Oh, Behave! Country Representation Dynamics Created by Feedback Loops in Music Recommender Systems

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Motivation

- Previous work shows that recommender systems may underrepresent music from certain countries in their recommendations.
- This may cause users receive less recommendations of their local (domestic) music and shift their preference towards global trends.
- This work investigates potential long-term effects of such underrepresentation.

Research questions

- RQ 1: How do different recommender systems affect representation of local and US-produced music in recommendations and user profiles in the long term?
- RQ 2: How do effects of feedback loops vary across different countries? Do different recommender systems treat individual countries differently?

Results RQ1

Representation of Local and US-produced Music in the Long Term

Right: Proportion of *local* and *US*-produced music in _____ recommendations (*Rec**) and _____ simulated user profiles (*Prof**) _____ at iteration 100, compared to _____ the respective proportions in _____ the original user profiles _____ (difference in %). Significant _____ changes are marked with * . ____

		Рор	ItemKNN	BPR	NeuMF	MultiVAE	LightGCN
k)	Rec _{local} Rec _{US}	-47.5* +15.6*	+0.4 +4.8*	$ -6.6^*$ +3.5*	-39.7* +17.0*	-21.2* +7.5*	-3.5 +0.5
י י ר	Prof _{local} Prof _{US}	-22.5* +19.5*	+2.1 +2.2*	$ -7.0^* +3.7^*$	-19.5* +7.6*	-10.5* +2.6*	-3.6 +0.8
	JSD _{Prof}	0.13	0.12	0.08	0.10	0.09	0.08
	nDCG ₁	0.03	0.26	0.13	0.08	0.12	0.14

- The proportion of local music declines in the recommendations and resulting profiles for most algorithms. The proportion of USproduced music generally increases.
- Varied impact patterns: LightGCN preserves proportions, least miscalibrated (JSD), ItemKNN – preserves proportions, most miscalibrated, NeuMF – distorts proportions, miscalibrated.
- Iteration 1: most models already show local-item proportion inconsistent with the initial profiles (dashed line, Right).
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Method

Feedback loop simulation:

- 1. Split current input data into train (75%), validation (20%), test (5%).
- 2. (Re)train the model, optimize for nDCG@10, produce 10 novel recommendations per user.
- 3. Simulate consumption of one item per user (randomly, higher probability for higher ranking items).
- 4. Combine simulated interactions with the current input. Go to step1.

Evaluation

- Average proportions of local and US music in recommendations and simulated consumptions histories (overall and per country).
- Jensen-Shannon Divergence (JSD) between 3-bin track popularity [*HighPop*, *MidPop*, LowPop] and country of origin (user-specific) [*local*, *US-produced*, *other*] distributions (miscalibration).
- Some models may converge or remain around the initial user profiles (ItemKNN, BPR, LightGCN).

Note on calibration

ltemKNN:

- Highest country miscalibration
- Lowest popularity miscalibration

No direct connection between popularity and country calibration



Dataset

	Tracks	Track Inte	eractions	Users	User		
		Total	Average		Interactions		
US	39,614	1,040,360	26.26	1,582	323,072		
UK	15,522	422,225	27.20	823	171,469		
DE	6,793	107,832	15.87	805	158,642		
SE	4,519	107,491	23.79	320	60,993		
CA	3,754	95,343	25.40	217	47,490		
FR	2,800	56,241	20.09	254	52,850		
AU	2,346	53,701	22.89	193	40,767		
FI	2,260	45,709	20.23	420	78,819		
NO	1,765	36,769	20.83	208	40,545		
BR	2,236	35,964	16.08	1,064	205,093		
NL	1,738	32,035	18.43	375	89,546		
PL	1,709	27,116	15.87	1,040	195,296		
RU	1,888	24,086	12.76	1,162	187,876		
JP	1,796	21,818	12.15	101	14,411		
IT	1,506	21,273	14.13	222	37,421		
other	9,651	159,769	16.55	2,990	583,442		
Total	99,897	2,287,732	22.90	11,776	2,287,732		

Effects of Feedback Loops Across Different Countries

Below: miscalibration between three-bin country distributions over the original user profiles and the profiles after 100 iterations (measured as JSD). Higher values indicate higher inconsistency.

Results RQ2

	US	UK	DE	SE	CA	FR	AU	FI	NO	BR	NL	PL	RU	JP	IT	all
Рор	0.073	0.126	0.146	0.133	0.108	0.144	0.101	0.186	0.108	0.133	0.096	0.16	0.163	0.196	0.153	0.128
ItemKNN	0.084	0.123	0.139	0.137	0.106	0.128	0.109	0.152	0.127	0.134	0.094	0.155	0.149	0.166	0.174	0.124
BPR	0.054	0.093	0.089	0.09	0.085	0.095	0.088	0.103	0.084	0.083	0.07	0.103	0.098	0.146	0.114	0.083
NeuMF	0.058	0.099	0.114	0.104	0.09	0.111	0.093	0.126	0.087	0.099	0.074	0.121	0.115	0.179	0.135	0.096
MultiVAE	0.062	0.098	0.101	0.099	0.092	0.1	0.094	0.114	0.091	0.1	0.074	0.111	0.108	0.132	0.136	0.092
LightGCN	0.053	<u>0.089</u>	0.079	0.081	0.08	<u>0.086</u>	0.091	0.088	0.084	0.069	0.068	<u>0.093</u>	<u>0.086</u>	<u>0.099</u>	0.13	0.077

Below: deviations in the average proportions of local and US music in user profiles at iteration 100 from the respective average proportions in the original user profiles before simulation (in %). Statistically significant deviations are marked with *. Per column: highest value in **bold**, lowest value <u>underlined</u>.

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- User listening activity on Last.fm in 2018-2019 (LFM-2b dataset sample).
- Artist country information from MusicBrainz.org.

Paper & code

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local proportion in user profiles																
Рор	-1.2	-4.2	-41.6*	-40.8*	-31.0*	-42.2*	-31.2*	-53.4*	-47.3*	-48.6*	-44.1*	-54.4*	-57.7*	-67.3*	-58.3*	-22.5*
ItemKNN	-1.5	<u>-15.0</u> *	+5.6	+2.4	-18.3	-20.1	-25.6	+16.0	-1.7	+7.2	-19.6	+28.5*	+10.7	+18.7	+49.6	+2.1
BPR	-2.3	-7.6*	-8.8	-17.9*	-23.7^{*}	-31.3*	-31.3^{*}	+3.4	-39.2*	+3.8	-35.1*	-1.2	-19.1*	-40.3	-5.6	-7.0^{*}
NeuMF	-3.3*	-6.1	<u>-33.6</u> *	<u>-31.4</u> *	-23.1^{*}	-40.2^{*}	<u>-33.8</u> *	-35.0^{*}	<u>-39.9</u> *	-29.5^{*}	<u>-37.3</u> *	-46.7^{*}	-49.0^{*}	-66.5^{*}	<u>-56.6</u> *	<u>-19.5</u> *
MultiVAE	-3.5^{*}	-7.5*	-7.0	-19.0*	-23.4^{*}	-34.8*	-29.8*	-19.0	-38.7*	-0.4	-36.3*	-23.6*	-12.7	-18.5	-53.1*	-10.5^{*}
LightGCN	-3.1	-10.3*	-3.4	-1.7	-21.3*	-3.7	-29.5^{*}	-0.8	-35.4*	-3.5	-31.5*	+12.3	-12.3	+10.0	+48.6*	-3.6
US proportion in user profiles																
Рор	-1.2	+14.9*	+28.8*	+23.8*	+2.3	+30.9*	+8.4*	+44.5*	+17.5*	+17.4*	+16.5*	+34.8*	+36.9*	+28.8*	+24.1*	+19.5*
ItemKNN	-1.5	+8.4*	+2.9	-0.7	-0.4	+7.7	+4.5	-6.3	+1.5	-0.4	+5.7	+3.8	+6.1	-10.8	-1.9	+2.2*
BPR	-2.3	+7.3*	+4.3	+6.0	+0.3	+9.8	+3.2	0.0	+7.0	+2.5	+4.8	+6.9*	+7.0*	-0.4	+3.1	+3.7*
NeuMF	-3.3*	+8.3*	+13.6*	+10.6*	-1.2	+15.4*	+2.9	+15.6*	+6.5	+9.7*	+6.7	+14.2*	+11.8*	+17.3*	+13.0*	+7.6*
MultiVAE	-3.5^{*}	+5.2	+2.4	+3.4	+0.1	+6.0	+2.6	+4.4	+5.4	+2.8	+4.4	+6.4*	+3.0	+0.7	+11.0*	+2.6*
LightGCN	-3.1	+5.2	+0.7	+0.2	-0.9	+3.0	+2.7	-2.6	+0.4	+1.9	+3.4	+3.2	+2.9	-5.6	<u>-8.2</u>	+0.8
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The list of significantly affected countries varies between algorithms.
Overall, on per-country level most changes are in local rather than US proportions.

- Users from countries less represented in the data (e.g., FI) are more likely to receive recommendations inconsistent with their original preferences. However, higher representation does not always guarantee the opposite (e.g., DE).
- Causes of miscalibration and under-representation call for more research.

https://github.com/hcai-mms/FeedbackLoops4RecSys