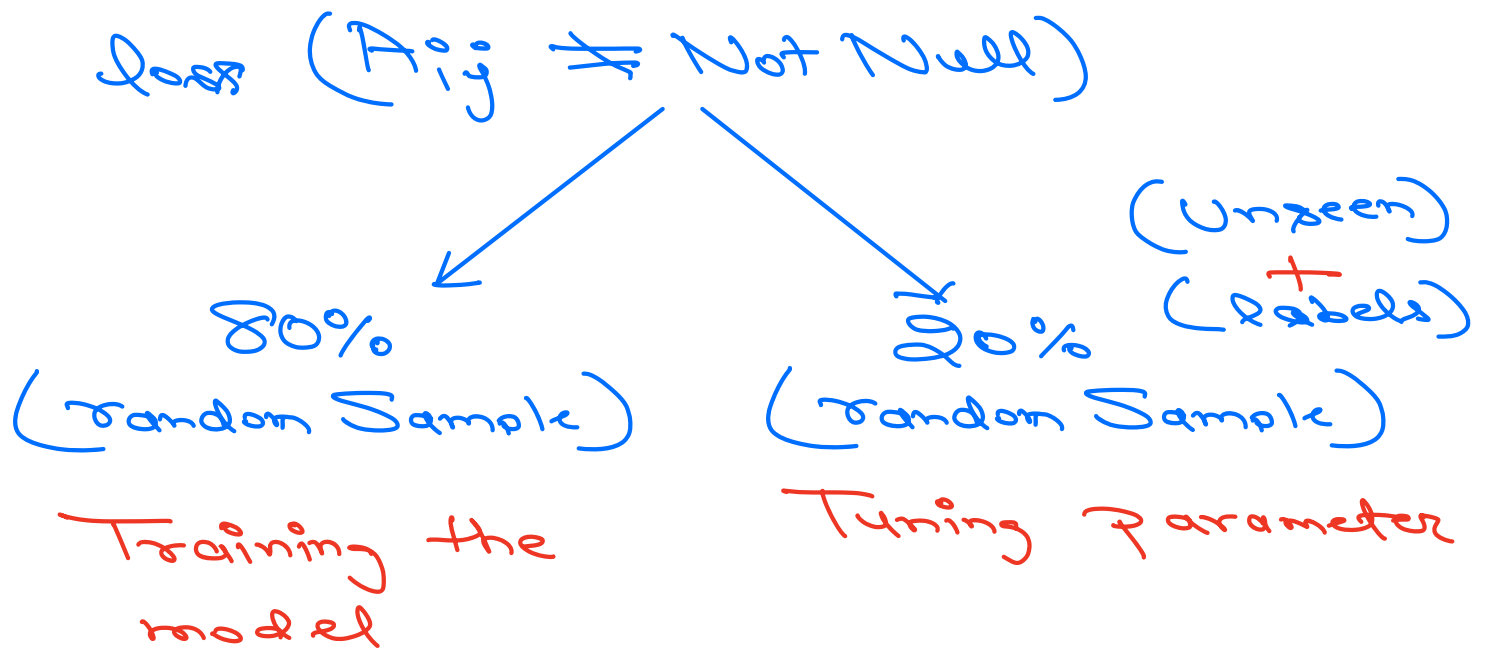
$$\min_{U, I} \left\| \underbrace{A}_{n \times m} - \underbrace{U}_{n \times d} \cdot \underbrace{I^T}_{d \times m} \right\|_F^2 \Rightarrow \text{Loss}$$

Q? How can we find best possible value of  $d$



\* The above Method can risk Overfitting

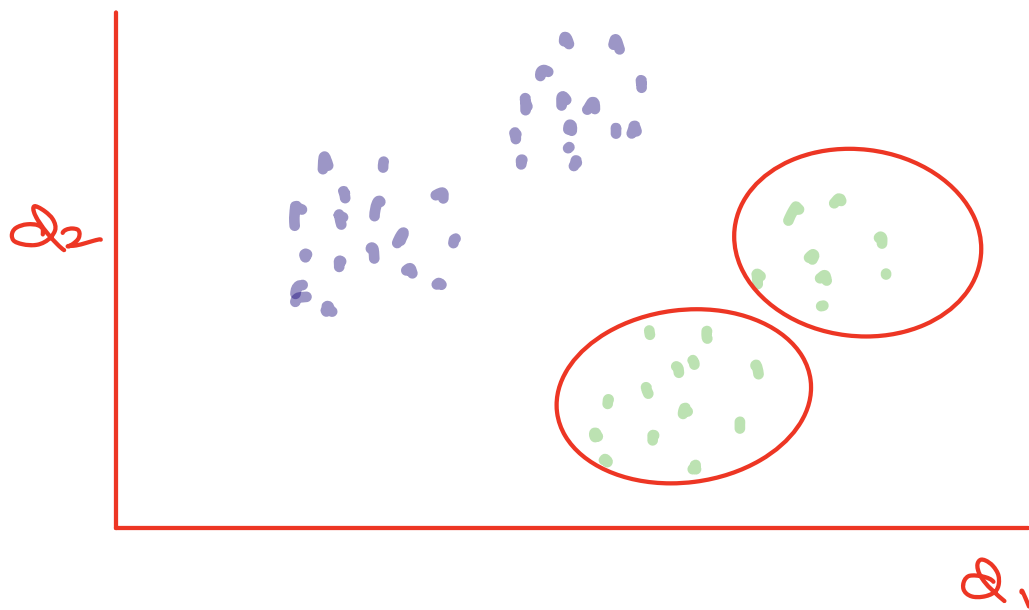
Option 2:



Can we interpret  $d$ ?

- ① No, there is no meaning or specific interpretation
- ② It's hidden dimension i.e. no dimension used for representing any item or User
- ③ Very High Value of  $d$  means you learn Every pattern and can risk Overfitting/slow Compute

$Q=Q$  : Can we interpret



Non Negative MF (NMF)

$$A_{n \times m} = B_{n \times d} \times C_{d \times m}$$

s.t.

$$B_{ij} \geq 0 \quad \forall i, j$$

$$C_{ij} \geq 0 \quad \forall i, j$$

Clustering :

$$X_{n \times d}$$

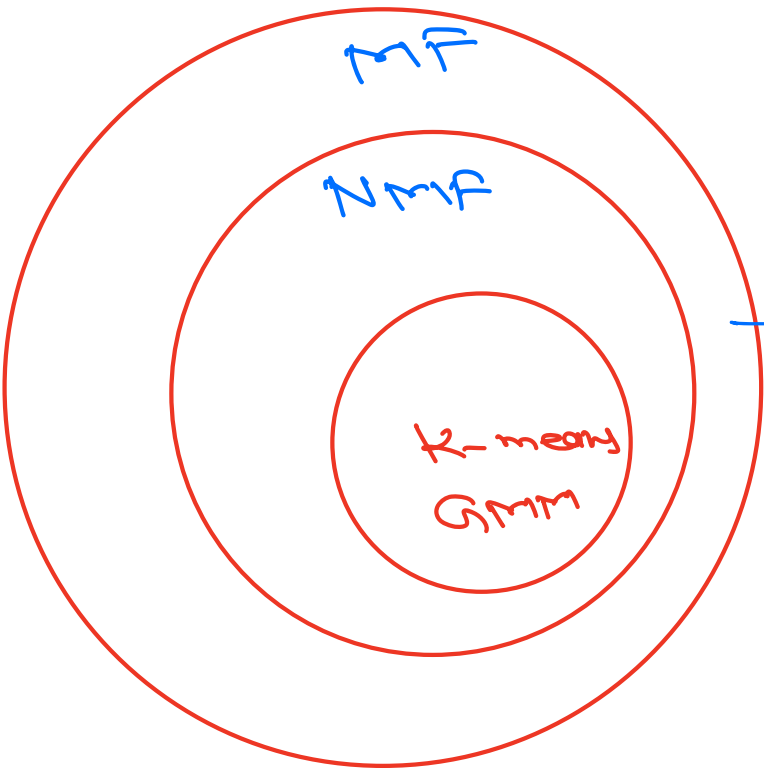
$$Z_{n \times k}^T \cdot C_{k \times d}$$

Non-Negative?

Non Negative?  
X

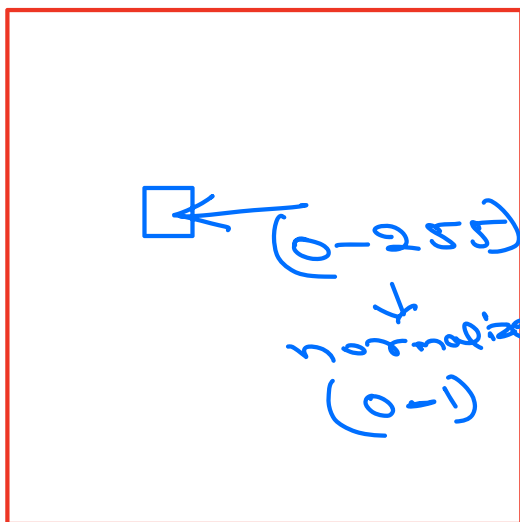


$z \in [0, 1]$  } Non-Negative



$\rightarrow MF \subseteq NMF \subseteq Clustering$

Images



NMF

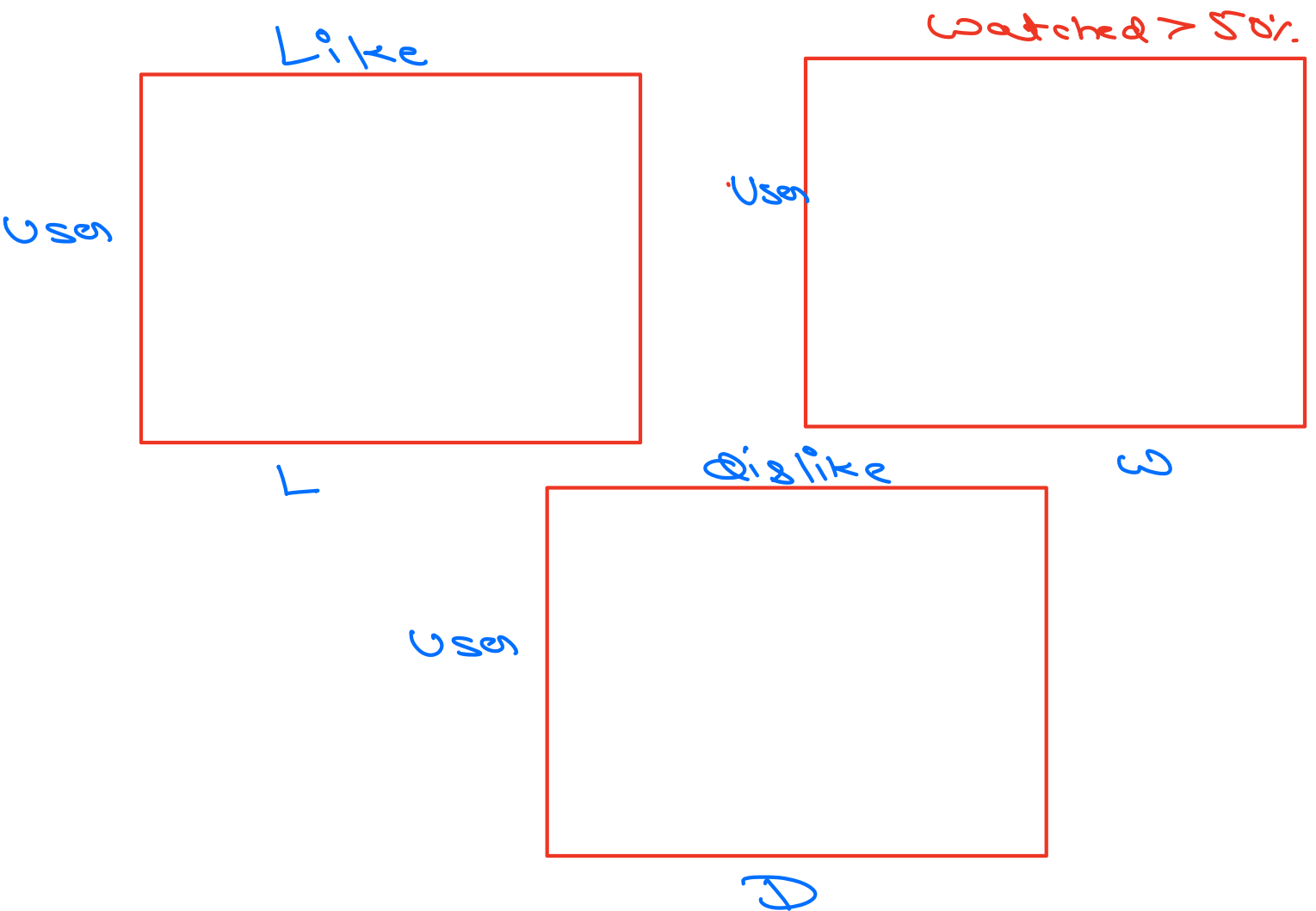
Extract features

Eigen Face

BW:  $L \times D$

Color:  $L \times D \times C$

$64 \times 64 \Rightarrow 256$  pixels



Option 1

$$A_{n \times m} = \alpha_1 L + \alpha_2 W + \alpha_3 D$$

$\swarrow$   $\downarrow$   $\swarrow$   
 $+1$   $+0.5$   $\begin{matrix} -1 \\ -2 \end{matrix} \}$  Negative

Option 2: Independent Decomposition

$$\begin{matrix} L \\ W \\ D \end{matrix} \rightarrow \begin{matrix} G \\ C \\ C \end{matrix}, \begin{matrix} H \\ H \\ H \end{matrix} \left. \vphantom{\begin{matrix} L \\ W \\ D \end{matrix}} \right\} \text{Cosine Sim}$$

9 Review : Final Winner Solution