

The problem:

We have an incomplete Matrix (R)

$$R = \begin{bmatrix} 2 & 1 & \boxed{?} & 5 & 3 \\ 1 & 4 & 3 & 2 & \boxed{?} \\ 2 & \boxed{?} & 4 & 3 & 3 \end{bmatrix}$$

Can we predict the incomplete entries?

- ⇒ We cannot develop any system that can provide an exact answer
- ⇒ But we can develop a protocol that can give very good estimates for the missing values
- ⇒ But How is that possible?
- ⇒ ~~Be~~, we presume that the entries of the matrix is an outcome of a physical process & not ~~beacuse~~ random numbers.
- ⇒ Therefore the idea is to find ~~the~~ the latent space of the physical process by the available entries. Then use ~~these~~ these latent variables & to estimate the missing entries.

SVD & Matrix Completion : Pg 2

This is a billin' \$ business!

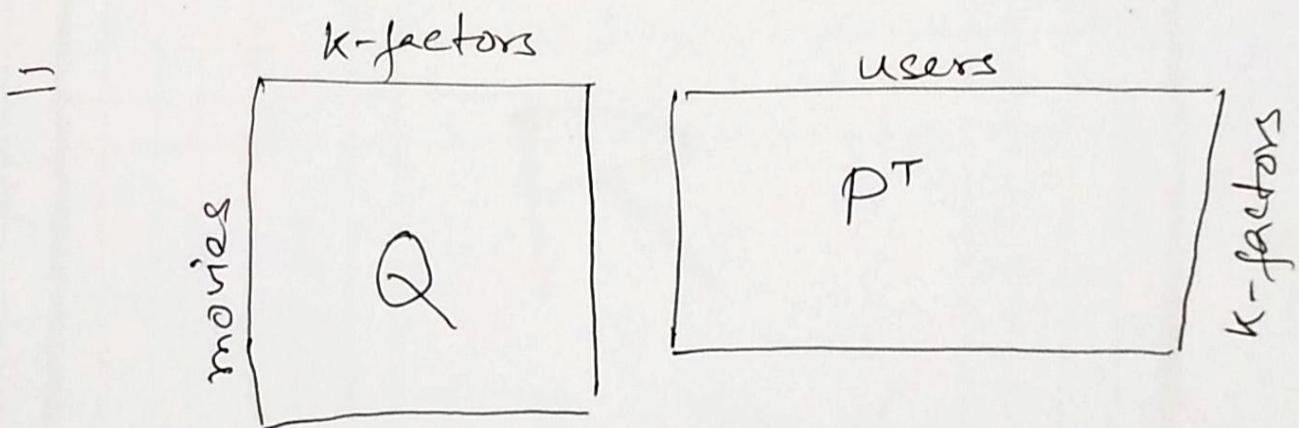
Netflix movie rating predictions:

R

480,000 users

17,700 movies

1	3	1	1	1
3	2	1	2	5
2	1	2	2	4
3	1	1	2	5

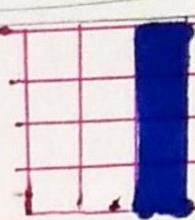


Thin k-factors

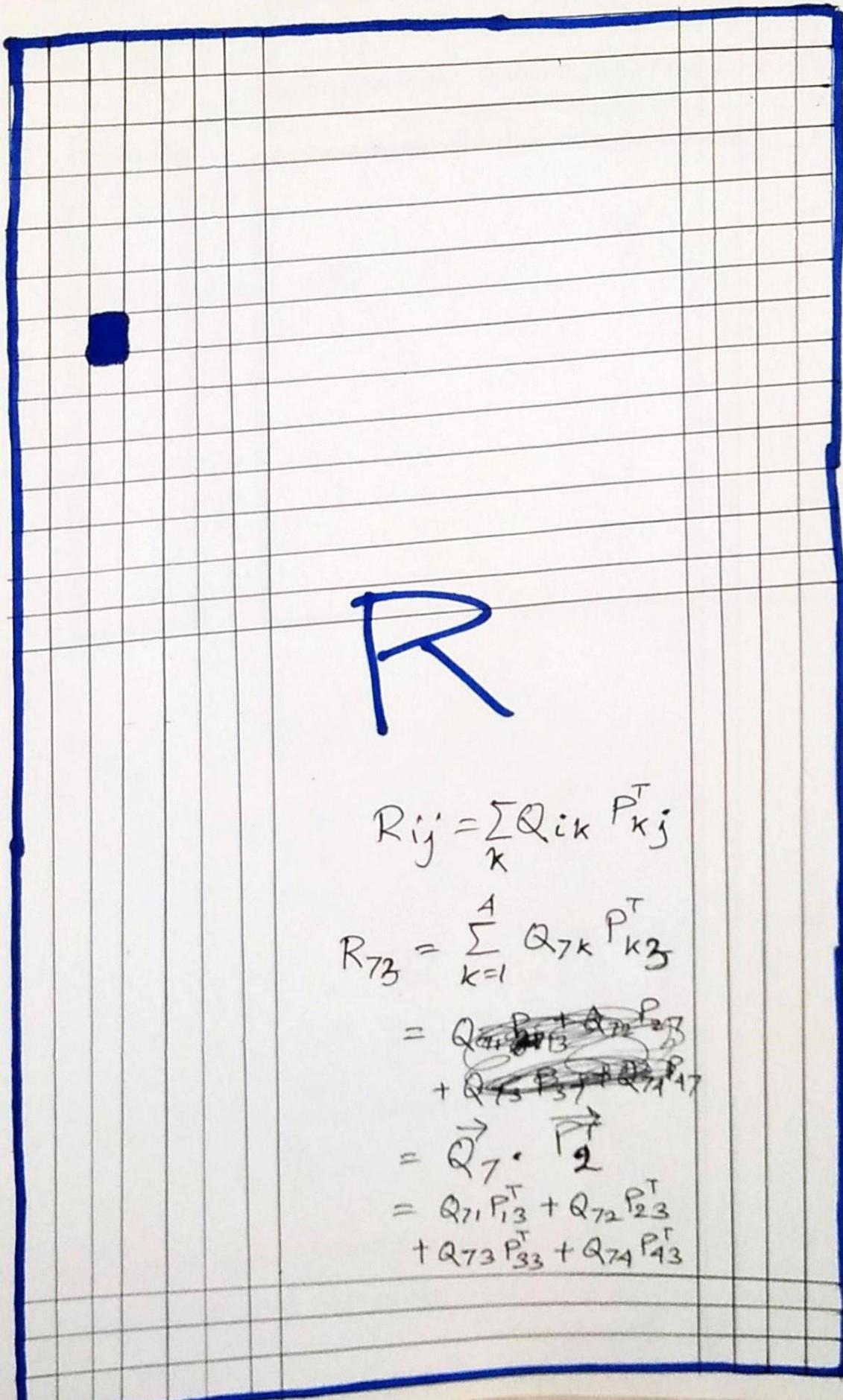
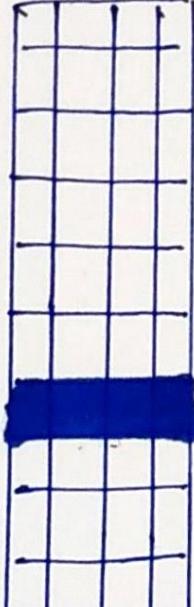
$$SVD: A = U \Sigma V^* \text{ or } U \Sigma V^T$$

Problem! we - have missing values in R

SVD Matrix completion: P_{73}^T , P & Q are thin matrices



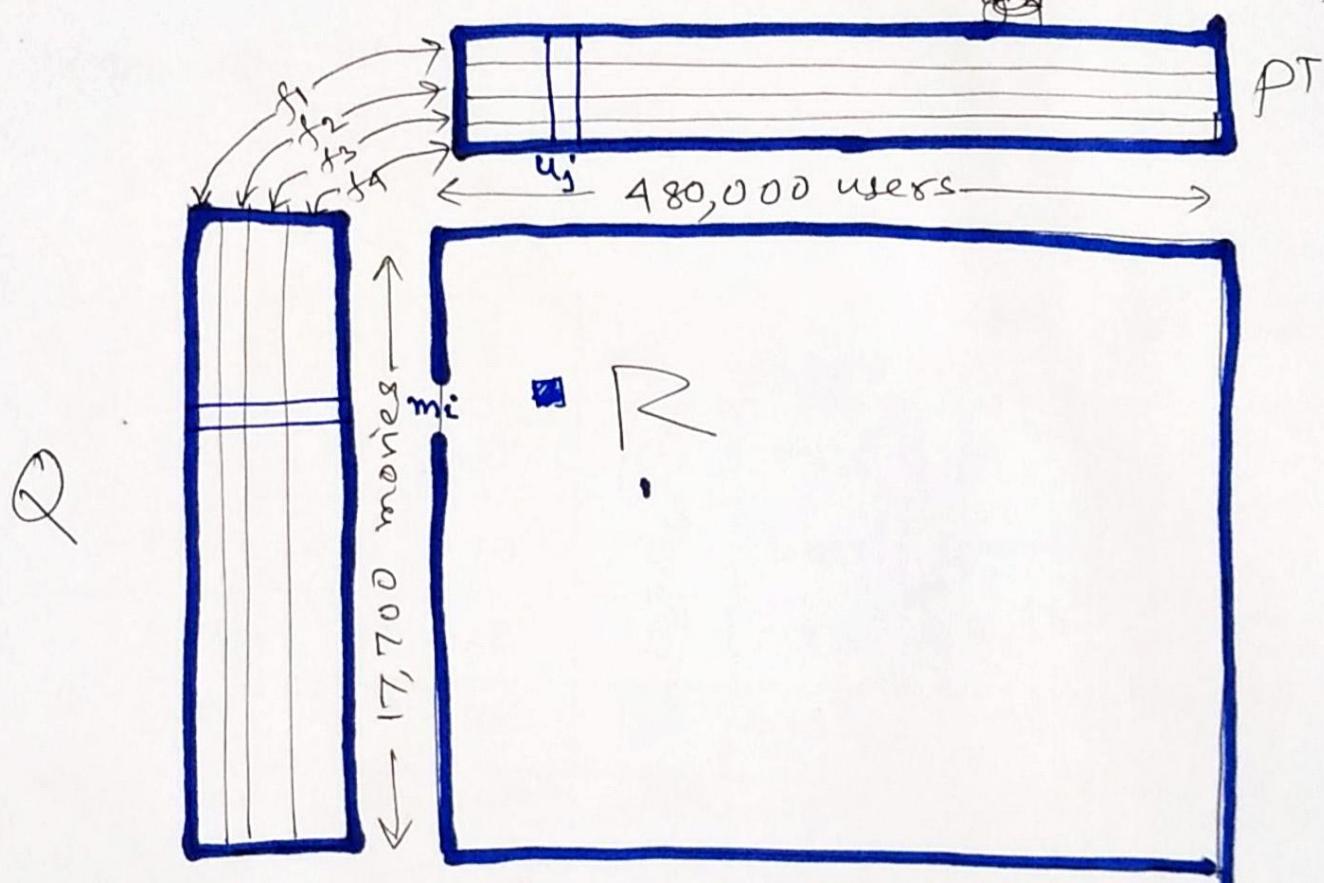
P^T



Q

SVD & Matrix Completion 2 Pg 4

Factors k: sci-fi, romance, comedy,
 f₁ crime - ~~thriller~~ etc.
 f₂ f₃
 f₄



$$R_{ij} = \sum_i \vec{q}_i \cdot \vec{p}_j^T$$

$$= \sum_k q_{ik} p_{kj}^T$$

$$= \sum_k q_{ik} p_{jk} \leftarrow \begin{array}{l} \text{Note: transpose} \\ \text{removed} \\ \text{'coz index} \\ \text{exchanged} \end{array}$$

But what to do with missing values
 in R ?

⇒ We can not perform the SVD in the traditional sense: Numerical methods will fall apart !!!

SVD & Matrix Completion: Pg 5

Let us perform a back-of-the-envelope calculation to understand whether it is a solvable problem or not!

$$\begin{array}{lll} \text{Size of } R \rightarrow m \times n \\ " " Q \rightarrow m \times k \\ " " P^T \rightarrow k \times n \end{array}$$

m	n	k	R	P	Q	Total Storage (P+Q)	Compression ratio R/(P+Q)	Comment
10^3	10^3	100	10^6	10^5	10^5	2×10^5	5	
10^6	10^6	100	10^{12}	10^8	10^8	2×10^8	5000	
10^9	10^9	100	10^{18}	10^{11}	10^{11}	2×10^{11}	5×10^6	
10^9	10^5	100	10^{14}	10^7	10^{11}	$\sim 10^{11}$	10^3	
10^{23}	10^{24}	100	10^{27}	10^{26}	10^{25}	$\sim 10^{25}$	10^2	
10^{23}	10^{23}	100	10^{46}	10^{25}	10^{25}	$\sim 10^{25}$	10^{21}	

→ Number of unknowns $\approx \sim 10^{11}$

Even if 1% Entries in R is available

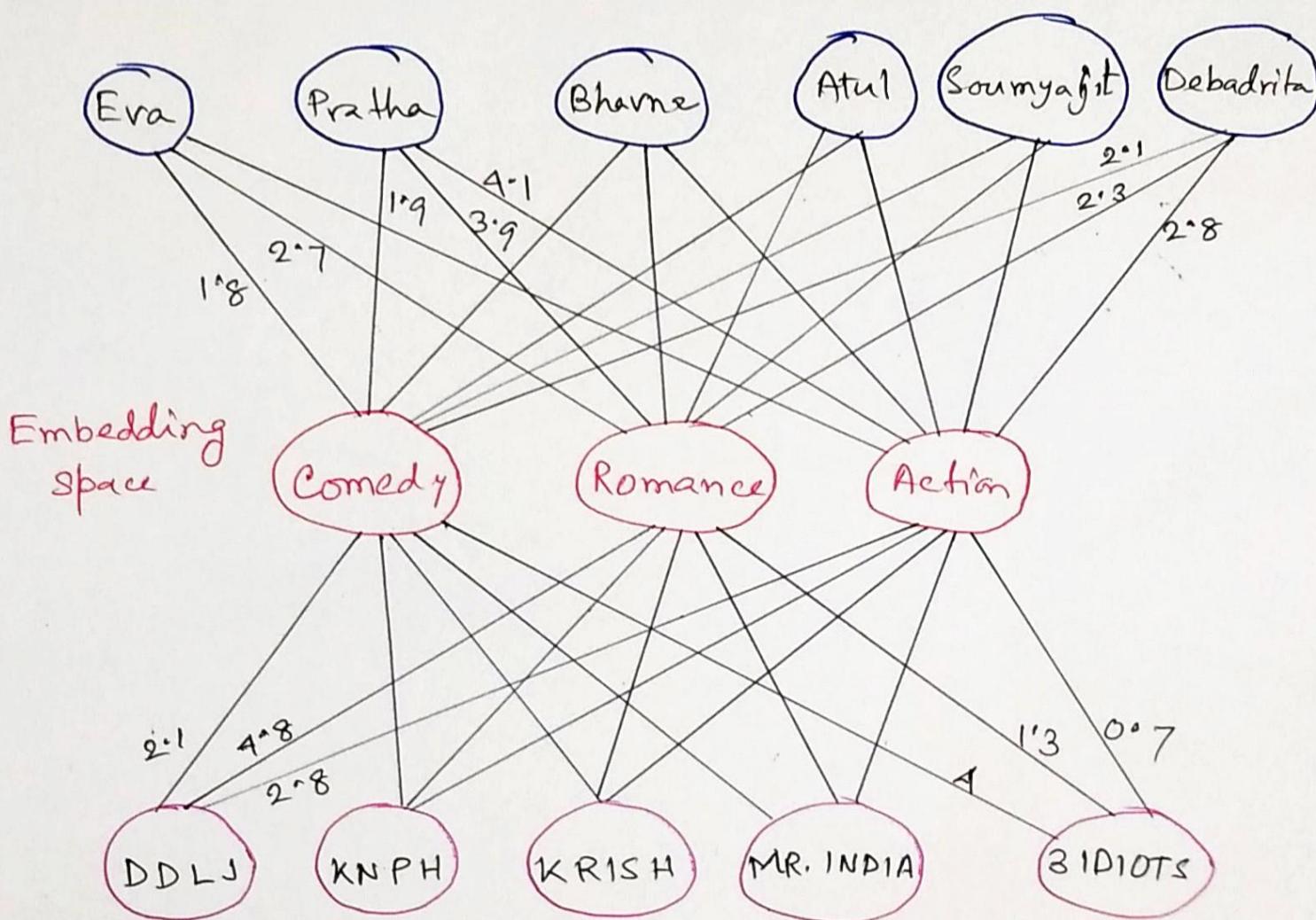
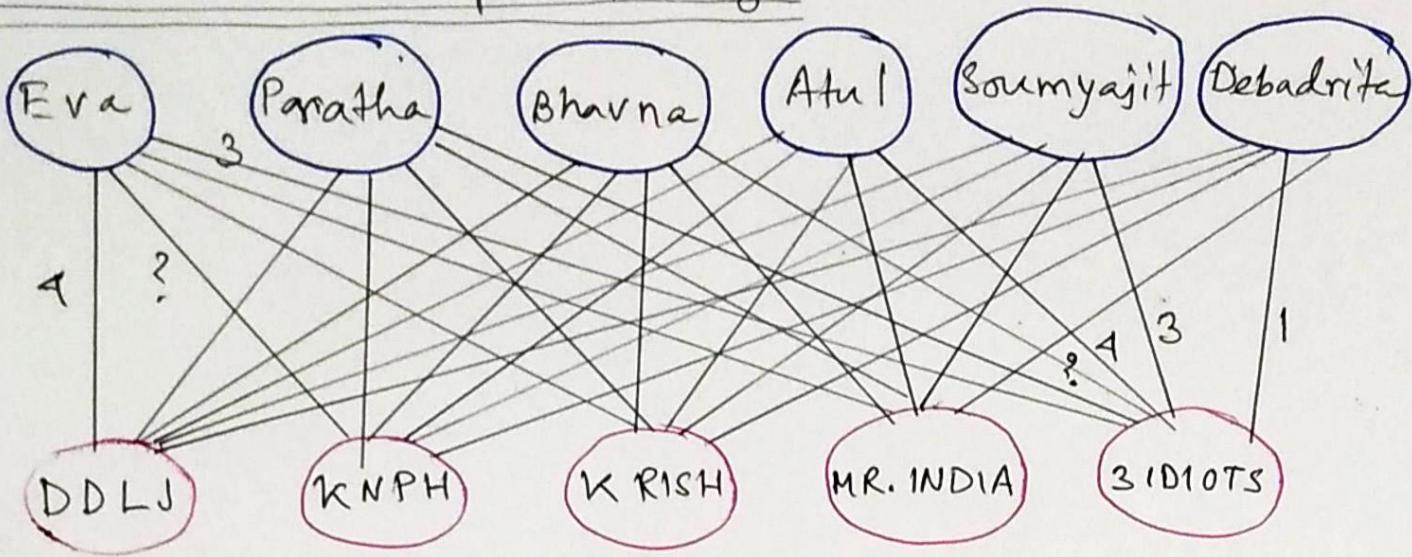
we have $10^{14} \times 10^{-2} = 10^{12}$ equations available

→ So ~~these~~ we still have nearly 10 times more equations available than needed

⇒ These systems of equations must be solvable.

⇒ Solution method: → one possible candidate: Gradient descent

SVD & Matrix completion : Pg 6



Discuss the connection with Auto-encoders.