PREDICTION FUNCTION

FUNCTION:

```
def predict(ratings,similarity,type='user'):
    if(type=='user'):
        mean_user_rating=ratings.mean(axis=1)
        #We use np.newaxis so that mean_user_rating has same format as ratings
        ratings_diff=(ratings-mean_user_rating[:,np.newaxis])
        pred=mean_user_rating[:,np.newaxis] + similarity.dot(ratings_diff) /
np.array([np.abs(similarity).sum(axis=1)]).T
    elif(type=='item'):
        pred = ratings.dot(similarity) / np.array([np.abs(similarity).sum(axis=1)])
    return pred
```

User-based Collaborative Filtering (type='user'):

Step 1: Compute the Mean Rating for Each User:

```
mean_user_rating = ratings.mean(axis=1)
```

Step 2:Reshapes the 1D array into a 2D column vector By adding [:, np.newaxis]:

```
mean_user_rating[:, np.newaxis]
```

Step 3: Center the Ratings by Subtracting the Mean:

```
ratings_diff = (ratings - mean_user_rating[:, np.newaxis])
```

Step 4:Apply np.abs(similarity) to take the absolute values of all elements:

```
np.abs(similarity)
```

Step 5: Sum Across Each Row (axis=1)

```
np.abs(similarity).sum(axis=1)
```

Step 6:Convert to a NumPy Array:

```
np.array([np.abs(similarity).sum(axis=1)])
```

Step 7:Compute the normalization term

```
np.array([np.abs(similarity).sum(axis=1)]).T
```

Step 8: Multiplies the similarity matrix by the ratings_diff matrix using matrix multiplication (the .dot() method).

```
similarity.dot(ratings_diff)
```

Step 9:Normalize the weighted sum:

similarity.dot(ratings_diff) / np.array([np.abs(similarity).sum(axis=1)]).T

Step 10:Add the user's mean rating back to get pred:

pred=mean_user_rating[:,np.newaxis] + similarity.dot(ratings_diff) /
np.array([np.abs(similarity).sum(axis=1)]).T

Item-based Collaborative Filtering (type='item')

Step 1:Compute the Weighted Sum of Ratings
multiplies the ratings matrix with the similarity matrix.

ratings.dot(similarity)

Step 2:Normalize the Weighted Sum of Ratings

pred = ratings.dot(similarity) / np.array([np.abs(similarity).sum(axis=1)])