The final predictions given by the recommendation system are the mean of two different predictions.

- 1. SVD prediction
- 2. Global Modeling prediction

A comprehensive description of both these methods is available on youtube. Search keywords "latent factor models".

SVD prediction:

SVD prediction is generated by decomposing the urm which is the user-rating matrix. urm is a ixj matrix where i and j are, respectively, number of users and number of films. Every (i,j)th element is the rating user i gave for the movie j. For a movie j that user i hasn't rated, the (i,j)th element is 0.

However, for implementing proper SVD, we cannot have empty, or zero, elements in the user-rating matrix. We fix this problem by mean normalization. For every movie the user hasn't rated, we consider its rating to be the average of the user's ratings for the movies he/she has watched.

We have two matrices by this time: original_matrix and urm. The O elements in urm have been normalized and the original_matrix still has all of its zeros. We do this so that we can later classify which ratings are real and which are predicted by our program.

SVD is performed on the normalized urm, which gives us U, S and V matrices. We predict ratings from these matrices.

Global Modeling prediction:

To predict the rating given by user i for film j, we first find ratings_mean, which is the mean of all ratings in original_matrix. Then we find the mean film_ratings_mean of the ratings given by different users for the film we have to predict rating for. We subtract ratings_mean from film_ratings_mean to see how well-rated this film is compared to all other movies in the database.

Similarly, we get the intuition on the user's taste based on how critical they are with their ratings. We find the mean user_ratings_mean of ratings the user has so far given. We subtract ratings_mean from user_ratings_mean to see how different the user's taste is from the rest of the database. Then we add all three:

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ratings_mean + (film_ratings_mean-ratings_mean) + (user_ratings_mean-ratings_mean)
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This gives us another naïve rating. The final prediction is the mean between this rating and the rating generated from SVD.