

# Implementing Recommendation Systems

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# Module Overview



Understand recommendation systems

Discuss content-based vs. collaborative filtering methods

Introduce matrix factorization algorithms

Implementation of a small-scale collaborative filtering system

Implementation of a larger-scale collaborative filtering system



# Understanding Recommendation Systems

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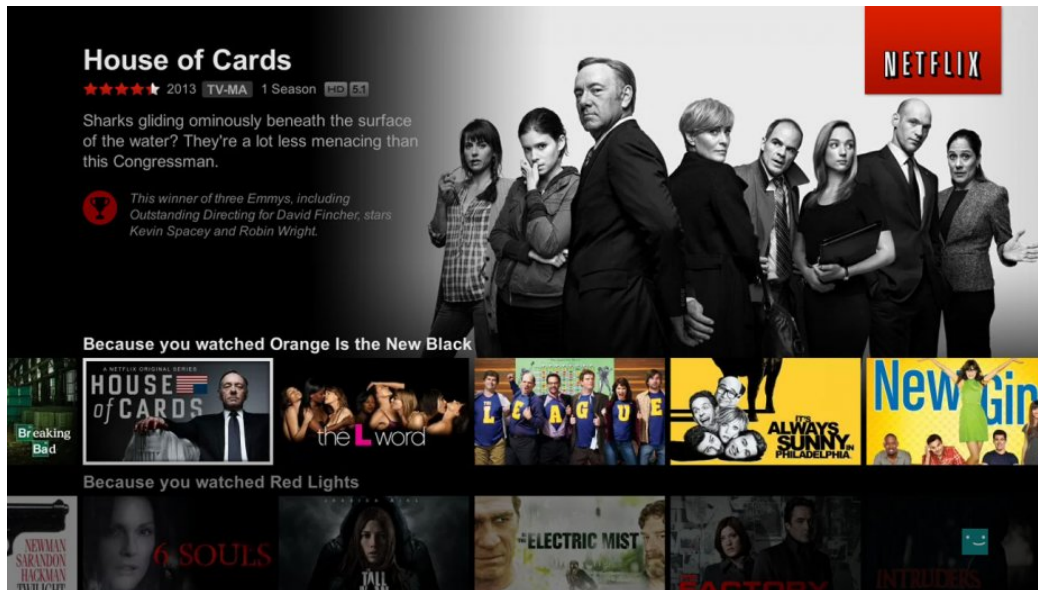


# Recommendation System

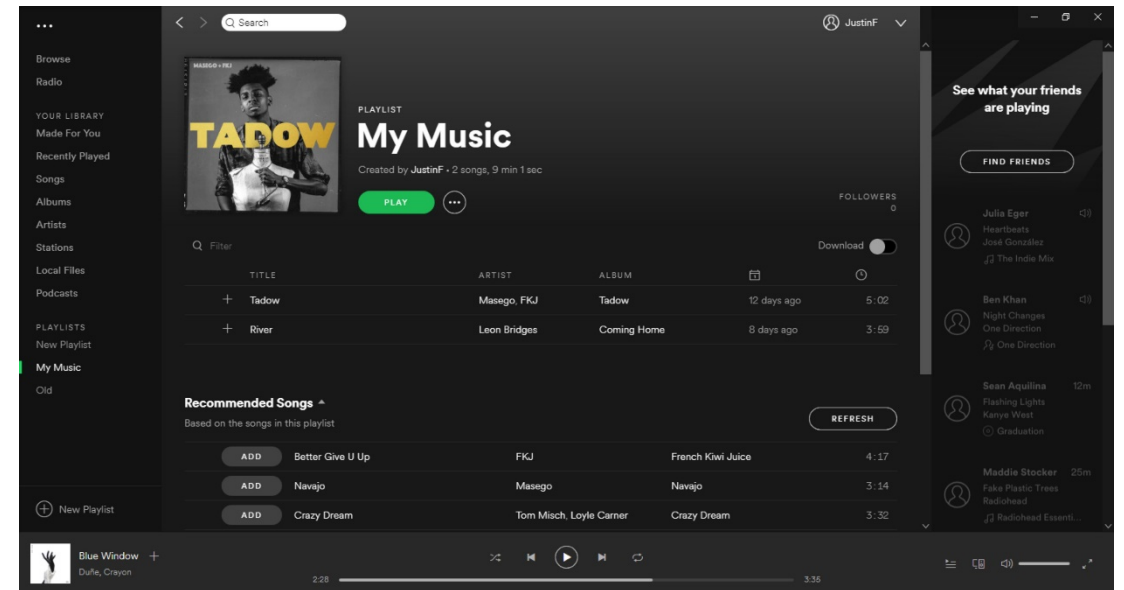
Information filtering system used to predict the rating or preference a user would give to an item. Giving the ability to recommend items.



# Recommendation Systems



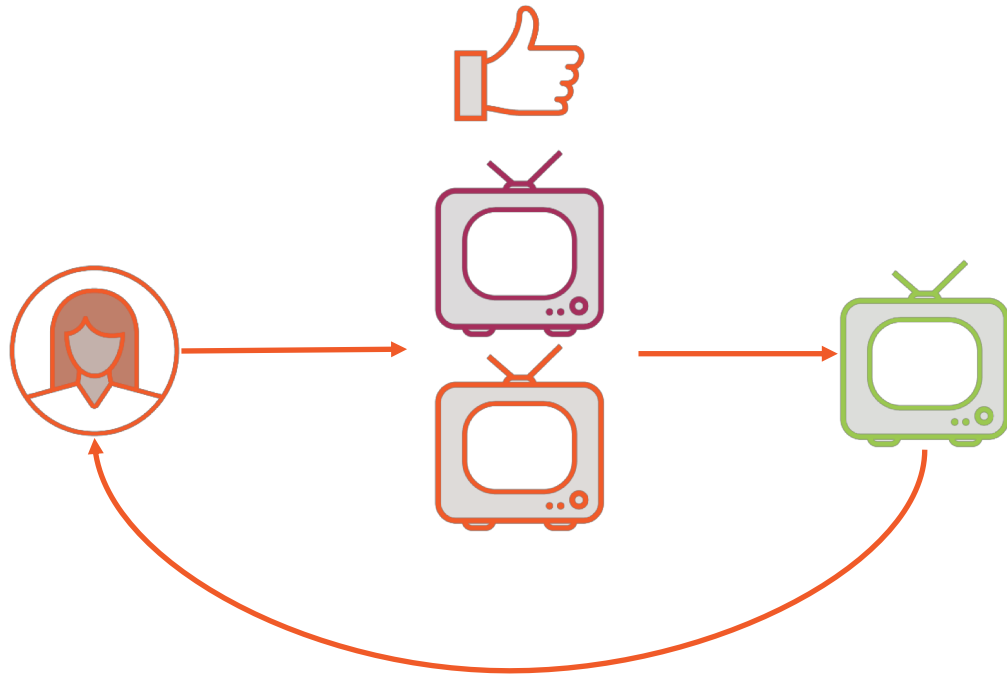
Netflix: Movie Recommendations



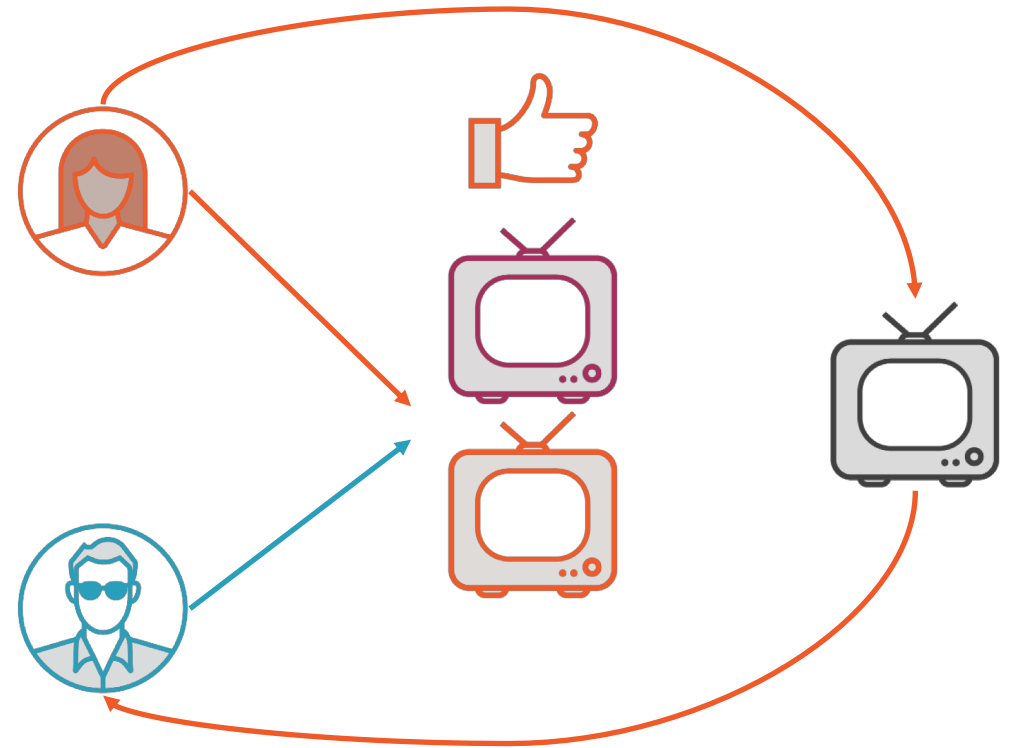
Spotify: Music Recommendations



# Content-based vs. Collaborative



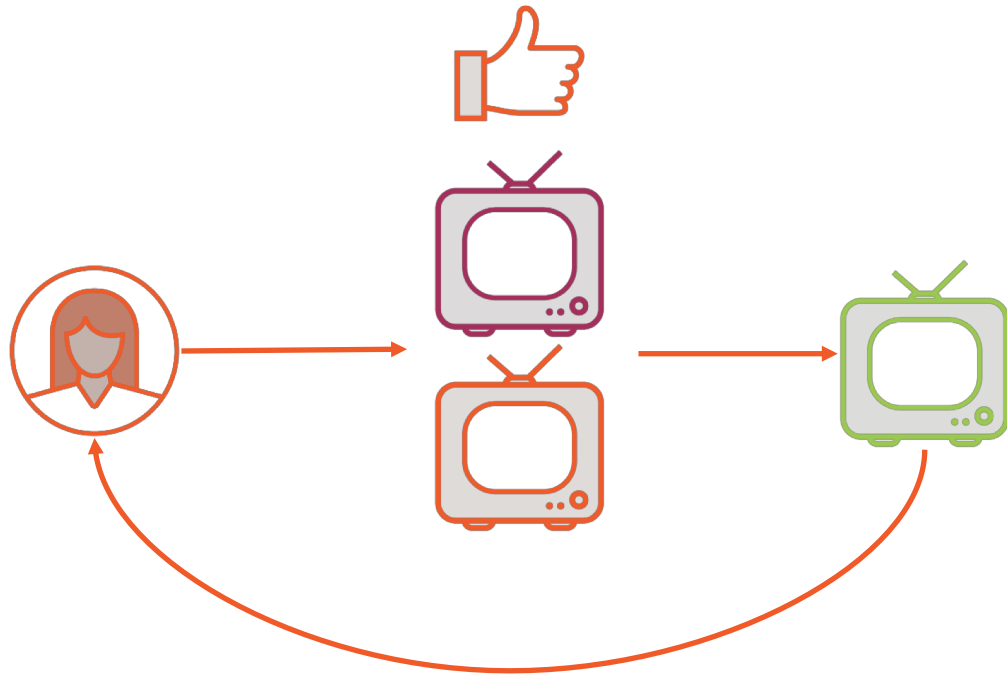
Content-based filtering



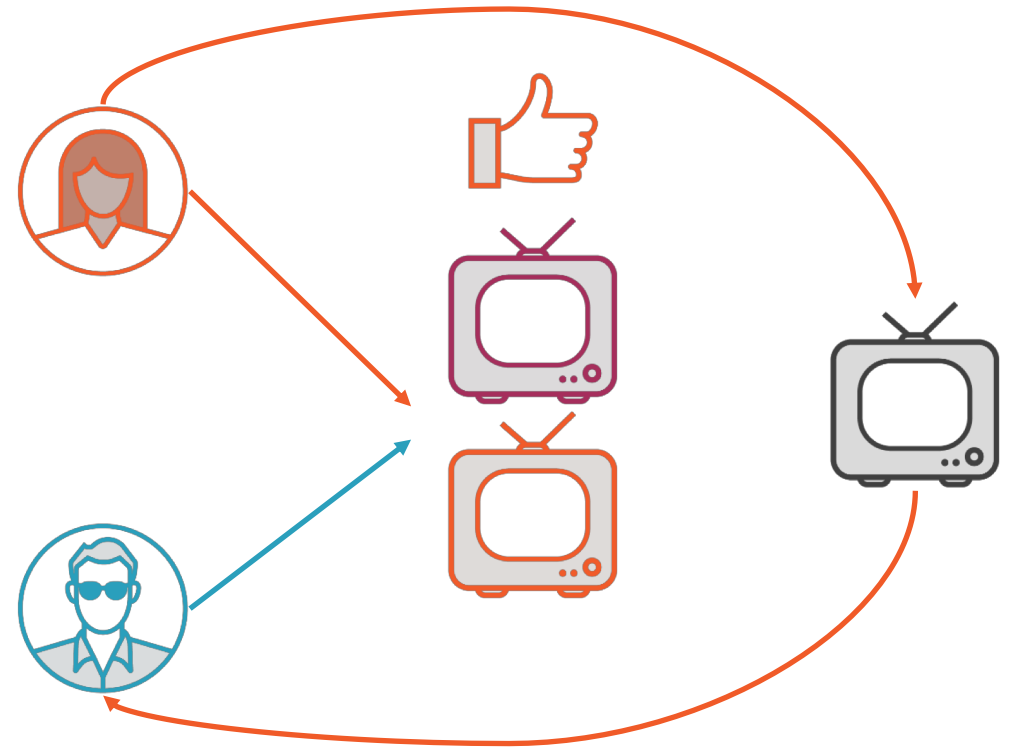
Collaborative filtering



# Content-based vs. Collaborative



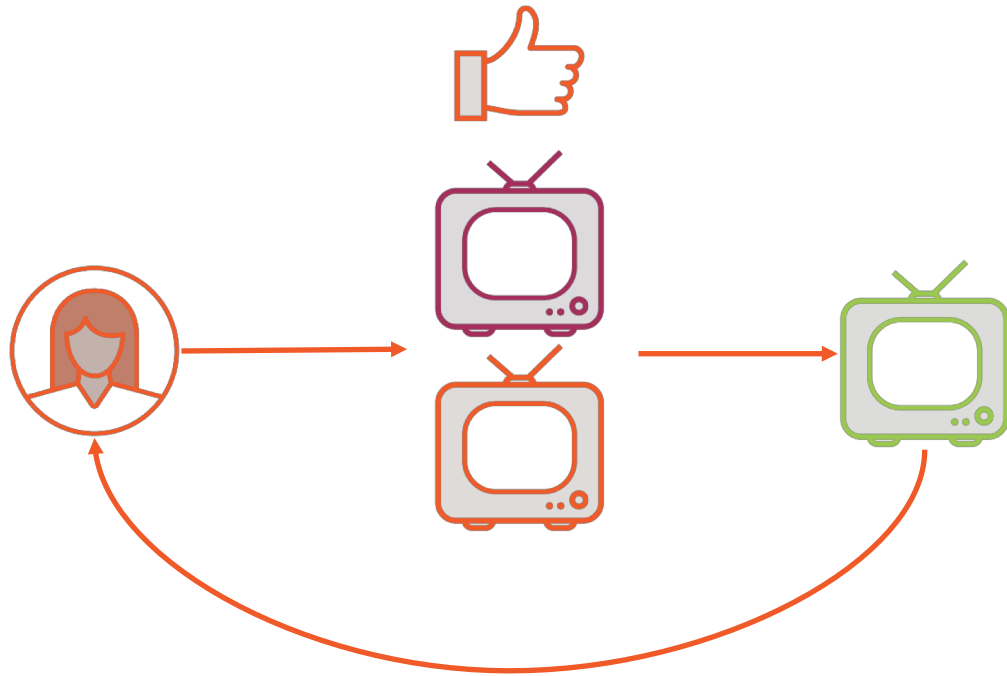
Content-based filtering



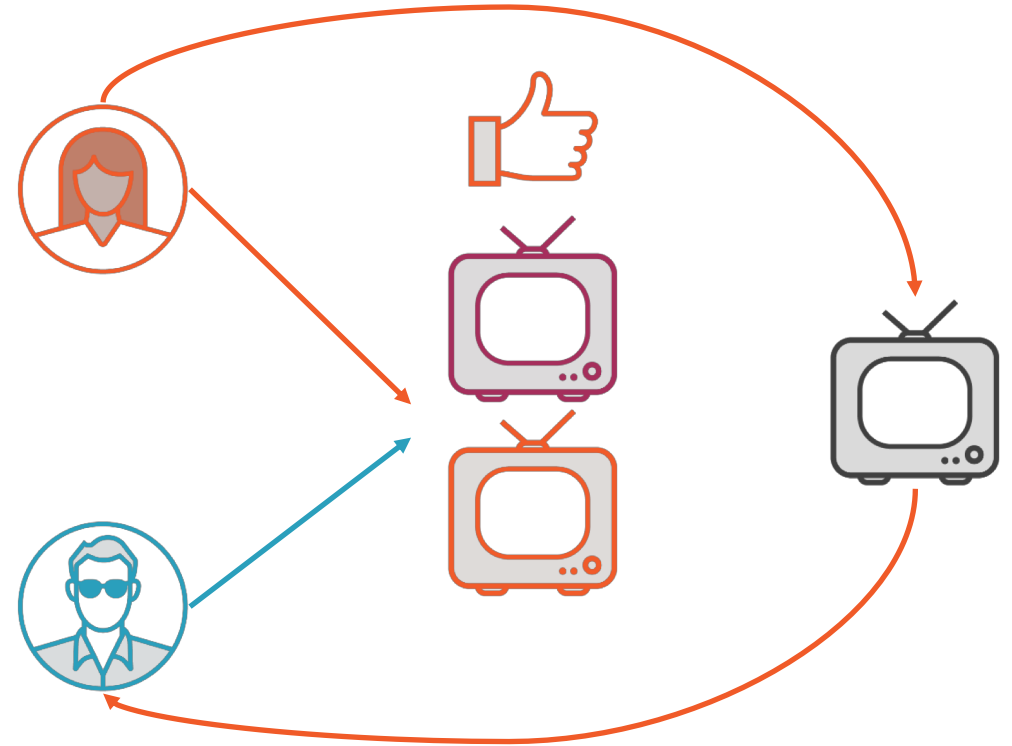
Collaborative filtering



# Content-based vs. Collaborative



Content-based filtering

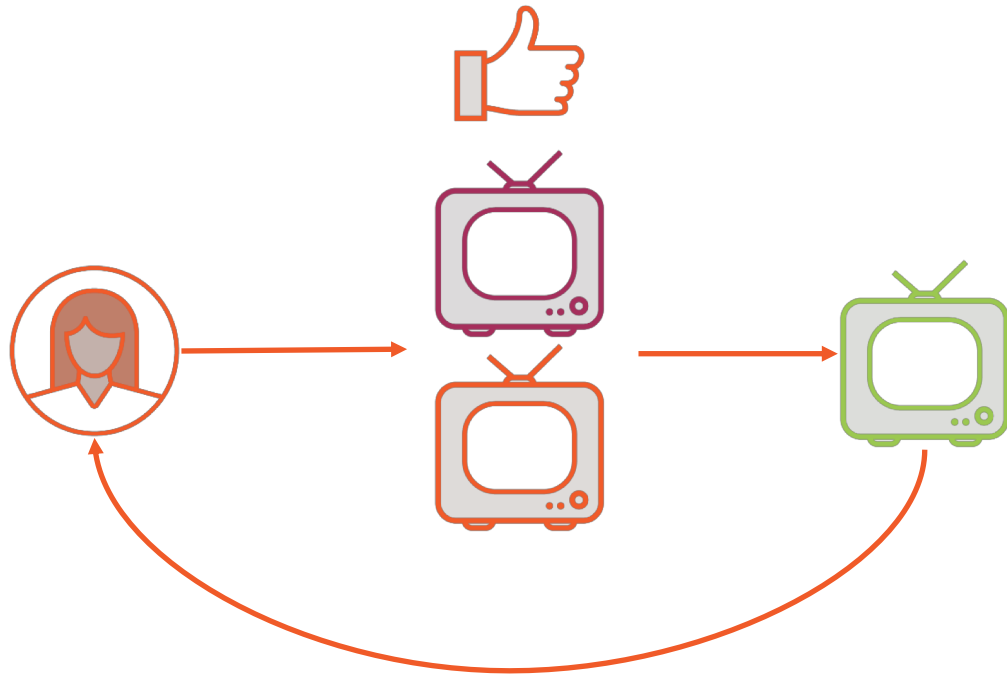


Collaborative filtering

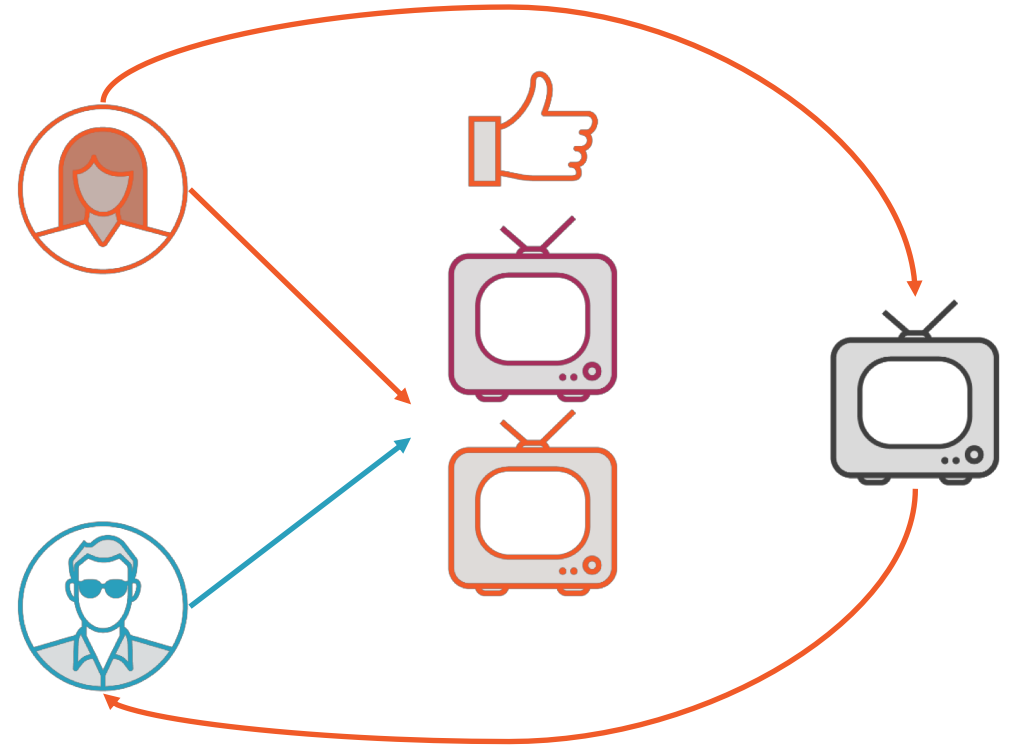




# Content-based vs. Collaborative



Content-based filtering



Collaborative filtering



# Content-based vs. Collaborative

## Content-based filtering

Based on known users preferences

Looks at item content

Recommendations inferred from similarities to known items

## Collaborative filtering

Based on similarities to other users

Does not look at item content

Recommendations inferred from the preferences of similar users



# Understanding Matrix Factorization

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# Matrix Factorization

Filtering algorithm which works by decomposing a user-item interaction matrix into the product of two lower dimension rectangular matrices.



# Matrix Factorization

	Die Hard	Die Hard 2	Toy Story	Toy Story 2	Toy Story 3
Justin	5	-	2	1	-
Mike	-	4	1	-	2
Claire	5	-	-	2	1
Steven	1	-	5	-	4
Susan	-	1	-	4	5



# Matrix Factorization

$R$



$[n \times m]$



# Matrix Factorization

$$\begin{matrix} R & & P & & Q \\ \left[ \right. & & \left[ \right. & \approx & \left[ \right. & & \left[ \right. \\ & & & & & & \\ & & & & & & \end{matrix}$$

$[n \times m]$        $[n \times k]$        $[k \times m]$



# Matrix Factorization

Prediction	$\hat{r} = pq$





# Matrix Factorization

Prediction	$\hat{r} = pq$
Cost	$\ r - \hat{r}\ $



# Matrix Factorization

<b>Prediction</b>	$\hat{r} = pq$
<b>Cost</b>	$\ r - \hat{r}\ $
<b>Regularization</b>	$\sum \lambda(\ p\  + \ q\ )$



# Matrix Factorization

<b>Prediction</b>	$\hat{r} = pq$
<b>Cost</b>	$\ r - \hat{r}\ $
<b>Regularization</b>	$\sum \lambda(\ p\  + \ q\ )$
<b>Total Cost</b>	$\sum \ r - \hat{r}\  + \lambda(\ p\  + \ q\ )$



# Implementing a Small-scale Collaborative Filtering System

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# Implementing a Larger-scale Collaborative Filtering System

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# Summary



Understood recommendation systems

Discussed content-based vs.  
collaborative filtering methods

Introduced matrix factorization  
algorithms

Implemented a small-scale collaborative  
filtering system

Implemented a larger-scale collaborative  
filtering system

