

Data Science

Module-10

Recommendation Engine



Agenda













A filtering system that seeks to predict and show the items of user interest.

It may or may not be accurate

Utilized in a variety of areas

Mostly used in the digital domain

Can significantly boost revenues, CTRs, conversions, and other important metrics







Data filtering tools that make use of algorithms and data to recommend the most relevant items to a particular use



An automated form of "SHOP COUNTER GUY"



Recommendation Engine Example





Reference: Amazon



Recommendation Engine Example



Frequently bought together



Some of these items ship sooner than the others. Show details

▼ This item: Samsung Galaxy J7 Neo (16GB) J701M/DS - 5.5", Android 7.0, Dual SIM Unlocked Smartphone... \$140.96

☑ Galaxy J7 Neo J701M/J7 Nxt J701F/J7 Core J701 Case, With Screen Protector & Stylus, Telegaming Dual... \$7.99

[3-PACK]-Mr Shield For Samsung "Galaxy J7 Neo" [Tempered Glass] Screen Protector [0.3mm Ultra Thin... \$6.95

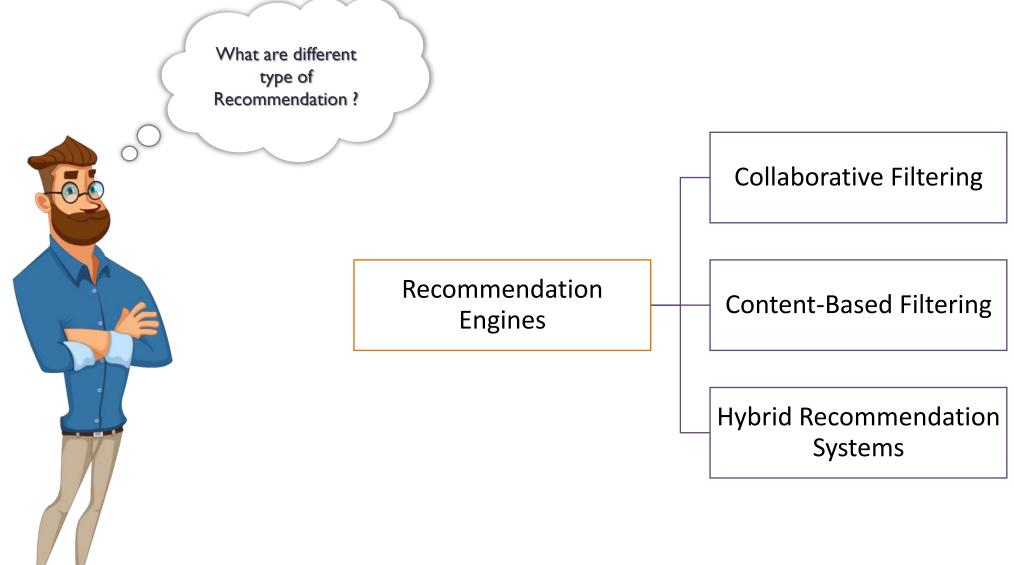
Reference: Amazon



Types of Recommendations

Types of Recommendation Engine







Collaborative filtering recommender systems

Collaborative filtering recommender systems



What is Collaborative Filtering Recommender Systems?

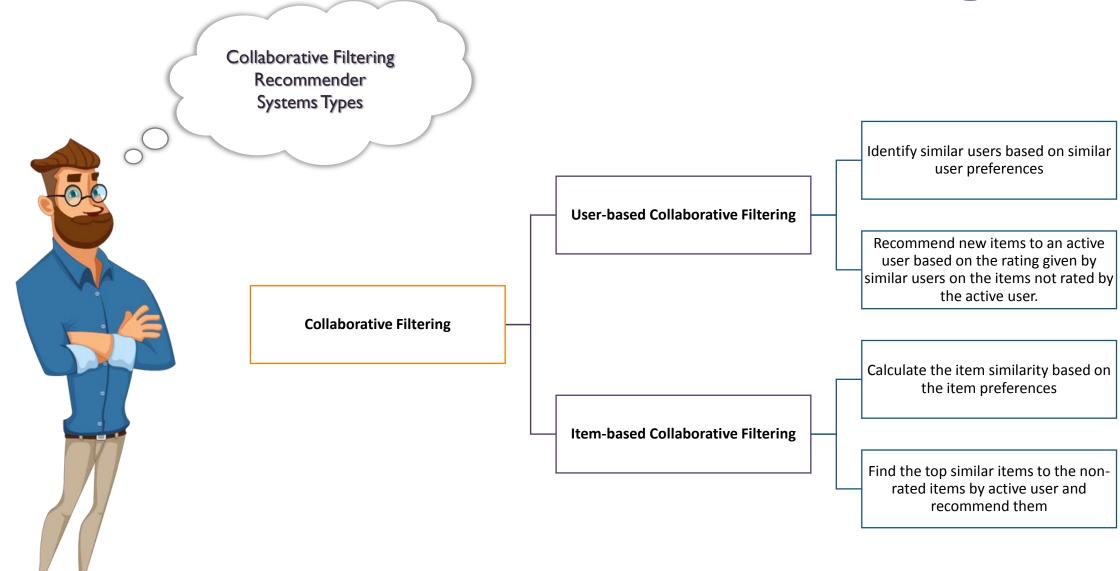
Filtering items from a large set of alternatives is done collaboratively by users' preferences





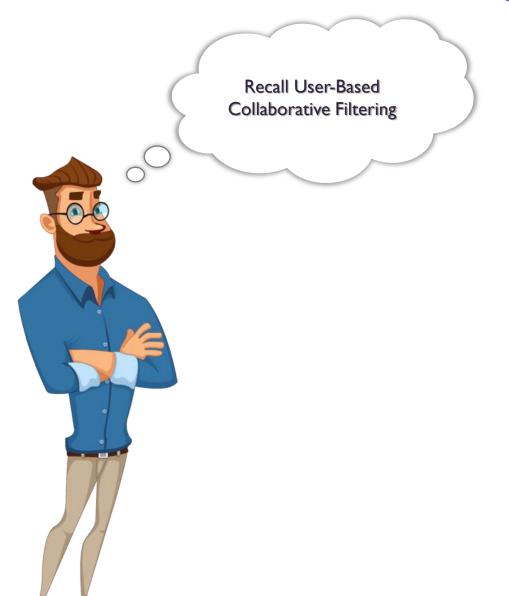
Collaborative filtering recommender systems

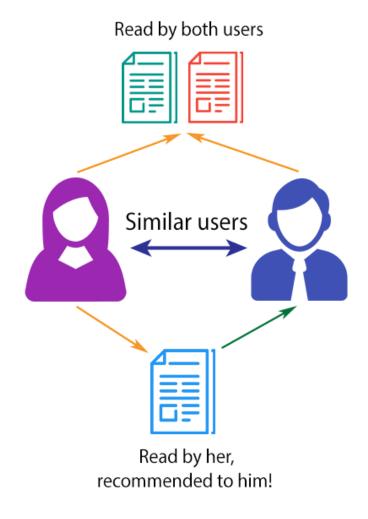




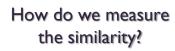














Pearson Correlation

$$u_{ik} = \frac{\sum_{j} (v_{ij} - v_i)(v_{kj} - v_k)}{\sqrt{\sum_{j} (v_{ij} - v_i)^2 \sum_{j} (v_{kj} - v_k)^2}} \qquad \cos(u_i, u_j) = \frac{\sum_{k=1}^{m} v_{ik} v_{jk}}{\sqrt{\sum_{k=1}^{m} v_{ik}^2 \sum_{k=1}^{m} v_{jk}^2}}$$

Cosine Similarity

$$\cos(u_{i}, u_{j}) = \frac{\sum_{k=1}^{m} v_{ik} v_{jk}}{\sqrt{\sum_{k=1}^{m} v_{ik}^{2} \sum_{k=1}^{m} v_{jk}^{2}}}$$

$$v_{ij}^* = K \sum_{v_{kj} \neq ?} u_{jk} v_{kj}$$



	The Avengers	Sherlock	Transformers	Matrix	Titanic	Me Before You	Similarity(i, E)
Α	2		2	4	5		NA
В	5		4			1	0.87
С			5		2		1
D		1		5		4	-1
Е	3.51*	3.81*	4	2.42*	2.48*	2	1
F	4	5		1			NA

Users' preference can change over time





	The Avengers	Sherlock	Transformers	Matrix	Titanic	Me Before You
A	2		2	4	5	2.94*
В	5		4			1
С			5		2	2.48*
D		1		5		4
Е			4			2
F	4	5		1		1.12*
Similarity	-1	-1	0.86	1	1	

Scalability

The worst-case complexity is O (mn)

Sparsity

Collaborative Filtering



Building collaborative filtering recommender systems



- How to calculate the similarity between users?
- How to calculate the similarity between items?
- How recommendations are generated?
- How to deal with new items and new users whose data is not known?

Cold Start problem



Thank You