## Recommender Systems

Quiz, 9 questions

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

https://www.coursera.org/learn/ml-foundations/exam/i1Jnh/recommender-systems

| Recommender Systems Recommending items using <b>featurized matrix factorization</b> can ( <i>check all that apply</i> ): |
|--|
| Quiz, 9 questions  |
| provide personalization  |
| capture context (e.g., time of day)  |
| none of the above  |
|  |
| 1  |
| point  |
| 5.   |
| Normalizing co-occurrence matrices is used primarily to account for:   |
| people who purchased many items  |
| items purchased by many people   |
| eliminating rare products  |
| none of the above  |
|  |
| 1  |
| 1 point  |
| 6.   |
| 0.   |

A store has 3 customers and 3 products. Below are the learned feature vectors for each user and Recommender Systems product. Based on this estimated model, which product would you recommend most highly to *User* Qui#99questions

| 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

1 point

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)















Enter answer here

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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)















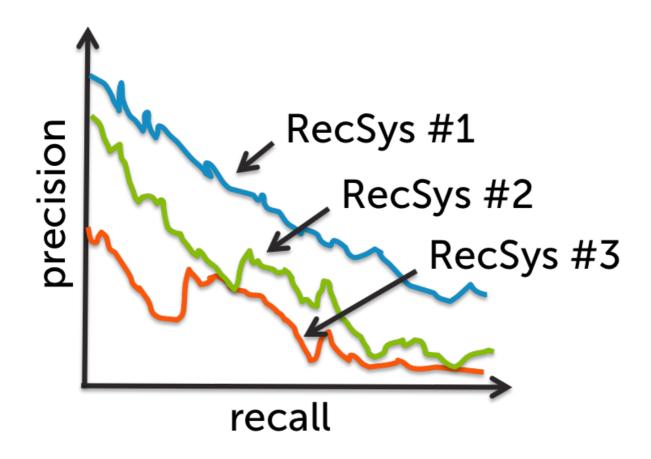
Enter answer here

1 point

9.

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

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| RecSvs | #1 |
|--------|----|

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RecSys #2

RecSys #3

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