

CS4033/5033: Assignment 5

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Topic 1: Low-Rank Matrix Factorization

Implement low-rank matrix factorization based on the alternate least square technique. Evaluate the technique on a real-world user-movie rating matrix. In this matrix, each row corresponds to a user and each column corresponds to a movie. Ratings take value in $\{1, 1.5, 2, 2.5, 3, 3.5, 4, 4.5, 5\}$; you should treat ‘0’ entries in the matrix as (truly) missing ratings and do not use it for either training or testing. For convenience, we have separated the set of observed ratings that should be used for training (stored in ‘rate_train.csv’) from the set of observed ratings that should be used for testing (stored in ‘rate_test.csv’).

Task 1. Learn a rank- k factorization of the rating matrix from the training set and evaluate it on the testing set. Report testing error versus the number of ALS updates in Figure 1. Pick the value for k , λ_1 , λ_2 and the maximum number of ALS updates by yourself.

Tip: you may round the predicted rates (given by UV) to the nearest value in $\{1, 1.5, 2, 2.5, 3, 3.5, 4, 4.5, 5\}$ to get a more meaningful prediction and possibly higher prediction accuracy.

Task 1 - Solution.

We model the rating matrix X as a low-rank product $X \approx UV$ with $U \in \mathbb{R}^{n \times k}$, $V \in \mathbb{R}^{k \times m}$. Let $O \subset \{1, \dots, n\} \times \{1, \dots, m\}$ be the set of observed (non-zero) entries in X . The ridge-regularized objective function is:

$$\min_{U,V} \sum_{(i,j) \in O} (X_{ij} - U_{i:} V_{:j})^2 + \lambda_1 \|U\|_F^2 + \lambda_2 \|V\|_F^2 \quad (1)$$

Holding V fixed, each user row $U_{i:}$ solves a regularized least-squares normal equation, and symmetrically for each movie column $V_{:j}$. We alternate these updates for a fixed number of iterations.

$$U_{i:} = \left(\sum_{j:(i,j) \in O} X_{ij} V_{:j}^T \right) \left(\sum_{j:(i,j) \in O} V_{:j} V_{:j}^T + \lambda_1 I \right)^{-1} \quad (2)$$

,

$$V_{:j} = \left(\sum_{i:(i,j) \in O} U_{i:}^T U_{i:} + \lambda_2 I \right)^{-1} \left(\sum_{i:(i,j) \in O} X_{ij} U_{i:}^T \right) \quad (3)$$

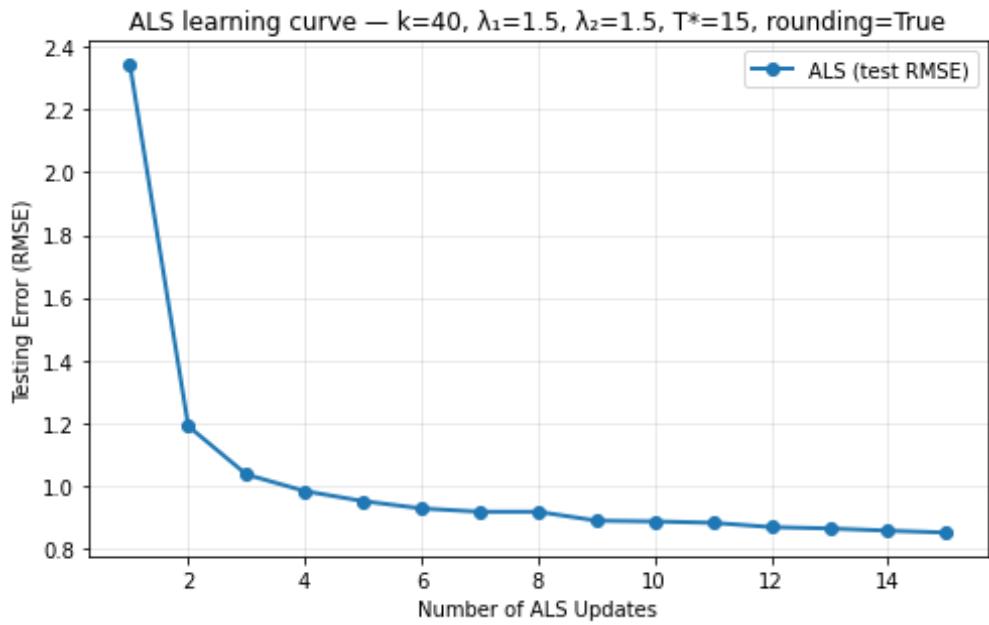


Fig. 1. Testing Error versus ALS Updates with $k = 40$

Task 2. Report testing error versus k in Figure 2. Pick five values of k and fix other configurations by yourself.

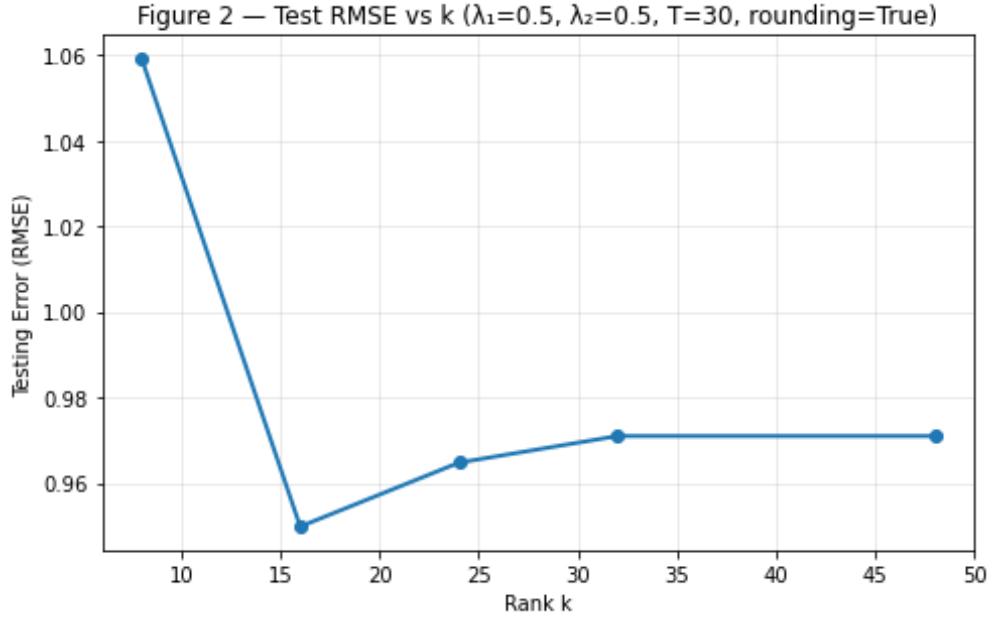


Fig. 2. Testing Error versus k

Topic 2: Markov Model

Suppose we want to learn a Markov model, and apply it to predict Sam's final grade as A or B. Our training data are shown in Figure 3. (The first row is Sam's record. Do not use it for training.)

Task 3. Assume 1st-order Markov chain. Evaluate the probability for Sam to get an A in the 5th semester. You need to elaborate on the estimation of key probabilities and the key formula such as Bayes theorem.

Task 3 - Solution. We model grades $X_t \in \{A, B\}$ as a time-homogeneous first-order Markov chain:

$$Pr(X_{t+1} = y | X_1, \dots, X_t = x_t) = Pr(X_{t+1} = y | X_t = x_t) \equiv p_{x_t y} \quad (4)$$

We need $Pr(X_5 = A | X_4 = B)$ for Sam. By the property this is exactly the transition probability p_{BA}

Estimate p_{xy} from the training rows (X1-X8 only) via MLE using bigram counts over adjacent semesters $t \rightarrow t + 1$ (for $t = 1, 2, 3, 4$):

$$N_{xy} = \#\{t : X_t = x, X_{t+1} = y\}, \quad N_x = \sum_{y \in \{A, B\}} N_{xy}, \quad \hat{p}_{xy} = \frac{N_{xy}}{N_x} \quad (5)$$

Equivalently, by Bayes' rule,

$$Pr(X_{t+1} = A | X_t = B) = \frac{Pr(X_{t+1} = A)}{Pr(X_t = B)} \approx \frac{N_{BA}/N}{N_B/N} = \frac{N_{BA}}{N_B} \quad (6)$$

where $N = \sum_{x,y} N_{xy}$ is the total number of observed transitions; the common N cancels.

From the training data, (rows $X_1 : X_8$, not using Sam), counting all $B \rightarrow \cdot$ transitions across the 4 semester steps gives

$$N_{BA} = 6, \quad N_{BB} = 4, \quad N_B = 10 \quad (7)$$

Hence

$$Pr(X_5 = A | X_4 = B) = \hat{p}_{BA} = \frac{6}{10} = 0.6 \quad (8)$$

So, under a 1st-order, time-homogeneous Markov model, the estimated probability that Sam gets an A in the 5th semester (given $X_4 = B$) is 0.6.

Estimate transition matrix

The estimated transition matrix is:

Note:

$$\hat{P} = \begin{bmatrix} \hat{p}_{AA} & \hat{p}_{AB} \\ \hat{p}_{BA} & \hat{p}_{BB} \end{bmatrix} = \begin{bmatrix} \frac{1}{3} & \frac{2}{3} \\ \frac{3}{5} & \frac{2}{5} \end{bmatrix} \quad (9)$$

Task 4. Assume 2nd-order Markov chain. Evaluate the probability for Sam to get an A in the 5th semester. You need to elaborate on the estimation of key probabilities and the key formula such as Bayes theorem.

Task 4 - Solution. Let $X_t \in \{A, B\}$ denote the grade in semester t . We model grades as a time-homogeneous second-order Markov chain:

$$Pr(X_{t+1} = z | X_t = y, X_{t-1} = x) = \theta_{xy \rightarrow z}, \quad x, y, z \in \{A, B\} \quad (10)$$

For Sam we need

$$Pr(X_5 = A | X_4 = B, X_3 = A) = \theta_{AB \rightarrow A} \quad (11)$$

since Sam's history is $X_1 = B, X_2 = A, X_3 = A, X_4 = B$.

[MLE of the 2nd-order transition]

From the training rows only (X1-X8), form all adjacent triples (X_t, X_{t+1}, X_{t+2}) for $t = 1, 2, 3$ in each row.

For each context pair (x, y) , count

$$N_{xy \rightarrow z} = \#\{t : X_t = x, X_{t+1} = y, X_{t+2} = z\}, \quad N_{xy} = \sum_{z \in \{A, B\}} N_{xy \rightarrow z} \quad (12)$$

The MLE is

$$\hat{\theta}_{xy \rightarrow z} = \frac{N_{xy \rightarrow z}}{N_{xy}} \quad (13)$$

Equivalently, by Bayes' rule,

$$Pr(X_{t+2} = z | X_{t+1} = y, X_t = x) = \frac{Pr(x, y, z)}{Pr(x, y)} \approx \frac{N_{xy \rightarrow z}}{N_{xy}} \quad (14)$$

[Counts for the needed context (A, B)]

From the table (rows X_1 to X_8 only), the triples give

$$N_{AB \rightarrow A} = 6, \quad N_{AB \rightarrow B} = 3, \quad N_{AB} = 9 \quad (15)$$

Hence

$$Pr(X_5 = A | X_4 = B, X_3 = A) = \hat{\theta}_{AB \rightarrow A} = \frac{6}{9} = \frac{2}{3} \approx 0.6667 \quad (16)$$

Semester	1st	2nd	3rd	4th	5th
Sam	B	A	A	B	?
X1	B	A	B	B	B
X2	A	B	A	B	A
X3	A	A	B	A	B
X4	B	B	A	B	A
X5	A	A	A	B	A
X6	A	B	B	A	A
X7	B	B	A	B	B
X8	B	A	B	A	B

Fig. 3. Sam's and other students' grade records

Submission Instruction

Please submit three files on Canvas.

- (i) Submit a ‘hw5.pdf’. It should contain your answers to all questions in this template.
- (ii) Submit a ‘hw5_als.py’. It should be the code that generates Figure 1.
- (iii) Submit a ‘hw5_k.py’. It should be the code that generates Figure 2.