1.	Recommending items based on global popularity can (check all that apply):	1 point
	provide personalization	
	capture context (e.g., time of day)	
	v none of the above	
2.	Recommending items using a classification approach can (check all that apply):	1 point
	v provide personalization	
	✓ capture context (e.g., time of day)	
	none of the above	
3.	Recommending items using a simple count based co-occurrence matrix can (<i>check all that apply</i>):	1 point
	✓ provide personalization	
	☐ capture context (e.g., time of day)	
	none of the above	

4.	Recommending items using featurized matrix factorization can (<i>check all that apply</i>):	1 point
	v provide personalization	
	✓ capture context (e.g., time of day)	
	none of the above	
5.	Normalizing co-occurrence matrices is used primarily to account for:	1 point
	O people who purchased many items	
	items purchased by many people	
	eliminating rare products	
	one of the above	
6.	A store has 3 customers and 3 products. Below are the learned feature vectors for each user and product. Based on this estimated model, which product would you recommend most highly to <i>User #2</i> ?	1 point

T	pon	ıι

User ID	Feature vector
1	(1.73, 0.01, 5.22)
2	(0.03, 4.41, 2.05)
3	(1.13, 0.89, 3.76)

Product ID	Feature vector
1	(3.29, 3.44, 3.67)

2	(0.82, 9.71, 3.88)
3	(8.34, 1.72, 0.02)

Product #1

Product #2

Product #3

7. For the liked and recommended items displayed below, calculate the recall and round to 2 decimal points. (As in the lesson, green squares indicate recommended items, magenta squares are liked items. Items not recommended are grayed out for clarity.) Note: enter your answer in American decimal format (e.g. enter 0.98, not 0,98)

1 point















0.33

8. For the liked and recommended items displayed below, calculate the precision and round to 2 decimal points. (*As in the lesson, green squares indicate recommended items, magenta squares are liked items. Items not recommended are grayed out for clarity.*) Note: enter your answer in American decimal format (e.g. enter 0.98, not 0,98)

1 point











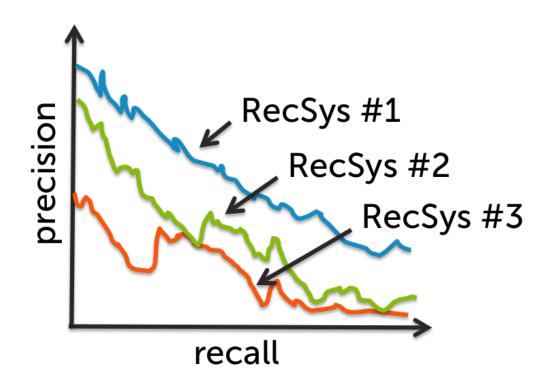




0.25

9. Based on the precision-recall curves in the figure below, which recommender would you use?

1 point



- RecSys #1
- RecSys #2
- RecSys #3

Coursera Honor Code Learn more



I, **Oleg Nyzhnyk**, understand that submitting work that isn't my own may result in permanent failure of this course or deactivation of my Coursera account.

Submit Six Saving...

 $\stackrel{\ \ \, }{\bigcirc}$ Like $\stackrel{\ \ \, }{\bigcirc}$ Dislike $\stackrel{\ \ \, }{\bigcirc}$ Report an issue