

Food Recommender Through Collaborative Filtering

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1 Load the model

```
1 from collaborative_filtering import main
2
3 def apply_cf():
4     main() # Main function of collaborative filtering
5     return
6
7 if __name__ == "__main__":
8     apply_cf()
```

2 Algorithm demonstration

Algorithm 1 Collaborative filtering

Input: $R \in \mathbb{R}^{m \times n}$ (Score matrix where each row represents a user's scores on foods), k (Number of similar users considered in prediction)

Output: R^{pred} (Score matrix where originally 0 scores are predicted by the algorithm)

- 1: For each pair of users, measure their similarity using cosine similarity $S_{ij} = \frac{R_i^\top R_j}{\|R_i\| \|R_j\|}$, where $i, j \in \{1, \dots, n\}$ and $i \neq j$.
 - 2: $R^{\text{pred}} \leftarrow \mathbf{0}_{m \times n}$. ▷ Initialize the prediction matrix.
 - 3: **for** $i \in \{1, \dots, m\}$ **do**
 - 4: **for** $j \in \{1, \dots, n\}$ **do**
 - 5: **if** $R_{ij} > 0$ **then**
 - 6: $R_{ij}^{\text{pred}} \leftarrow R_{ij}$. ▷ The score already exists so does not need prediction.
 - 7: **else**
 - 8: For each user $i \in \{1, \dots, m\}$, find the most similar k users as $\mathcal{N}_i = \{u_1, \dots, u_k\}$ where $S_{iu_1} \geq S_{iu_2} \geq \dots \geq S_{iu_k}$.
 - 9: Among the the most similar k users of user i , select the similar users that has scores on food j , as $\hat{\mathcal{N}}_i = \{\hat{u}_1, \dots, \hat{u}_{k'}\}$ where k' denotes the size of $\hat{\mathcal{N}}_i$.
 - 10: Predict the score of user i on food j : $R_{ij}^{\text{pred}} \leftarrow \frac{1}{k'} \sum_{u=1}^{k'} S_{iu} R_{uj}$. ▷ Mean of weighted sum of scores by users in $\hat{\mathcal{N}}_i$.
 - 11: $R_{ij}^{\text{pred}} \leftarrow \text{round}(\text{clamp}(R_{ij}^{\text{pred}}, 1, 5))$. ▷ Ensure that the predicted score is an integer in $[1, 5]$.
 - 12: **end if**
 - 13: **end for**
 - 14: **end for**
 - 15: **return** R^{pred}
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3 Test results of the best model

RMSE = 1.1905, MAE = 0.9173.