!pip install turicreate

from google.colab import drive
drive.mount('/content/gdrive')

Mounted at /content/gdrive

Build a song recommender system

import turicreate

▼ Load some music data

song_data = turicreate.SFrame('/content/gdrive/My Drive/Turicreate/Week 5/song_data.sframe/')

▼ Explore our data

song_data

user id	song_id	listen count	title
b80344d063b5ccb3212f76538 f3d9e43d87dca9e	SOAKIMP12A8C130995	1	The Cove
b80344d063b5ccb3212f76538 f3d9e43d87dca9e	SOBBMDR12A8C13253B	2	Entre Dos Aguas
b80344d063b5ccb3212f76538 f3d9e43d87dca9e	SOBXHDL12A81C204C0	1	Stronger
b80344d063b5ccb3212f76538 f3d9e43d87dca9e	SOBYHAJ12A6701BF1D	1	Constellations
b80344d063b5ccb3212f76538 f3d9e43d87dca9e	SODACBL12A8C13C273	1	Learn To Fly
b80344d063b5ccb3212f76538 f3d9e43d87dca9e	SODDNQT12A6D4F5F7E	5	Apuesta Por El Rock 'N' Roll
b80344d063b5ccb3212f76538 f3d9e43d87dca9e	SODXRTY12AB0180F3B	1	Paper Gangsta

Show the most popular songs in the dataset

†3d9e43d8/dca9e	I		kosmisch
song_data['song'].show()			
1	1	1	vour

Count the number of unique users in the data

```
users = song_data['user_id'].unique()
| stronger - kanye vvest |
len(users)

66346
| Apuesta Por El Rock 'N' |
```

Create a song recommender

```
train_data,test_data = song_data.random_split(.8,seed=0)
```

Create a very simple popularity recommender

Warning: Ignoring columns song_id, listen_count, title, artist;

▼ Use the popularity model to make some predictions

Pata prepared in: 1 //105/s
popularity_model.recommend(users=[users[0]])

user_id	song	score	rank
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Sehr kosmisch - Harmonia	4754.0	1
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Undo - Björk	4227.0	2
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	You're The One - Dwight Yoakam	3781.0	3
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Dog Days Are Over (Radio Edit) - Florence + The	3633.0	4
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Revelry - Kings Of Leon	3527.0	5
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Horn Concerto No. 4 in E flat K495: II. Romance	3161.0	6
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Secrets - OneRepublic	3148.0	7
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Fireflies - Charttraxx Karaoke	2532.0	8
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Tive Sim - Cartola	2521.0	9
c66c10a9567f0d82ff31441a9 fd5063e5cd9dfe8	Drop The World - Lil Wayne / Eminem	2053.0	10

[10 rows x 4 columns]

popularity_model.recommend(users=[users[1]])

user_id	song	score	rank
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Sehr kosmisch - Harmonia	4754.0	1
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Undo - Björk	4227.0	2
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	You're The One - Dwight Yoakam	3781.0	3
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Dog Days Are Over (Radio Edit) - Florence + The	3633.0	4
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Revelry - Kings Of Leon	3527.0	5
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Horn Concerto No. 4 in E flat K495: II. Romance	3161.0	6
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Secrets - OneRepublic	3148.0	7
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Hey_ Soul Sister - Train	2538.0	8
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Fireflies - Charttraxx Karaoke	2532.0	9
02f015d32ac2cd1e52d26e3ec 36048711dd5711b	Tive Sim - Cartola	2521.0	10

[10 rows x 4 columns]

Build a recommender with personalization

Warning: Ignoring columns song_id, listen_count, title, artist;

To use one of these as a rating column, specify the column name to be used as target and use a method that allows the use of a target.

Preparing data set.

Data has 893580 observations with 66085 users and 9952 items.

Data prepared in: 1.49533s

Training model from provided data.

Gathering per-item and per-user statistics.

Elapsed Time (Item Statistics)	% Complete
1.407ms	1.5 100

Setting up lookup tables.

Processing data in one pass using dense lookup tables.

Elapsed Time (Constructing Lookups)	Total % Complete	Items Processed
256.656ms 2.01s	0 100	0

Finalizing lookup tables.

Generating candidate set for working with new users.

Finished training in 3.09063s

▼ Apply personalized model to make song recommendations

personalized_model.recommend(users=[users[0]])

user_id	song	score	rank
c66c10a9567f0d82ff31441a9	Cuando Pase El	0.019450453611520622	1
fd5063e5cd9dfe8	Temblor -		
	Soda Stereo		
c66c10a9567f0d82ff31441a9	Fireflies - Charttraxx	0.014473731701190654	2
fd5063e5cd9dfe8	Karaoke		

personalized_model.recommend(users=[users[1]])

uson id	song	S.COWO.	ronk
user_id	song	score	rank
02f015d32ac2cd1e52d26e3ec	Where The Boat	0.06153608858585358	1
36048711dd5711b	Leaves		
	From (Album) - Zac		
	Brown		
02f015d32ac2cd1e52d26e3ec	Different Kind Of Fine	0.06052837520837784	2
36048711dd5711b	(Album) - Zac Brown		
	Band		
02f015d32ac2cd1e52d26e3ec	Jolene (Album) - Zac	0.05786820501089096	3
36048711dd5711b	Brown Band		
02f015d32ac2cd1e52d26e3ec	Sic 'Em On A Chicken	0.05518669635057449	4
36048711dd5711b	(Album) - Zac Brown	0.03310003033037113	
300407110437110	Band		
00/045 100 0 14 50 100 0		0.05206222222454520	_
02f015d32ac2cd1e52d26e3ec	Who's Kissing You	0.05306333303451538	5
36048711dd5711b	Tonight		
	- Jason Aldean		
02f015d32ac2cd1e52d26e3ec	What Country Is -	0.037490807473659515	6
36048711dd5711b	Luke		
	Bryan		
02f015d32ac2cd1e52d26e3ec	Highway 20 Ride	0.03733157366514206	7
36048711dd5711b	(Album) -		
300 107 110 037 110	Zac Brown Band		
02/045 122 2 14 52 126 2		0.022077250400522026	
02f015d32ac2cd1e52d26e3ec	Do I - Luke Bryan	0.033077359199523926	8

Apply model to find similar songs in the data set

personalized_model.get_similar_items(['With Or Without You - U2'])

song	similar	score	rank
With Or Without You - U2	I Still Haven't Found What I'm Looking For	0.04285717010498047	1
With Or Without You - U2	Hold Me_ Thrill Me_ Kiss Me_ Kill Me - U2	0.033734917640686035	2
With Or Without You - U2	Window In The Skies - U2	0.03283584117889404	3
With Or Without You - U2	Vertigo - U2	0.03007519245147705	4
With Or Without You - U2	Sunday Bloody Sunday - U2	0.02713179588317871	5
With Or Without You - U2	Bad - U2	0.02517986297607422	6
With Or Without You - U2	A Day Without Me - U2	0.023715436458587646	7
With Or Without You -	Another Time Another	0.020325183868408203	8

personalized_model.get_similar_items(['Chan Chan (Live) - Buena Vista Social Club'])

song	similar	score	rank	
Chan Chan (Live) - Buena	Murmullo - Buena Vista	0.1881188154220581	1	
Vista Social Club	Social Club			
Chan Chan (Live) -	La Bayamesa - Buena	0.1871921420097351	2	
Buena	Vista	0.1071321420037331	_	
Vista Social Club	Social Club			
Chan Chan (Live) -	Amor de Loca Juventud -	0.18483412265777588	3	
Buena	Buena Vista Social Club			
Vista Social Club				
Chan Chan (Live) -	Diferente - Gotan Project	0.021459221839904785	4	
Buena				
Vista Social Club				
Chan Chan (Live) -	Mistica - Orishas	0.020576119422912598	5	
Buena				
Vista Social Club				
Chan Chan (Live) -	Hotel California - Gipsy	0.019304990768432617	6	
Buena	Kings			
Vista Social Club				
Chan Chan (Live) -	Nací Orishas - Orishas	0.019157111644744873	7	
Buena				
Vista Social Club				
Chan Chan (Live) -	Gitana - Willie Colon	0.018796980381011963	8	

Compare the models quantitatively

We now formally compare the popularity and the personalized models using precision-recall curves.

%matplotlib inline

model_performance = turicreate.recommender.util.compare_models(test_data, [popularity_model, person

 \Box

compare_models: using 2931 users to estimate model performance

PROGRESS: Evaluate model MO

recommendations finished on 1000/2931 queries. users per second: 2880.67 recommendations finished on 2000/2931 queries. users per second: 2827.94 WARNING:root:Model trained without a target. Skipping RMSE computation.

Precision and recall summary statistics by cutoff

_	L	L	1.		
	cutoff	 mean_precision		mean_recall	
	1 2	0.02388263391334016 0.02200614124872059	T . 	0.006151279820675928 0.011831884708036184	-T
	3	0.021835551006482472		0.018200163875701256	
	4 5	0.019617877857386577 0.018287273967929066	 	0.022059848108978125 0.025565896062313683	
	6	0.017229614466052538		0.02918379494482813	
	7	0.016376663254861864		0.032297225590018665	
	8	0.01586489252814739		0.03537798330230148	
	9 10	0.015201486030554625 0.014602524735585164	 -	0.037579592016192646 0.03973713158417237	1
_	10 	0.014002324733363104 	 -		 -

[10 rows x 3 columns]

The table shows that the personalized model provides much better performance.

FLECTSTOLL ALLO LEGATE SUMMMALY STATISTICS BY CUTOTE

cutoff mean_precision mean_recall 1 0.1047424087342204 0.03105546520186173 2 0.08512453087683383 0.04995969537784218 3 0.07380871147503701 0.06378640659452275 4 0.0655066530194473 0.07525315547171171 5 0.05991129307403611 0.08455082357183623 6 0.0540202433754123 0.09211122064093345 7 0.049617390456694464 0.09838824378080202 8 0.04652848857045376 0.104057632203517 9 0.04367110201296487 0.10908059173005889 10 0.04152166496076429 0.11476113602370451			
2 0.08512453087683383 0.04995969537784218 3 0.07380871147503701 0.06378640659452275 4 0.0655066530194473 0.07525315547171171 5 0.05991129307403611 0.08455082357183623 6 0.0540202433754123 0.09211122064093345 7 0.049617390456694464 0.09838824378080202 8 0.04652848857045376 0.104057632203517 9 0.04367110201296487 0.10908059173005889	cutof	f mean_precision	mean_recall
9 0.04367110201296487 0.10908059173005889	4 5 6 7	0.08512453087683383 0.07380871147503701 0.0655066530194473 0.05991129307403611 0.0540202433754123 0.049617390456694464	0.04995969537784218 0.06378640659452275 0.07525315547171171 0.08455082357183623 0.09211122064093345 0.09838824378080202
	8	0.04652848857045376	0.104057632203517
10 0.04152166496076429 0.11476113602370451	9	0.04367110201296487	0.10908059173005889
	10	0.04152166496076429	0.11476113602370451

[10 rows x 3 columns]

✓ 3초 오후 11:12에 완료됨