

Adaptive Hierarchical Clustering of Message Flows in a Multicast Data Dissemination System

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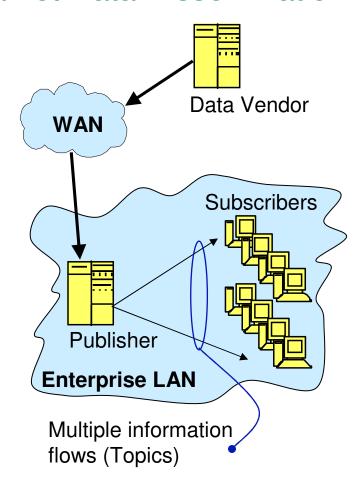
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Outline

- Introduction
 - ♦ The System Pub/Sub Messaging for Data Dissemination
 - Multicast Technology
- Multicast Mapping
- Clustering Algorithms
 - Modified K-Means, Hierarchical Clustering
- Real-Life Messaging-Load Model
- Experiments & Results
- An Adaptive System
- Future Directions
- Summary

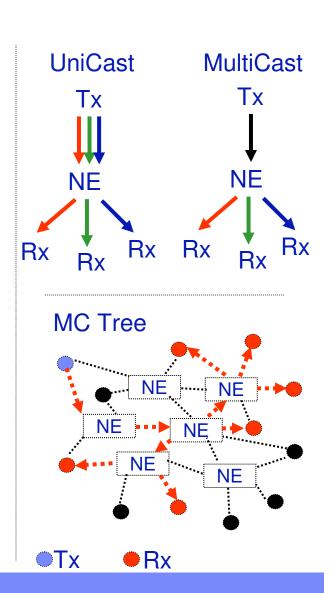
The Basic Scenario – Pub/Sub for Market Data Dissemination

- Publisher divides data feed into a large number information flows (topics), (~100K) e.g. stock symbols, futures, commodities
- Many stand-alone subscribers (~1K)
- Subscribers display interest heterogeneity are interested in different yet overlapping subsets of the topics
- Any single topic may be delivered to a large number of subscribers (hot / cold topics)
- Unicast duplicate transmissions
- Flooding (Broadcast) receivers burdened by unwanted incoming traffic



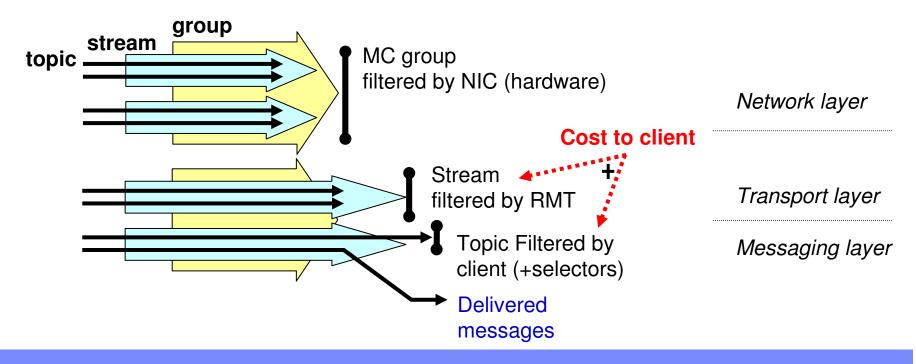
Multicast Technology

- ♦ IP multicast, Network layer
 - A single packet sent by a transmitter reaches all the hosts that joined a certain Multicast Group
 - Unreliable, no traffic control, no ordering
- ♦ Reliable Multicast Transport (RMT) Protocols
 - Reliability, Ordering, Flow & Congestion control
 - * "Session" or "Stream" transport layer entity
- Cannot allocate a group (or stream) per topic
- Limited number of usable multicast groups (NE state problem, receiver resources)
- Limited number of reliable multicast streams
- ♦ # Flows >> # RMT Streams >= # IP MC Groups
 - => Mapping Flows to Streams
 - => Mapping Streams to Groups



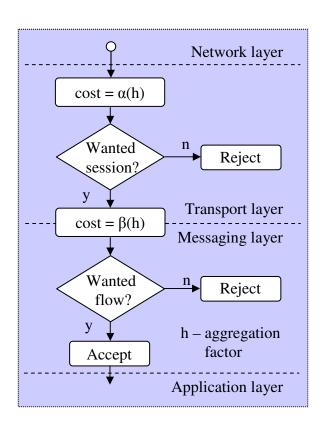
Map Structure and Filtering Cost

- Each Topic is mapped to a single RMT stream
- Each RMT Stream is mapped to a single multicast group
- Client filtering is a must
- The cost to the client depends on implementation details

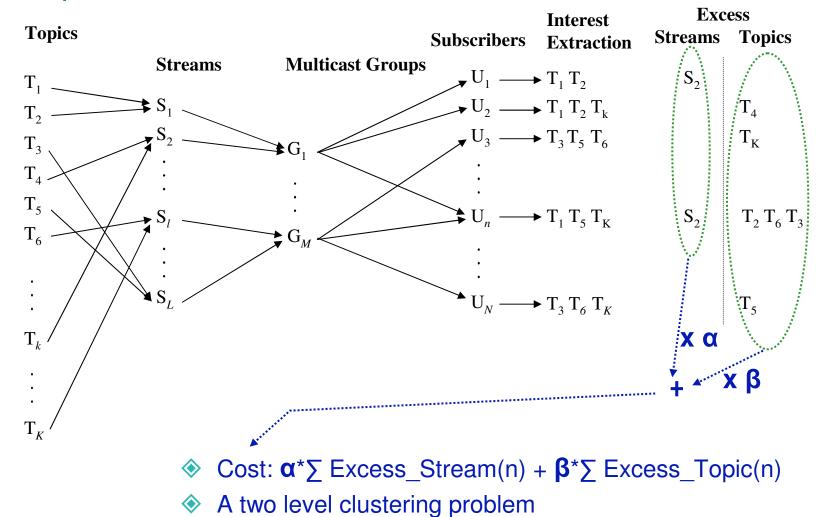


Message Aggregation and Filtering Cost

- Aggregation multiple messages from the same RMT stream share the same packet
- At transport layer
 - Some processing for each packet
 - Some processing for each message
 - Amortization of packet-level processing across multiple messages, increases performance
- ♦ At messaging layer processing per message
- Depends on implementation
- We estimated the effect of message aggregation and included it in the cost function



Example

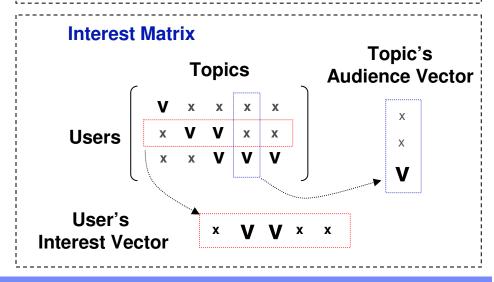


Time level electering problem

Algorithm Input - Messaging Statistics

- Publication
 - The list of published topics
 - The publication rate of each topic
- Subscription
 - The list of topics each client required
 - Client are anonymous
- Interest Matrix
 - A binary matrix indicating the interest of the clients
- Publication Rate Vector

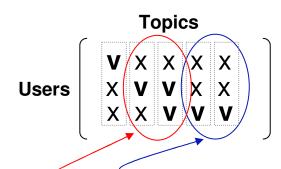
#	Topic	Rate	
	[TopicSpace:Topic]	[msg/s]	
#1	Cars:Toyota/Hilux	10	
#2	Cars:Honda/Civic	20	
	Comp:IBM/pSeries	30	



Mapping Algorithm

- ♦ Input
 - interest matrix, topic rate vector
- Basic insight
 - Put "similar" topics in the same group
 - "Similar" topics have a similar audience
 - A group with a homogenous audience causes less filtering to the audience
- Take the rate into account
 - The cost of putting two topics in the same group
 - The cost of adding a new topic to a group of topics

Interest Matrix



→ Topics with identical audience

Topics with similar audience

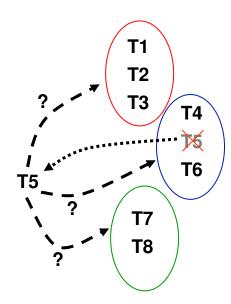
		Topics		Filtering Cost	
		1	2		
Users	1	V	Χ	R2	
	2	V	V	0	
	3	Χ	V	R1	
	4	X	Χ	0	
				R1+ R2	

Rk – the rate of topic k

Iterative Clustering Algorithm (K-means)

- Init: Topics are assigned into a fixed number of groups
- Move: In each step, remove a single topic, and move it to the best group – the one producing the lowest cost
- Cost: After each epoch, compute total filtering cost
- Stop: time elapsed | cost does not improve | exceeded max number of iterations | number of topics moved

Topic group				Group audience vector	Candidate topic 5	The cost of adding topic 5 to topic group {1,2,3}
Users	V	Х	V	V	V	0
	V	V	Χ	V	V	0
	V	V	V	V	V	0
	Χ	Χ	V	V	X	R5
	Χ	Χ	Χ	X	V	R1+R2+R3
	X	X	Χ	X	X	0
	L					R1+R2+R3+R5



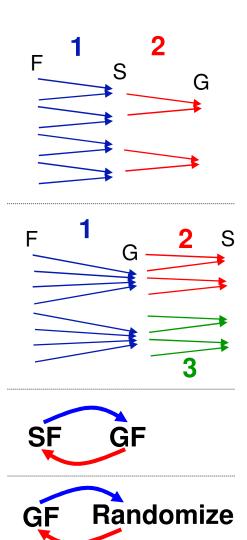
The best group for topic K

is the group

with the lowest cost

Hierarchical Clustering Algorithms

- Streams First (SF)
 - Cluster flows to streams
 - Cluster the resulting streams into groups
- ♦ Group First (GF)
 - Cluster flows into groups
 - Within each group separately, cluster flows into streams.
- An Iterative Approach (IT)
 - Iterative invocation of GF and SF
 - ◆Taking the best map from all the iterations
- Random Restart with Annealing (RRA)
 - Random reassignment of a diminishing percentage of flows to streams,
 - ◆Do a GF step
 - ◆Taking the best map from all the iterations



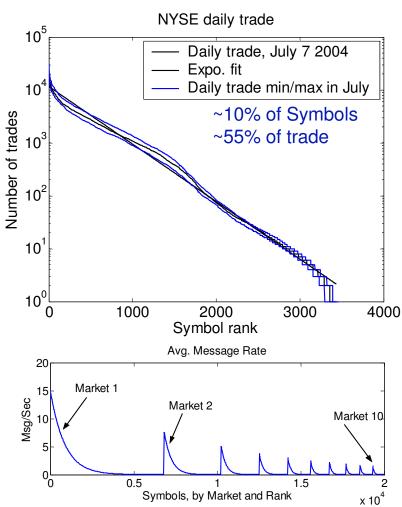
Messaging Load Model – Based on Market Research

Financial front office

- Hundreds of users, requiring stock quotes and financial information from several markets
- ♦ Up-stream action (from brokers to market buy/sell) is reflected in the Down-stream traffic (from market to broker stock quotes)

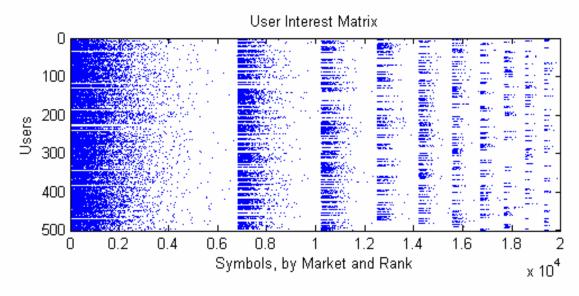
♦ Topic space structure

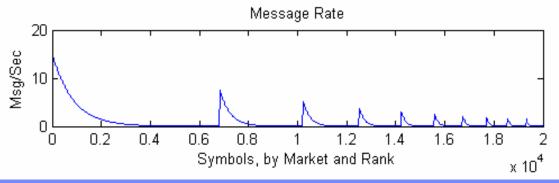
- Within each market, symbol popularity and rate are exponentially distributed (NYSE market research)
- ♦ Several different markets, with Avg. popularity and size prop. ~1/m (assumption).



Real Life Messaging Load Model

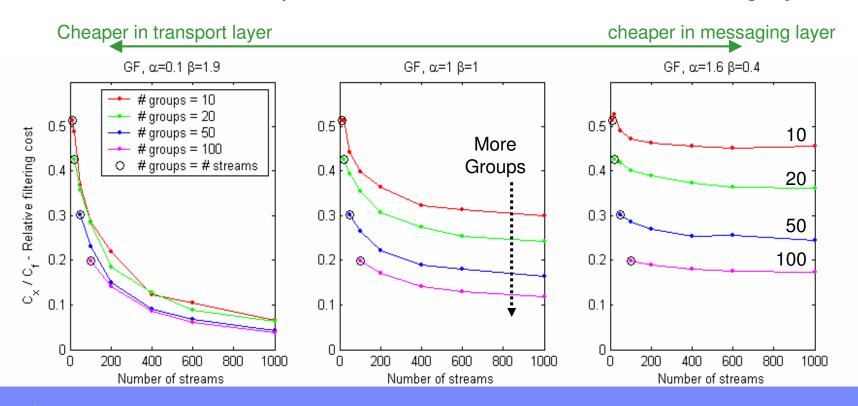
- Based on statistical analysis of NYSE daily trade data
- 20K Topics
- ♦ 500 Subscribers
- ♦ Avg. ~70 topics / user
- Min 15 topics / user
- ♦ Max 115 topics / user
- Avg. message fan out ~10.1 clients
- Multicast message is transmitted once
- Unicast transmitter data rate is x10 of multicast!





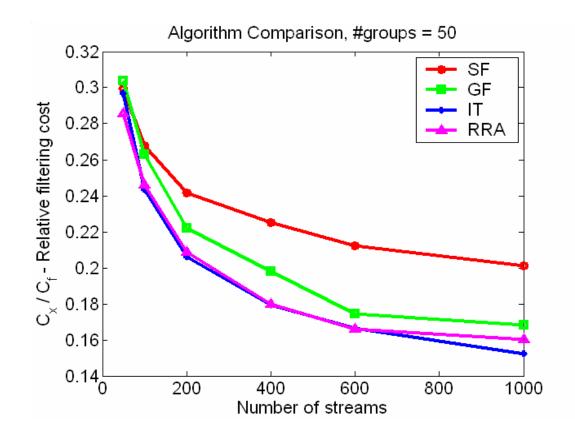
The Effect of the Number of Groups and Streams

- Increasing the number of streams and groups always improves performance
- Hierarchical filtering is more efficient than non-hierarchical
- Relative effectiveness depends on the amount of work in each filtering layer



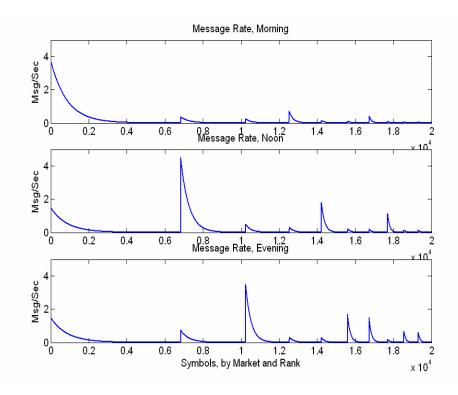
Algorithm Comparison

- GF is better than SF
- GF is fastest (not shown)
- Iterative algorithms
 - produce better results
 - take longer to execute (not shown)
- \Leftrightarrow GF / Random = 0.4 0.6



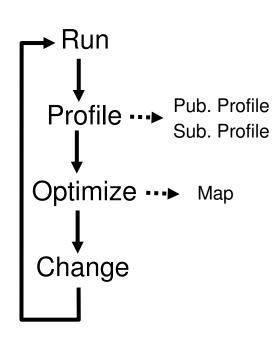
The Case For Adaptive Mapping

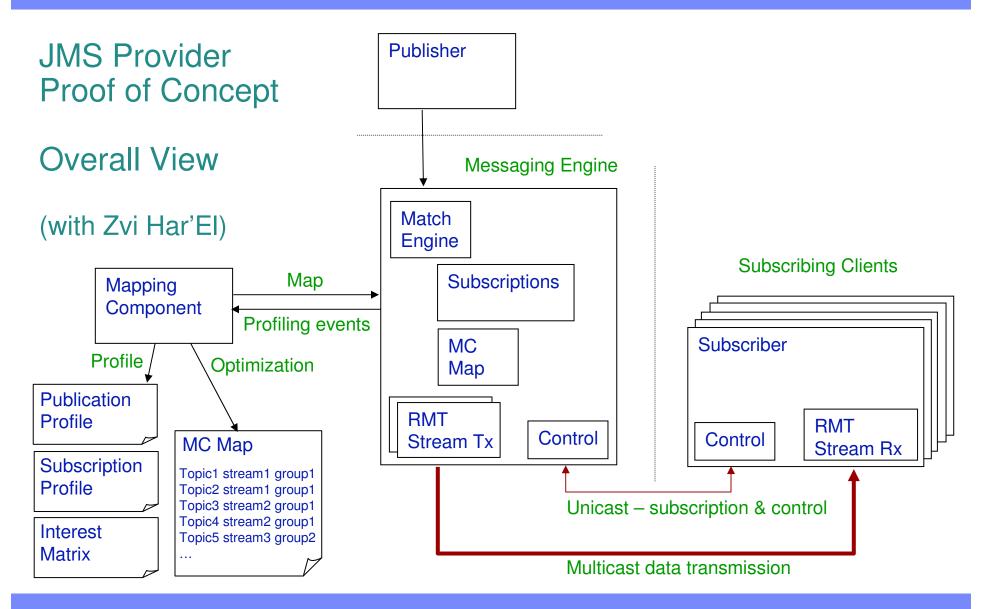
- User interest & message rate change during the day
 - Across markets
 - Within a market
 - ♦ In response to world events
 - ◆Trading hours
- Manual management
 - ♦ Expensive, intractable
 - ◆Error prone
- The "average" map
 - Of yesterday or a few days back
- Dynamic, Adaptive
 - Adapts to interests and rate
 - Runtime migration mechanism



Adaptive Multicast Infrastructure

- Run: running a messaging load in a given configuration.
- Profile: profiling publications and subscription.
- Optimization: the profiling results are fed into the optimization algorithm. The result is a map.
- Change: change publisher and subscriber configuration to the new map.
- The optimization starts from a previous map (fast)
- The adaptation time scale can be days, hours, minutes
- Change process is automatic, subject to QoS requirements
- Manual override process control, map editing, pub/sub profiles





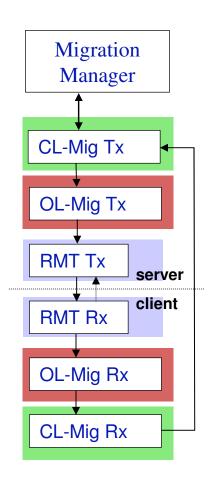
Migration Protocol

General requirements

- Preserve flow message sequencing
- Avoid duplicate transmissions
- Conform with the multicast reliability guarantees
- Fast reasonable time from start to finish
- Scalable number of clients / subscriptions

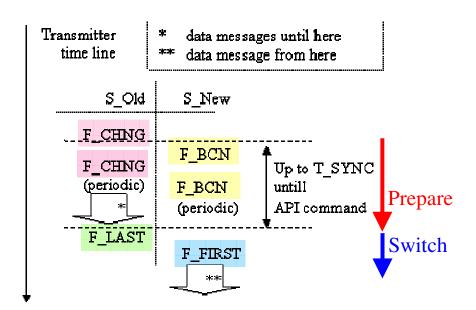
Efficient protocol

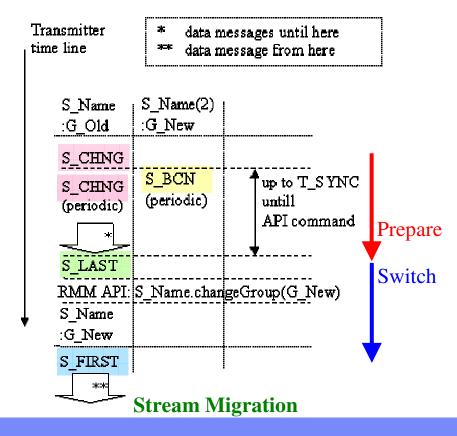
- ◆ Not "stop the world", no pipeline drainage
- Messaging activity and throughput is hardly affected
- Use existing RMT API with minimal changes
- A layered approach
 - Isolation of lower level protocols
 - Allow for two levels of quality of service



Flow & Stream Migration - Open Loop

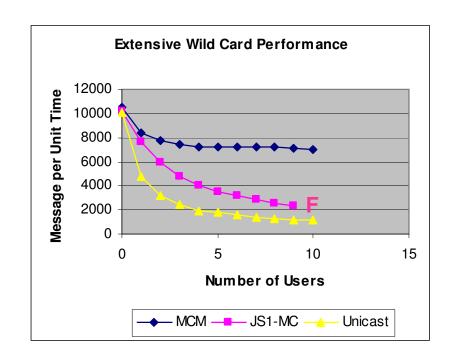
- Based on standard reliable multicast transport protocol / infrastructure (e.g. PGM)
- No feedback from receivers, reliability based on timing, receiver detects failures
- Prepare phase two signals (Change + Beacon)
- Switch phase two signals (Last + First)





Overall Performance – Extensive Wild Card

- JMS messaging provider POC
- Hierarchic topic-based
- Subscriptions are unique and overlapping, e.g.
 - ♦ /*/b/*/d
 - ♦ /a/*/*/d
- ♦ ~7 topics/user @ 10 users
- Unicast approx. 1/n
- JS1-MC stream per unique subscription: causes data duplication – performance degradation is almost like unicast
- Multicast mapping is scalable, and applicable



Current & Future Work (With Gregory Chockler, Roie Melamed)

- Distributed Large Scale Pub/Sub
 - ♦ A large number of topics (x1000)
 - ♦ A large number of users (x10000)
 - Correlated user interests (x100 / User)
 - High churn
 - No IP multicast
- Based on overlay network, P2P
 - That takes into account the user interest
- How do we
 - Define abstract dissemination channels
 - Map topics to abstract dissemination channels
 - Migrate topics between channels

Summary

- Large scale multicast Pub/Sub
 - A huge number of topics
 - ♦ A limited number of RMT streams, IP multicast groups
 - Hierarchic approach
- Cost function hierarchic filtering, message aggregation
- Estimated the relative cost of transport vs. messaging layer filtering
- Iterative clustering algorithm based on K-means
- Several hierarchic clustering algorithms
- Real-life messaging load based on NYSE market research
- Hierarchic filtering is better then flat
- Advantage for efficient filtering at transport layer
- The challenges of an adaptive fully distributed system