A Study of Heterogeneity in Recommendations for a Social Music Service

Alejandro Bellogín, Iván Cantador, Pablo Castells {alejandro.bellogin, ivan.cantador, pablo.castells}@uam.es

Universidad Autónoma de Madrid Escuela Politécnica Superior Information Retrieval Group http://ir.ii.uam.es







Social Music Service: Last.fm

w! Festival reco	mmendations based on your taste »	Ø) English He	Music search	٩	
Artist	The Beatles	Т	op Trac	ks		
iography	264,585,910 plays (2,259,243 listeners)	La	ast Week La	ast 6 months		
tures	Listening now: polly_xq, FedorMeloman, mauro6006		1 Ac	ross the Universe	13110	
eos	= 27,101 shouts	8. 1	7 Co	me Together	11 605	
	🕂 Add to my Library 🛛 🖾 Share		3 Let	tit Be	11.024	
ims	Liverpool, England (1960 – 1970)		4 Hel	lo!	10,652	
:ks		Far	5 Ye	sterday	10,465	
vte	 The Beatles were an iconic rock group from Liverpool, England They are frequently cited as the most commercially. 		6 Hei	re Comes the Sun	10,413	
1.5	successful and critically acclaimed band in modern history.		7 Soi	mething	9,343	
s	with innovative music, a cultural impact that helped define		8 Al	You Need Is Love	9,139	
ts	the 1960s and an enormous influence on music that is still	2/121	9 Ele	anor Rigby	8,875	
	acts to sell more than 1 billion records with only Elvis		10 He	y Jude	8,763	
lar Artists	Presley having been able to achieve the same feat.	7.8	11 Ye	llow Submarine	8,691	
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cilcis	classic rock, rock, british, 60s, pop		14 Cai	n't Buy Me Love	7,574	
nal	See more	1	15 Str	awberry Fields Forever	7,525	
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			Scrob	_xq bling now from WMP eatles – I'm a Loser		
	John Lennon George Paul Paul Ringo Starr V Harrison McCartney McCartney &	fings Paul & Linda McCartney	Fedor Scrob The B	r Meloman bling now from The Last.fm S eatles – The Night Before	crobbler	
		see more 🥥	mirin	dA92		

Scrobbling now from foobar2000 The Beatles – Searchin'

22

bigdaddyets

Top Listener

dsendecki

michaelwsimpson

Top Listener

formundo

- R-

1st research question

Which sources of information in social systems are more valuable for recommendation?





Tags?

Music Radio Events Charts

New! Festival recommendations based on your taste »

The Beatles

Artist
Biography
Pictures
Videos
Albums
Tracks
Events
News
Charts
Similar Artists
Tags
Listeners
Journal
Groups

264,585,910 plays (2,259,243 listeners)	
Listening now: polly_xq, FedorMeloman, mauro6006	E
🕂 Add to my Library 🛛 🖾 Share	-70
Liverpool, England (1960 — 1970)	0
The Beatles were an iconic rock group from Liverpool, England. They are frequently cited as the most commercially successful and critically acclaimed band in modern history, with innovative music, a cultural impact that helped define the 1960s and an enormous influence on music that is still felt today. Currently, The Beatles are one of the two musical acts to sell more than 1 billion records, with only Elvis Presley having been able to achieve the same feat. Read more	
Lagged us.	Play The
classic rock, rock, british, 60s, pop See more	4.5
Similar Artists	



🔞 English | Help Music search

Last Week Last 6 months

Top Tracks

1	Across the Universe	13,119
2	Come Together	11,605
З	Let It Be	11,024
4	Help!	10,652
5	Yesterday	10,465
6	Here Comes the Sun	10,413
7	Something	9,343
8	All You Need Is Love	9,139
9	Eleanor Rigby	8,875
10	Hey Jude	8,763
11	Yellow Submarine	8,691
12	Lucy in the Sky With Diamonds	8,624
13	While My Guitar Gently Weeps	7,775
14	Can't Buy Me Love	7,574
15	Strawberry Fields Forever	7,525

Join Login

Q

Listeners

dsendecki

	polly_xq Scrobbling now fr The Beatles – I'm a	om WMP a Loser		
5	FedorMeloman Scrobbling now fr The Beatles – The	om The La Night Bef	ast.fm Scrobbler ore	•
ę	mirindA92 Scrobbling now fr The Beatles – Sea	om foobai archin'	2000	•
ĥ	bigdaddyets Top Listener		michaelwsim Top Listener	pson

formundo



George Harrison

Paul Paul McCartney & McCartney



Community



Wings

See more 🔘

McCartney







Track listenings?

Festival recom	mendations based on your taste »	😵 English Help 🛛 Music search	٩
st	The Beatles	Top Tracks	
raphy	264,585,910 plays (2,259,243 listeners)	Last Week Last 6 months	
es	Listening new, polly xa, Federmeloman, mauro6006		\wedge
	📮 27,101 shouts	1 Across the Universe	(3,119
S	+ Add to myllibrary Share	2 Come Together	11,605
is		3 Let It Be	11,024
	Liverpool, England (1960 – 1970)	4 Help!	10,652
5	The Beatles were an iconic rock group from Liverpool,	5 Yesterday	10,465
s	England. They are frequently cited as the most commercially	6 Here Comes the Sun	10,413
	successful and critically acclaimed band in modern history,	7 Something	9,343
	with innovative music, a cultural impact that helped define	8 All You Need Is Love	9,139
ts	felt today. Currently: The Beatles are one of the two musical	9 Eleanor Rigby	8,875
ar Artists	acts to sell more than 1 billion records, with only Elvis	10 Hey Jude	8,763
i Ai doto	Presley having been able to achieve the same feat.	11 Yellow Submarine	8,691
	Read more 🖉 Edit See :	12 Lucy in the Sky With Diamonds	8,624
ners	Targed as:	13 While My Guitar Gently Weeps	7,775
	classic rock, rock, british, 60s, pop	14 Can't Buy Me Love	7,574
al	See more	15 Strawberry Fields Forever	525
os			
	Similar Artists	Listeners	
		Scrobbling now om WMP	
	John Lennon George Paul Paul Ringo Starr Wings Paul	L& Linda Scrobbling now from The Last.fm	Scrobbler
	Harrison McCartney McCartney & Mc	See more Q	
		mirindA92	

biggaadayets Top Listener

dsendecki

michaelwsimpson

Top Listener

formundo

22

Profile

Library

Charts

Events

Friends

Neighbours

Groups

Journal

Tags



araine5

araine tayle

deathbyfailure

10minstolive Peter, 23, Male, United States Last track: Zero 7 - In the Waiting Line



Antiak

Andrew, 23, Male, United States Last track: Minus the Bear - Part 2



Crimson-Eyed

Alex, 22, Male, United Kingdom Last track: Grizzly Bear - Lullabye



deadwomb

ISHANK SHAKEDOWN, 20, Female, Canada Last track: MSTRKRFT - Breakaway f. Jahmal (The Carps)



currentdreams Jeremy, 19, Male, United States Last track: Gucci Mane - Socialite

Last track. Treija - The Mirac

United Kingdom



devilinthiscity

Monica, 20, Female, United States Last track: The Devil Wears Prada - Revive



DiaBoLiCcC Billy Deichmeister, 20, Male, United States III Listening: No Age - Glitter

dee to the ee en why ess, 20, Male, Canada

Last track: Crystal Castles - Pap Smear



dunkD

Last track: Method Man & Redman - Father's Day



Disco42

Gabriel, 18, Male, Canada Last track: Elliott Brood - Woodward Avenue.



kateapplin

Kate, 22, Female, Canada Last track: Rufus Wainwright - Slideshow





nomak559 Eric, 24, Male, United States Last track: Showbread - The Missing Wife



OMouse Rudolf, 23, Male, Canada Last track: David Bowie - Suffragette City









nomak559 Eric, 24, Male, United States Last track: Showbread – The Missing Wife



OMouse Rudolf, 23, Male, Canada Last track: David Bowie – Suffragette City



Profile		
Library Events Listening Now Loved by	· Friends	
Charts gio1984 Jorge		
Events The Pigeon Detective	es – I'm Always Right 2 hours ago	
Pixies – The Happen	ing Yesterday 1:13am	
Friends Kings of Leon – Tap	er Jean Girl - Explicit Yesterday 1:10am	
Neighbours	Whittle	
Groups	2 haur 200	
Europe - Cherokee	2 nouis ago	
Sour Hall	2 hours ago	
Tags	Track 2 hours ago	
gawibowo Gato	t Ari Wibowo	(The Carns)
Amorphis – Sky Is M	ine Shours ago	(The carps)
Amorphis – Elegy Me	adley (Against Widows / Cares / On Rich And Poor) 3 hours ago	
Amorphis – My Kant	ele 3 hours ago	
neglox		
Movie Sounds Unlim	ted – Battle Of The Heroes 4 hours ago	
Movie Sounds Unlim	ted – The World Is Not Enough - Main Title From James Bond 4 hours ago	
Movie Sounds Unlim	ted – Ballad Of Ira Hayes 4 hours ago	
Ruth_84 Ruth		
Keane – Stop For A	Minute 4 hours ago	
Keane – Is It Any W	onder? 4 hours ago	
Kate Bush – Wuther	ing Heights 4 hours ago	
willow_21 Parad	lise Seeker	
Lady Gaga – Fashio	n Shours ago	
In Flames – Cloud Co	nnected 5 hours ago	

Last track, showbread - the wissing vale







Profile	Friends Listening Now		
Library			
Charts	La Casa Azul – Galletas - Demo	0 1	
Events	See all loved tracks = Leave a shout		
Friends	Rammstein – Rammlied	n vour library (12 plays)	
Neighbours	Loved by Andiferum Yesterday 1:44am		
Groups	See all loved tracks 📮 Leave a shout		
Journal	Killem – Biolypse	¢ .∗	
Tags	Loved by srbardo Yesterday 8:42pm		
	Blind Guardian – War of the Thrones	in your library (1 play)	The Carps)
	Loved by Andiferum Sep 12:31am		
	Blind Guardian – A Voice In The Dark	In your library (3 plays)	
	Loved by Andiferum 14 Sep 12:31am		
	KISS – God Gave Rock n' Roll to You	Ø *	
	Loved by Andiferum 14 Sep 12:30am		
	See all loved tracks Cave a shout		
	💿 KISS – I Was Made For Lovin' You	In your library (3 plays)	
	Loved by Andiferum 14 Sep 12:30am		
	See all loved tracks 📮 Leave a shout		
	Zipi Zape – Sueños Inventados	(Q x)	
	Loved by Kasabian_Madrid 13 Sep 10:12am		
	Last track: Showbread - The Missing Wife	Last track: David Bowie – Suffragette City	





How can we address the problem?

- **RQ1**: Which sources of information in social systems are more valuable for recommendation?
 - Performance metrics
 - Precision
 - Recall
 - Discounted Cumulative Gain





2nd research question

Do recommenders in social systems really offer heterogeneous item suggestions, from which hybrid strategies could benefit?





How can we address this problem?

- **RQ2**: Do recommenders in social systems really offer heterogeneous item suggestions, from which hybrid strategies could benefit?
 - Non performance metrics
 - Coverage
 - Overlap
 - Diversity
 - Novelty





Methodology

- Implement different recommenders
 - Content-based (CB) \leftarrow collaborative tags
 - Collaborative-filtering (CF) ← track listenings
 - Social-based \leftarrow social contacts
- Evaluate the implemented recommenders
 - Performance metrics
 - Non-performance metrics





Evaluated recommenders

- Content-based recommenders (CB) ← collaborative tags
 - TF-based recommender
 - BM25-based recommender
 - TF-IDF cosine-based recommender
 - BM25 cosine-based recommender
- Collaborative filtering recommenders (CF) ← track listenings
 - User-based recommender (N=15)
 - Item-based recommender
- Social recommenders \leftarrow social contacts
 - Social recommender: friends as neighbours
 - Social+CF recommender





Performance metrics

- Precision
 - Recommended items that are relevant for the user
 - P@N (considering items in the top N results)
- Recall
 - Relevant items that are recommended
 - R@N (considering items in the top N results)
- Discounted cumulative gain
 - Relevant items should appear higher in the result list





Non-performance metrics (I)

- Coverage
 - Fraction of items a recommender can provide predictions for
 - E.g., CF cannot deal with new items, CB with untagged items, ...
- Diversity
 - (Relevant) Items recommended that are not very popular nor very unpopular
 - Other diversity definitions have to be investigated
- Novelty
 - Relevant but non popular items
 - Other novelty definitions have to be investigated





Non-performance metrics (II)

- Overlap
 - Proportion of (relevant) recommended items provided by two recommenders
 - Two metrics: Jaccard-based, Ranking-based
- Relative diversity
 - (Relevant) Items recommended by a recommender once the user has already seen another result list





Evaluation protocol

- 1. Split the track set for each user (5-fold cross validation)
 - 80% for training set
 - 20% for test set
- 2. Build recommenders using training set
- 3. Evaluate all recommenders for each user:
 - 3.1. Predict a score for all items in the test set
 - 3.2. Rank the items according to the predicted score
 - 3.3. Compute performance and non-performance metrics





Results (I)

- Performance values
 - Best: CB
 - Worst: user based-CF (too much sparsity)
- Non performance values
 - Best coverage: CB
 - Highest diversity: social
 - Highest novelty: social / CF

Recommender	MAP	NDCG
BM25 Cosine	0.014	0.212
TF-IDF Cosine	0.012	0.220
User based CF	0.002	0.076

Recommender	Coverage	Diversity	Novelty
BM25 Cosine	0.017	0.015	0.003
TF-IDF Cosine	0.017	0.018	0.004
User based CF	0.015	0.005	0.001
Social	0.013	0.054	0.005





Results (I) – New experiments!

Decommondor	Coverage	D:		Novelty	Recommen			
Recommender	Coverage		ersity	noverty	BM25 Cos			
BM25 Cosine	0.208	3.67		5.66	TE IDE Cosina			
TF-IDF Cosine	0.208		3.88	5.74		OF	0.012	0.220
User based CF	0.061	(5.65	6.27	User based	CF	0.002	0.076
Social	0.074	(5.72	6.26				
Item based CF	0.008	ial	2.75	6.97		_		
			Recommender		Coverage	Div	ersity	Novelty
BM25 Cosine		0.017	0.	015	0.003			
TF-IDF Cosine		0.017	0.	018	0.004			
User based CF		0.015	0.	005	0.001			
				ocial	0.013	0.	054	0.005





Results (II)

- Non performance values (cont'd)
 - Overlap: only among CBs and between CF and social
 - Not too much between social and CF
 - Cosine seems to be more influential than the weighting function

Jaccard overlap	TF	BM25	BM25 Cosine	TF-IDF Cosine
TF		0.005	0.005	0.009
BM25			0.011	0.008
BM25 Cosine				0.015
TF-IDF Cosine				

- Relative diversity: only among CBs and between CF and social
 - Not conclusive, further analysis required





Conclusions

- **RQ1**: Which sources of information in social systems are more valuable for recommendation?
 - **Tags** provide very effective recommendations
- **RQ2**: Do recommenders in social systems really offer heterogeneous item suggestions, from which hybrid strategies could benefit?
 - Yes! And each source of information captures a different characteristic
 - Tags \rightarrow Coverage
 - − Friends \rightarrow Diversity
 - Track listenings \rightarrow Novelty





Future work

- Use the obtained results and conclusions to build hybrid recommenders
 - Well performing, with good coverage, offering diverse and novel item suggestions... (a perfect recommender?)
 - Every source of information has to be used
- Compare the non performance metric definitions with others in the literature
 - Check different approximations for our definitions
- Extend our empirical study
 - Different datasets
 - More recommenders





Thank you





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Research questions

- RQ1. Which sources of information available in social systems are more valuable for recommendation?
 - Performance metrics (precision and recall)

- RQ2. Do recommendation approaches exploiting different sources of information in social systems really offer heterogeneous item suggestions, from which hybrid strategies could benefit?
 - Non-performance metrics (coverage, overlap, diversity and novelty)





Performance metrics (with definitions)

- Precision
 - Recommended items that are relevant for the user
 - P@N (considering items in the top N results)

precision = |relevant items retrieved| |retrieved items|

- Recall
 - Relevant items that are recommended
 - R@N (considering items in the top N results)

 $recall = \frac{|relevant items retrieved|}{|relevant items|}$

- Discounted cumulative gain
 - Relevant items should appear higher in the result list

NDCG_N =
$$N_k \sum_{m=1}^k \frac{2^{rel(d_m)} - 1}{\log_2(1+m)}$$





Non-performance metrics (I)

- Coverage
 - Fraction of items a recommender can provide predictions
 - E.g., CF cannot deal with new items, CB with untagged items, ...

$$cvg(a) = \frac{|S_a|}{|I|}$$
 $cvg^R(a) = \frac{|S_a^R|}{|\bigcup_{u \in U} R_u|}$

- Diversity
 - (Relevant) Items recommended which are not very popular nor very unpopular

$$div(a) = \frac{1}{|\mathcal{U}|} \sum_{u \in \mathcal{U}} div_u(a) \qquad p_{u,i} = \frac{\sum_{a \in \mathcal{A}} \delta(a, u, i)}{|\mathcal{A}|},$$
$$div_u(a) = H_u(a) = -\sum_{i \in \overline{S}_{a,u}^R} p_{u,i} \cdot \log p_{u,i} \qquad \text{where } \delta(a, u, i) = 1 \text{ iff } \in \overline{S}_{a,u}^R, \text{ and } 0 \text{ otherwise.}$$

- Novelty
 - Relevant but non popular items

$$p_i = \frac{|\{v \in \mathcal{U}: i \in R_v\}}{|\mathcal{U}|}$$

$$nov(a) = \frac{1}{|\mathcal{U}|} \sum_{u \in \mathcal{U}} nov_u(a)$$
$$nov_u(a) = H_u(a) = -\sum_{i \in \overline{S}_{a,u}^R} p_i \cdot \log p_i,$$





Non-performance metrics (II)

• Overlap

- Proportion of (relevant) recommended items provided by two recommenders
- Two metrics: Jaccard-based, Ranking-based

$$ove_jacc(a,b) = \frac{1}{|\mathcal{U}|} \sum_{u \in \mathcal{U}} ove_jacc_u(a,b) \qquad ove_rank(a,b) = \frac{1}{|\mathcal{U}|} \sum_{u \in \mathcal{U}} ove_rank_u(a,b)$$
$$ove_jacc_u(a,b) = \frac{\left|\overline{S}_{a,u}^R \cap \overline{S}_{b,u}^R\right|}{\left|\overline{S}_{a,u}^R \cup \overline{S}_{b,u}^R\right|} \quad ove_rank_u(a,b) = \frac{1}{N} \sum_{i \in \overline{S}_{a,u}^R \cap \overline{S}_{b,u}^R} \left(1 - \frac{\left|\tau_{a,u}(i) - \tau_{b,u}(i)\right|}{N - 1}\right)$$

- Relative diversity
 - (Relevant) Items recommended by a recommender once the user has already seen another result list

$$div(a,b) = \frac{1}{|\mathcal{U}|} \sum_{u \in \mathcal{U}} div_u(a,b) \qquad p_{a,u,i} = \frac{1}{\left|\overline{S}_{a,u}^R\right|}$$
$$div_u(a,b) = H_u(a|b) = \sum_{i \in \overline{S}_{a,u}^R \cap \overline{S}_{b,u}^R} p_{a,u,i} \cdot \log \frac{p_{a,u,i}}{p_{b,u,i}} \qquad p_{a,u,i} = \frac{1}{\left|\overline{S}_{a,u}^R\right|}$$





Notation

Let R_u be the set of items relevant for user u, and let A be the set of recommendation algorithms to be evaluated.

We define $L_{a,u}$, the ranked list of recommendations provided to user u by algorithm $a \in A$, as:

$$L_{a,u} = \{(u, i, \tau): i \in I, \tau > 0\},\$$

where τ is the ranking position of item *i* in the recommendation list based on the predicted item utility $g_a(u, i)$, having $\tau_{a,u}(i) < \tau_{a,u}(j) \Rightarrow g_a(u, i) \ge g_a(u, j), \forall i, j \in I$.

We denote by $S_{a,u}$ the set of items that belong to $L_{a,u}$:

$$S_{a,u} = \left\{ i: (u, i, \cdot) \in L_{a,u} \right\}$$

Finally, we define $S_{a,u}^R$ as the set of those items belonging to $S_{a,u}$ that are relevant for user u. That is:

$$S_{a,u}^{R} = S_{a,u} \cap R_{u} = \{i: (u, i, \cdot) \in L_{a,u}, i \in R_{u}\}$$

The previous definitions $S_{a,u}$ and $S_{a,u}^R$ for a given recommendation algorithm *a* are extended to consider all users with the following expressions:

$$S_a = \bigcup_{u \in U} S_{a,u}, \ S_a^R = \bigcup_{u \in U} S_{a,u}^R$$

Since some of the non-performance metrics explained below only depend on the top N recommendations provided by each algorithm a, we define $\overline{S}_{a,u}$, $\overline{S}_{a,u}^R$, \overline{S}_a and \overline{S}_a^R as, respectively, $S_{a,u}$, $S_{a,u}^R$, S_a and S_a^R on the set $L_{a,u}^N$ of top N recommendations for user u, where:

$$L_{a,u}^N = \left\{ (\cdot, \cdot, \tau) \in L_{a,u}, \tau \le N \right\}$$





Evaluated recommenders (I)

- Content-based recommenders
 - TF-based recommender

$$g(u_m, i_n) = tf_u(u_m, i_n) = \frac{\sum_{l:i_{n,l}>0} tf_{u_m}(t_l)}{\max_{u \in \mathcal{U}, t \in \mathcal{T}} (tf_u(t))}$$

• BM25-based recommender

$$g(u_m, i_n) = bm25_u(u_m, i_n) = \sum_{(l|i_{n,l}>0)} bm25_{u_m}(t_l)$$

• TF-IDF cosine-based recommender

$$g(u_m, i_n) = \cos_{tf \text{-}idf}(u_m, i_n) = \frac{\sum_l tf_{u_m}(t_l) \cdot iuf(t_l) \cdot tf_{i_n}(t_l) \cdot iif(t_l)}{\sqrt{\sum_l \left(tf_{u_m}(t_l) \cdot iuf(t_l)\right)^2} \cdot \sqrt{\sum_l \left(tf_{i_n}(t_l) \cdot iif(t_l)\right)^2}}$$

BM25 cosine-based recommender

$$g(u_m, i_n) = \cos_{bm25}(u_m, i_n) = \frac{\sum_l \left(bm25_{u_m}(t_l) \cdot bm25_{i_n}(t_l) \right)}{\sqrt{\sum_l \left(bm25_{u_m}(t_l) \right)^2} \cdot \sqrt{\sum_l \left(bm25_{i_n}(t_l) \right)^2}}$$





Evaluated recommenders (II)

- Collaborative filtering recommenders
 - User-based recommender (N=15)

 $g(u_m, i_n) = C \sum_{v \in N[u_m, k]} sim(u_m, v) \times rat(v, i_n)$ N[u_m, k] denotes the set (with size k) of neighbours of u_m

$$sim(u,v) = \frac{\sum_{i} (rat(u,i) - \overline{rat}(u)) (rat(v,i) - \overline{rat}(v))}{\sqrt{\sum_{i} (rat(u,i) - \overline{rat}(u))^{2}} \sqrt{\sum_{i} (rat(v,i) - \overline{rat}(v))^{2}}}$$

• Item-based recommender

$$g(u_m, i_n) = C \sum_{j \in I_m} sim(i_n, j) \times rat(u, j)$$

where I_m is the set of items rated by user u_m





Evaluated recommenders (III)

- Social recommenders
 - Only social recommender: friends as neighbours

 $N[u_m, k] = N[u_m] = \{v \in \mathcal{U}: v \text{ is friend of } u_m\}$

• Social+CF recommender

 $N[u_m, k] = \{v \in \mathcal{U}: v \text{ is friend of } u_m\} \cup \{v \in \mathcal{U}: sim(u_m, v) \ge \rho_m\}$

where $\rho_m > 0$ is the minimum similarity to be satisfied between the active user and his/her most similar neighbours





Results (II) – New experiments!

- Non performance values (cont'd)
 - Overlap: only among CBs and between CF and social
 - Not too much between social and CF

BM25 cosine compares the best

- Cosine is more influential than the weighting function
- Relative diversity: only among CBs and between CF and social

Jaccard overlap	TF	BM2 5	BM25 Cosine	TF-IDF Cosine
TF		0.26	0.26	0.44
BM25	-		0.30	0.26
BM25 Cosine				0.39
TF-IDF Cosine				

TF-IDF Cosine	Relative diversity	TF	BM25	BM25 Cosine	TF-IDF Cosine
0.44	TF		-0.04	0.08	0.15
0.26	BM25	0.02		0.07	0.05
0.39	BM25 Cosine	-0.18	-0.27		-0.29
	TF-IDF Cosine	-0.36	-0.15	0.16	



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