CCBDA HW#5

● **How to run**: 使用 jupyter-notebook 或 google colab 逐格 執行即可

Method:

使用 model 為 pyspark. ml. recommendation 中的 ALS algorithm.

```
1 from pyspark.ml.tuning import ParamGridBuilder, CrossValidator
```

在一開始的時候我嘗試使用 CrossValidator, 想找出 best model, 使用的參數為

```
1 # Build cross validation using CrossValidator
2 cv = crossValidator(estimator=als, estimatorParamMaps=param_grid, evaluator=evaluator, numFolds=5)
```

執行結果

```
22/12/31 00:17:31 WARN DAGScheduler: Broadcasting large task binary with size 1345.7 KiB
22/12/31 00:17:31 WARN DAGScheduler: Broadcasting large task binary with size 1401.7 KiB
22/12/31 00:17:31 WARN DAGScheduler: Broadcasting large task binary with size 1400.4 KiB
22/12/31 00:17:32 WARN DAGScheduler: Broadcasting large task binary with size 1575.0 KiB
22/12/31 00:17:34 WARN DAGScheduler: Broadcasting large task binary with size 1594.9 KiB
22/12/31 00:17:36 WARN DAGScheduler: Broadcasting large task binary with size 1595.8 KiB
RMSE=1.2254533350717227
22/12/31 00:17:36 WARN DAGScheduler: Broadcasting large task binary with size 1400.4 KiB
22/12/31 00:17:36 WARN DAGScheduler: Broadcasting large task binary with size 1401.7 KiB
22/12/31 00:17:36 WARN DAGScheduler: Broadcasting large task binary with size 1346.1 KiB
22/12/31 00:17:37 WARN DAGScheduler: Broadcasting large task binary with size 1576.0 KiB
[Stage 209:----> (182 + 16) / 200]
      iteml
                  user|rating|item index|user index|prediction|
|B000SQTJJ0|A31BD4RXCON7Q0| 4| 148.0| 7634.0| 3.4450731|
|B000SQTJJ0|A2TTHN1UM082VY| 5| 148.0| 329.0| 3.5525923|
|B000SQTJJ0|A3LUYUZNKG378S| 5| 148.0| 12556.0| 4.1586866|
|B000SQTJJ0|A1RF9YK4BK5TRH| 3| 148.0| 206.0| 3.292471|
|B000SQTJJ0|A265B1IZE5RVG6| 2| 148.0|
                                           429.0 3.5757942
                                  |B000SQTJJO| A2W34ZSDB0PC6|
                                  148.0| 1725.0| 4.1937737|
148.0| 695.0| 3.5518086|
|B000SQTJJO|A3DYBTW1TEZL3M|
                                148.0
|B000SQTJJO|A2HELIKP5RV27F|
                            4| 148.0| 8079.0| 3.588092|
|B000SQTJJ0|A3J2YU2D9BH2J7|
                            5| 148.0| 14912.0| 4.0|
| B000SOTJJO| A1YCWZWOXLUAYS| | | | |
|B000SQTJJO| AQFFA5JFDLQRS| 1| 148.0| 13656.0| 2.9760876|
|B000SQTJJO|A2M0RGVSV6YCMZ| 4| 148.0| 4599.0| 3.601359|
|B000SQTJJO|A2065HBMYDXJ15| 3| 148.0| 29.0| 3.5952232|
|B000SQTJJ0|A17437N1L775IJ| 4| 148.0| 1465.0| 3.9412987|
|B000SQTJJO|A1RIU1AAU4ZPEC| 1|
                                   148.0 | 10299.0 | 3.0508318 |
|B000SQTJJO| AY1I85LLDMETC| 4|
                                  148.0
                                           1879.0 3.6905727
|B000SQTJJO| AZECTOVTVA5Z4|
                                   148.0 | 13974.0 | 3.5253327 |
|B000SQTJJ0|A2IJ54FX1L83WK|
                                   148.0
                                            208.0 3.9424765
                                   463.0 10161.0 3.5867653
|B00000GBQJ|A1N7BFJSBP75A8|
|B00000GBQJ| ARF6NZ2PH6MCB|
                                   463.0 17323.0 4.127616
only showing top 20 rows
```

• Reference:

Collaborative Filtering - Spark 2.2.0 Documentation (apache.org)

Building a Recommendation System with Spark ML and Elasticsearch | by Lijo

Abraham | Towards Data Science

https://miro.medium.com/max/828/1*D34HqTvyzuvrCerHWCZSHQ.webp