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1. (1%)請比較有無 normalize 的差別。並說明如何 normalize. (ref: 去年手把手)

MF model(no normalize): public: 0.88594 private: 0.88139

MF model(normalize): public: 0.87487 private: 0.86755

Output Shape	Param #	Connected to
(None, 1)	θ	
(None, 1)	θ	
(None, 1, 100	604100	input_5[0][0]
(None, 1, 100	395300	input_6[0][0]
(None, 100)	θ	embedding_9[0][0]
(None, 100)	θ	embedding_10[0][0]
(None, 1, 1)	6041	input_5[0][0]
(None, 1, 1)	3953	input_6[0][0]
(None, 1)	θ	flatten_8[0][0] flatten_9[0][0]
(None, 1)	θ	embedding_11[0][0]
(None, 1)	θ	embedding_12[0][0]
(None, 1)	θ	dot_3[0][0] flatten_10[0][0] flatten_11[0][0]
	(None, 1) (None, 1, 100 (None, 1, 100 (None, 1, 100 (None, 100) (None, 100) (None, 1, 1) (None, 1, 1) (None, 1) (None, 1)	(None, 1) 0 (None, 1) 0 (None, 1, 100) 604100 (None, 1, 100) 395300 (None, 100) 0 (None, 100) 0 (None, 1, 1) 6041 (None, 1, 1) 3953 (None, 1) 0 (None, 1) 0 (None, 1) 0

Total params: 1,009,394 Trainable params: 1,009,394 Non-trainable params: 0

方法:rating = (rating - mean(rating)) / std(rating)

2. (1%)比較不同的 embedding dimension 的結果。(ref: 去年手把手)

NN model(dim = 500): public: 0.87449 private: 0.86563

Layer (type)	0utput	Shape	Param #	Connected to
input_7 (InputLayer)	(None,	1)	θ	
input_8 (InputLayer)	(None,	1)	θ	
embedding_11 (Embedding)	(None,	1, 500)	3020500	input_7[0][0]
embedding_12 (Embedding)	(None,	1, 500)	1976500	input_8[0][0]
flatten_9 (Flatten)	(None,	500)	θ	embedding_11[0][0]
flatten_10 (Flatten)	(None,	500)	θ	embedding_12[0][0]
concatenate_2 (Concatenate)	(None,	1000)	θ	flatten_9[0][0] flatten_10[0][0]
dense_4 (Dense)	(None,	150)	150150	concatenate_2[0][0]
batch_normalization_1 (BatchNorm	(None,	150)	600	dense_4[0][0]
leaky_re_lu_1 (LeakyReLU)	(None,	150)	θ	batch_normalization_1[0][0]
dropout_3 (Dropout)	(None,	150)	θ	leaky_re_lu_1[0][0]
dense_5 (Dense)	(None,	150)	22650	dropout_3[0][0]
batch_normalization_2 (BatchNorm	(None,	150)	600	dense_5[0][0]
leaky_re_lu_2 (LeakyReLU)	(None,	150)	θ	batch_normalization_2[0][0]
dropout_4 (Dropout)	(None,	150)	θ	leaky_re_lu_2[0][0]
dense_6 (Dense)	(None,	50)	7550	dropout_4[0][0]
batch_normalization_3 (BatchNorm	(None,	50)	200	dense_6[0][0]
leaky_re_lu_3 (LeakyReLU)	(None,	50)	θ	batch_normalization_3[0][0]
dropout_5 (Dropout)	(None,	50)	θ	leaky_re_lu_3[0][0]
dense_7 (Dense)	(None,	1)	51	dropout_5[0][0]

Total params: 5,178,801 Trainable params: 5,178,101 Non-trainable params: 700

NN model(dim = 100): public: 0.86752 private: 0.85880

Layer (type)	Output	Chana	Param #	Connected to
Layer (type)	output	511ape		
input_5 (InputLayer)	(None,	1)	θ	
input_6 (InputLayer)	(None,	1)	θ	
embedding_5 (Embedding)	(None,	1, 100)	604100	input_5[0][0]
embedding_6 (Embedding)	(None,	1, 100)	395300	input_6[0][0]
flatten_5 (Flatten)	(None,	100)	θ	embedding_5[0][0]
flatten_6 (Flatten)	(None,	100)	θ	embedding_6[0][0]
concatenate_3 (Concatenate)	(None,	200)	θ	flatten_5[0][0] flatten_6[0][0]
dropout_9 (Dropout)	(None,	200)	θ	concatenate_3[0][0]
dense_9 (Dense)	(None,	150)	30150	dropout_9[0][0]
batch_normalization_7 (BatchNorm	(None,	150)	600	dense_9[0][0]
leaky_re_lu_7 (LeakyReLU)	(None,	150)	θ	batch_normalization_7[0][0]
dropout_10 (Dropout)	(None,	150)	θ	leaky_re_lu_7[0][0]
dense_10 (Dense)	(None,	150)	22650	dropout_10[0][0]
batch_normalization_8 (BatchNorm	(None,	150)	600	dense_10[0][0]
leaky_re_lu_8 (LeakyReLU)	(None,	150)	θ	batch_normalization_8[0][0]
dropout_11 (Dropout)	(None,	150)	θ	leaky_re_lu_8[0][0]
dense_11 (Dense)	(None,	50)	7550	dropout_11[0][0]
batch_normalization_9 (BatchNorm	(None,	50)	200	dense_11[0][0]
leaky_re_lu_9 (LeakyReLU)	(None,	50)	θ	batch_normalization_9[0][0]
dropout_12 (Dropout)	(None,	50)	θ	leaky_re_lu_9[0][0]
dense_12 (Dense)	(None,	1)	51	dropout_12[0][0]
Total params: 1 061 201				

Total params: 1,061,201 Trainable params: 1,060,501 Non-trainable params: 700

Dim = 500 的情況下,參數量有點過多,導致 overfitting。

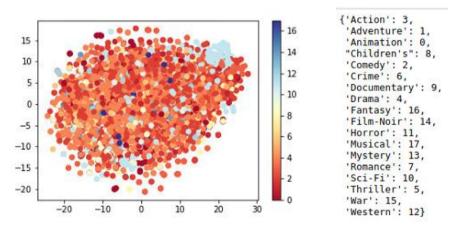
3. (1 %)比較有無 bias 的結果。(ref: 去年手把手)

MF model(有 bias): public: 0.87487 private: 0.86755

MF model(無 bias): public: 0.87545 private: 0.86861

可以發現有 bias 雖然成績會比較好,但沒有影響那麼多。

4. (1 %)請試著將 movie 的 embedding 用 tsne 降維後,將 movie category 當作 label 來作圖。(ref: 去年手把手)



PS: 當有超過一個 category 時,取第一個 category。(MF model with bias) 效果並沒有很好,沒有把不同的 category 分堆,可能是因為只取第一個 category 的關係。

5. (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。

方法:將 movie. csv 裡面的 Genres 拿來用,因為每一個 movie 可能會有很多種種類,所以將此做成一個 shape=(18,1)的 one hot encoding 的 np. array 加入 nn model 的 input layer 去 train。

Layer (type)	Output	Shape	Param #	Connected to
input_4 (InputLayer)	(None,	1)	Θ	
input_5 (InputLayer)	(None,	1)	Θ	
embedding_3 (Embedding)	(None,	1, 100)	604100	input_4[0][0]
embedding_4 (Embedding)	(None,	1, 100)	395300	input_5[0][0]
flatten_3 (Flatten)	(None,	100)	0	embedding_3[0][0]
flatten_4 (Flatten)	(None,	100)	θ	embedding_4[0][0]
input_6 (InputLayer)	(None,	18)	Θ	
concatenate_2 (Concatenate)	(None,	218)	θ	flatten_3[0][0] flatten_4[0][0] input_6[0][0]
dropout_5 (Dropout)	(None,	218)	0	concatenate_2[0][0]
dense_5 (Dense)	(None,	150)	32850	dropout_5[0][0]
batch_normalization_4 (BatchNorm	(None,	150)	600	dense_5[0][0]
leaky_re_lu_4 (LeakyReLU)	(None,	150)	Θ	batch_normalization_4[0][0]
dropout_6 (Dropout)	(None,	150)	Θ	leaky_re_lu_4[0][0]
dense_6 (Dense)	(None,	150)	22650	dropout_6[0][0]
batch_normalization_5 (BatchNorm	(None,	150)	600	dense_6[0][0]
leaky_re_lu_5 (LeakyReLU)	(None,	150)	Θ	batch_normalization_5[0][0]
dropout_7 (Dropout)	(None,	150)	0	leaky_re_lu_5[0][0]
dense_7 (Dense)	(None,	50)	7550	dropout_7[0][0]
batch_normalization_6 (BatchNorm	(None,	50)	200	dense_7[0][0]
leaky_re_lu_6 (LeakyReLU)	(None,	50)	θ	batch_normalization_6[0][0]
dropout_8 (Dropout)	(None,	50)	Θ	leaky_re_lu_6[0][0]
dense_8 (Dense)	(None,	1)	51	dropout_8[0][0]

Total params: 1,063,901 Trainable params: 1,063,201 Non-trainable params: 700

## 結果:

public score:  $0.86752 \rightarrow 0.86103$  private:  $0.85880 \rightarrow 0.84892$