```
!pip install pyspark
    Collecting pyspark
     Downloading pyspark-3.5.0.tar.gz (316.9 MB)
                                           - 316.9/316.9 MB 2.7 MB/s eta 0:00:00
      Preparing metadata (setup.py) ... done
    Requirement already satisfied: py4j==0.10.9.7 in /usr/local/lib/python3.10/dist-packages (from pyspark) (0.10.9.7)
    Building wheels for collected packages: pyspark
      Building wheel for pyspark (setup.py) ... done
      Stored in directory: /root/.cache/pip/wheels/41/4e/10/c2cf2467f71c678cfc8a6b9ac9241e5e44a01940da8fbb17fc
    Successfully built pyspark
    Installing collected packages: pyspark
    Successfully installed pyspark-3.5.0
from pyspark.sql import SparkSession
spark = SparkSession.builder.master("local").appName("als").config('spark.ui.port','4050').getOrCreate()
from google.colab import drive
drive.mount('/content/drive')
    Mounted at /content/drive
```

1. Carga de datos

```
from zipfile import ZipFile
import os
# Ruta del archivo zip
games_zip = "/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/games.csv.zip"
# Ruta de destino para la extracción
destino_extraccion = "/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/Descomprimido"
# Descomprimir el archivo zip
with ZipFile(games_zip, 'r') as zip_ref:
       zip_ref.extractall(destino_extraccion)
# Ruta del archivo zip
games_metadata_zip = "/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/games_metadata.json.zip"
# Ruta de destino para la extracción
destino_extraccion2 = "/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/Descomprimido"
# Descomprimir el archivo zip
with ZipFile(games_metadata_zip, 'r') as zip_ref:
       zip ref.extractall(destino extraccion2)
# Ruta del archivo zip
recommendations\_zip = "/content/drive/MyDrive/Datos \ Masivos/Proyectos/Proyecto\_final/Datos/recommendations.csv.zip" = "/content/drive/MyDrive/Datos \ Masivos/Proyectos/Proyecto_final/Datos/recommendations.csv.zip" = "/content/drive/MyDrive/Datos \ Masivos/Proyectos/Proyectos_final/Datos/recommendations.csv.zip" = "/content/drive/MyDrive/Datos \ Masivos/Proyectos_final/Datos/recommendations.csv.zip" = "/content/drive/MyDrive/Datos \ Masivos_final/Datos/Recommendations.csv.zip" = "/content/drive/MyDrive/MyDrive/Datos \ Masivos_final/Datos/Recommendations.csv.zip" = "/content/drive/MyDrive/MyDrive/Datos \ Masivos_final/Datos/Recommendations.csv.zip" = "/content/drive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDrive/MyDriv
# Ruta de destino para la extracción
destino_extraccion3 = "/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/Descomprimido"
# Descomprimir el archivo zip
with ZipFile(recommendations_zip, 'r') as zip_ref:
        zip_ref.extractall(destino_extraccion3)
# Ruta del archivo zip
users_zip = "/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/users.csv.zip"
# Ruta de destino para la extracción
destino_extraccion4 = "/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/Descomprimido"
# Descomprimir el archivo zip
with ZipFile(users zip, 'r') as zip ref:
        zip_ref.extractall(destino_extraccion4)
juegos = spark.read.csv("/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/Descomprimido/games.csv",header=True)
recomendaciones = spark.read.csv("/content/drive/MyDrive/Datos \ Masivos/Proyectos/Proyecto_final/Datos/Descomprimido/recommendations.csv') \\
usuarios = spark.read.csv("/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/Datos/Descomprimido/users.csv",header=True)
```

juegos.show()

app_id title	date_release win	mac linux		rating	positive_ratio	user_reviews	price_final	price_original d
13500 Prince of Persia:	2008-11-21 true	false false	Very	Positive	84	2199	9.99	9.99
22364 BRINK: Agents of	2011-08-03 true	false false		Positive	85	21	2.99	2.99
113020 Monaco: What's Yo	2013-04-24 true	true true	Very	Positive	92	3722	14.99	14.99
226560 Escape Dead Island	2014-11-18 true	false false		Mixed	61	873	14.99	14.99
249050 Dungeon of the EN	2014-10-27 true	true false	Very	Positive	88	8784	11.99	11.99
250180 METAL SLUG 3	2015-09-14 true	false false	Very	Positive	90	5579	7.99	7.99
253980 Enclave	2013-10-04 true	true true	Mostly	Positive	75	1608	4.99	4.99
271850 Men of War: Assau	2014-05-16 true	false false		Mixed	61	199	6.99	6.99
282900 Hyperdimension Ne	2015-01-29 true	false false	Very	Positive	94	9686	14.99	14.99
19810 The Sum of All Fears	2008-10-10 true	false false	Mostly	Positive	75	33	9.99	9.99
15270 Cold Fear™	2008-05-13 true	false false	Very	Positive	85	800	9.99	9.99
21130 LEGO® Harry Potte	2010-06-25 true	false false	Very	Positive	85	5169	19.99	19.99
22130 Hearts of Iron 2	2009-01-23 true	false false	Very	Positive	85	462	14.99	14.99
29180 Osmos	2009-08-18 true	true true	Very	Positive	88	532	9.99	9.99
32750 Comanche 4	2009-06-18 true	false false	Very	Positive	90	222	9.99	9.99
241620 Inquisitor	2013-08-01 true	false false	Mostly	Positive	70	390	9.99	9.99
408520 FORM	2017-06-01 true	false false	Very	Positive	91	934	14.99	14.99
244910 Homesick	2015-05-28 true	false false	Mostly	Positive	77	1139	14.99	14.99
245950 Borderlands 2: He	2014-02-11 true	true true	Very	Positive	84	294	0.89	2.99
250460 Bridge Constructor	2013-10-16 true	true true	Mostly	Positive	77	716	2.39	19.99

only showing top 20 rows

recomendaciones.show()

+ app_id	+ helpful	+ funny	date	+ is_recommended	hours	 user_id	review_id
975370	0	+ 0	 2022-12-12	true	 36.3	 51580	+ 0
304390	4	0	2017-02-17	false	11.5	2586	1
1085660	2	0	2019-11-17	true	336.5	253880	2
703080	0	0	2022-09-23	true	27.4	259432	3
526870	0	0	2021-01-10	true	7.9	23869	4
306130	0	0	2021-10-10	true	8.6	45425	5
238960	0	0	2017-11-25	true	538.8	88282	6
730	0	0	2021-11-30	false	157.5	63209	7
255710	0	0	2021-05-21	true	18.7	354512	8
289070	0	0	2020-05-26	true	397.5	454422	9
431960	0	0	2020-10-14	true	30.3	199725	10
1086940	0	0	2020-10-07	true	50.0	85822	11
1938090	0	0	2022-11-16	true	46.7	161081	12
1286830	2	0	2020-07-26	true	19.3	113279	13
1172620	0	0	2020-11-04	true	89.1	122640	14
306130	0	0	2021-05-12	true	61.1	75422	15
635260	. 0	0	2022-01-30	true	177.0	76583	16
1151340		0	2020-07-01	true	86.3	124924	17
289070	0	0	2020-05-29	true	244.1	261528	18
392160	3	0	2018-12-26	false	320.5	408750	19

only showing top 20 rows

usuarios.show()

+		+
user_id p	roducts re	views
+		+
7360263	359	0
14020781	156	1
8762579	329	4
4820647	176	4
5167327	98	2
5664667	145	5
5889167	447	2
7281762	1083	1
7445952	273	1
7462927	51	1
7922733	108	2
9201535	166	1
9514331	237	4
9972262	250	3
10184828	187	1
10714376	1824	2
11237958	12	1
11274058	102	4
11300174	210	17
11463309	318	5
+		+

only showing top 20 rows

2. Limpieza de datos

Cambio a entero al dataframe juegos

```
juegos.printSchema()
      |-- app_id: string (nullable = true)
      |-- title: string (nullable = true)
      |-- date_release: string (nullable = true)
       -- win: string (nullable = true)
      -- mac: string (nullable = true)
       -- linux: string (nullable = true)
      |-- rating: string (nullable = true)
      -- positive_ratio: string (nullable = true)
      |-- user_reviews: string (nullable = true)
       -- price_final: string (nullable = true)
      |-- price_original: string (nullable = true)
       -- discount: string (nullable = true)
      |-- steam_deck: string (nullable = true)
from pyspark.sql.functions import col,explode
juegos = juegos.\
         withColumn('app_id',col('app_id').cast('integer')).\
         withColumn('positive_ratio',col('positive_ratio').cast('integer')).\
         withColumn('user_reviews',col('user_reviews').cast('integer'))
juegos.printSchema()
     root
      |-- app_id: integer (nullable = true)
      -- title: string (nullable = true)
      -- date_release: string (nullable = true)
      |-- win: string (nullable = true)
      |-- mac: string (nullable = true)
       -- linux: string (nullable = true)
       -- rating: string (nullable = true)
       -- positive_ratio: integer (nullable = true)
      -- user_reviews: integer (nullable = true)
       -- price_final: string (nullable = true)
      |-- price_original: string (nullable = true)
       -- discount: string (nullable = true)
      |-- steam_deck: string (nullable = true)
```

Cambio a entero al dataframe recomendaciones

```
recomendaciones.printSchema()
      |-- app_id: string (nullable = true)
      |-- helpful: string (nullable = true)
       -- funny: string (nullable = true)
      -- date: string (nullable = true)
      |-- is_recommended: string (nullable = true)
       -- hours: string (nullable = true)
      |-- user id: string (nullable = true)
      |-- review_id: string (nullable = true)
from pyspark.sql.functions import col,explode
recomendaciones = recomendaciones.\
          withColumn('app_id',col('app_id').cast('integer')).\
          withColumn('helpful',col('helpful').cast('integer')).\
          withColumn('funny',col('funny').cast('integer')).
          withColumn('user id',col('user id').cast('integer')).\
          withColumn('review_id',col('review_id').cast('integer')).\
          withColumn('hours',col('hours').cast('integer'))
recomendaciones.printSchema()
      |-- app_id: integer (nullable = true)
       -- helpful: integer (nullable = true)
      |-- funny: integer (nullable = true)
       -- date: string (nullable = true)
      |-- is_recommended: string (nullable = true)
      |-- hours: integer (nullable = true)
|-- user_id: integer (nullable = true)
```

```
|-- review_id: integer (nullable = true)
```

Cambio a entero al dataframe usuarios

Eliminar filas con valores nulos

```
# Elimina filas con valores nulos
juegos_cleaned = juegos.na.drop()
# Muestra los primeros registros del DataFrame limpio
juegos_cleaned.show()
```

app_id title	date_release	win	mac	linux		rating	positive_ratio	user_reviews	price_final	price_original
13500 Prince of Persia:	+ 2008-11-21	true	false	+ false	 Verv	Positive	84	2199	+ 9.99	+ 9.99
22364 BRINK: Agents of						Positive			1	
113020 Monaco: What's Yo						Positive	92	3722	14.99	14.99
226560 Escape Dead Island	2014-11-18	true	false	false	ĺ	Mixed	61	873	14.99	14.99
249050 Dungeon of the EN	2014-10-27	true	true	false	Very	Positive	88	8784	11.99	11.99
250180 METAL SLUG 3	2015-09-14	true	false	false	Very	Positive	90	5579	7.99	7.99
253980 Enclave	2013-10-04	true	true	true	Mostly	Positive	75	1608	4.99	4.99
271850 Men of War: Assau	2014-05-16	true	false	false		Mixed	61	199	6.99	6.99
282900 Hyperdimension Ne	2015-01-29	true	false	false	Very	Positive	94	9686	14.99	14.99
19810 The Sum of All Fears	2008-10-10	true	false	false	Mostly	Positive	75	33	9.99	9.99
15270 Cold Fear™	2008-05-13	true	false	false	Very	Positive	85	800	9.99	9.99
21130 LEGO® Harry Potte	2010-06-25	true	false	false	Very	Positive	85	5169	19.99	19.99
22130 Hearts of Iron 2	2009-01-23	true	false	false	Very	Positive	85	462	14.99	14.99
29180 Osmos	2009-08-18	true	true	true	Very	Positive	88	532	9.99	9.99
32750 Comanche 4	2009-06-18	true	false	false	Very	Positive	90	222	9.99	9.99
241620 Inquisitor	2013-08-01	true	false	false	Mostly	Positive	70	390	9.99	9.99
408520 FORM	2017-06-01	true	false	false	Very	Positive	91	934	14.99	14.99
244910 Homesick	2015-05-28	true	false	false	Mostly	Positive		1139	14.99	14.99
245950 Borderlands 2: He					,	Positive			1	'
250460 Bridge Constructor	2013-10-16	true	true	true	Mostly	Positive	77	716	2.39	19.99

only showing top 20 rows

```
# Elimina filas con valores nulos
recomendaciones_cleaned = recomendaciones.na.drop()
```

Muestra los primeros registros del DataFrame limpio recomendaciones_cleaned.show()

+	+	+	+	+-	+	+	+
app_id he	lpful fu	nny	date is_r	ecommended h	ours	user_id	review_id
+		+			+	+	+
975370	0	0 2022-	12-12	true	36	51580	0
304390	4	0 2017 -	02-17	false	11	2586	1
1085660	2	0 2019-	11-17	true	336	253880	2
703080	0	0 2022-	09-23	true	27	259432	3
526870	0	0 2021-	01-10	true	7	23869	4
306130	0	0 2021-	10-10	true	8	45425	5
238960	0	0 2017-	11-25	true	538	88282	6
730	0	0 2021-	11-30	false	157	63209	7
255710	0	0 2021-	05-21	true	18	354512	8

```
289070
                   0 | 2020 - 05 - 26 |
                                           true| 397| 454422|
              0
                                                   30 | 199725 |
50 | 85822 |
431960
                    0 2020 - 10 - 14
                                                                        10
              0
                                            truel
10869401
              01
                    0 | 2020 - 10 - 07 |
                                            truel
                                                                       11
                   0 | 2022-11-16 |
                                           true
19380901
               0
                                                   46 | 161081 |
19 | 113279 |
                                                                        12
1286830
              2
                    0 | 2020 - 07 - 26 |
                                            true
                                                                        13
                                           true| 89| 122640|
|1172620|
              0 | 0 | 2020 - 11 - 04 |
                                                                        14
 306130
              0
                    0 2021-05-12
                                                    61
                                                         75422
                                                                        15
                                            true
                                           true | 177 | 76583 |
 635260
              0 | 0 | 2022 - 01 - 30 |
                                                                        16
                                           true
1151340
              0
                     0 | 2020 - 07 - 01 |
                                                    86 124924
                                                                        17
                                             true 244 261528
289070
                     0 | 2020 - 05 - 29 |
                                                                        18
392160
              3 |
                   0 | 2018-12-26 |
                                           falsel 320 408750
                                                                       19
+-----
```

only showing top 20 rows

usuarios_cleaned.show()

```
# Elimina filas con valores nulos
usuarios_cleaned = usuarios.na.drop()
# Muestra los primeros registros del DataFrame limpio
```

only showing top 20 rows

3. Analisis Exploratorio

plt.show()

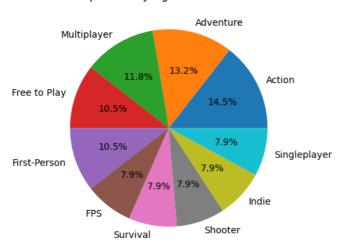
```
usuario_pandas = usuarios_reduce.toPandas()
entrenamiento = juegos_cleaned.toPandas()
rating = recomendaciones_reduce.toPandas()
```

Recopilar los datos de juegos y sus etiquetas correspondientes, identificar los juegos con la mayor cantidad de reseñas y organizar en orden descendente según su ratio positivo.

```
\verb|metadata| = spark.read.json("/content/drive/MyDrive/Datos Masivos/Proyectos/Proyectos_final/Datos/Descomprimido/games_metadata.json")|
from pyspark.sql.functions import col
# Realizar la unión de los dos DataFrames en PySpark
gamesdf = juegos_cleaned.join(metadata, on='app_id')
# Estas son las columnas que te interesan
relevant_cols = ['app_id', 'title', 'positive_ratio', 'user_reviews', 'tags']
gamesdf = gamesdf.select(*relevant_cols)
# Ordenar por user_reviews para mantener los mejores 100 juegos revisados
gamesdf = gamesdf.orderBy('user_reviews', ascending=False)
gamesdf = gamesdf.limit(100)
# Calcular el promedio de user_reviews
review_avg = gamesdf.agg({'user_reviews': 'mean'}).collect()[0][0]
# Mostrar el DataFrame resultante
gamesdf.show()
                          title|positive_ratio|user_reviews|
                                                                             tags
         730|Counter-Strike: G...|
                                       88 | 7494460 |
                                                                               [1]
                                           57 | 2217226 |
82 | 2045628 |
      578080 | PUBG: BATTLEGROUNDS |
                          Dota 2
                                                                               []
       271590 Grand Theft Auto V
                                                    1484122
                                            86 |
86 |
      359550 Tom Clancy's Rain...
                                                     993312
         440 | Team Fortress 2
                                            93 |
97 |
                                                      985819
                                                                               []
      105600
                 Terraria
Garry's Mod
                                                     943413
                                            96
                                                     853733
        4000
     | 252490| Rust|
|1172470| Apex Legends™|
                                                      786668
                                            80
                                                     713182
                                                                               []
      292030 | The Witcher® 3: W...
                                                      668455
                                                                                []
      431960| Wallpaper Engine
                                                     637341
                                                                               []
      945360
                        Among Us
                                             92
                                                      587821
                  Left 4 Dead 2
                                            97
         550
                                                     574470
      1085660
                     Destiny 2
                                                      562723
                                             81
                 Cyberpunk 2077
     1091500
                                             80
                                                     557051
                                                     542198
      230410
                         Warframe
                                             86
                   Wartrame|
|ELDEN RING
     112456201
                                             92
                                                     528702
                                                  515016|[Free to Play, Su...
     3049301
                        Unturned
                                             91
                Stardew Valley
                                                      505882
      413150
     only showing top 20 rows
from pyspark.sql.functions import explode
# Obtener la columna de tags y explotarla para obtener una fila por cada tag
tags_df = gamesdf.select('tags').withColumn('tag', explode('tags'))
# Contar la frecuencia de cada tag
tag_counts = tags_df.groupBy('tag').count()
# Ordenar los resultados en orden descendente
tag_counts = tag_counts.orderBy('count', ascending=False)
# Mostrar información sobre la serie
print("Series Size: ", tag_counts.count())
print("Average Common Tags: ", tag_counts.agg({'count': 'mean'}).collect()[0][0])
# Crear un gráfico para visualizar los resultados
tag slice = tag counts.limit(10).toPandas()
labels = tag_slice['tag'].tolist()
fig, ax = plt.subplots()
plt.title('Etiquetas de Juegos Más Revisadas')
pchart = ax.pie(tag_slice['count'], labels=labels, autopct='%1.1f%%')
```

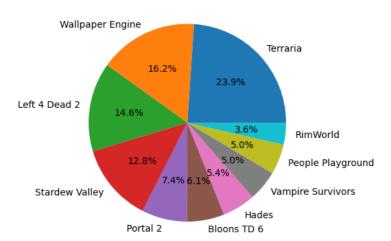
Series Size: 110 Average Common Tags: 2.1818181818181817

Etiquetas de Juegos Más Revisadas

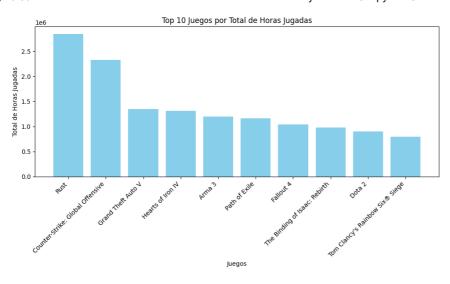


```
from pyspark.sql.functions import desc
# Ordenar el DataFrame por 'positive_ratio' en orden descendente
gamesdf = gamesdf.orderBy(desc('positive_ratio'))
# Seleccionar los 10 juegos más positivamente revisados
positivedf = gamesdf.limit(10)
# Contar la frecuencia de cada título
title_counts = positivedf.groupBy('title').agg({'user_reviews': 'sum'})
# Ordenar los resultados en orden descendente
title_counts = title_counts.orderBy(desc('sum(user_reviews)'))
# Mostrar información sobre la serie
print("Series Size: ", title_counts.count())
print("Average Common Titles: ", title_counts.agg({'sum(user_reviews)': 'mean'}).collect()[0][0])
# Crear un gráfico para visualizar los resultados
title slice = title counts.limit(10).toPandas()
labels = title_slice['title'].tolist()
fig, ax = plt.subplots()
plt.title('Títulos de Juegos Más Positivamente Revisados')
pchart = ax.pie(title_slice['sum(user_reviews)'], labels=labels, autopct='%1.1f%%')
plt.show()
     Series Size: 10
     Average Common Titles: 394130.5
```

Títulos de Juegos Más Positivamente Revisados



```
from pyspark.sql.functions import sum
# Realizar la unión de los DataFrames usando la columna "app id"
merged_df = juegos_cleaned.join(recomendaciones_reduce, on='app_id')
# Sumar las horas jugadas por cada app_id
sum_hours_df = merged_df.groupBy("app_id", "title").agg(sum("hours").alias("total_hours"))
print(" Dimensiones de los datos", "\n",
      "Fila recomendaciones ", merged_df.count(), " - ", "Columnas recomendaciones", len(merged_df.columns))
      Dimensiones de los datos
      Fila recomendaciones 200000 - Columnas recomendaciones 20
sum_hours_df.show()
     | app_id|
                          title|total hours|
      -----
      975370 Dwarf Fortress
                                       39529
      304390 | FOR HONOR™ |
1085660 | Destiny 2 |
703080 | Planet Zoo |
526870 | Satisfactory
                                       375559
      1085660
                                      383769
                                       99713
                                      400595
       306130 | The Elder Scrolls...
                                       583084
      238960 Path of Exile
                                    1160204
         730 Counter-Strike: G...
                                      2325331
      255710 | Cities: Skylines
                                       271443
      289070 Sid Meier's Civil...
                                       605952
      | 431960| Wallpaper Engine|
|1086940| Baldur's Gate 3|
                                       475610
                                       109263
     19380901
                   Call of Duty®
                                       480229
      |1286830|STAR WARS™: The O...|
                                       197683
      |1172620|Sea of Thieves 20...|
                                       464919
      635260 CarX Drift Racing...
                                       45163
      1151340|
                     Fallout 76
                                       235533
                  X4: Foundations
      392160
                                        71815
                        Dota 2
                                       897305
     | 534380|Dying Light 2 Sta...|
                                      118955
     only showing top 20 rows
# Obtener los datos del DataFrame como listas de Python
top_titles = sum_hours_df.orderBy("total_hours", ascending=False).limit(10).select("title").rdd.flatMap(lambda x: x).collect()
top_total_hours = sum_hours_df.orderBy("total_hours", ascending=False).limit(10).select("total_hours").rdd.flatMap(lambda x: x).collect
# Crear el gráfico de barras
plt.figure(figsize=(10, 6))
plt.bar(top_titles, top_total_hours, color='skyblue')
plt.xlabel('Juegos')
plt.ylabel('Total de Horas Jugadas')
plt.title('Top 10 Juegos por Total de Horas Jugadas')
plt.xticks(rotation=45, ha='right') # Rotar las etiquetas del eje x para mayor legibilidad
plt.tight_layout()
# Mostrar el gráfico
plt.show()
```



4. Procesamiento

```
from pyspark.ml.evaluation import RegressionEvaluator
from pyspark.ml.recommendation import ALS
from pyspark.ml.tuning import ParamGridBuilder, CrossValidator
```

recomendaciones_reduce.printSchema()

```
oot
|-- app_id: integer (nullable = true)
|-- helpful: integer (nullable = true)
|-- funny: integer (nullable = true)
|-- date: string (nullable = true)
|-- is_recommended: string (nullable = true)
|-- hours: integer (nullable = true)
|-- user_id: integer (nullable = true)
|-- review_id: integer (nullable = true)
```

ratings_union = recomendaciones_reduce.join(juegos_cleaned, on='app_id')
Mostrar algunas filas del nuevo DataFrame
ratings_union.show()

app_id	helpful	funny	da	te	is_recommended	hours	user_id	review_id	title	date_release	win	mac	linux	
975370	0	0	2022-12-	L2	true	36	51580	0	Dwarf Fortress	2022-12-06	true	false	false	Overwhe]
304390	4	0	2017-02-3	L7	false	11	2586	1	FOR HONOR™	2017-02-13	true	false	false	
1085660	2	0	2019-11-3	L7	true	336	253880	2	Destiny 2	2019-10-01	true	false	false	\
703080	0	0	2022-09-2	23	true	27	259432	3	Planet Zoo	2019-11-05	true	false	false	\
526870	0	0	2021-01-3	10	true	7	23869	4	Satisfactory		true	false	false	Overwhe]
306130	0	0	2021-10-3	10	true	8	45425	5	The Elder Scrolls	2017-05-22	true	true	false	\
238960	0	0	2017-11-2	25	true	538	88282	6	Path of Exile	2013-10-23	true	true	false	\
730	0	0	2021-11-3	30	false	157	63209	7	Counter-Strike: G	2012-08-21	true	true	true	\
255710	0	0	2021-05-2	21	true	18	354512	8	Cities: Skylines	2015-03-10	true	true	true	\
289070	0	0	2020-05-2	26	true	397	454422	9	Sid Meier's Civil	2016-10-20	true	true	true	\
431960	0	0	2020-10-3	L4	true	30	199725	10	Wallpaper Engine	2018-11-01	true	false	false	Overwhe]
1086940	0	0	2020-10-0	97	true	50	85822	11	Baldur's Gate 3	2023-08-03	true	true	false	Overwhe]
1938090	0	0	2022-11-2	16	true	46	161081	12	Call of Duty®	2022-10-27	true	false	false	
1286830	2	0	2020-07-2	26	true	19	113279	13	STAR WARS™: The O	2020-07-21	true	false	false	\
1172620	0	0	2020-11-0	94	true	89	122640	14	Sea of Thieves 20	2020-06-03	true	false	false	\
306130	0	0	2021-05-3	12	true	61	75422	15	The Elder Scrolls	2017-05-22	true	true	false	\
635260	0	0	2022-01-3	30	true	177	76583	16	CarX Drift Racing	2017-11-17	true	false	false	Overwhe]
1151340	0	0	2020-07-0	11	true	86	124924	17	Fallout 76	2020-04-14	true	false	false	Mos
289070	0	0	2020-05-2	29	true	244	261528	18	Sid Meier's Civil	2016-10-20	true	true	true	\
392160	3	0	2018-12-2	26	false	320	408750	19	X4: Foundations	2018-11-30	true	false	true	Mos

only showing top 20 rows

```
num filas = ratings union.count()
num_columnas = len(ratings_union.columns)
print(f"Número de filas: {num filas}")
print(f"Número de columnas: {num_columnas}")
     Número de filas: 200000
     Número de columnas: 20
ratings_cleaned = ratings_union.select("user_id","app_id","title","hours","is_recommended")
# Mostrar el esquema del nuevo DataFrame
ratings_cleaned.printSchema()
# Mostrar algunas filas del nuevo DataFrame
ratings_cleaned.show()
       |-- user_id: integer (nullable = true)
        -- app id: integer (nullable = true)
        -- title: string (nullable = true)
       |-- hours: integer (nullable = true)
       |-- is_recommended: string (nullable = true)
      |user_id| app_id|
                                         title|hours|is recommended|
       | S1580 | 975370 | Dwarf Fortress | 36 | 2586 | 304390 | FOR HONOR** | 11 | 253880 | 1085660 | Destiny 2 | 336 | 259432 | 703080 | Planet Zoo | 27 | 23869 | 526870 | Satisfactory | 7 | 45405 | 306130 | The Elder Scrolls... | 8 |
                                                                   false
                                                                 true|
true|
                                                                    truel
         45425 | 306130 | The Elder Scrolls...
                                                                   true
        88282 | 238960 | Path of Exile | 538 |
                                                                    true
         63209
                   730 Counter-Strike: G... | 157
                                                                   false
        354512 | 255710 | Cities: Skylines | 18
                                                                   true
        454422 | 289070 | Sid Meier's Civil...
                                                    397
                                                                    true
       199725 | 431960 | Wallpaper Engine | 30 |
85822 | 1086940 | Baldur's Gate 3 | 50 |
161081 | 1938090 | Call of Duty® | 46 |
                                                                    true
        113279 | 1286830 | STAR WARS™: The O...
                                                     19
                                                                    true
                                                    89|
       122640 1172620 Sea of Thieves 20...
                                                                    true
         75422| 306130|The Elder Scrolls...|
                                                     61
                                                                    true
         76583 | 635260 | CarX Drift Racing... | 177 |
                                                                    true
       124924|1151340| Fallout 76| 86|
                                                                    truel
       261528 | 289070 | Sid Meier's Civil...
                                                    244
                                                                    true
      | 408750| 392160| X4: Foundations| 320|
                                                                   false
     only showing top 20 rows
```

Se ha agregado una nueva columna denominada "rating" al conjunto de datos, que evalúa la cantidad de horas dedicadas a un juego y si el usuario lo recomendó. Esta columna clasifica los juegos en una escala de 1 a 5, asignando calificaciones más altas a aquellos con mayores horas jugadas y recomendaciones positivas.

```
from pyspark.sql import SparkSession
from pyspark.sql.functions import when

# Definir las condiciones y asignar un rating en consecuencia
ratings_cleaned = ratings_cleaned.withColumn(
    "rating",
    when((ratings_cleaned.hours >= 0) & (ratings_cleaned.hours <= 499) & (ratings_cleaned.is_recommended == "true"), 4)
    .when((ratings_cleaned.hours >= 500) & (ratings_cleaned.hours <= 999) & (ratings_cleaned.is_recommended == "true"), 5)
    .when((ratings_cleaned.hours >= 0) & (ratings_cleaned.hours <= 299) & (ratings_cleaned.is_recommended == "false"), 1)
    .when((ratings_cleaned.hours >= 300) & (ratings_cleaned.hours <= 599) & (ratings_cleaned.is_recommended == "false"), 2)
    .when((ratings_cleaned.hours >= 600) & (ratings_cleaned.hours <= 999) & (ratings_cleaned.is_recommended == "false"), 3)
    .otherwise(0) # Manejar cualquier otro caso, si es necesario
)

# Muestra el DataFrame resultante con la nueva columna "rating"
ratings_cleaned.show()

# Muestra el DataFrame resultante con la nueva columna "rating"
ratings_cleaned.show()

# Sis80 975370 | Dwarf Fortress| 36| true| 4|</pre>
```

```
FOR HONOR™
   2586 | 304390 |
                                                      false
                                         11
 253880 | 1085660 |
                                                                 41
                           Destiny 2
                                        336
                                                       true
 259432 | 703080 |
                          Planet Zool
                                         27
                                                       true
                                                                 41
  23869 | 526870 |
                        Satisfactory
                                                       true
                                                                 41
  45425 | 306130 | The Elder Scrolls...
                                          8
                                                       true
                                                                 4
  88282 | 238960 |
                       Path of Exile
                                        538
                                                       true
            730 Counter-Strike: G...
  63209
                                        157
 354512 | 255710 | Cities: Skylines
                                         18
                                                      true
 454422 | 289070 | Sid Meier's Civil...
                                        397
                                                       true
 199725 | 431960 |
                 Wallpaper Engine
                                         30
                                                       true
  85822 | 1086940 |
                    Baldur's Gate 3
                                         50
                                                                 41
                                                       truel
                       Call of Duty®
                                                                 4
 161081 | 1938090 |
                                         46
                                                       truel
 113279 | 1286830 | STAR WARS™: The O... |
                                                                 41
                                         19
                                                       true
 122640 | 1172620 | Sea of Thieves 20... |
                                         89
                                                       true
  75422 | 306130 | The Elder Scrolls...
                                         61
                                                       true
                                                                 4
  76583 | 635260 | CarX Drift Racing...
                                        177
                                                       truel
                                                                 4
 124924 | 1151340 |
                          Fallout 76
                                                                 4
                                        86
 261528 | 289070 | Sid Meier's Civil...
                                         244
                                                       true
| 408750| 392160| X4: Foundations| 320|
                                                      false|
```

only showing top 20 rows

Dividir el DataFrame ratings_cleaned en tres conjuntos de datos: entrenamiento (train_data), prueba (test_data) y validación (validation).

```
train_data, test_data, validation = ratings_cleaned.randomSplit([0.7, 0.2, 0.1], seed=42)
#train data.write.option("header",True).csv('/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto final/ratings/rating train-csv')
#test_data.write.option("header",True).csv('/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/ratings/rating_test-csv')
#validation.write.option("header",True).csv('/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/ratings/rating_validation-csv
print("Tamaño de dato de entrenamiento: ", train_data.count())
print("Tamaño de dato de prueba: ", test_data.count())
print("Tamaño de dato de validacion: ", validation.count())
     Tamaño de dato de entrenamiento: 139838
     Tamaño de dato de prueba: 39960
     Tamaño de dato de validacion: 20202
als = ALS(userCol="user_id", itemCol="app_id", ratingCol="rating",
          nonnegative=True, implicitPrefs=False, coldStartStrategy="drop")
type(als)
     pyspark.ml.recommendation.ALS
```

Se definen combinaciones de parámetros (rank y regParam) para evaluar modelos ALS, usando métrica RMSE.

```
param_grid = ParamGridBuilder() \
              .addGrid(als.rank,[1]) \
              .addGrid(als.regParam,[0.3]) \
              .build()
evaluator = RegressionEvaluator(metricName='rmse', labelCol='rating', predictionCol='prediction')
print("Numero de modelos a evaluar", len(param_grid))
     Numero de modelos a evaluar 1
cv = CrossValidator(estimator = als. estimatorParamMaps=param grid, evaluator=evaluator, numFolds=5)
print(cv)
     CrossValidator_eda4f03fb204
```

5. Entrenamiento

```
model = cv.fit(train data)
best_model = model.bestModel
```

```
from pyspark.ml.tuning import CrossValidatorModel
#Guardamos el modelo para un uso en futuro
#model.write().overwrite().save('/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/modelo')
#model = CrossValidatorModel.load(path='/content/drive/MyDrive/Datos Masivos/Proyectos/Proyecto_final/modelo')
#best_model = model.bestModel
print("Mejor modelo 5")
print("Rank:", best_model._java_obj.parent().getRank())
print("MaxIter:", best_model._java_obj.parent().getMaxIter())
print("RegParam:", best_model._java_obj.parent().getRegParam())
test_predicciones = best_model.transform(test_data)
RMSE = evaluator.evaluate(test_predicciones)
print("RMSE: ",RMSE)
     Mejor modelo 5
     Rank: 1
     MaxIter: 10
     RegParam: 0.3
     RMSE: 1.4277035596730039
Mejor modelo 1: Rank: 7 MaxIter: 10 RegParam: 0.6 RMSE: 2.5076880277785705
Mejor modelo 2: Rank: 6 MaxIter: 10 RegParam: 0.3 RMSE: 2.4081241623658416
Mejor modelo 3: Rank: 4 MaxIter: 10 RegParam: 0.3 RMSE: 2.254908497737335
Mejor modelo 4: Rank: 5 MaxIter: 10 RegParam: 0.1 RMSE: 2.050853303677055
Mejor modelo 5: Rank: 1 MaxIter: 10 RegParam: 0.3 RMSE: 1.4277035596730039
from pyspark.sql.functions import max, min
# Calcular el máximo y mínimo de las calificaciones
max_rating = ratings_cleaned.agg(max("rating")).collect()[0][0]
min_rating = ratings_cleaned.agg(min("rating")).collect()[0][0]
# Calcular el rango
rating_range = max_rating - min_rating
print("Máxima calificación:", max_rating)
print("Mínima calificación:", min_rating)
print("Rango de calificación:", rating_range)
     Máxima calificación: 5
     Mínima calificación: 1
     Rango de calificación: 4
```

test_predicciones.show(50)

+	·				·	++
user_id	app_id	title	hours	is_recommended	rating	prediction
+	h					++
10999	578080	PUBG: BATTLEGROUNDS	247	false	1	0.84161645
15232	311210	Call of Duty®: Bl	2	true	4	3.561506
17736	250900	The Binding of Is	147	true	4	0.99319416
23597	275850	No Man's Sky	26	false	1	0.8691543
23597	289070	Sid Meier's Civil	31	false	1	0.8845645
30684	359550	Tom Clancy's Rain	170	true	4	3.8211203
59192	1222670	The Sims™ 4	66	true	4	3.8914433
67667	374320	DARK SOULS™ III	64	true	4	3.648781
70586	493520	GTFO	0	true	4	3.635912
119812	107410	Arma 3	675	true	5	3.8867872
123608	440900	Conan Exiles	324	true	4	0.9341673
143745	1151340	Fallout 76	344	true	4	4.4028106
144903	275850	No Man's Sky	22	true	4	3.6405146
151218	1063660	Bendy and the Dar	15	true	4	3.6788993
153960	244850	Space Engineers	858	true	5	3.8345459
158748	582660	Black Desert	31	true	4	3.5240164
166848	346110	ARK: Survival Evo	0	true	4	3.782125
185688	1222670	The Sims™ 4	176	true	4	3.9393508
185909	431960	Wallpaper Engine	29	true	4	3.7171109
198725	1172620	Sea of Thieves 20	74	true	4	3.5483317
198728	635260	CarX Drift Racing	96	true	4	4.1441545
227273	1332010	Stray	6	true	4	3.7661552
233010	1789480	Marauders	7	true	4	3.5389218
241669	394360	Hearts of Iron IV	237	false	1	1.0212784
269607	346110	ARK: Survival Evo	112	true	4	3.871374
306166	1466860	Age of Empires IV	4	true	4	3.7306745
332605	686810	Hell Let Loose	714	false	3	3.626688
357092	271590	Grand Theft Auto V	5	false	1	1.793386
393538	346110	ARK: Survival Evo	124	true	4	4.057405
413062	815370	Green Hell	3	true	4	3.8128362

ī	422863	1 633360	Risk of Rain 2	20	false	1	3.6236563
ł	427316		RISK OF KAIN 2 Brawlhalla		true		
!				-	'		
	463623	367520	Hollow Knight	14	true	4	3.7590837
	467085	386360	SMITE®	241	true	4	3.4917452
	473439	250900	The Binding of Is	25	true	4	3.8006465
	484319	1938090	Call of Duty®	17	false	1	3.2900813
İ	486616	236390	War Thunder	19	false	1	4.4714994
	503801	431960	Wallpaper Engine	14	true	4	3.7727282
ĺ	518353	784080	MechWarrior 5: Me	2	false	1	3.62443
	566987	4000	Garry's Mod	90	true	4	3.6960185
	575207	739630	Phasmophobia	46	true	4	3.6846583
	576682	22380	Fallout: New Vegas	12	true	4	3.613783
	592658	489830	The Elder Scrolls	132	true	4	3.4210224
	594470	440900	Conan Exiles	831	true	5	3.759099
	619739	570	Dota 2	17	false	1	4.0226054
	622213	374320	DARK SOULS™ III	33	true	4	2.3202872
	623642	39210	FINAL FANTASY XIV	125	true	4	4.070985
	626881	493340	Planet Coaster	0	false	1	3.8794112
	632096	602960	Barotrauma	30	true	4	3.8319683
	635407	1063660	Bendy and the Dar	19	true	4	3.5424058
+		+	++	+	+	+	+

only showing top 50 rows

6. Generar Recomendaciones

nrecommendations = best_model.recommendForAllUsers(10)
nrecommendations.limit(10).show()

++ user_id	recor	+ mmendations
2 171 225 249 478 529 641 800	[{548430, [{548430, [{548430, [{548430, [{548430, [{548430, [{548430, [{548430, [{548430, [{548430, [{548430,	3.99887 3.05394 3.98865 4.04594 1.02585 3.98865 1.03484 3.89781 4.13278 5.11745
++		+

ratings_cleaned.show()

rating	is_recommended	hours	title	app_id	user_id
4	true	36	Dwarf Fortress	975370	51580
1	false	11	FOR HONOR™	304390	2586
4	true	336	Destiny 2	1085660	253880
4	true	27	Planet Zoo	703080	259432
4	true	7	Satisfactory	526870	23869
4	true	8	The Elder Scrolls	306130	45425
5	true	538	Path of Exile	238960	88282
1	false	157	Counter-Strike: G	730	63209
4	true	18	Cities: Skylines	255710	354512
4	true	397	Sid Meier's Civil	289070	454422
4	true	30	Wallpaper Engine	431960	199725
4	true	50	Baldur's Gate 3	1086940	85822
4	true	46	Call of Duty®	1938090	161081
4	true	19	STAR WARS™: The O	1286830	113279
4	true	89	Sea of Thieves 20	1172620	122640
4	true	61	The Elder Scrolls	306130	75422
4	true	177	CarX Drift Racing	635260	76583
4	true	86	Fallout 76	1151340	124924
4	true	244	Sid Meier's Civil	289070	261528
2	false	320	X4: Foundations	392160	408750

only showing top 20 rows

data = nrecommendations.join(ratings_cleaned, on='user_id').show()

+		+		
user_id	recommendations app_id	title	nours is	_recommended rating
+		+-		
2 [{ 5484	430, 3.99887 291550	Brawlhalla	12	true 4
171 [{5484	430, 3.05394 582660	Black Desert	854	false 3
225 [{ 5484	430, 3.98865 108600	Project Zomboid	40	true 4
249 [{ 5484	430, 4.04594 1286830 STAI	R WARS™: The O	0	true 4
478 [{ 5484	430, 1.02585 376210	The Isle	2	false 1

	529 [{ 548430,	3.98865	108600	Project Zomboid	112	truel	41
i	1 5 0			Marauders			1
i	1 5 0			The Binding of Is			
	1 5 0			Sea of Thieves 20			41
	1 6 6			Counter-Strike: G			5
				No Man's Sky			1
!	1 5 0					!	+ !
	1116 [{548430,	5.36593	438100	VRChat	983	true	5
	1254 [{ 548430,	4.17939	1599340	Lost Ark	166	true	4
	1308 [{548430,	3.90419	431960	Wallpaper Engine	7	true	4
	1427 [{ 548430 ,	4.33210	1938090	Call of Duty®	236	true	4
	1431 [{548430,	5.11745	730	Counter-Strike: G	999	true	5
	1558 [{ 548430,	3.84531	252490	Rust	356	true	4
	1740 [{548430,	4.25107	1151640	Horizon Zero Dawn	53	true	4
	1889 [{548430,	5.11745	730	Counter-Strike: G	617	true	5
	2017 [{548430,	1.08302	1938090	Call of Duty®	70	false	1
+	+		+	h	+	+	+
on1	y showing top 2	a nows					
JIII	y 3110WING COP 2	0 1000					

ratings_cleaned.filter(ratings_cleaned.user_id == 1427).show()

+				+
user_id app_id	title	hours is_	recommended ra	ating
+				+
1427 1938090 Call of	Duty®	236	true	4
4				+

ratings_cleaned.filter(ratings_cleaned.user_id == 478).show()

+	+	++
user_id app_id title hou	rs is_recommend	ded rating
+	+	++
478 376210 The Isle	2 fal	lse 1
	+_	+

ratings_cleaned.filter(ratings_cleaned.app_id == 1938090).show()

+				·	+
user_id app_id		title	hours	is_recommended	rating
+					+
161081 1938090 Ca	all of	Duty®	46	true	4
664486 1938090 Ca	all of	Duty®	7	true	4
903527 1938090 Ca	all of	Duty®	98	true	4
1271336 1938090 Ca	all of	Duty®	124	true	4
7909145 1938090 Ca	all of	Duty®	17	false	1
11131971 1938090 Ca	all of	Duty®	123	false	1
13330978 1938090 Ca	all of	Duty®	88	true	4
188838 1938090 Ca	all of	Duty®	302	true	4
5929717 1938090 Ca	all of	Duty®	192	false	1
8951161 1938090 Ca	all of	Duty®	150	false	1
1072234 1938090 Ca	all of	Duty®	17	false	1
4697968 1938090 Ca	all of	Duty®	206	true	4
6880727 1938090 Ca	all of	Duty®	54	true	4
7179102 1938090 Ca	all of	Duty®	39	false	1
1863686 1938090 Ca	all of	Duty®	135	true	4
6072048 1938090 Ca	all of	Duty®	57	false	1
8672245 1938090 Ca	all of	Duty®	93	true	4
4374667 1938090 Ca	all of	Duty®	311	true	4
4717532 1938090 Ca	all of	Duty®	16	true	4
6758969 1938090 Ca	all of	Duty®	198	true	4
+					+

only showing top 20 rows

ratings_cleaned.sort('user_id', ascending=False).show(10, truncate=False)

+	+	+	++
user_id app_id title	hours	is_recommended	rating
+	+	+	++
14305844 270880 American Truck Simulator	121	true	4
14305769 252490 Rust	600	false	3
14305581 1449850 Yu-Gi-Oh! Master Duel	106	false	1
14305542 1426210 It Takes Two	15	true	4
14305385 1245620 ELDEN RING	136	true	4
14305333 289070 Sid Meier's Civilization® VI	56	true	4
14305263 244850 Space Engineers	586	true	5
14305218 397540 Borderlands 3	21	true	4
14304939 1086940 Baldur's Gate 3	101	true	4
14304908 292030 The Witcher® 3: Wild Hunt	136	true	4
+	+	+	++

only showing top 10 rows

```
#Crear un dataframe con un nuevo usuario
df_usuarios = spark.createDataFrame([(14305845,1938090,'Call of Duty^0'),(14305845,376210,'The Isle')], ['user_id', 'app_id','title'])
df usuarios.show()
      | user_id| app_id| title|
      |14305845|1938090|Call of Duty®
      |14305845| 376210| The Isle
ratings_cleaned.filter(ratings_cleaned.user_id == 8993770).show()
                                 title|hours|is_recommended|rating|
      |user_id| app_id|
                                                     55| true|
93| true|
      |8993770|1044720| Farthest Frontier|
      |8993770|1250410|Microsoft Flight ...| 93|
      # Usuario que tiene mayor recomendaciones
usu_r = ratings_cleaned.filter(ratings_cleaned['user_id'] == 8993770)
usu r.show()
      |user_id| app_id| title|hours|is_recommended|rating|

    |8993770|1044720|
    Farthest Frontier
    55|
    true

    |8993770|1250410|Microsoft Flight ...
    93|
    true

    |8993770|1184370|Pathfinder: Wrath...
    16|
    true

    |8993770|1240440|
    Halo Infinite
    197|
    true

    |8993770|1496790|
    Gotham Knights
    4|
    false
```

falsel

Recomendar juegos al usuario con la id 8993770

+-----+----+----

```
from pyspark.sql import functions as F
user_id = 8993770
num recommendations = 10
# Obtén todas las recomendaciones para todos los usuarios
all_recommendations = best_model.recommendForAllUsers(num_recommendations)
# Filtra las recomendaciones para el usuario específico
user_recommendations = all_recommendations.filter(F.col("user_id") == user_id)
# Explora las recomendaciones para el usuario
user_recommendations.show()
     |user id| recommendations|
     |8993770|[{548430, 4.25449...|
from pyspark.sql.functions import col
# Unión de DataFrames basada en la columna 'app id'
user_recommendations_with_titles = user_recommendations_final.join(
   ratings_cleaned,
   user_recommendations_final.app_id == ratings_cleaned.app_id
   user_recommendations_final["user_id"],
   user_recommendations_final["app_id"],
   ratings_cleaned["title"],
   user_recommendations_final["rating"]
).dropDuplicates()
# Muestra las recomendaciones con títulos sin duplicados
user_recommendations_with_titles.show()
                                 title| rating|
     |user_id|app_id|
```

+	·+
8993770 4000	Garry's Mod 4.1865788
8993770 252490	Rust 4.12766
8993770 526870	Satisfactory 4.0907865
8993770 570	Dota 2 4.2405243
8993770 105600	Terraria 4.2408776
8993770 294100	RimWorld 4.1383023
8993770 107410	Arma 3 4.1369596
8993770 394360	Hearts of Iron IV 4.1782303
8993770 548430	Deep Rock Galactic 4.2544966
8993770 39210	FINAL FANTASY XIV 4.2015495

Conclusión

Basándose en los resultados obtenidos, parece que el modelo de recomendación ha sugerido juegos con altas calificaciones ("rating") para el usuario con el ID 8993770. Los títulos recomendados incluyen juegos populares y bien valorados, como "Garry's Mod", "Rust", "Satisfactory", "Dota 2", "Terraria" y otros. La calificación asociada con cada recomendación indica la estimación de gusto del usuario para ese juego en particular.

En cuanto al éxito, se puede afirmar que el modelo ha demostrado ser efectivo al proporcionar recomendaciones que están alineadas con los gustos y preferencias del usuario. La consistencia

 de las calificaciones estimadas, que son generalmente altas, sugiere que el modelo ha logrado capturar de manera acertada los patrones de preferencia del usuario, ofreciendo así recomendaciones significativas y relevantes.

Empieza a programar o a crear código con IA.