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1. (1%)請比較有無 normalize(rating)的差別。並說明如何 normalize.

(collaborator:)

標準 model: latent dimension=64, batch size=256, epochs=100

Normalized model: 設定跟上面相同,將 rating 直接除以 5

	Kaggle (public)
標準	0.86086
Normalized	2.85070

因為我在澳洲參加 conference,因此提早準備 report,以 kaggle public 分數為討論依據。可以看出 normalized 的效果非常差,我原本預期做完 normalized 效果會比較好,或許換別的 normalization 方法效果會比較好。

2. (1%)比較不同的 latent dimension 的結果。

(collaborator:)

以第一題的標準 model 為主,改變 latent dimension

Latent dimension	Kaggle (public)
32	0.86144
64	0.86068
128	0.86162
256	0.88420

3. (1%)比較有無 bias 的結果。

(collaborator:)

	Kaggle (public)
標準(have bias)	0.86086
No Bias	0.92461

4. (1%)請試著用 DNN 來解決這個問題,並且說明實做的方法(方法不限)。並比較 MF 和 NN 的結果,討論結果的差異。

(collaborator:)

MF model(標準): 將 user 跟 movie 做 embedding,轉換成 vector(Flatten),最後將兩者相乘(dot)。

DNN model: 把上述的 Model 中的 dot 改成 concatenate, 然後接 DNN。

	Kaggle (public)
標準(MF)	0.86086
DNN	0.86243

兩者結果相近。

## MF model:

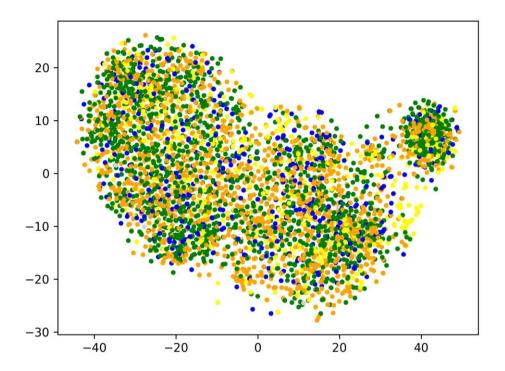
Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 1)	0	
input_2 (InputLayer)	(None, 1)	0	
embedding_1 (Embedding)	(None, 1, 64)	386560	input_1[0][0]
embedding_2 (Embedding)	(None, 1, 64)	252928	input_2[0][0]
flatten_1 (Flatten)	(None, 64)	0	embedding_1[0][0]
flatten_2 (Flatten)	(None, 64)	0	embedding_2[0][0]
dropout_1 (Dropout)	(None, 64)	0	flatten_1[0][0]
dropout_2 (Dropout)	(None, 64)	0	flatten_2[0][0]
embedding_3 (Embedding)	(None, 1, 1)	6040	input_1[0][0]
embedding_4 (Embedding)	(None, 1, 1)	3952.	input_2[0][0]
dot_l (Dot)	(None, 1)	0	dropout_1[0][0] dropout_2[0][0]
flatten_3 (Flatten)	(None, 1)	0	embedding_3[0][0]
flatten_4 (Flatten)	(None, 1)	0	embedding_4[0][0]
add_1 (Add)	(None, 1)	0	dot_1[0][0] flatten_3[0][0] flatten_4[0][0]
Total params: 649,480 Trainable params: 649,480 Non-trainable params: 0			

## DNN model:

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	(None, 1)	0	
input_2 (InputLayer)	(None, 1)	0	
embedding_1 (Embedding)	(None, 1, 64)	386560	input_1[0][0]
embedding_2 (Embedding)	(None, 1, 64)	252928	input_2[0][0]
flatten_1 (Flatten)	(None, 64)	0	embedding_1[0][0]
flatten_2 (Flatten)	(None, 64)	0	embedding_2[0][0]
dropout_1 (Dropout)	(None, 64)	0	flatten_1[0][0]
dropout_2 (Dropout)	(None, 64)	0	flatten_2[0][0]
concatenate_1 (Concatenate)	(None, 128)	0	dropout_1[0][0] dropout_2[0][0]
dense_1 (Dense)	(None, 128)	16512	concatenate_1[0][0]
dense_2 (Dense)	(None, 64)	8256	dense_1[0][0]
dense_3 (Dense)	(None, 32)	2080	dense_2[0][0]
dense_4 (Dense)	(None, 1)	33	dense_3[0][0]
Total params: 666,369 Trainable params: 666,369			

Total params: 666,369 Trainable params: 666,369 Non-trainable params: 0 5. (1%)請試著將 movie 的 embedding 用 tsne 降維後,將 movie category 當作 label 來作圖。

(collaborator:)



Orange	['Animation' "Children's" 'Comedy' 'Adventure']
Green	['Romance' 'Drama' 'Documentary' 'Musical']
Blue	['Fantasy' 'Action' 'Sci-Fi' 'War' 'Western']
Yellow	['Crime' 'Thriller' 'Horror' 'Film-Noir']
Lightgray	['other']

6. (BONUS)(1%)試著使用除了 rating 以外的 feature, 並說明你的作法和結果, 結果 好壞不會影響評分。

(collaborator:) 我用 DNN model 加上額外的 features(user occupation & movie genre) 一起放進 DNN,效果意外的好: 0.86243→0.85019

Layer (type)	Output Shape	Param #	Connected to
input 1 (InputLayer)	(None, 1)	0	
input_2 (InputLayer)	(None, 1)	0	
embedding 1 (Embedding)	(None, 1, 128)	773248	input 1[0][0]
embedding 2 (Embedding)	(None, 1, 128)	505984	input_2[0][0]
input_3 (InputLayer)	(None, 21)	0	
input 4 (InputLayer)	(None, 18)	0	
flatten_1 (Flatten)	(None, 128)	0	embedding_1[0][0]
flatten 2 (Flatten)	(None, 128)	0	embedding 2[0][0]
dense_1 (Dense)	(None, 128)	2816	input_3[0][0]
dense 2 (Dense)	(None, 128)	2432	input 4[0][0]
dropout 1 (Dropout)	(None, 128)	0	flatten_1[0][0]
dropout 2 (Dropout)	(None, 128)		flatten 2[0][0]
dropout 3 (Dropout)	(None, 128)	0	dense 1[0][0]
dropout 4 (Dropout)	(None, 128)		dense 2[0][0]
concatenate 1 (Concatenate)	(None, 512)		dropout 1[0][0]
concatenate_1 (concatenate)	(None, 512)	U	dropout 2[0][0]
			dropout_3[0][0] dropout_4[0][0]
dense_3 (Dense)	(None, 128)	65664	concatenate_1[0][0]
dense_4 (Dense)	(None, 64)	8256	dense_3[0][0]
dense_5 (Dense)	(None, 32)	2080	dense_4[0][0]
dense_6 (Dense)	(None, 16)	528	dense_5[0][0]
dense_7 (Dense)	(None, 1)	17	dense_6[0][0]
Total parama: 1 261 025			

Total params: 1,361,025 Trainable params: 1,361,025 Non-trainable params: 0