

%sql

Table

SELECT * FROM movie_recommendation.default.movies_csv LIMIT 5;

▶ ■ _sqldf: pyspark.sql.dataframe.DataFrame = [movield: integer, title: string ... 1 more field]

	1 ² 3 movield	A ^B C title	A ^B C genres		
1	1	Toy Story (1995)	Adventure Animation Children Comedy Fantasy		
2	2	Jumanji (1995)	Adventure Children Fantasy		
3	3	Grumpier Old Men (1995)	Comedy Romance		
4	4	Waiting to Exhale (1995)	Comedy Drama Romance		
5	5	Father of the Bride Part II (1995)	Comedy		

5 rows

1 This result is stored as _sqldf and can be used in other Python and SQL cells.

2

%sql

SELECT * FROM movie_recommendation.default.ratings_csv LIMIT 5;

▶ ■ _sqldf: pyspark.sql.dataframe.DataFrame = [userld: integer, movield: integer ... 2 more fields]

Table

Q7 I D

Q7ID

	1 ² 3 userId	1 ² 3 movield	1.2 rating	1 ² 3 timestamp
1	1	296	5	1147880044
2	1	306	3.5	1147868817
3	1	307	5	1147868828
4	1	665	5	1147878820
5	1	899	3.5	1147868510

5 rows

1 This result is stored as _sqldf and can be used in other Python and SQL cells.

3

df_movies = spark.read.option("header", True).option("inferSchema", True).csv("/FileStore/tables/movies.csv")

df_movies.show()

▶ ■ df_movies: pyspark.sql.dataframe.DataFrame = [movield: integer, title: string ... 1 more field]

3|Grumpier Old Men ...| Comedy|Romance| 4|Waiting to Exhale...|Comedy|Drama|Romance| 5|Father of the Bri...| Comedy Heat (1995) | Action | Crime | Thri... | 7| Sabrina (1995)| Comedy|Romance| 8| Tom and Huck (1995)| Adventure|Children| 9| Sudden Death (1995)| Action| 10| GoldenEye (1995)|Action|Adventure|...| 11|American Presiden...|Comedy|Drama|Romance| 12|Dracula: Dead and...| Comedy|Horror| 13| Balto (1995)|Adventure|Animati...| 14| Nixon (1995)| Drama| 15|Cutthroat Island ...|Action|Adventure|...| Casino (1995)| Crime|Drama| 17|Sense and Sensibi...| Drama|Romance| 18| Four Rooms (1995)| Comedy | 19|Ace Ventura: When...| Comedy | 20| Money Train (1995)|Action|Comedy|Cri...|

only showing top 20 rows

```
df_ratings = spark.read.option("header", True).option("inferSchema", True).csv("/FileStore/tables/ratings.csv")
    df_ratings.show()
▶ ■ df_ratings: pyspark.sql.dataframe.DataFrame = [userld: integer, movield: integer ... 2 more fields]
           307| 5.0|1147868828|
665| 5.0|1147878820|
899| 3.5|1147868510|
      11
      1|
      1|
      1 | 1088 | 4.0 | 1147868495 |
      1| 1175| 3.5|1147868826|
     1| 1260| 3.5|1147877857|
      1| 1653| 4.0|1147868097|
      1| 2011| 2.5|1147868079|
     1 | 2012 | 2.5 | 1147868068 | 1 | 2068 | 2.5 | 1147869044 |
      1| 2161| 3.5|1147868609|
     1| 2351| 4.5|1147877957|
     1 | 2573 | 4.0 | 1147878923 |
1 | 2632 | 5.0 | 1147878248 |
1 | 2692 | 5.0 | 1147869100 |
only showing top 20 rows
    df_ratings.select("userId", "movieId",
    "rating").write.format("delta").mode("overwrite").save("/dbfs/delta/ratings_delta")
```

```
from pyspark.ml.recommendation import ALS

ratings_df = spark.read.format("delta").load("/dbfs/delta/ratings_delta")

als = ALS(
    userCol="userId",
    itemCol="movieId",
    ratingCol="rating",
    nonnegative=True,
    coldStartStrategy="drop",
    implicitPrefs=False,
    rank=10,
    maxIter=10,
    regParam=0.1
)

model = als.fit(ratings_df)
```

► ■ ratings_df: pyspark.sql.dataframe.DataFrame = [userld: integer, movield: integer ... 1 more field]

```
model.save("/dbfs/models/movie_recommender")
```

```
user_recs = model.recommendForAllUsers(10)
user_recs.show(truncate=False)
```

▶ ■ user_recs: pyspark.sql.dataframe.DataFrame = [userld: integer, recommendations: array]

```
|[{203080, 0.222300}, {151989, 0.0580005}, {205882, 5.850087}, {194454, 5.7407120}, {183947, 5.501555}, {194534, 5.4094
304}, {205277, 5.4652166}, {194883, 5.416352}, {166812, 5.265595}, {184299, 5.213943}]
               [{194334, 6.9836917}, {192089, 6.4453025}, {194883, 6.196027}, {203882, 6.1777706}, {194332, 6.164963}, {155923, 6.063
3526}, {159467, 6.0409594}, {203086, 6.0151234}, {183947, 5.93982}, {194434, 5.936064}]
               \lfloor \{205277,\ 7.259401\},\ \{151989,\ 6.27239\},\ \{144202,\ 6.1628313\},\ \{135099,\ 5.992943\},\ \{169606,\ 5.9657702\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\ \{194268,\ 5.94386\},\
5}, {158747, 5.866911}, {192261, 5.8125367}, {194434, 5.791329}, {203086, 5.7466793}]
             | [{151989, 6.374255}, {194434, 6.1283813}, {203882, 5.891994}, {203086, 5.887548}, {183947, 5.794066}, {196787, 5.52803
176
4}, {157791, 5.5205073}, {157789, 5.5205073}, {192089, 5.5080566}, {205277, 5.4857564}]
|78
               [{151989, 6.6395993}, {194434, 6.616261}, {203086, 6.44422}, {203882, 6.409883}, {192089, 6.4095297}, {200930, 6.28702
93}, {145871, 6.277875}, {194334, 6.2171636}, {157791, 6.1864796}, {157789, 6.1864796}]
              |[{151989, 4.6866813}, {207640, 4.4423537}, {193257, 4.4281144}, {203086, 4.390354}, {203882, 4.3540764}, {194434, 4.24
60127}, {161048, 4.2433467}, {183947, 4.235991}, {200930, 4.221349}, {196787, 4.2117767}] |
               [{203086, 5.725191}, {194334, 5.6194167}, {193257, 5.610821}, {151989, 5.5646186}, {205453, 5.512639}, {184299, 5.4478
28, {151615, 5.431345}, {200930, 5.397193}, {176729, 5.364778}, {166812, 5.3621187}]
only showing top 20 rows
         from pyspark.sql.functions import explode
         recs_exploded = user_recs.select(
                  "userId",
                  explode("recommendations").alias("rec")
         )
  ▶ ■ recs_exploded: pyspark.sql.dataframe.DataFrame = [userld: integer, rec: struct]
```

```
recs_flat = recs_exploded.select(
    "userId",
    recs_exploded.rec.movieId.alias("movieId"),
    recs_exploded.rec.rating.alias("predicted_rating")
)

recs_flat: pyspark.sql.dataframe.DataFrame = [userId: integer, movieId: integer ... 1 more field]
```

```
recs_with_name = recs_flat.join(df_movies, on="movieId", how="left")
```

▶ ■ recs_with_name: pyspark.sql.dataframe.DataFrame = [movield: integer, userld: integer ... 3 more fields]

```
recs_with_name.select("userId", "movieId", "title", "predicted_rating").show(truncate=False)
|1
       |194334 |Les Luthiers: El Grosso Concerto (2001)
                                                                    |5.341385
       |183947 |NOFX Backstage Passport 2
                                                                   15.3345475
11
|1
       |203086 |Truth and Justice (2019)
                                                                   |5.3135343
|1
       |203882 |Dead in the Water (2006)
                                                                    15.2937703
11
       |192089 |National Theatre Live: One Man, Two Guvnors (2011)|5.199268
|1
       |200930 |C'est quoi la vie? (1999)
                                                                    |5.1535807
11
       |155923 |Sing (1989)
                                                                   15.1285763
       |157791 |.hack Liminality In the Case of Kyoko Tohno
|1
                                                                   15.107959
                                                                    |6.462091
16
       |151989 |The Thorn (1971)
       |194434 |Adrenaline (1990)
16
                                                                    16.415434
|6
       |202231 |Foster (2018)
                                                                    |5.8129597
       |183947 |NOFX Backstage Passport 2
16
                                                                   15.8091736
|6
       |203882 |Dead in the Water (2006)
                                                                   15.6967783
                                                                    |5.643959
16
       |203086 |Truth and Justice (2019)
|6
       |201821 |Civilisation (1969)
                                                                   |5.6336117
|6
       |157791 |.hack Liminality In the Case of Kyoko Tohno
                                                                    |5.5766034
       |157789 |.hack Liminality In the Case of Yuki Aihara
16
                                                                   15.5766034
       |188569 |Elizabeth at 90: A Family Tribute (2016)
16
                                                                   |5.5162764
only showing top 20 rows
```

```
recs_with_name.filter("userId == 1").show(truncate=False)
```

movieId user	Id predicted_rat	ing title	genres
 151989 1	+ 5.5198274		+ Comedv
194434 1	5.3604155	Adrenaline (1990)	(no genres listed)
194334 1	5.341385	Les Luthiers: El Grosso Concerto (2001)	(no genres listed)
183947 1	5.3345475	NOFX Backstage Passport 2	(no genres listed)
203086 1	5.3135343	Truth and Justice (2019)	Drama
203882 1	5.2937703	Dead in the Water (2006)	Horror
192089 1	5.199268	National Theatre Live: One Man, Two Guvnors (20	11) Comedy
200930 1	5.1535807	C'est quoi la vie? (1999)	Drama
155923 1	5.1285763	Sing (1989)	(no genres listed)
157791 1	5.107959	.hack Liminality In the Case of Kyoko Tohno	(no genres listed)

```
watched_list_u1 = df_ratings.select("userId","movieId","rating").filter("userId == 1")
```

▶ ■ watched_list_u1: pyspark.sql.dataframe.DataFrame = [userld: integer, movield: integer ... 1 more field]

```
new_recommendations = recs_with_name.filter("userId == 1").join(watched_list_u1.select("movieId"), on="movieId",
how="left_anti")
```

new_recommendations.show(truncate=False)

▶ 🔳 new_recommendations: pyspark.sql.dataframe.DataFrame = [movield: integer, userld: integer ... 3 more fields]

movieId use	rId predicted_ra	ting title	genres
151989 1	5.5198274	The Thorn (1971)	Comedy
194434 1	5.3604155	Adrenaline (1990)	(no genres listed)
194334 1	5.341385	Les Luthiers: El Grosso Concerto (2001)	(no genres listed)
183947 1	5.3345475	NOFX Backstage Passport 2	(no genres listed)
203086 1	5.3135343	Truth and Justice (2019)	Drama
203882 1	5.2937703	Dead in the Water (2006)	Horror
192089 1	5.199268	National Theatre Live: One Man, Two Guvnors (20	11) Comedy
200930 1	5.1535807	C'est quoi la vie? (1999)	Drama
155923 1	5.1285763	Sing (1989)	(no genres listed)
157791 1	5.107959	.hack Liminality In the Case of Kyoko Tohno	(no genres listed)

recs_with_name.write.format("delta").mode("overwrite").save("/delta/user_recommendations")

```
▶ ■ df: pyspark.sql.dataframe.DataFrame = [movield: integer, userld: integer ... 3 more fields]
```

```
[Row(movieId=151989, userId=1, predicted_rating=5.519827365875244, title='The Thorn (1971)', genres='Comedy'),
Row(movieId=194434, userId=1, predicted_rating=5.360415458679199, title='Adrenaline (1990)', genres='(no genres listed)'),
Row(movieId=194334, userId=1, predicted_rating=5.3413848876953125, title='Les Luthiers: El Grosso Concerto (2001)', genres
='(no genres listed)'),
Row(movieId=183947, userId=1, predicted_rating=5.334547519683838, title='NOFX Backstage Passport 2', genres='(no genres liste
d)'),
Row(movieId=203086, userId=1, predicted_rating=5.313534259796143, title='Truth and Justice (2019)', genres='Drama'),
Row(movieId=203882, userId=1, predicted_rating=5.2937703132629395, title='Dead in the Water (2006)', genres='Horror'),
Row(movieId=192089, userId=1, predicted_rating=5.199267864227295, title='National Theatre Live: One Man, Two Guvnors (2011)',
genres='Comedy'),
Row(movieId=200930, userId=1, predicted_rating=5.153580665588379, title='C'est quoi la vie? (1999)", genres='Drama'),
Row(movieId=155923, userId=1, predicted_rating=5.128576278686523, title='Sing (1989)', genres='(no genres listed)'),
Row(movieId=157791, userId=1, predicted_rating=5.107958793640137, title='.hack Liminality In the Case of Kyoko Tohno', genres='(no genres listed)')]
```