Apache S2Graph(incubating) At Kakao A Large Scale Distributed Graph Database



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1. Technical Challenges.

2. What is S2Graph.

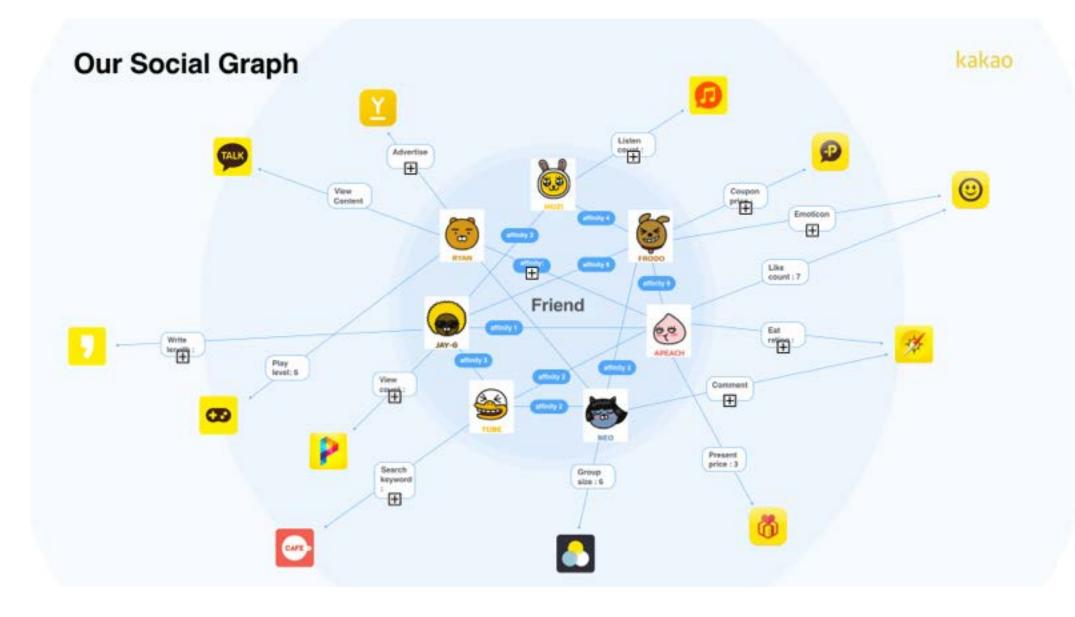
3. Use Cases.

4. Real World Use Cases.

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Technical Challenges

Why We Got into Graph Databases



Graph Databases

- **Relationship-Oriented**
- **Expressive yet Simple**



+30 services at Kakao are powered by S2Graph

Our Data

- **Highly Connected**
- **Complex Relations**



Social Graph?

Relations + Activities = Social Graph

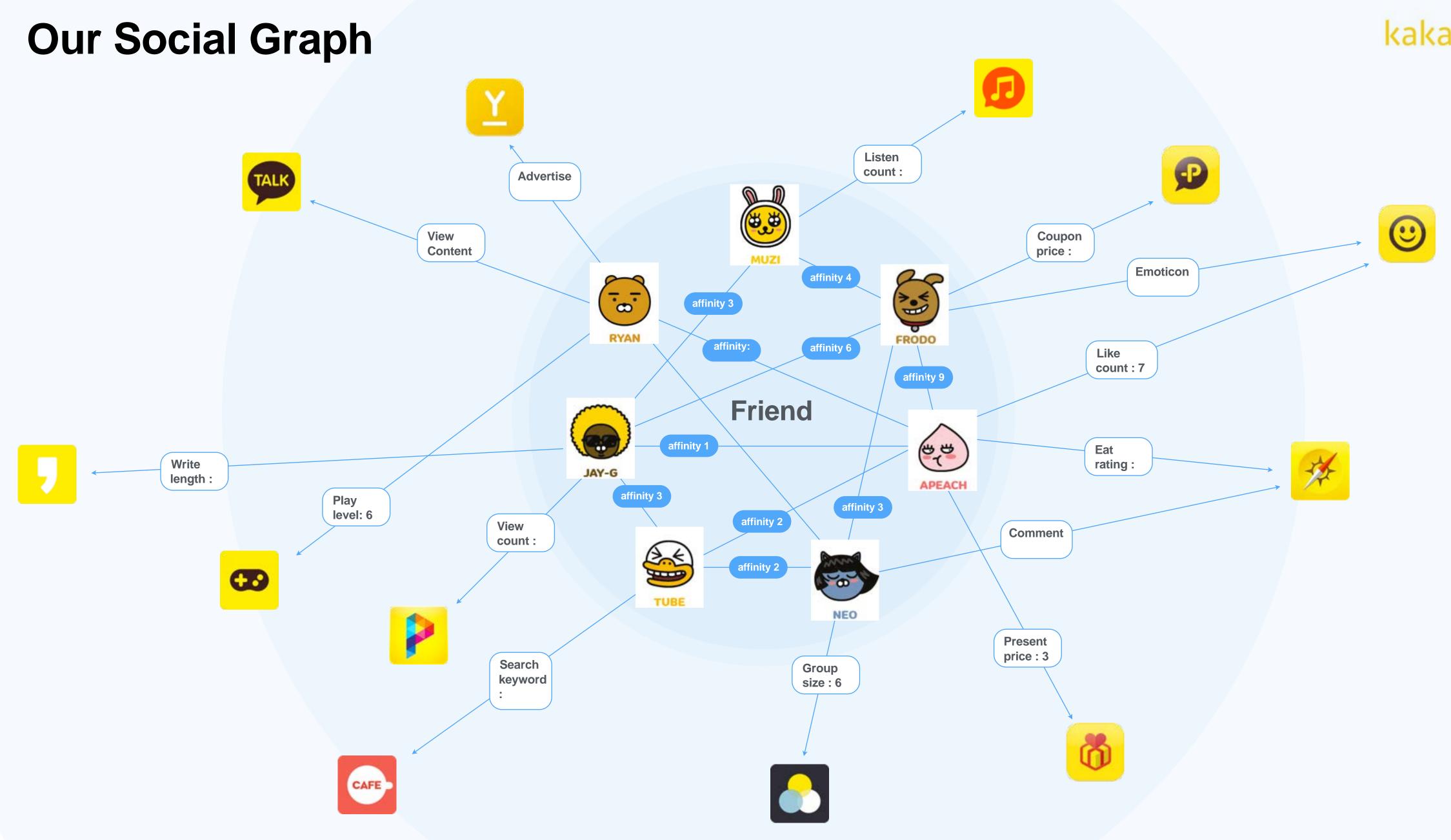
















Technical Challenges

1.Large graph constantly changing.

- 1. Total # of Edge: + 1 trillion and growing.
- 2. Social Network: more than 10 billion edges, 200 million vertices, 50 million realtime update on social network.
- 3. User activities(Click, Like, Share, Buy): 2.5 ~ 3 billion real-time incoming edges, 50 billion batch processed edges.

2. Low latency for breadth first search traversal on connected data.

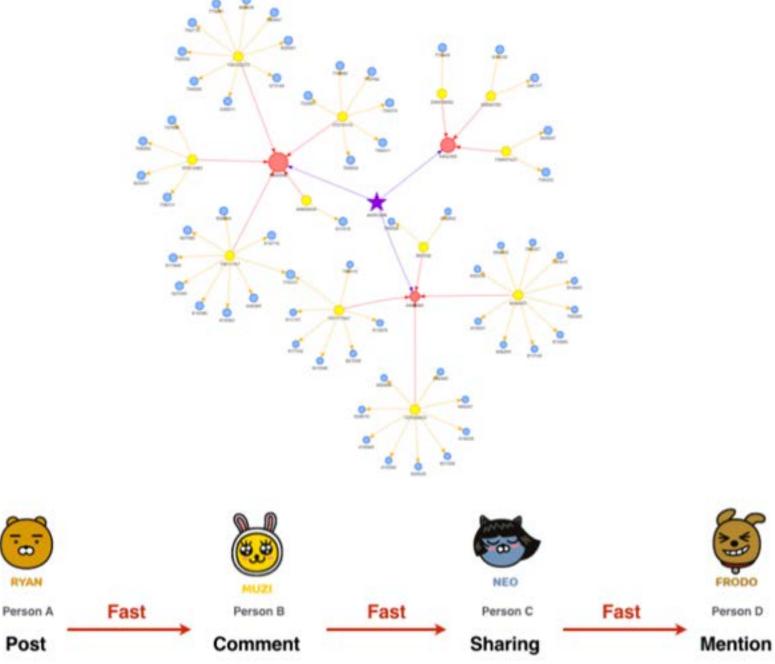
- 1. Peak graph-traversing query per minute: 4 million
- 2. Average response time: 50 ms

3. Update should be applied into query result in real-time.

4. Support for Dynamic Ranking logic

- 1. push strategy: hard to change data ranking logic dynamically.
- 2. pull strategy: can try various data ranking logic





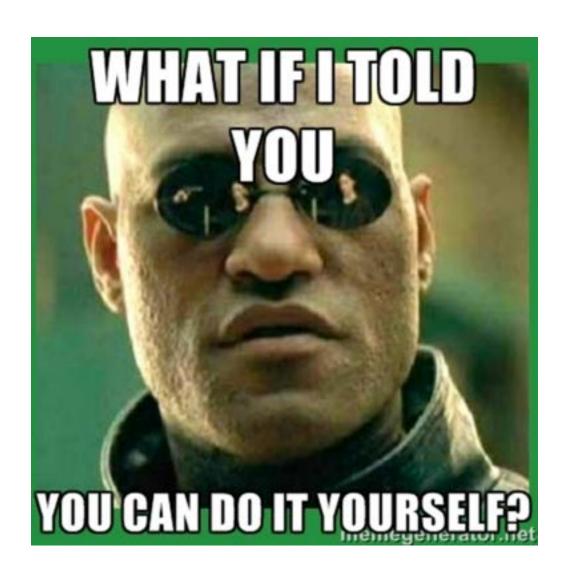
Why Did We Build It?

Existing solutions weren't performant enough for our needs.

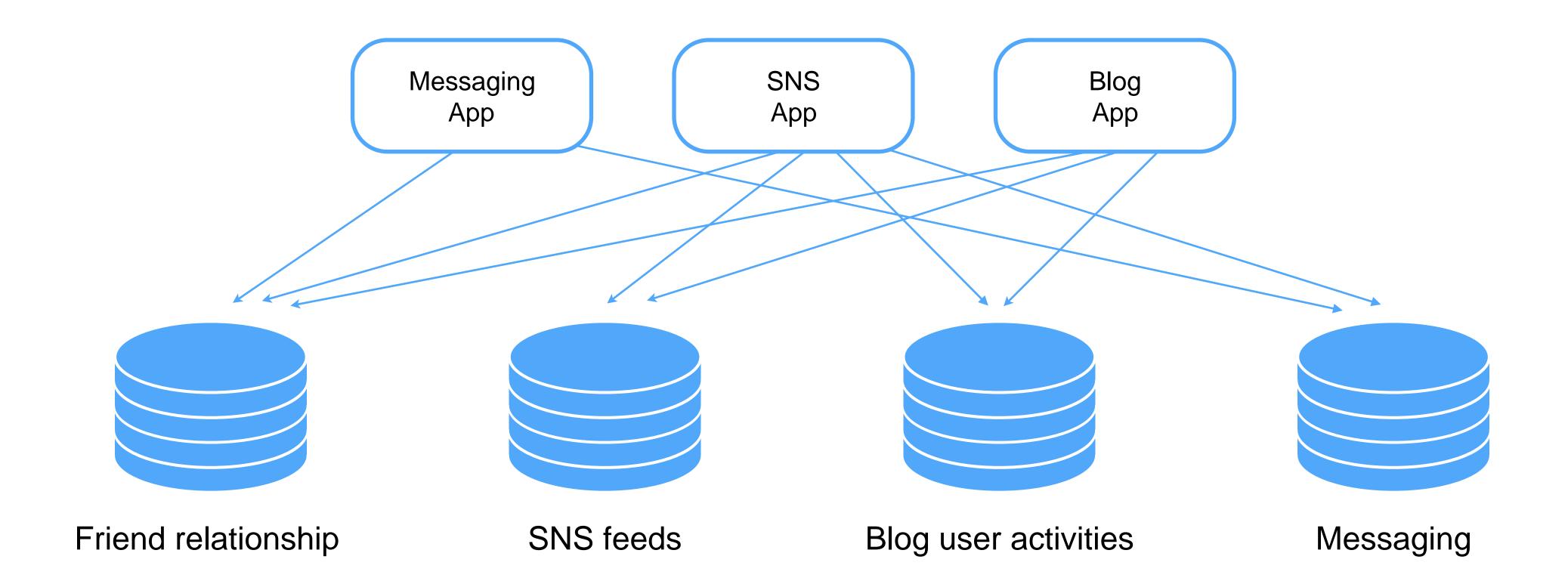
Especially,

Maintaining a mutable graph at scale was not supported. (i.e. Updates/ Deletes were slow!)
 Breadth First Search traversal was not fast enough.

So we built our own!

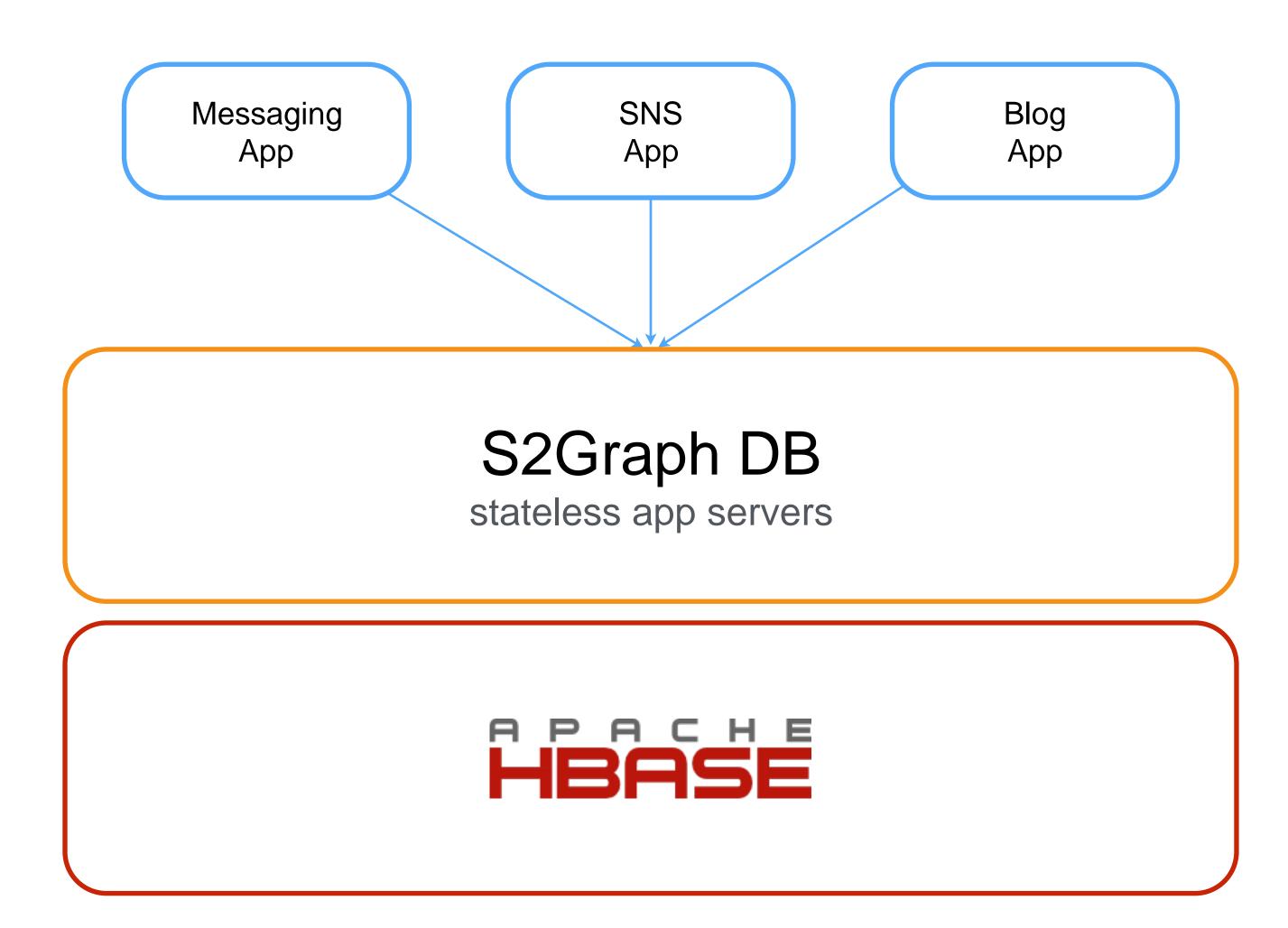


Before



- Each app server should know each DB's sharding logic.
- Highly inter-connected architecture

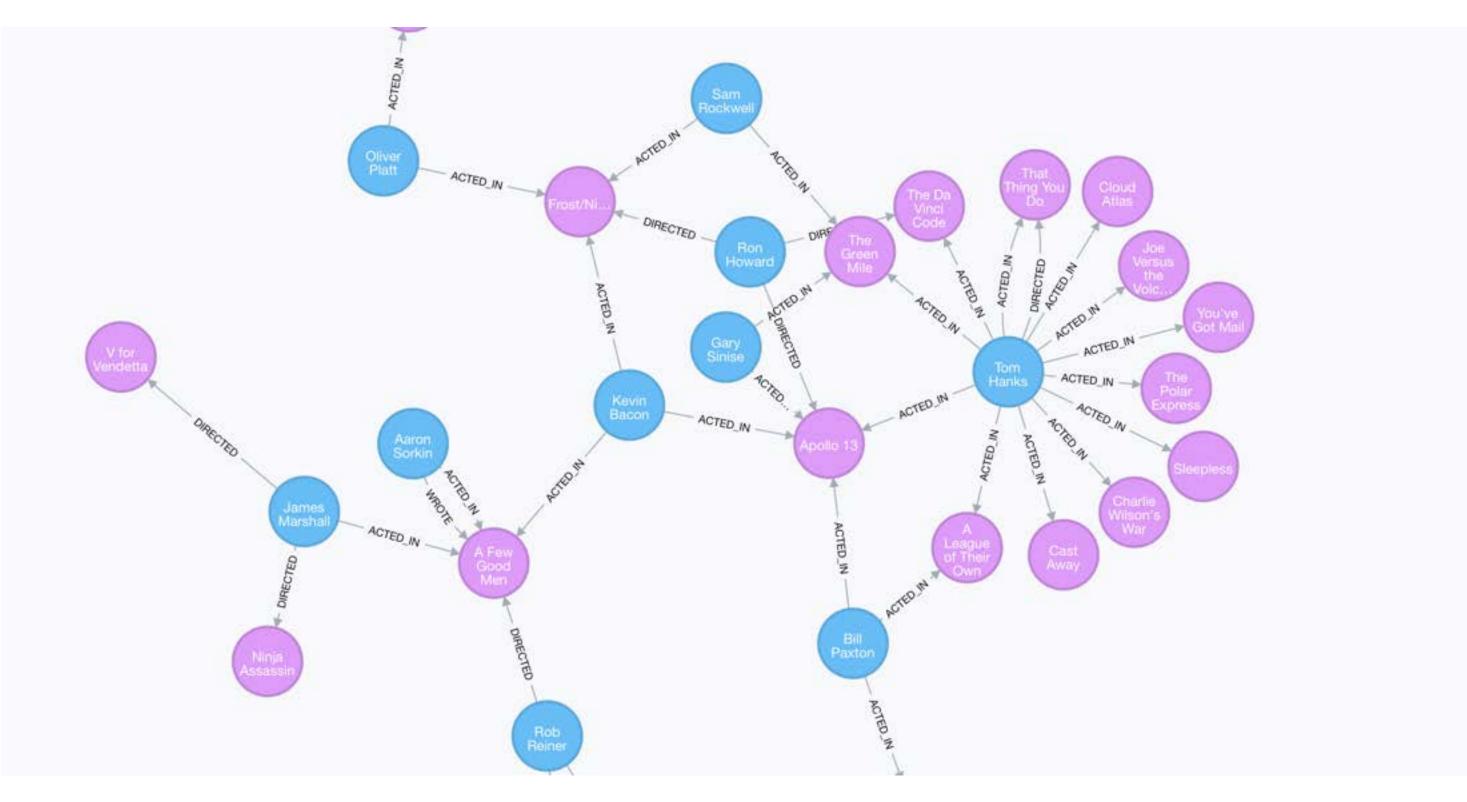
After



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What is S2Graph?

What is S2Graph?



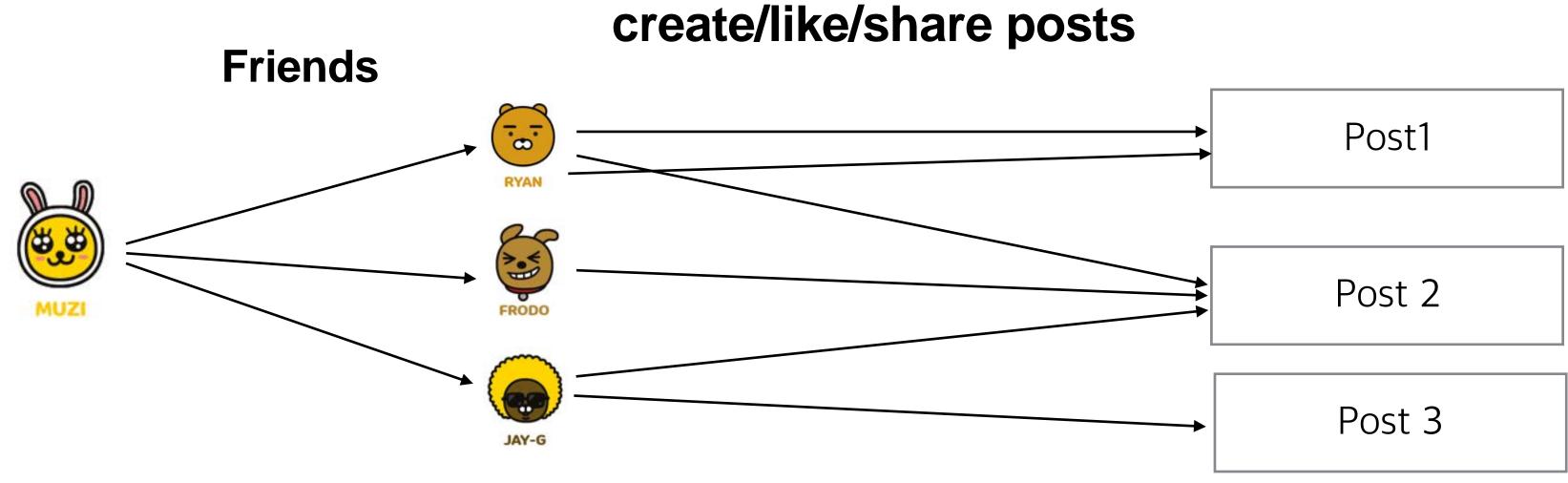
A property graph of actor Kevin Bacon (http://propellerhead.ca/2016/02/fun-with-graph-databases)

- Property Graph Model: Vertices + Edges + Properties
- S2Graph = Property Graph Model + Scalability + Fast CRUD Operations
- Graph-processing layer atop HBase

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- Properties bility + Fast CRUD Operations

Example: News Feed (cont)



Posts that my friends interacted.

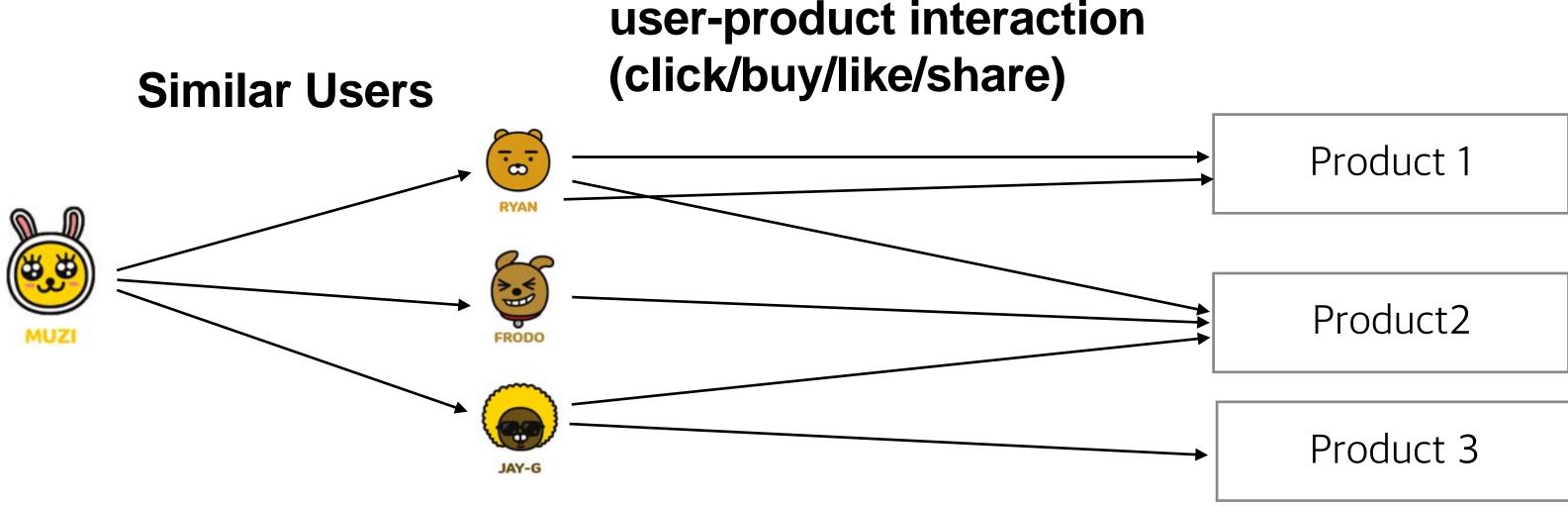
SELECT a.*, b.*	curl –XP
FROM friends a, user_posts b	{
WHERE a.user_id = b.user_id WHERE b.updated_at >= yesterday	"src
and b.action_type in ('create', 'like', 'share')	"ste
	[{
	[{
]
	}
	1. A.

```
KPOST localhost:9000/graphs/getEdges _H 'Content_Type: Application/json' _d '
```

```
cVertices": [{"serviceName": "s2graph", "columnName": "user_id", "id":1}],
teps": [
[{"label": "friends", "direction": "out", "limit": 100}], // step
[{"label": "user_posts", "direction": "out", "limit": 10, "where": "created_at >= yesterday"}]
```



Example: Recommendation(User-based CF) (cont)



Products that similar user interact recently.

SELECT a.* , b.* FROM similar_users a, user_products b WHERE a.sim_user_id = b.user_id AND b.updated_at >= yesterday



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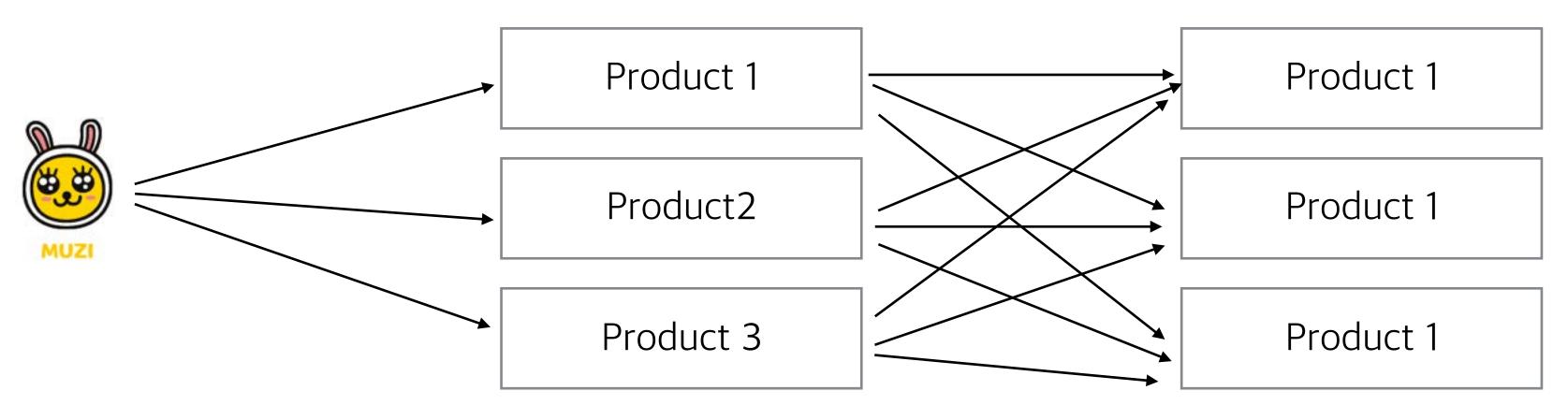
curl -XPOST localhost:9000/graphs/getEdges -H 'Content-Type: Application/json' -d '

```
"filterOut": {"srcVertices": [{"serviceName": "s2graph", "columnName": "user_id", "id": 1}],
  "steps": [[{"label": "user_products_interact"}]]
```

```
"srcVertices": [{"serviceName": "s2graph", "columnName": "user_id", "id":1}],
  [{"label": "similar_users", "direction": "out", "limit": 100, "where": "similarity > 0.2"}], // step
  [{"label": "user_products_interact", "direction": "out", "limit": 10,
    "where": "created_at >= yesterday and price >= 1000"}]
```

Example: Recommendation(Item-based CF) (cont)

user-product interaction (click/buy/like/share)



Products that are similar to what I have interested.

SELECT a.* , b.* FROM similar_ a, user_products b WHERE a.sim_user_id = b.user_id AND b.updated_at >= yesterday

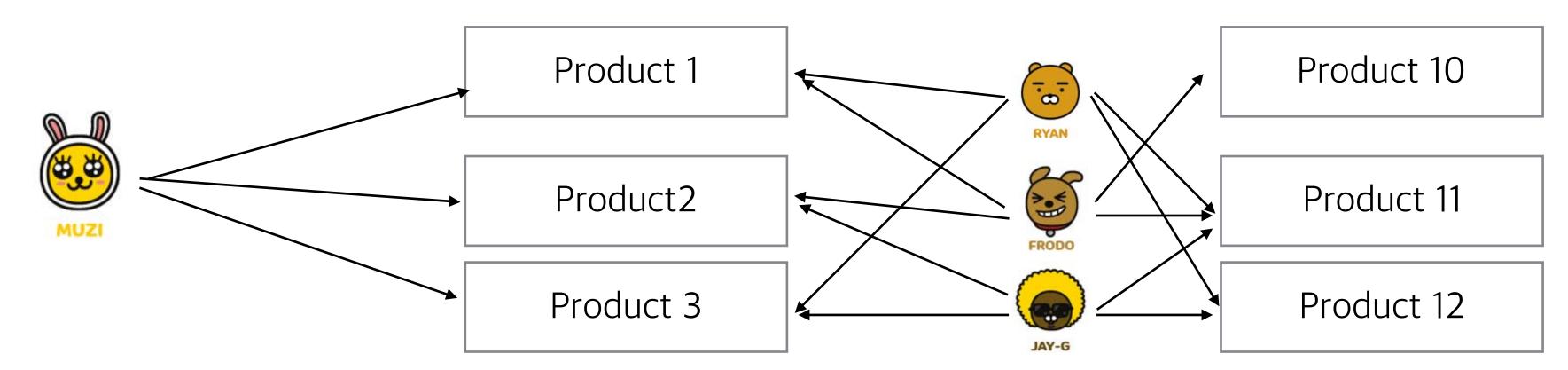
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Similar Products

curl -XPOST localhost:9000/graphs/getEdges -H 'Content-Type: Application/json' -d '

Example: Recommendation(Spreading Activation) (cont)





Products that is interacted by users who interacted on products that I interact

```
SELECT b.product_id, count(*)
FROM user_products a, user_products b
WHERE a.user_id = 1
AND a.product_id = b.product_id
GROUP BY b.product_id
```

```
curl -XPOST localhost:9000/graphs/getEdges -H 'Content-Type: Application/json' -d '
   "srcVertices": [{"serviceName": "s2graph", "columnName": "user_id", "id":1}],
   "steps": [
     [{"label": "user_products_interact", "direction": "out", "limit": 100, "where": "created_at >= yesterday and price >= 1000"}],
       [{"label": "user_products_interact", "direction": "in", "limit": 10, "where": "created_at >= today"}],
       [{"label": "user_products_interact", "direction": "out", "limit": 10, "where": "created_at >= 1 hour ago"}],
```

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user-product interaction (click/buy/like/share)



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Use Cases

1. Storage for user activities and relationships

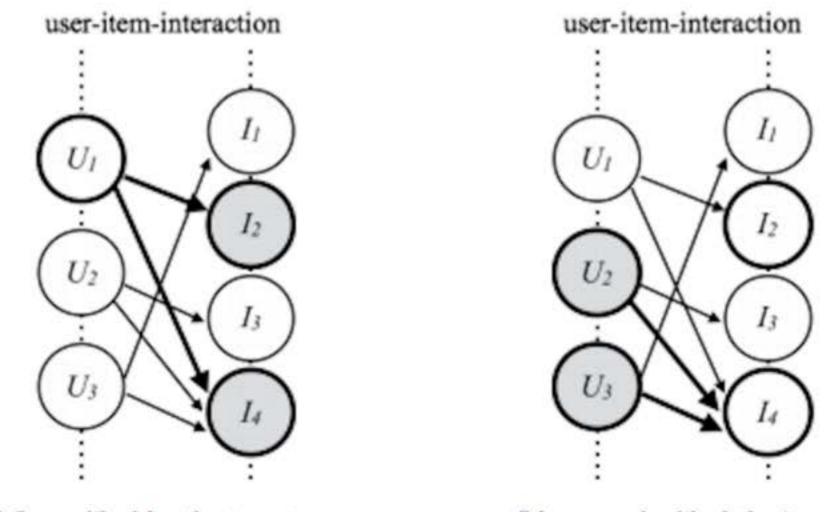
Friends, clicks, purchases, likes, shares, comments, etc.





2. Real-time Recommendation (Spreading Activation)

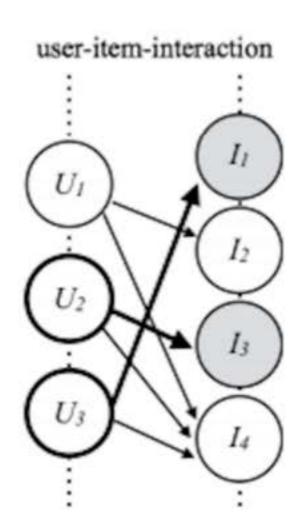
1. Find items a user has reacted to (clicked, purchased..) 2. Find other users who reacted to the same items. **3.** Find other items that those users reacted to.



(a) Items liked by the target user

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(b) users who liked the items



(c) items liked by the users

Figure 5: Real-time collaborative filtering by Spreading Activation

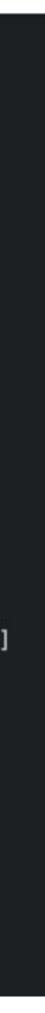
2. Real-time Recommendation (cont)

Composite multiple queries via weighted sum

EX) Ensemble of recommendation algorithms

1.Item-based collaborative filtering (CF) 2.User-based CF **3.Matrix Factorization** 4.Demographical MP items (for cold-start problem)

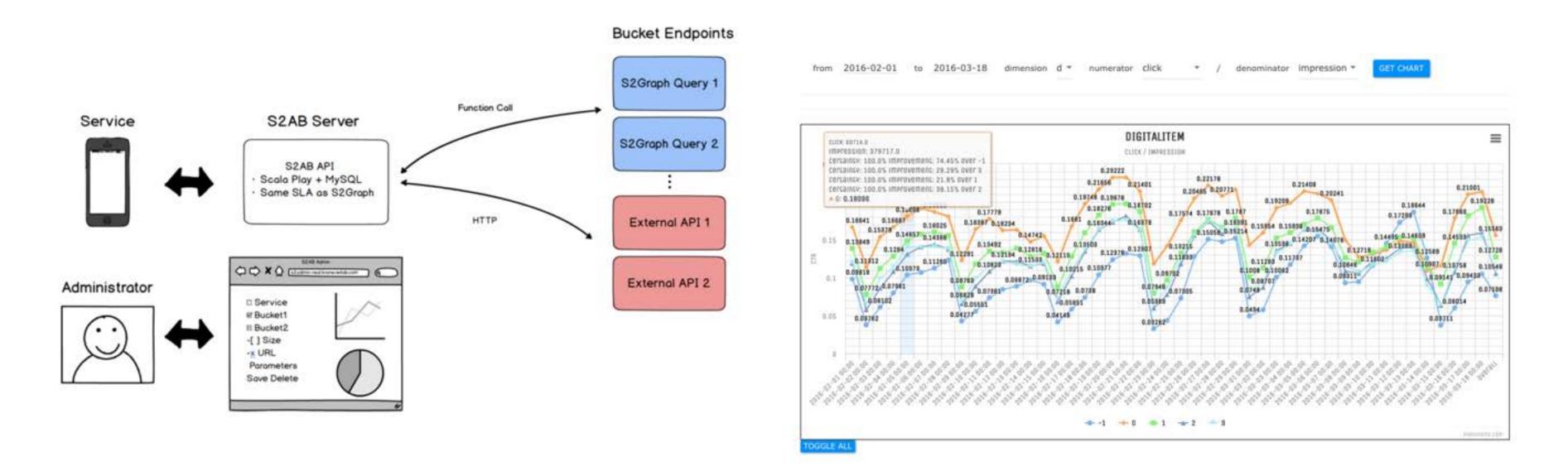
```
"weights": [
            1000,
            100,
            10
        1,
        "queries": [
                "srcVertices": [...], // some user
                "steps": [[{"label": "item_based_cf_label"}]]
            },
            {
                "srcVertices": [...], // some user
                "steps": [[{"label": "user_based_cf_label"}]]
            },
                "srcVertices": [...], // some user
                "steps": [[{"label": "matrix_factorization_label"}]]
            },
                "srcVertices": [...], // some user
                "steps": [
                  [{"label": "user_demography_label"}],
                  [{"label": "demographical_top_k_label"}]
28
```



3. Native A/B (Bucket) Testing

1.Bucket = Query

2.Track performance of each bucket realtime: CTR, conversion rate.. 3.Maximize performance.

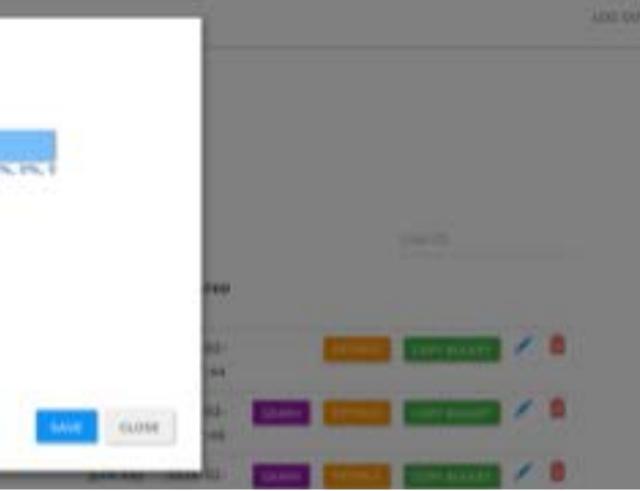


Real World Use Cases

Experiment example

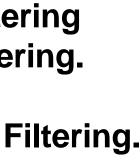
		i modular i 30 onpecer i 454 1		Self-date	Editing Buckets Ratio			
							Auropea Ste, Ste, Ste, Ste, Ste, Ste, Ste, Ste,	55
					11-0479-/8	1-5		
					10.000	6~10		
					10.04	$0.1 \leftrightarrow 0.3$		
	12403	ANPERSONN'			ri-sa, hybrid	16-100		
		225 105 re-default 5	base from	read, hybrid, with, math	0-0			
83					10-04P	0-0		
	163							
		100		parrietata	1. LANKA - MARK - A	_	All and a second second	_

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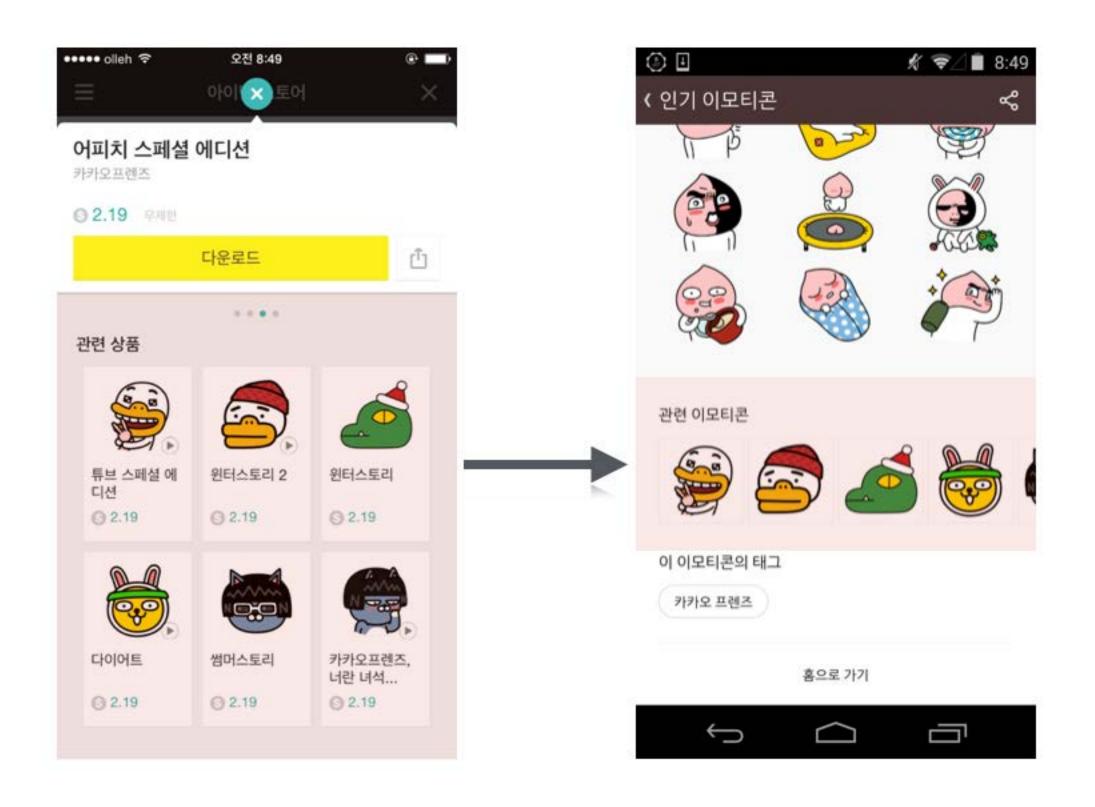


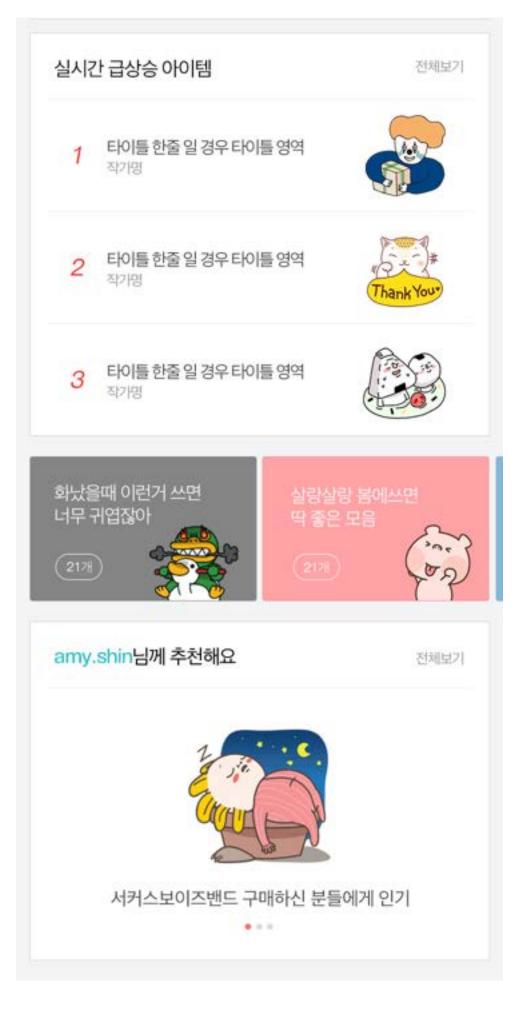
- 1. User Based Collaborative Filtering
- 2. Item Based Collaborative Filtering.
- 3. Matrix Factorization.
- 4. Content Based Collaborative Filtering.
- 5. Most Popular.
- 6. Segmented Most Popular.
- 7. Spreading Activation.
- 8. Social Recommendation.

Weighted sum of each algorithms are also possible.

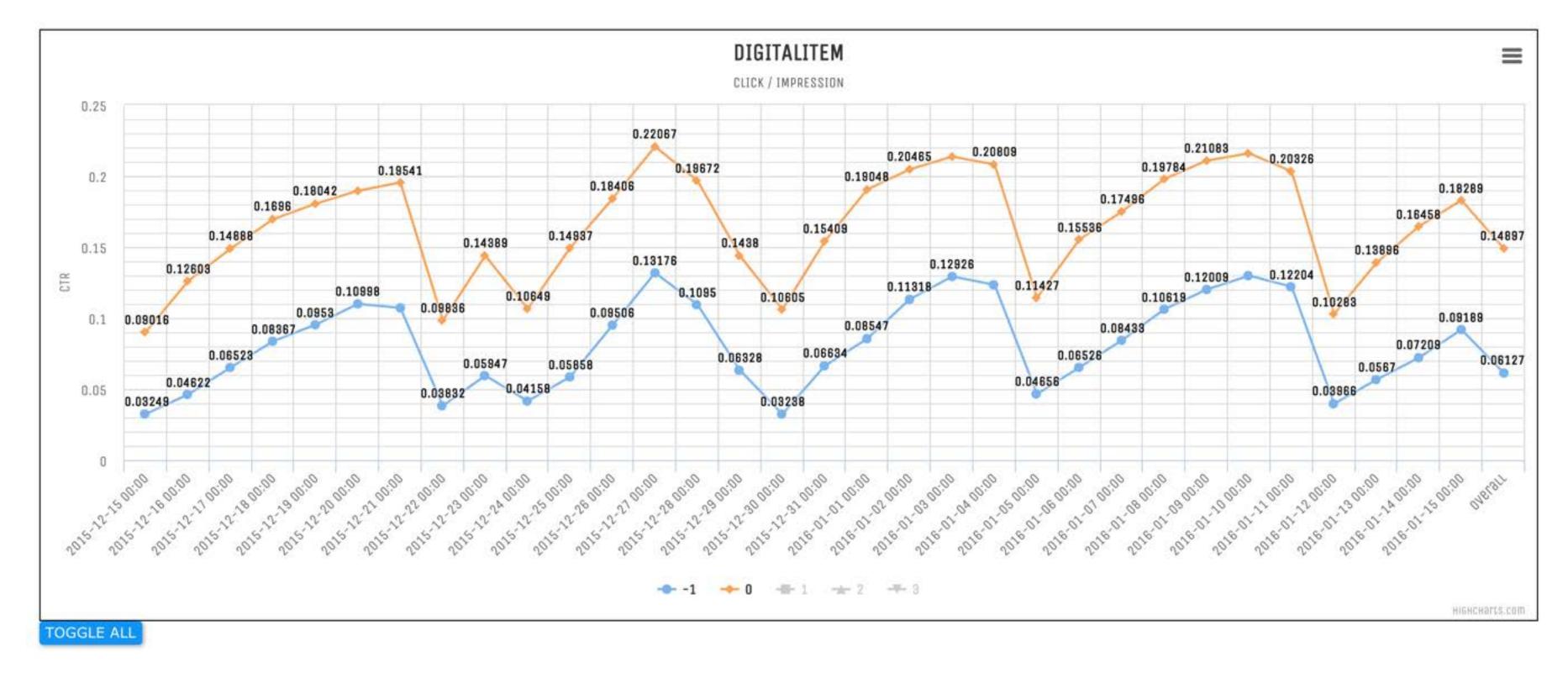


1. Emoticon Store





1. Emoticon Store (cont)



Baseline(most popular) vs S2Graph

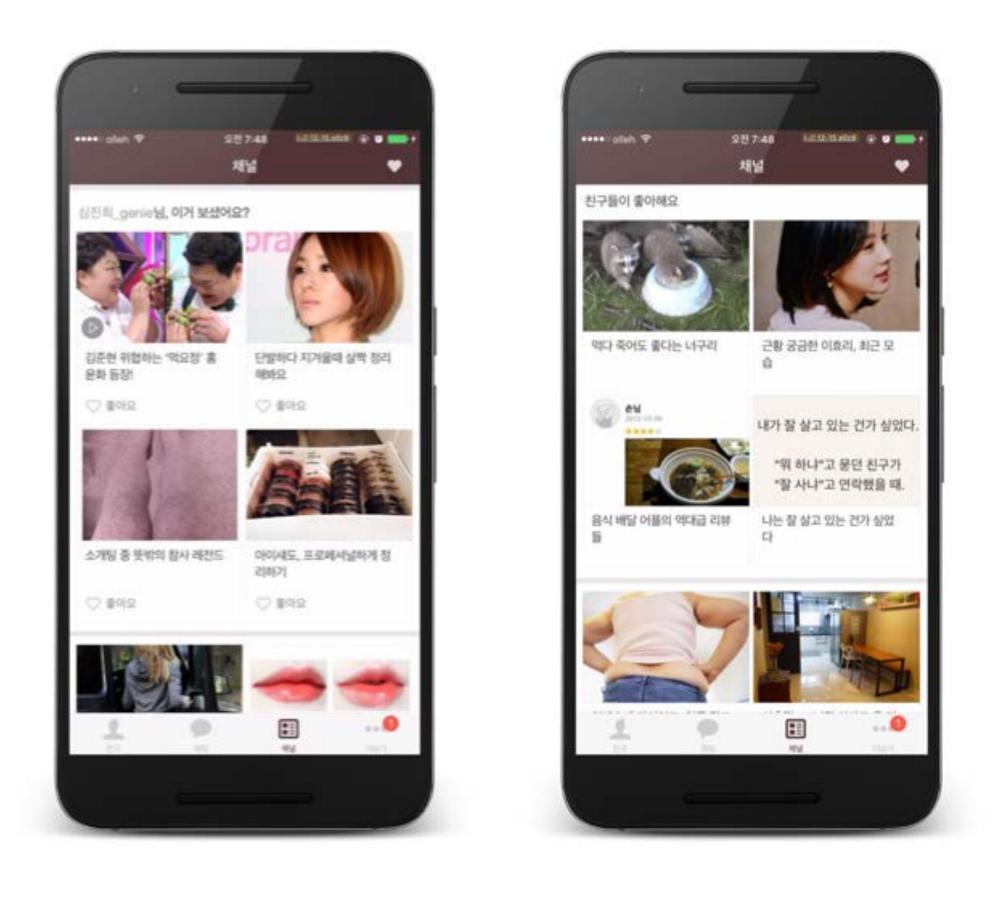
CTR: +137.52% Purchase/Click: +70.8% Purchase/Impression: +304%

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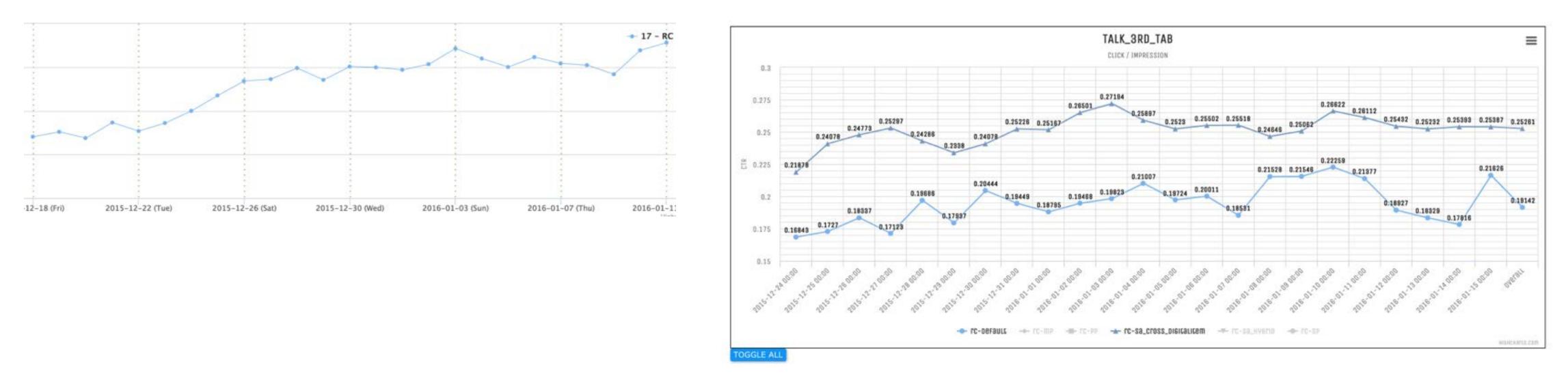
Before -> After

```
Total Item Click: x3.8
Unique User with Item Click: x3.5
```

2. Kakao Talk Channel: Recommend Card, Social Card



2. Kakao Talk Channel: Recommend Card (cont)



Baseline(most popular) vs S2Graph

Total Click: +123% CTR: +36.6% Unique # of contents with clicks: x20

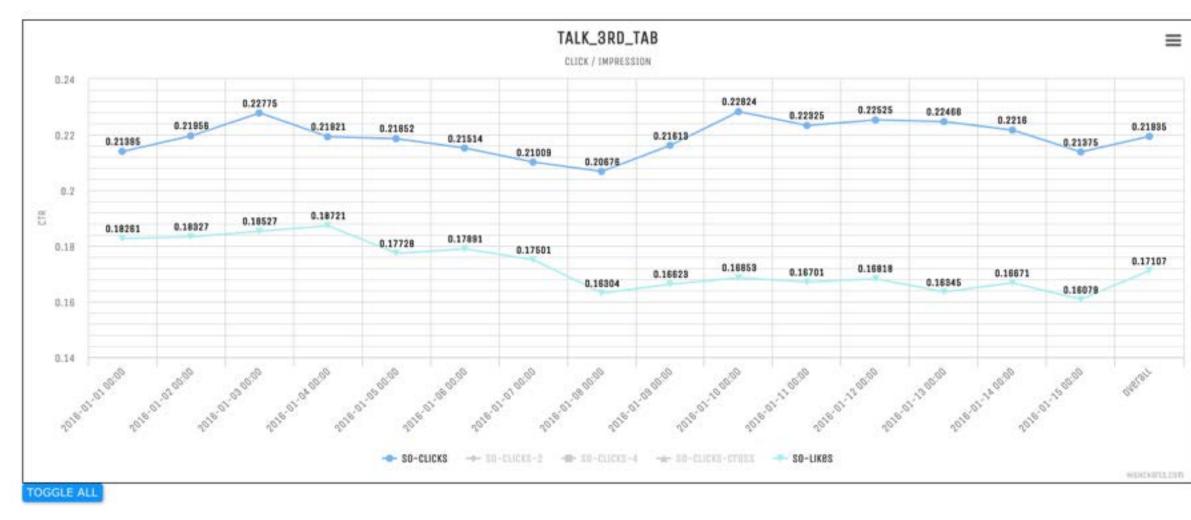


2. Kakao Talk Channel: Social Card (cont)



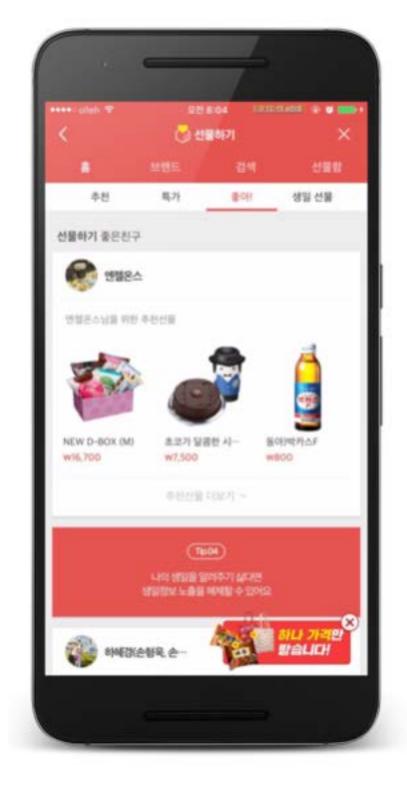
Baseline vs S2Graph

Total Click: +591% CTR: +23.65%

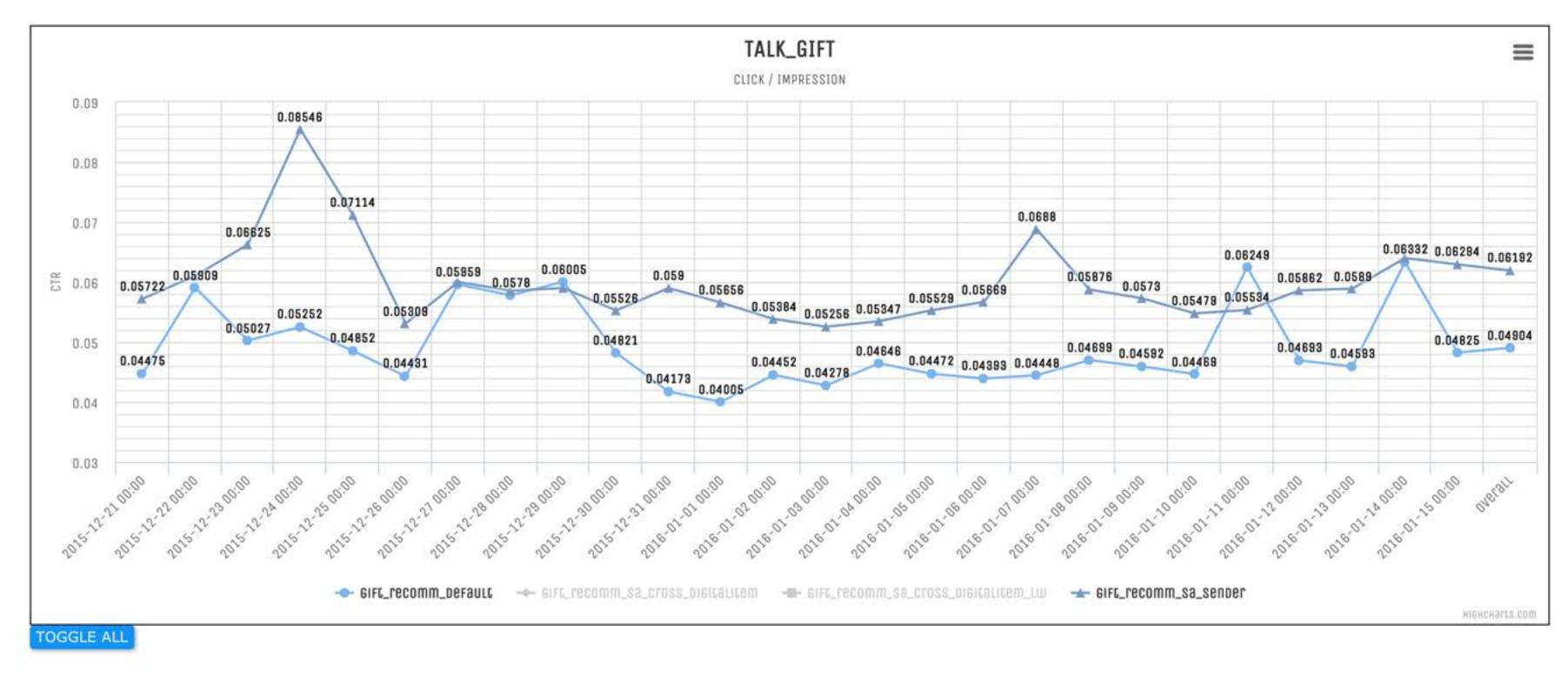




3. Kakao Talk Gift Shop



3. Kakao Talk Gift Shop (cont)



Baseline(most popular) vs S2Graph

CTR: +26.03% Purchase/Click: +344% Purchase/Impression: +459%

Powered By S2Graph

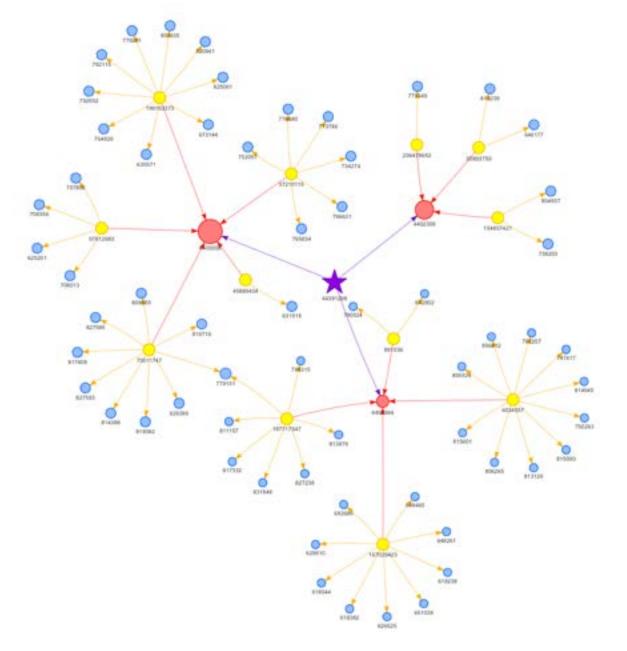


Statistics

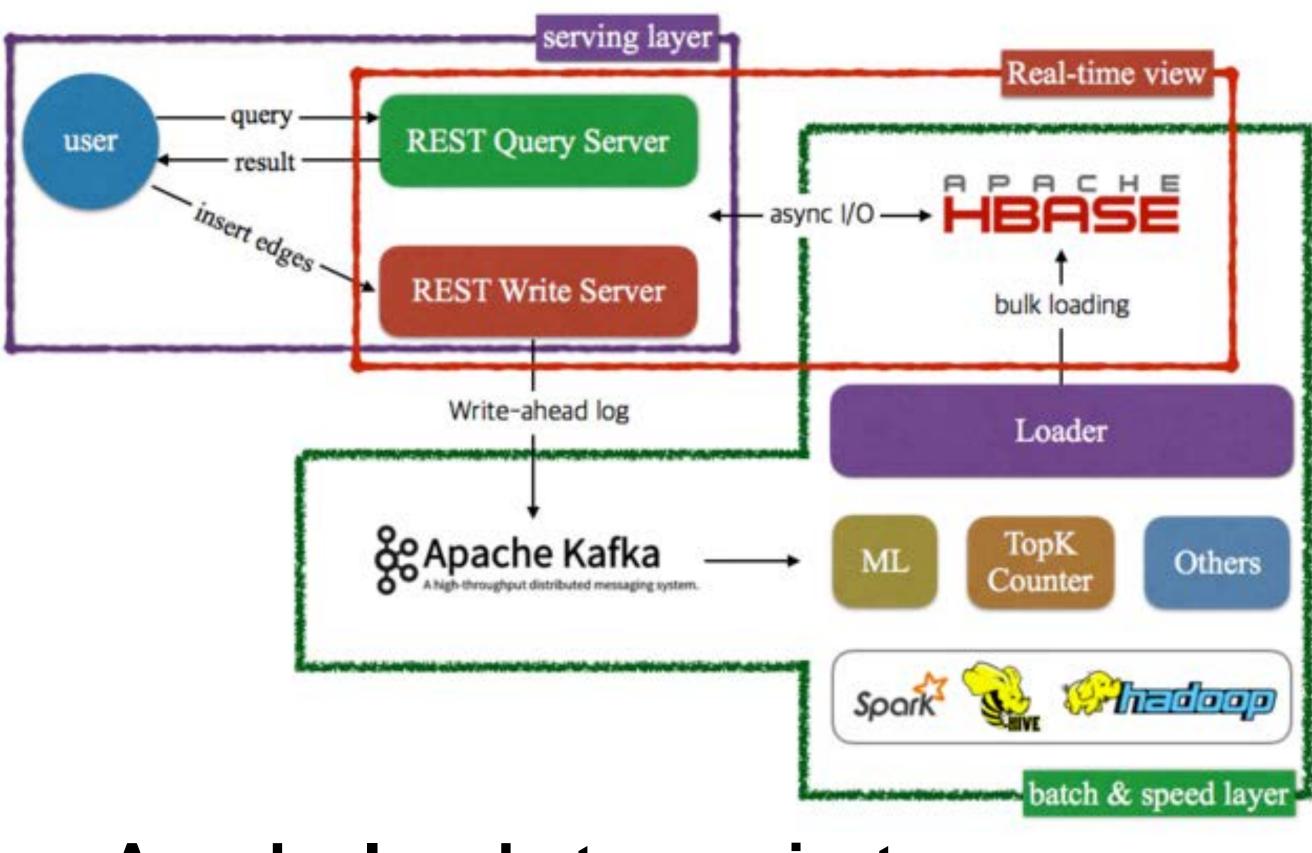
- 1. total # of edges(activities + relations) >= 1 trillion
- 2. daily, new incoming edges in realtime >= 3 billion
- 3. daily, new edges that processed from batch process >= 50 billion
- 4. average query per minute >= 200 million. peak >= 400 million. under 50 ms.
- 5. 40% queries are 3 step query, 40% are 2 step, 20% are 1 step.

Operations

- 1. # of HBase region server = 40
- 2. # of query server = 70
- 3. # of write server = 20



Why S2Graph: Apache V2 License



Entire projects are Apache Incubator project. This means S2Graph is open to anyone.