Hybrid and Advanced Techniques Quiz 测验, 6 个问题

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1。 What i	s the main point of hybrid algorithms?
	To handle cases where a recommender is trying to balance objectives, e.g., to recommend good products for individual users, but also to make sure that each product gets recommended to enough different users to get sold.
•	To take advantage of situations where no single algorithm provides the best recommendations by combining different algorithms together to achieve a better result.
	To speed up the computation of recommendations which too often are slow using non-hybrid algorithms. Hybrid algorithms also are easier to optimize for parallel execution.
each s	ching hybrid selects only one recommender algorithm for ituation. Which of the following is the best example of a on where a switching hybrid algorithm would be most
\bigcirc	When we have several algorithms each of which does a great job with the top 4-5 recommendations, but is worse deeper down in the list.

Hybrid and Advanced Techniques Oul Z recommending to users with few ratings vs. to users with many ratings.

When we have several different algorithms, but don't really like the results from any of them.

point

3.

How does SVD++ create a form of hybrid recommender within a matrix factorization framework?

- SVD++ incorporates a latent feature space representation for observable user- and item-features (e.g., user age, film genre) as a way to hybridize traditional ratings-based matrix factorization collaborative filtering with recommendation based on user properties (demographics) or item properties (content-based).
- SVD++ performs two separate factorizations of the ratings matrix, one biased towards user coherence (which emulates user-user collaborative filtering) and one biased towards item coherence (which emulates item-item collaborative filtering). It then combines the highest singular values from each factorization into a single vector representation that represents a hybrid of the two techniques.
- SVD++ generates recommendations lists from two different dimensionality reduction recommenders (traditional SVD-based matrix factorization and the shortcut gradient descent method) and then merges the results from those lists creating a more robust mixture hybrid recommender.

point

4。 **Hybrid and Advanced Tembriques to Wiz**trix completion 测验, 6 个问题 techniques fail for top-n recommendation. What is that reason?

These techniques depend on the assumption that the
data we observe is randomly distributed from all data
(e.g., you've rated movies at random from among all
movies), and this assumption is almost never true in real
recommendation environments

Accurate matrix completion techniques generally require
between 10 and 20% observed values, but most
recommender applications have much less available
data. The result is that these techniques are unbiased,
but have a very large noise component that hurts our
recommendations

We have accurate methods for matrix completion, but
they are computationally intractible (their time is
exponential in the rank of the matrix), so we can't use
them). Instead we use approximation techniques that we
know are biased towards filling in values too close to the
mean.

1 point

5。

In an interview with Arindam Banerjee, we learned about multidimensional matrix factorization as an approach to hybrid recommendation. All of the following are benefits of such an approach except one. Which of the following **is not** a benefit of hybrid matrix factorization?

Multidimensional matrix factorization is significantly
faster than traditional techniques

\bigcirc	Multidimensional matrix factorization can lead to
	meaningful interpretations of the latent vector space
	used to describe users and items

Multidimensional matrix factorization can use more **Hybrid and Advanced Techniquese Chiz** better predictions or 测验, 6 个问题 recommendations.

	recommendations.
	Multidimensional matrix factorization can reveal relationships among factors such as customer attributes and product attributes (e.g., older customers prefer larger cars).
1 point	
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	liscussed the Netflix Competition. Which of the following ents about the competition and the winning solution is most?
•	The winning algorithm involved a complex hybrid algorithm that used statistical/machine learning techniques to mix together a variety of general-purpose and special-purpose algorithms, in the end resulting in a significantly improved prediction performance for the competition data.
	The winning algorithm focused on making the best possible top-100 predictions, and thus turned out to be very useful for on-screen recommendation even though it wasn't as useful at predicting "deeper down" in the set of movies.
	Because the competition limited algorithms to using only ratings data, the winning algorithm wasn't nearly as good as an ordinary algorithm that also factors in user and item attributes like age, genre, or cast.



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