

some times like in the movie recommendation system the system recommends the movies on the base of people who have also watched so while recommending this movies the system doesn't necessary to know what is the name of movie etc etc still it can recommend users some movies on the basis of what movies he have watched

here we are doing the analysis of movie rating and the user from the list of user in pandas dataframe which is available there

for merging the movie names with code of the dataframe we can use

```
ratings = ratings.merge(movies)
```

and after that we can get the movies along with the names code which it's getting passed

for loading the data with the dataloader api we can do it using

```
dls = CollabDataLoaders.from_df(ratings, item_name = 'title', bs = 64)
```

and we can use the `dls.show_batch()` for showing the batch

now we will need to create the number of users and the number of movies and for predicting them we need to have some number of factors which are gonna act as the weights for the model

for getting the latent factor we have to get the value by using the dot product of instruction

looking up at the index can be represented as the matrix products

if we replace the indices with the one hot encoded vector it is identical to looking up at the index

we are multiplying the one hot encoded number to the user factors if we did this with the matrix multiplication is similar as the `dot_product`

`embedding` : embedding is matrix multiplication with the one hot encoding where you don't have to do one hot encoding cause it already knows the gradients of one hot encode it's so far the efficient way for doing the matrix multiplication

Creating the collaborative filtering from scratch:

init method is the special method it can be used to initiate the function when you want to store some keywords while initialization

inheriting the properties of other class for using it in

we can do it like `class DotProduct(Module)` here we have inherited properties of Module class from python to the dot product class

the Module class is the pytorch class which is little bit modified by the fast ai

here we are defining the

```
self.user_factors = Embedding(n_users, n_factors)
```

here embedding layer is doing the matrix multiplication with the dot product in more efficient way

while we call any function in pytorch it automatically calls the forward here we can use forward function help for doing the multiplication and the addition

it's the special pytorch method this is where we are doing the actual computation method here we are passing the user id's and the movie id's

here we can pass the output as the sigmoid range, pass the user bias etc things to normalize the model but it won't prevent overfitting. for preventing overfitting we have to do `regularization`

Regularization:

regularization is the set of techniques which allows us to use the models with lot of parameters and train them for lot of period of time and penalize them effectively for preventing over fitting cause them to stop overfitting

←-----→ lesson 6 over ←-----→