

Recommender Systems

Ing. Tomáš Řehořek

tomas.rehorek@fit.cvut.cz



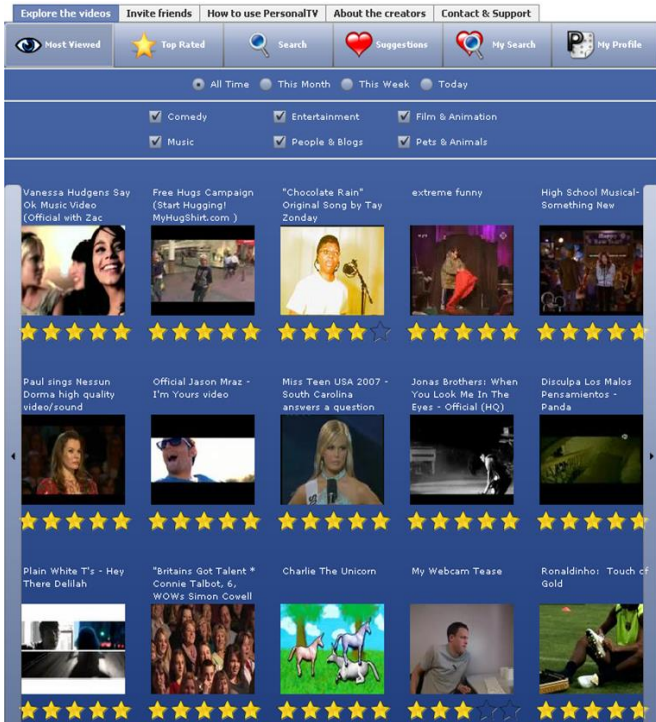
Computational Intelligence Group (CIG)
Department of Theoretical Computer Science
Faculty of Information Technology (FIT)
Czech Technical University (CTU) in Prague



Modgen, s.r.o.

Recommender Systems: Quick Intro

- Software systems providing **suggestions** for **items** to be of use to a **user**
 - Based on her past behavior (item purchases, ratings with stars, views...)
 - Based on behavior of other users of the system



Content-based Recommendation

- Uses meta-data about users/items in the system
- Attempts to recommend items similar to those liked by the user in the past

Collaborative Filtering

- Most prominent family of algorithms in Recommender Systems
- Analyses **purchase/rating history** in the whole userbase, recommends items liked by similar users

Collaborative Filtering

	Movie 1	Movie 2	Movie 3	...	Movie n
User 1	2	?	?		?
User 2	?	5	4		?
User 3	4	?	1		3
⋮					
User m			2		

INPUT

- Set of **users** $U = \{u_1, \dots, u_m\}$,
- Set of **items** $I = \{i_1, \dots, i_n\}$,
- Set of **ratings** $R = \{r_1, \dots, r_k\}$
 - $r_i \in U \times I \times \mathbb{R}$ (explicit ratings) or
 - $r_i \in U \times I \times \{0,1\}$ (implicit ratings)
- **Target user** $u_x \in U$

OUTPUT

- **Top-N recommendations**
 - Set $\{i_1^{rec}, \dots, i_N^{rec}\} \subset I$ of items that the user will most likely to appreciate
- **Rating predictions**
 - Predict ratings of unknown couples from $U \times I$

Our Research

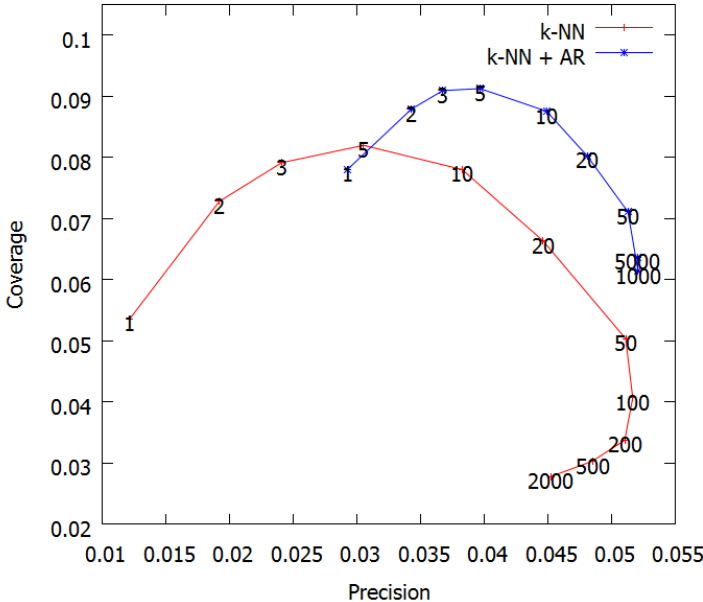
II. Matrix-Based Algorithms

- User-based k -Nearest Neighbors
- Item-based k -Nearest Neighbors
- Matrix Factorization

UserID	Item1	Item2	Item3	Item4	Item5	Item6	Item7	...
1	?	3	?	?	4	1	3	...
2	3	?	4	?	5	?	1	...
3	?	2	5	5	3	1	?	...
4	2	?	?	?	2	?	5	...
5	3	?	?	5	4	?	2	...
6	1	4	?	4	?	3	?	...
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮

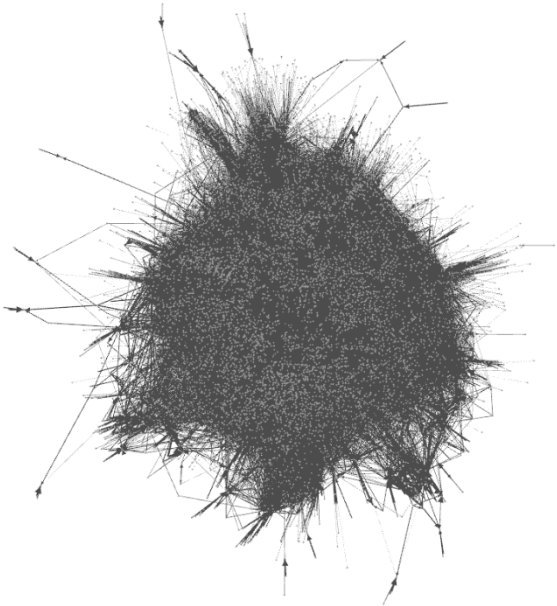
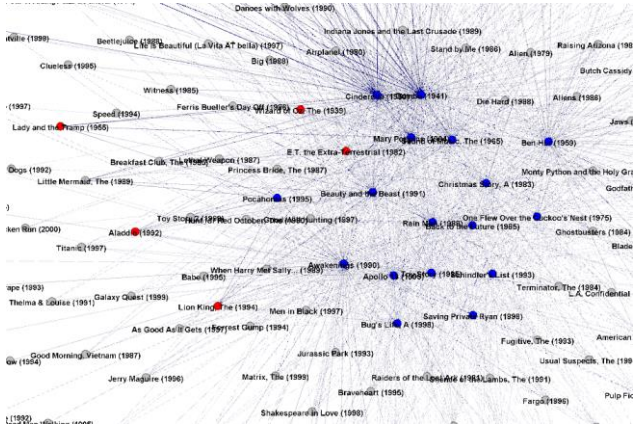
I. Multi-Objective Optimization

- Trade-offs between the accuracy and the diversity of items being recommender



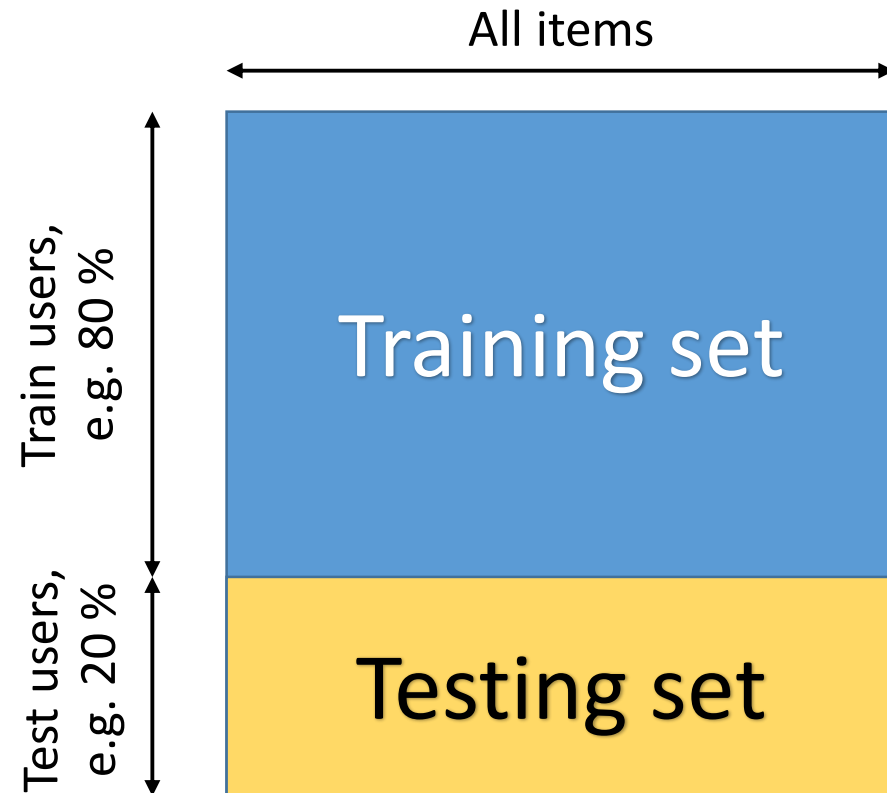
III. Rule-Based and Graph-based Algorithms

- Association Rules
- Sequential Patterns



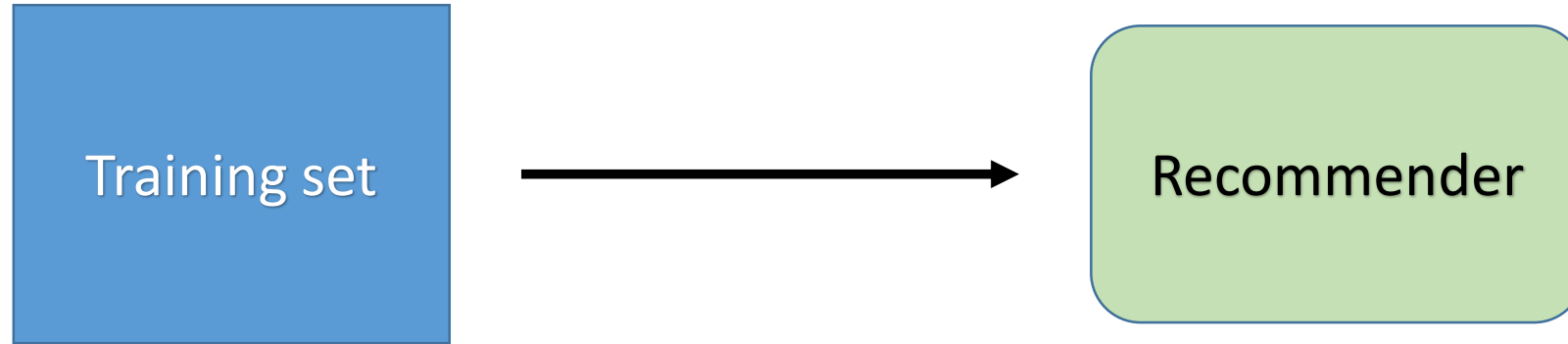
Split Validation of a Recommender System

- To validate the Recommender, one may divide the **users** into:
 - **Training set**, which is fully submitted to the Recommender
 - **Testing set**, submitted only partially and used to evaluate the Recommender



Split Validation of a Recommender System

- From the **training set**, all the items are submitted to the system

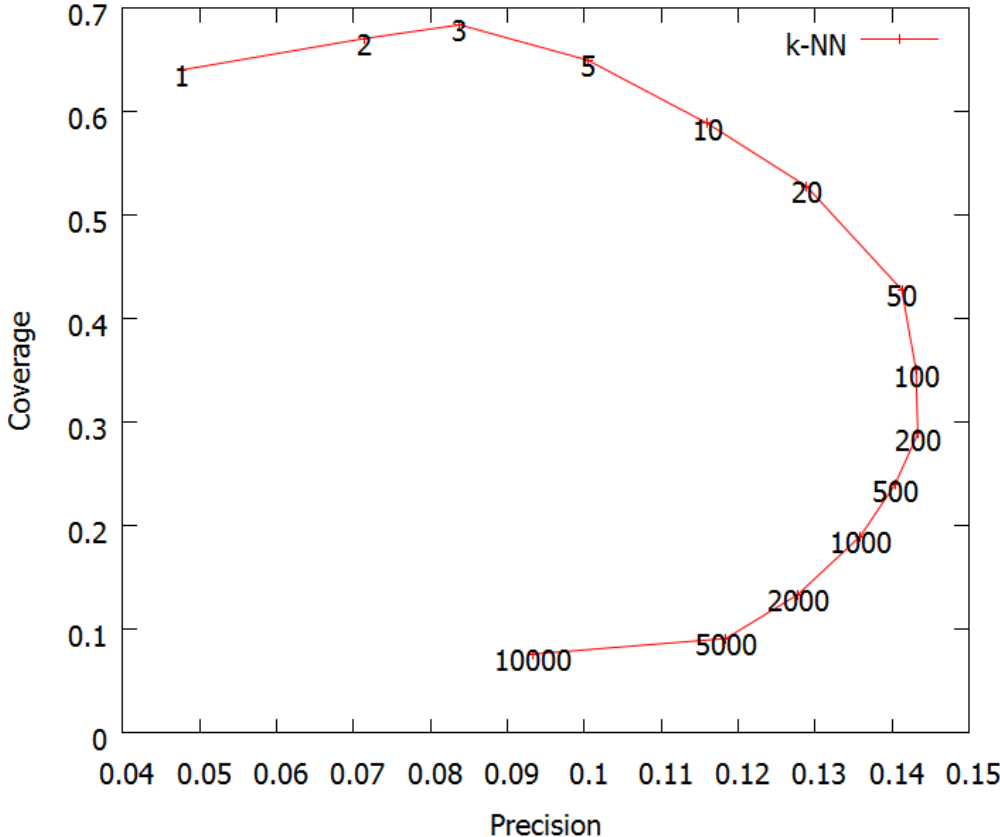


- From the **testing set**, for **each user**, the set of ratings is divided into:
 - **Observation subset** – the ratings/purchases **submitted to the system**
 - **Testing subset** – the ratings/purchases used to **evaluate the system**

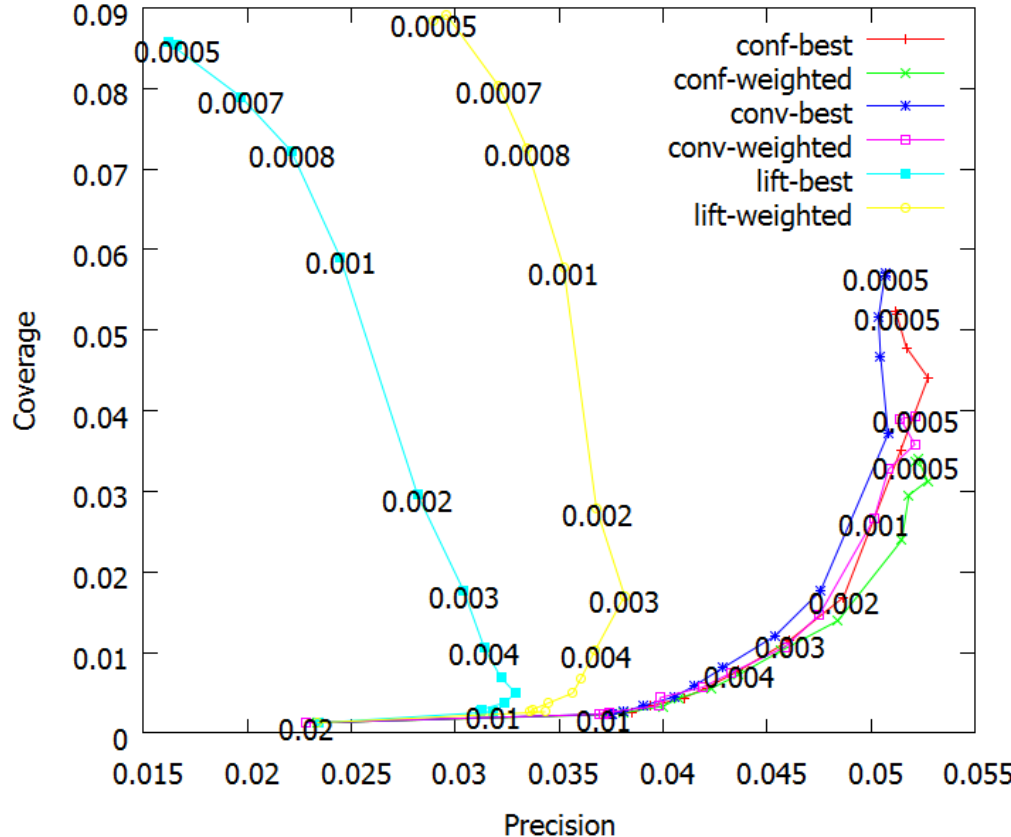


Our Research: Multi-Objective Optimization

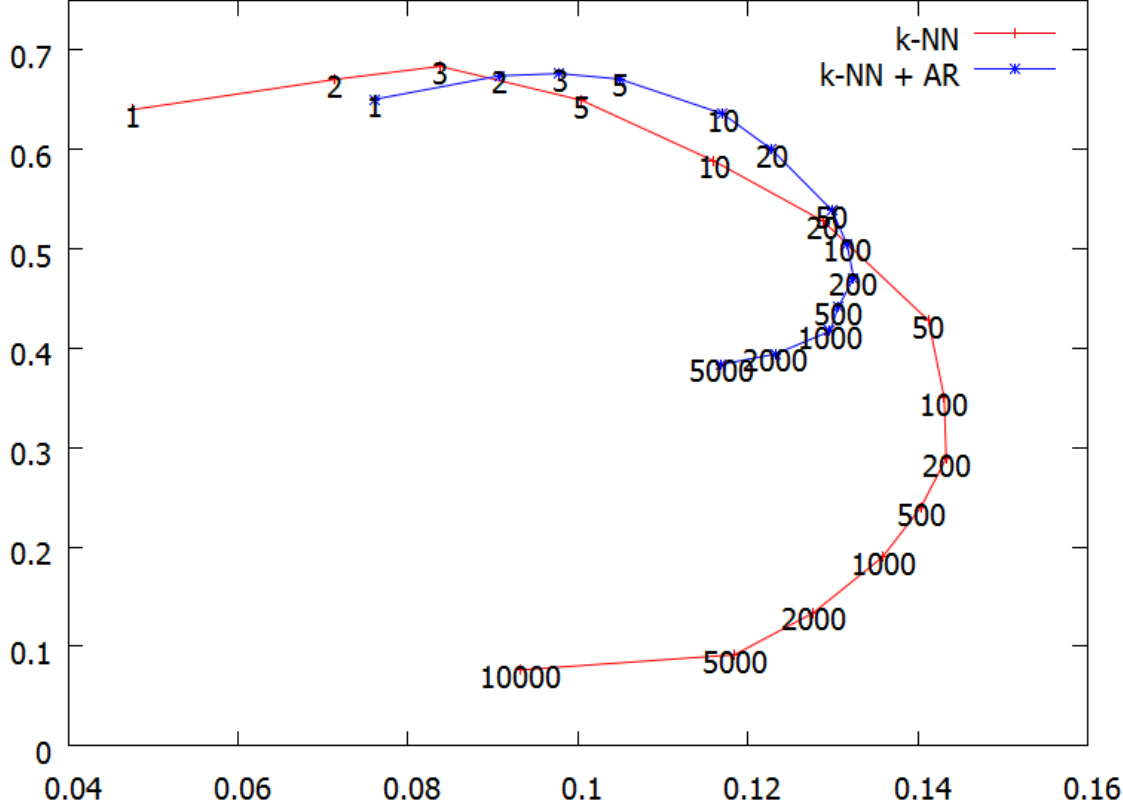
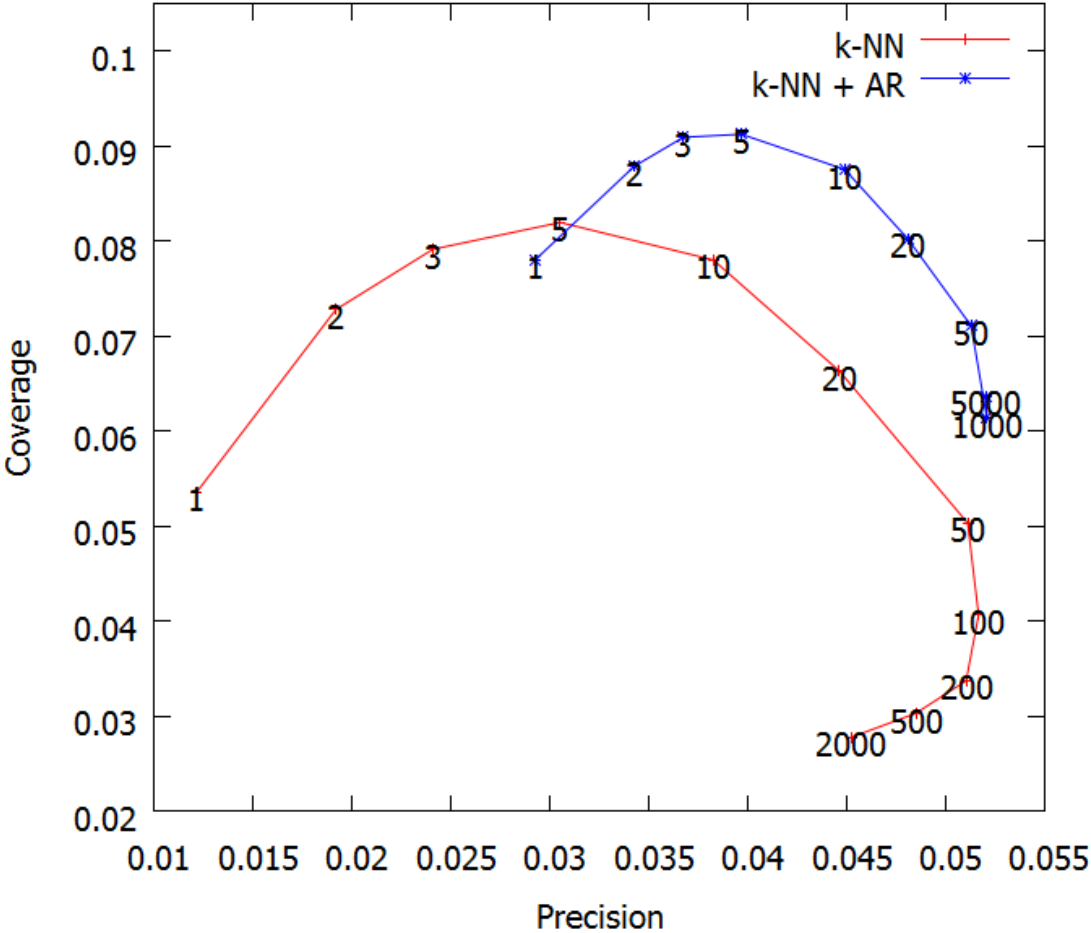
k-Nearest Neighbors



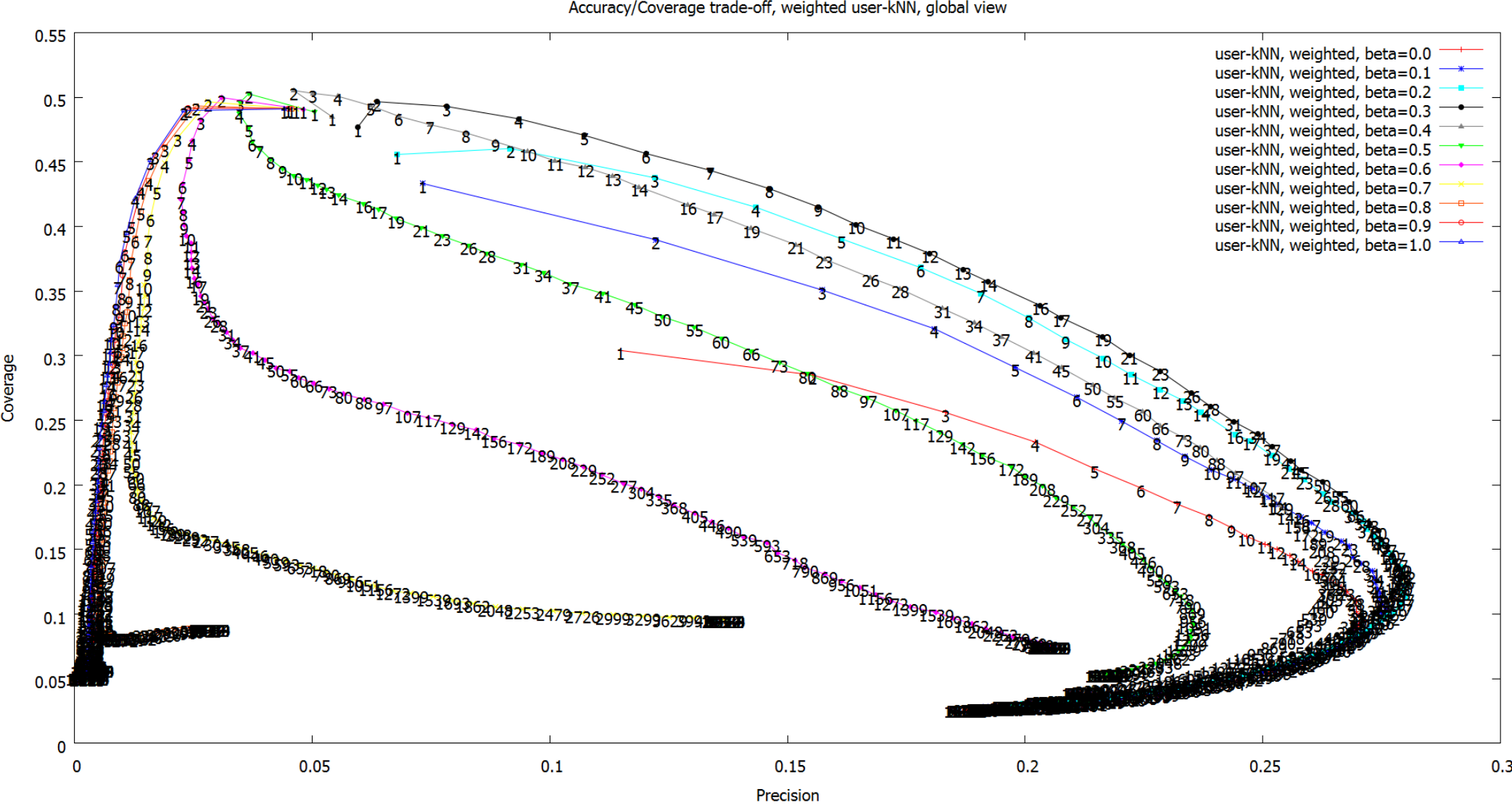
Association Rules



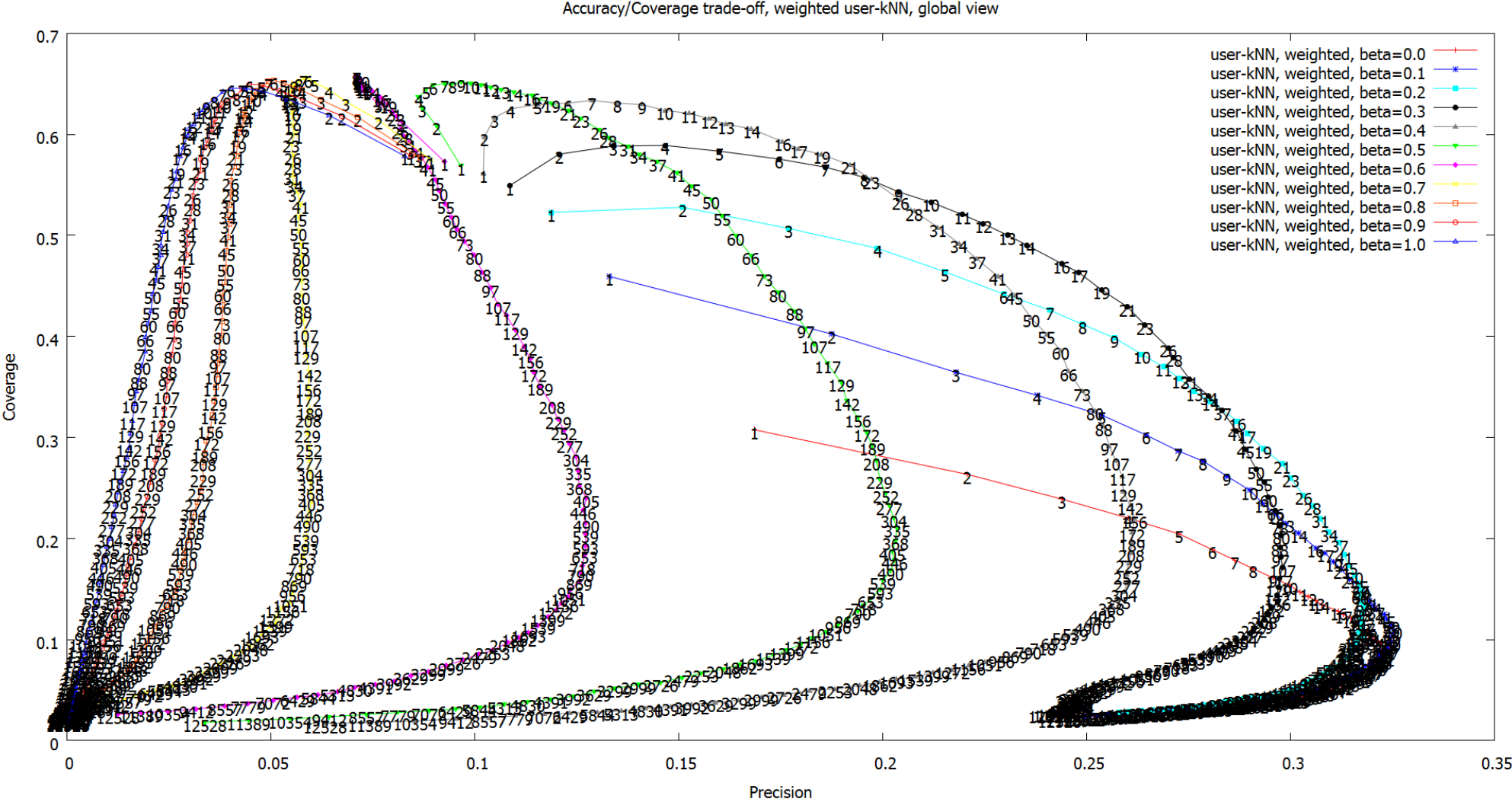
Our Research: Model Ensembles



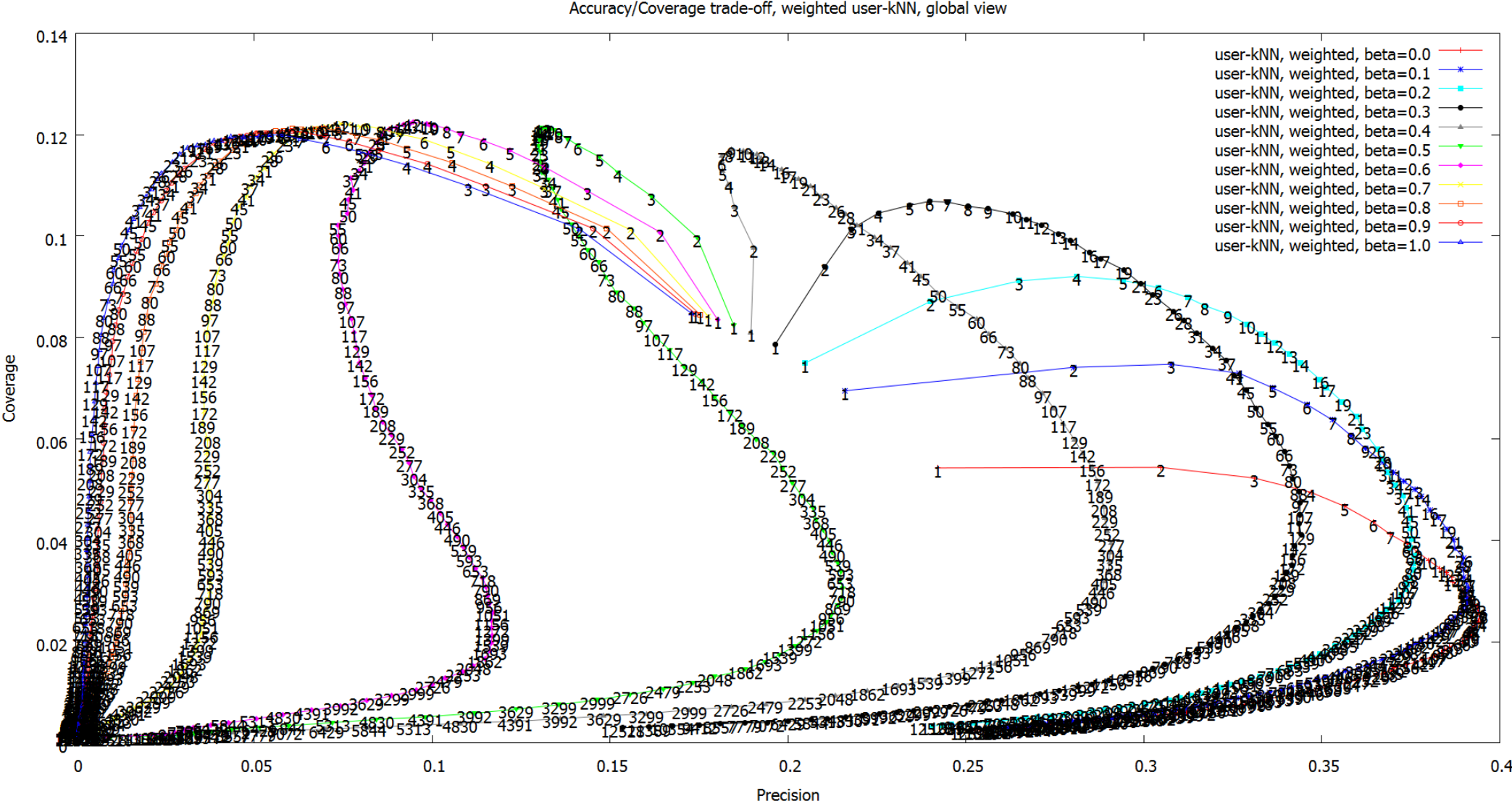
Our Research: Bestseller Penalization (MovieLens1M)



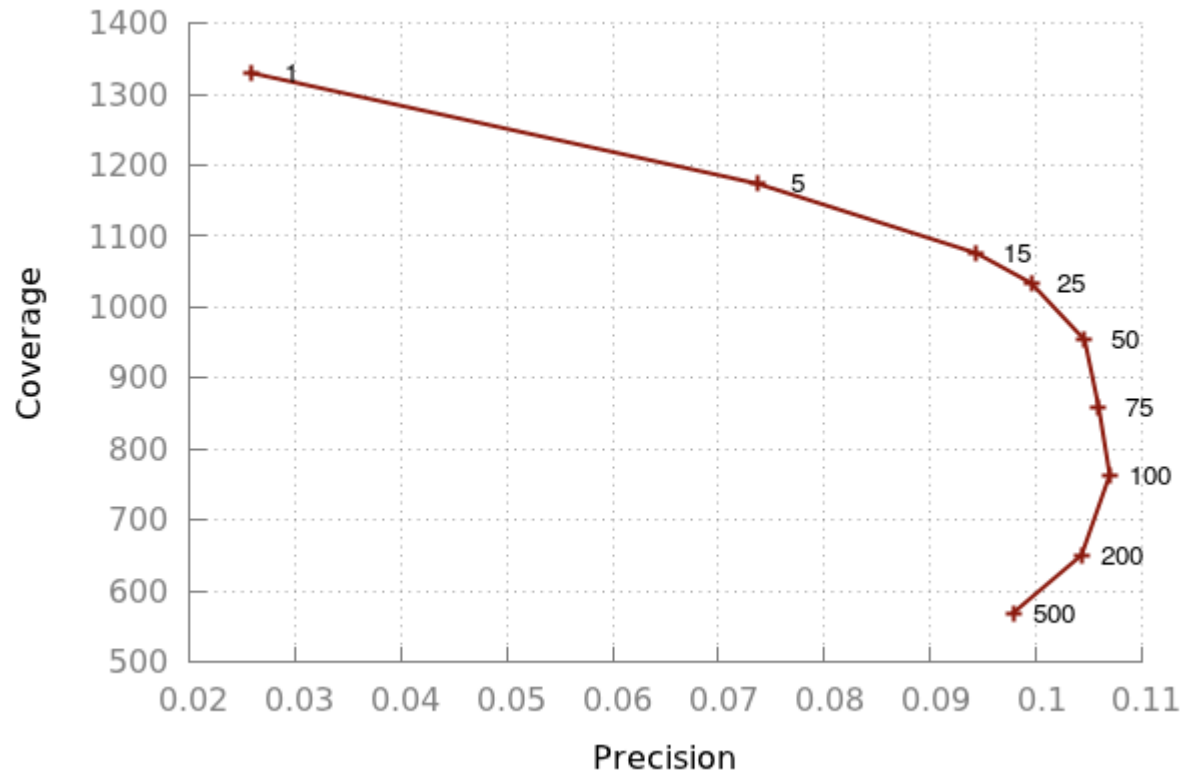
Our Research: Bestseller Penalization (MovieLens10M)



Our Research: Bestseller Penalization (Libimseti.cz)



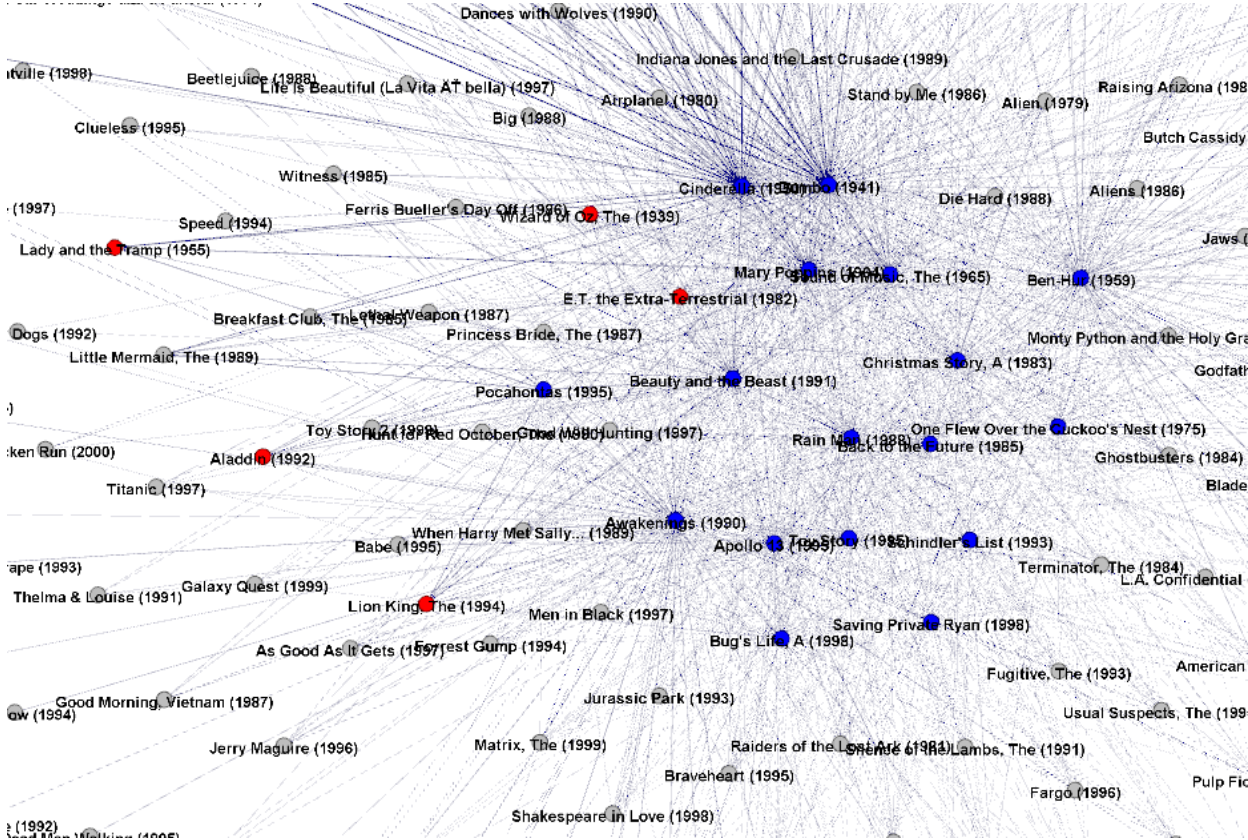
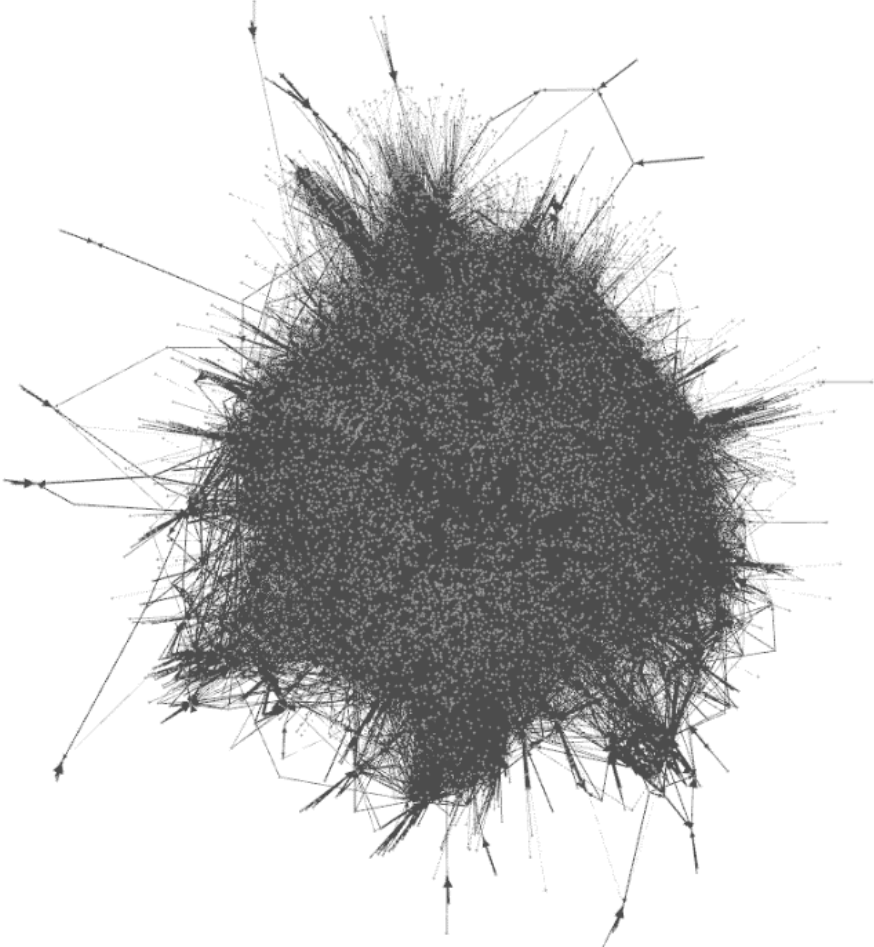
Business Impact?



2. n25	339 (8.79%)
3. n100	209 (5.42%)
4. n500	207 (5.37%)
5. n5	120 (3.11%)
6. old	99 (2.57%)

isthereanydeal.com

Live demo: Visualizing Association Rules







Recommendation API

- Recommendation as a cloud service
- Domain-independent
 - IPTV, VoD, Eshops, Cultural events, Sport facilities...
- Fast and scalable implementation of CF algorithms
 - Parallel Association Rules
 - Parallel k-NN
 - Parallel Matrix Factorization
- Real-time model updates
- Adjusting recommendation to fit business needs
 - Filtering, Boosting

Live demo: Boosting and Filtering

<https://modgen.net:8766/goout/>

User 86118115275

Items purchased

Count: 19

!itemId	!Purchase time	!Total purchases	name	schedule	score	tag	type
event-163790	2013-11-21 17:47:29	690	Lucie, větší než malé množství lásky	1395615540.0	None	{muzikal, 'divadlo', 'show'}	event
event-165498	2013-11-21 17:47:53	42	Strakonický dudák	1400194740.0	None	{drama, 'divadlo'}	event
event-17737	2013-11-21 17:48:49	323	Vražda v salónním coupé	1386889140.0	None	{top, 'komedie', 'divadlo'}	event
event-17749	2013-11-21 17:49:03	308	Drobečky z perníku	1392677940.0	None	{divadlo, 'tragikomedie'}	event
event-17737	2013-11-21 17:50:47	323	Vražda v salónním coupé	1386889140.0	None	{top, 'komedie', 'divadlo'}	event
event-5810	2013-11-21 17:51:03	131	39 stupňů	1386889140.0	None	{drama, 'divadlo'}	event
event-212343	2013-11-21 17:51:31	8	Zahradní slavnost	1386889140.0	None	{drama, 'divadlo'}	event
event-184929	2013-11-21 17:52:10	225	Jedenácté přikázání	1403737140.0	None	{komedie, 'divadlo'}	event
event-184929	2013-11-21 17:53:47	225	Jedenácté přikázání	1403737140.0	None	{komedie, 'divadlo'}	event
event-5810	2013-11-21 17:53:50	131	39 stupňů	1386889140.0	None	{drama, 'divadlo'}	event
event-100659	2013-11-21 17:53:59	560	Kapka medu pro Verunku	1401663540.0	None	{pro-deti, 'muzikal', 'pohadka', 'divadlo'}	event
event-166156	2013-11-21 17:54:05	232	Sejdeme se Na Cibulce	1387321140.0	None	{debata, 'divadlo', 'show'}	event
event-137165	2013-11-21 17:54:31	185	Růžové brýle	1391900340.0	None	{drama, 'divadlo'}	event
event-134719	2013-11-21 17:54:47	21	Viny	1390258740.0	None	{drama, 'divadlo', 'show'}	event
event-8363	2013-11-21 17:54:57	31	Příběhy ze Starého zákona II	1386889140.0	None	{drama, 'divadlo', 'literatura'}	event
event-111031	2013-11-21 17:55:08	256	Aspects of Alice	1388444340.0	None	{cerne, 'divadlo'}	event
event-140055	2013-11-21 17:55:35	155	Faust – Muž mezi Bohem a Ďáblem	1393109940.0	None	{cerne, 'divadlo'}	event
event-209912	2013-11-21 17:56:41	68	Strauss Gala	1386889140.0	None	{klasicka, 'koncerty'}	event
performer-15694	2013-11-21 17:56:41	174	Symfonický orchestr Českého rozhlasu	None	0	{klasicka, 'kapely'}	performer

Items recommended

Filter

```
'schedule' > 1386279637 and "komedie" not in 'tag'
```

Booster

```
if "drama" in 'tag' then 2 else 1
```

Submit

Result

Count: 10

Time: 155.82 ms

!itemId	!Total purchases	name	schedule	score	tag	type
event-176612	105	Dva	1390863540.0	None	{drama, 'divadlo'}	event
event-5147	47	César a Drana	1387234740.0	None	{drama, 'divadlo'}	event
event-5368	30	Šoa	1390517940.0	None	{drama, 'divadlo'}	event
event-11271	30	Polední úděl	1390863540.0	None	{drama, 'divadlo'}	event
event-198430	81	Vánoční koncert Strauss gala	1386889140.0	None	{klasicka, 'koncerty'}	event
event-108968	40	Amerikana III	1400799540.0	None	{balet, 'tanec', 'divadlo'}	event
event-142810	113	Babička	1390690740.0	None	{drama, 'divadlo'}	event
event-202294	29	Ateliér Open Space 1	1387493940.0	None	{workshop, 'divadlo'}	event
event-8444	66	Carmen	1399330740.0	None	{opera, 'divadlo'}	event
event-30166	42	Česká mše vánoční – Dětská opera Praha	1387666740.0	None	{klasicka, 'divadlo', 'koncerty'}	event

Parallel Computing Challenges in Recommender Systems

- Online Recommender Systems must satisfy many tough requirements:
 - Handling click-streams of **millions of users** in **real-time**
 - 100 ms to build recommendations based on database of 1.000.000 users?
 - Datasets for online portals are as large as 250 GB of compressed rating data!

Sample problems

Find Nearest Neighbors

Given vector $u \in \mathbb{R}^n$ and set $U = \{u'_1, \dots, u'_m\}$, find subset of k vectors from U which are most similar to u

Find Frequent Itemsets

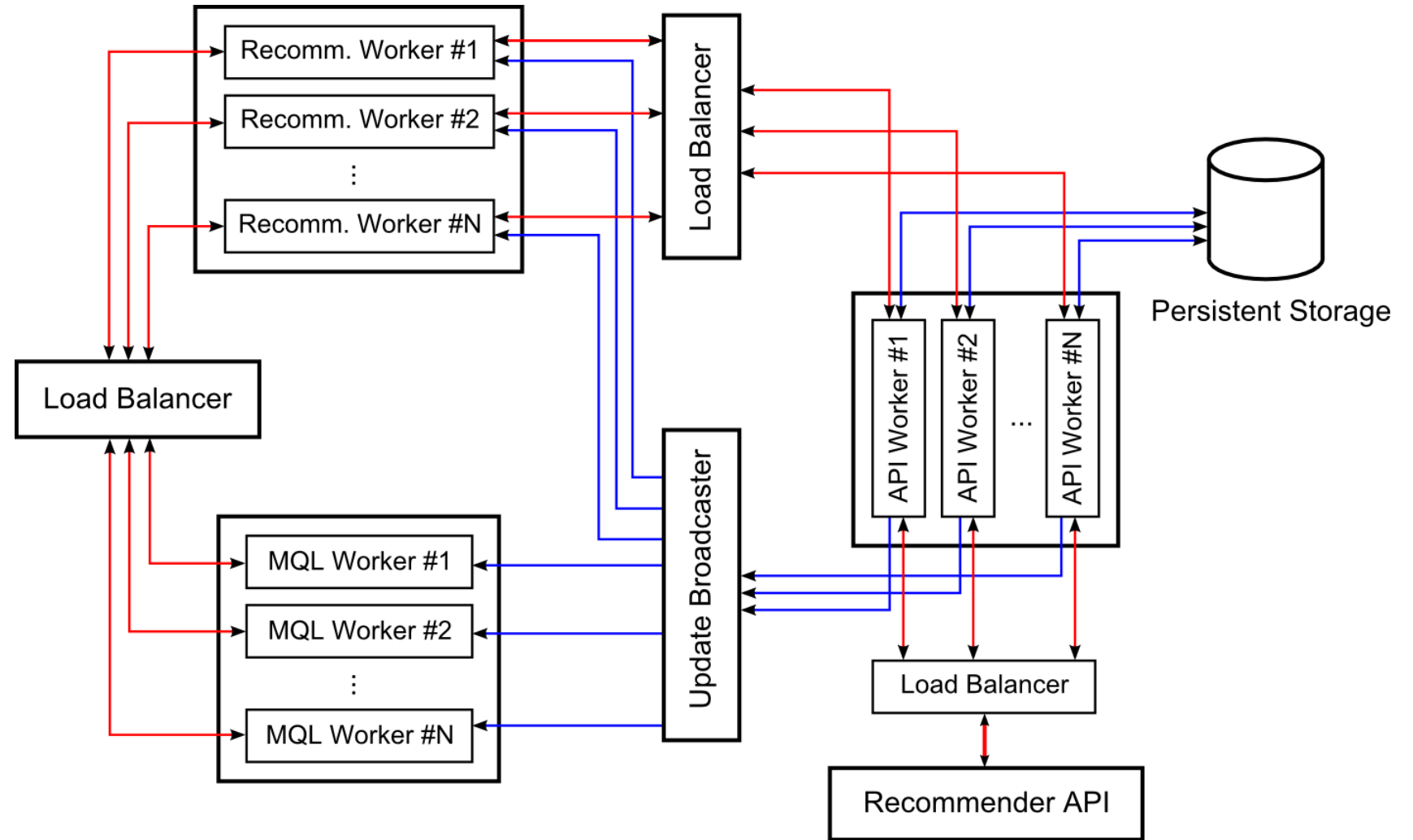
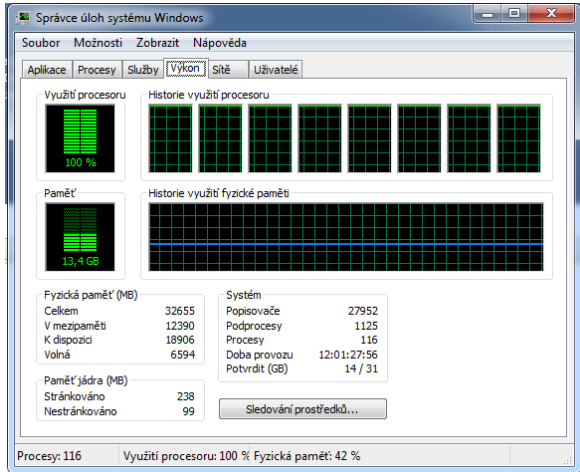
Given sets $I = \{i_1, \dots, i_n\}$ and $U = \{u_1, \dots, u_m\}$, $u_i \subseteq I$, find set $X \subseteq 2^I$ such that $x \in X$ iff

$$\frac{|\{u \in U \mid x \subseteq u\}|}{|I|} \geq s_{min}$$

for some $s_{min} \in [0,1]$

- for $n, m = 10^6$ or even more (Youtube, last.fm)
- within small, fixed amount of time (several 10s of milliseconds?)

Parallel Computing: Modgen Architecture



Thank you for your attention!

Ing. Tomáš Řehořek
tomas.rehorek@modgen.net