# Collaborative Filtering

#### Stavros Korokithakis

Intelligent Systems in Business

# Problem?

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- Too little time.
- How can I choose?
- Collaborative filtering, that's how!

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- Consider the entire userbase and only the userbase
- Do not know anything about the item

- Harder to implement, but provide more advantages
- Build a model of the item and use its aspects to provide recommendations
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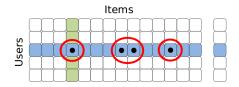
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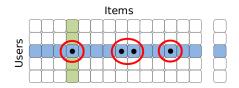
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#### • Utilises a memory-based technique

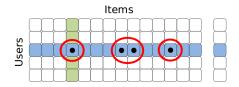
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- Considers users as vectors and uses their cosine to gauge taste compatibility.
- Runs relatively fast off-line and offers good performance.



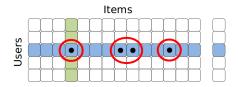
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- Google need to be able to filter thousands of items in near-realtime, with items having a lifetime of a few days at most.
- No clear vote of preference from the user.
- Approach must be content-agnostic so it can be applied to other types of media as well.

∪.к. ⊻	Business	
Top Stories Starred ☆ World U.K.	The Guardian	WCL student's ISB presentation "mindblowing"     The Guardian - 1 hour later     Students who were lucky enough to attend Stavros Korokithakis' ISB presentation described it     as "mindblowing" and "nothing short of spectacular.     B Korokithakis' presentation ∰ ITN NEWS
Business Sci/Tech Entertainment		<u>"It was only OK", naysayers complein</u> BBC News This is Money - Redures - Wall Street Journal all 1.283 news articles . C∂ Email this story ∵ Vauxhall given £270 million lifeline by Government
Sports Health Spotlight Most Popular	BBC News	Telegraph.co.uk - <u>Andrew Porter</u> - <b>3 hours ago</b> Car manufacturer Yuxahi, which has been under the threat of closure for a year, was last night offered a lifeline by the Government. <u>UK agrees loan guarantee for GM Europe</u> Reuters <u>Yauxhall and Ope jed st 270m UK loans guarantee</u> BBC News
		Times Online - Financial Times - Wall Street Journal - ShareCast all 258 news articles » Email this story

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#### Amazon

- Why not group relevant items instead of users with the same tastes?
- If many users have bought pairs of items, consider them similar.
- Can be calculated off-line.
- Needs almost no data about the user.

**Customers Who Bought This Item Also Bought** 



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# Questions?

# This presentation was amazing.

