

## Step-by-Step Conversion

### 1. Inherit from `BaseEstimator`:

- We will inherit from `BaseEstimator` to get basic functionality like hyperparameter setting and getting.

### 2. Define the `fit` and `predict` methods:

- The `fit` method will compute the global mean of the training ratings.
- The `predict` method will return the global mean for each test pair.

Here's the example code for global mean RS methods:

```
import numpy as np
from sklearn.base import BaseEstimator

class GlobalMeanRS(BaseEstimator):
    def __init__(self):
        # model parameters
        self.glb_mean_ = 0

    def fit(self, rating):
        # fit parameter
        self.glb_mean_ = np.mean(rating)

    def predict(self, pair):
        # Ensure the estimator is fitted
        check_is_fitted(self, 'glb_mean_')
        # Return the global mean for each test pair
        r_pred = np.ones(len(pair))
        return r_pred*self.glb_mean_
```

## Explanation

### 1. Initialization (`__init__` method):

- Initialize `glb_mean_` to 0. The trailing underscore indicates that this is an attribute set during the `fit` method.

### 2. Fitting the Model (`fit` method):

- Compute the global mean of the training ratings and store it in `glb_mean_`.

### 3. Making Predictions (`predict` method):

- Ensure the estimator has been fitted using `check_is_fitted`.
- Return the global mean for each test pair.

## Usage

The example usage demonstrates how to use this custom estimator with Scikit-learn's familiar `fit`, `predict` methods. This allows you to leverage Scikit-learn's powerful tools for model selection, evaluation, and preprocessing while using your tailored algorithms.

By following this approach, you can create custom machine learning estimators that integrate seamlessly with Scikit-learn's ecosystem.