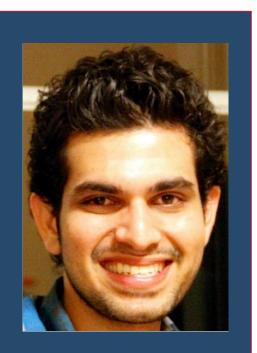
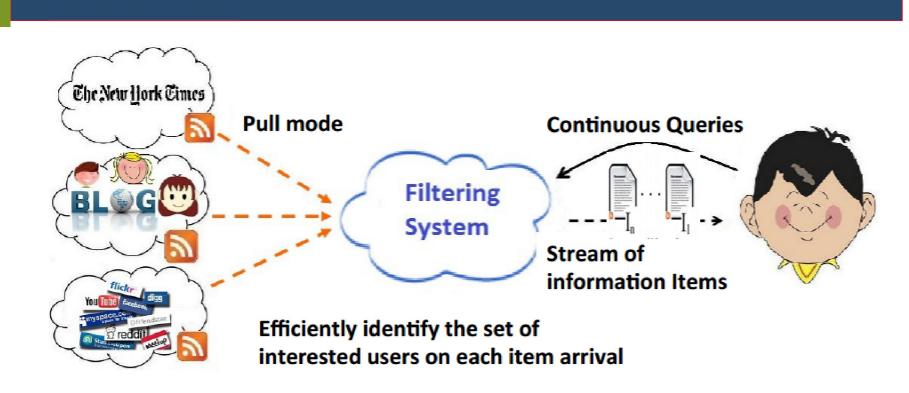


Cloud based publish/subscribe model for Top-k matching over continuous data-streams Y.S. Horawalavithana, Dr. D.N. Ranasinghe University of Colombo School of Computing



Introduction



Why publish/subscribe?

- ☐ Is the backbone among many modern day large scale applications emerged with "live" info production, too numerous applications behind:
 - ☐ *The Web*: issue subscriptions for pages' updates, etc.
 - ☐ *The network/system*: track requests with specific IPs,
 - ☐ (Multi-player Strategic Computer) Games
- ☐ Decoupling of producers and consumers of info is valuable for more
 - ☐ Flexible, Lightweight, and Scalable systems

Boolean publish/subscribe model

Drawbacks

- ☐ A subscriber may be either overloaded with publications or receive too few publications
- ☐ Impossible to compare different matching publications as ranking functions are not defined,
- ☐ Partial matching between subscriptions and publications is not supported.

☐ Pub/Sub Matching

☐ Top-k pub/sub

☐ Pub/Sub Indexing

personalized

subscriptions

scoring or ranking

☐ Indexing to support

☐ Indexing to support

continuous Top-k

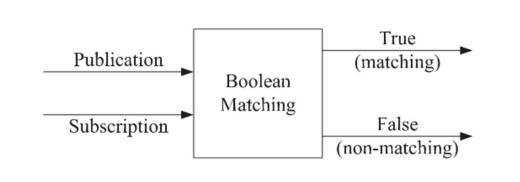
publications retrieval

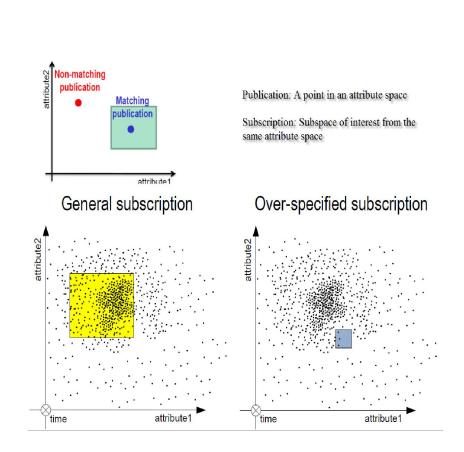
Top-k publish/subscribe

publications at a given instance

☐ Expressive stateful query processing systems

 \Box User defined parameter k restricts the delivered





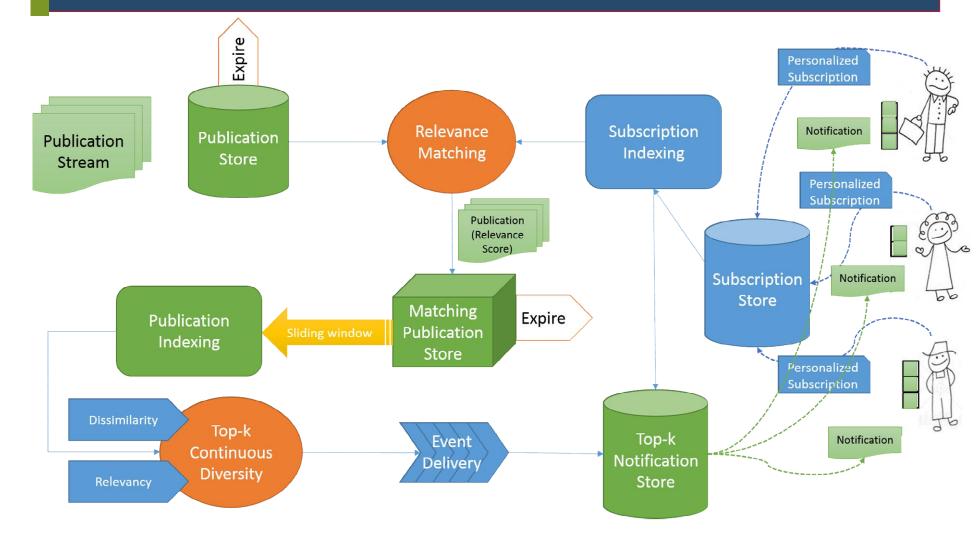
Overlay Layer

Unstructured Hybrid

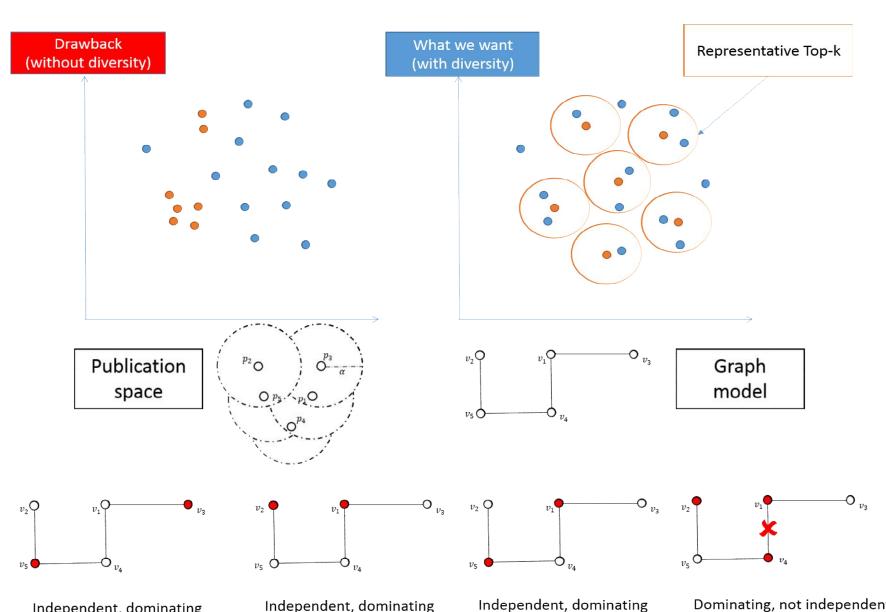
Network Layer

Persistent

Design & Architecture



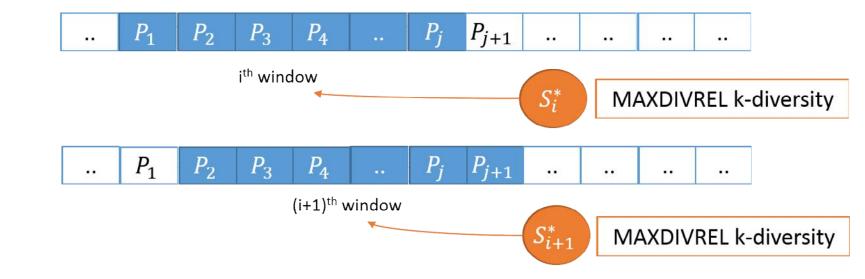
Top-k representative set



■ MAXDIVREL continuous k-diversity

☐ Sliding window Top-k computation to handle streaming publications

Matching publication stream



□ NP-Hard: *MAXDIVREL* is mapped to minimum independent domination set problem in graph theory

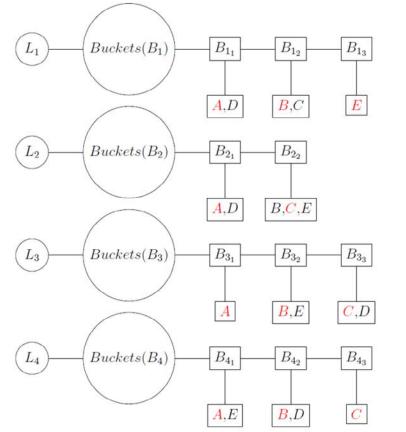
☐ Indexing streaming publications

- ☐ Avoid the curse of re-calculating neighborhood
- ☐ Based on Locality Sensitive Hashing (LSH)
- ☐ Fast Min-Hashing
 - Minhash signatures to represent publications

$minhash_i$	publication X	publication Y	publication Z
h_1	$h_1(X)$	$h_1(Y)$	$h_1(Z)$
h_2	$h_2(X)$	$h_2(Y)$	$h_2(Z)$
h_m	$h_m(X)$	$h_m(Y)$	$h_m(Z)$

Signature Matrix

- ☐ Map the signatures into,
 - ☐ L Hash-Tables
 - ☐ With arbitrary b number of buckets
- □ *Voting mechanism to retrieve Top-k publications*

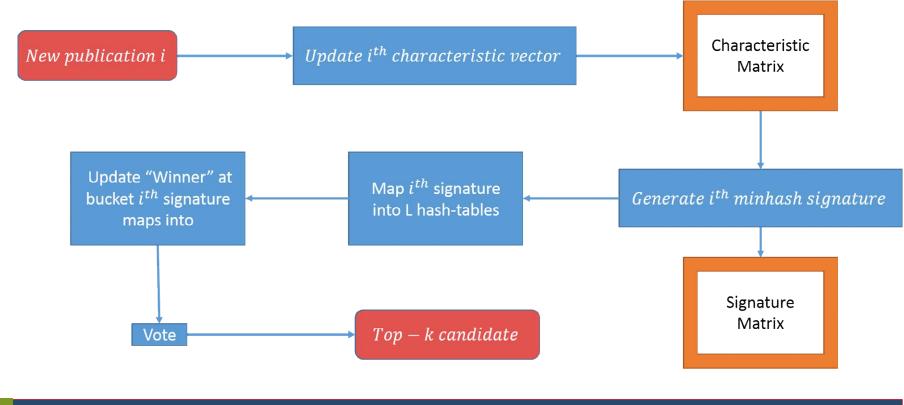


P_A	P_B	P_C	P_D	P_E	P_F	P_G	P_H	
		th						
p	ublic	atio	n (F	P)	A	B	C	E
	V	otes	S		4	3	3	1
☐ Bucket "Winner" — a publication which has the								

- □ Bucket "Winner" a publication which has the highest relevancy score
 ✓ Winner is dominant to represent it's bucket
- neighborhood

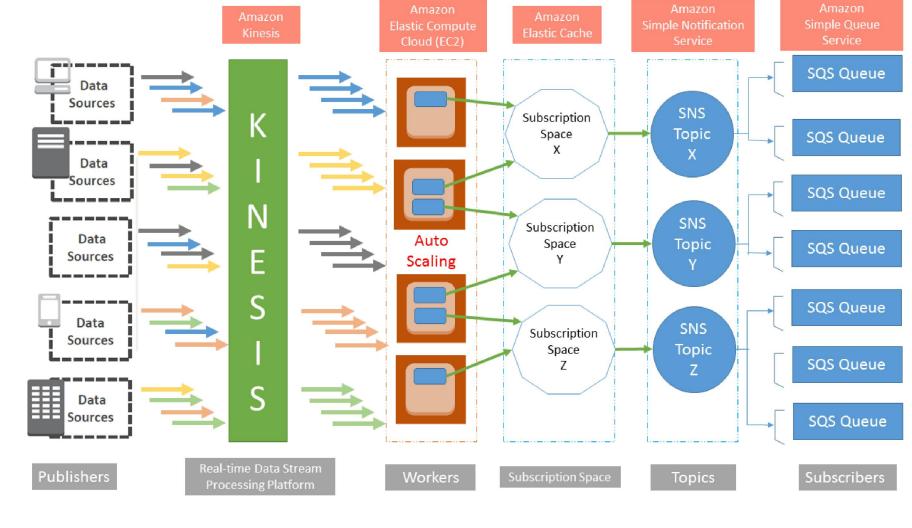
 ☐ Top-k "winners" that have a majority of votes

 ✓ k winners are independent
- Incremental LSH Top-k computation

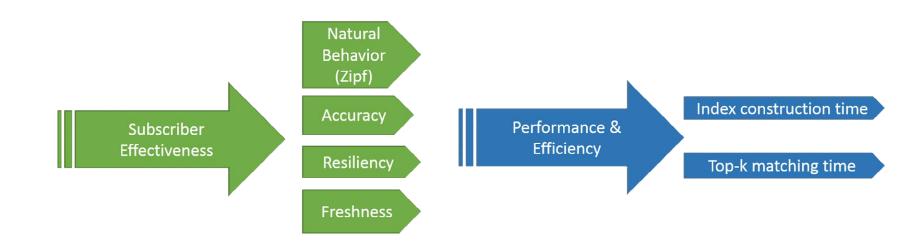


Implementation & Evaluation

☐ Implementation on top of Amazon Web Services

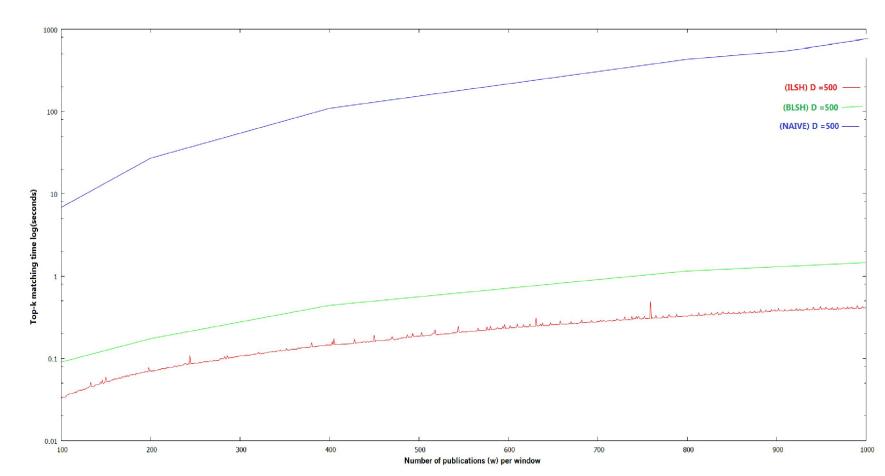


☐ Evaluation Methodology



		Τ	
Diversity method k	MAXMIN	DisC	MAXDIVREL
10	4.6123	3.4632	2.4883
50	12.2535	2.7392	2.4851
250	46.1347	2.5381	2.1956
500	50.3878	2.1023	1.9420
1000	62.5921	2.2003	1.9591

Average zipf law exponent in a comparison with other methods



A comparison of incremental LSH indexing with naïve & batch method: ranking time on number of publications with dimension D=500