

Boosting Topic-Based Publish-Subscribe Systems with Dynamic Clustering

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Outline

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- Implementation
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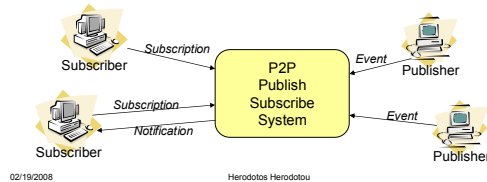
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Motivation

- **Challenges:**
 - Dynamic Nature of P2P Pub-Sub Environment
 - Decentralized Nature of P2P Pub-Sub Environment
- **Goal:**
 - Maximize performance



Background: Pub-Sub Systems

- **Three Components:**
 - Subscribers
 - Express interest in events
 - Publishers
 - Create topics and generate events
 - Infrastructure
 - Match events with interests and deliver events

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Background: Pub-Sub Systems

- **Categories:**
 - Topic-Based
 - Subscribers specify interests as “topics”
 - Publishers label events with topics
 - Content-Based
 - Subscribers specify interests through event filters
 - Published event content matched against filters

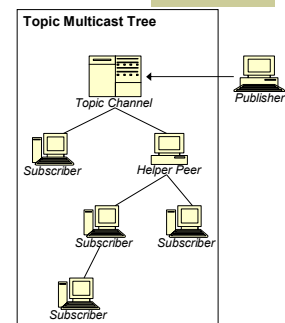
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Topic-Based Pub-Sub Systems

- **Operations:**
 - Create
 - Publish
 - Subscribe
 - Unsubscribe
- **Costs:**
 - Transmission Cost
 - Maintenance Cost



Topic Clusters

- **How to decrease maintenance cost?**
 - Group topics with similar sets of subscribers into a "virtual" topic
 - => Unify underlying supporting structures
 - => Eliminate redundancy in structures of individual topics
- **Side effects:**
 - Users can subscribe to topics or topic-clusters
 - Need local filters to eliminate irrelevant events

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Cost Model

Maintenance Cost:

$$MC = \sum_{t \in T} cost_m(size(t)) + \sum_{c \in C} (cost_m(size(c)) + cost_a(|c|))$$

Dissemination Cost:

$$DC = \sum_{t \in T} freq(t) \times [(cost_n() + cost_d(size(t))) + \sum_{\{c | t \in c\}} (cost_n() + cost_d(size(c)))]$$

Overall Cost:

$$OC = MC + DC$$

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Dynamic Clustering Algorithm

- **Cluster Update Operations:**
 - Topic 1 + Topic 2 → New Cluster
 - Topic + Cluster → New Cluster
 - Cluster 1 + Cluster 2 → New Cluster
 - Old Cluster → New Cluster + Topic
 - Old Cluster → Topic 1 + Topic 2
- **Algorithm:**
 - Perform Update if and only if it is beneficial

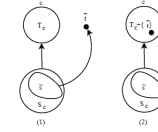
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Adding Topic to Existing Cluster 1/2

Benefit Estimation:



Update is beneficial when:

$$|\tilde{s}| > \frac{1 + freq(\tilde{t})(|S_c| + 1)}{1 + freq(t)}$$

Triggering Benefit Estimation:

- Frequency of events in topic
- Number of subscribers to cluster
- Size of subscriber subset registered to topic

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Adding Topic to Existing Cluster 2/2

- **Trigger Benefit Estimation Probabilistically:**
 - When user subscribes to topic t , perform benefit estimation with probability ω .
 - Conduct survey among c cluster's subscribers.
 - $|\tilde{s}| \approx \left(\text{Fraction of common subscribers to } t \text{ \& } c \right) * (\text{cluster size})$
- **Notes:**
 - $\omega > 1 - (1 - \phi)^{\frac{1}{|\tilde{s}|}}$
where ϕ = configuration parameter

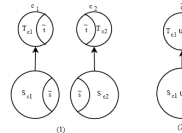
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Merging two Clusters 1/3

Benefit Estimation:



Update is beneficial when:

$$|\tilde{s}| > \frac{\alpha |T_{c1} \setminus \tilde{t}| \cdot |S_{c2} \setminus \tilde{s}| + \alpha |T_{c2} \setminus \tilde{t}| \cdot |S_{c1} \setminus \tilde{s}| - |\tilde{t}|(\alpha + 1)}{\alpha |\tilde{t}| + 1}$$

Triggering Benefit Estimation:

- Frequency of events in topics
- Number of topics in each cluster
- Number of subscribers in each cluster
- Number of common topics and subscribers

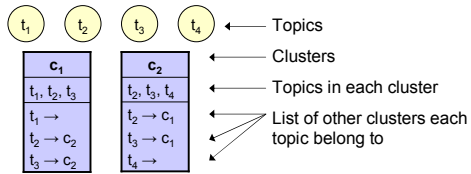
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Merging two Clusters 2/3

- Estimate number of common topics:



- Number of common topics = number of times c_2 appears in the lists of c_1
- Lists maintenance through cluster update notifications

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Merging two Clusters 3/3

- Estimate number of common subscribers:

- Maintain overall number of subscribers
- Sample subscribers periodically to learn which other clusters they are subscribed to
- $|S| \approx \left(\frac{\text{Fraction of common}}{\text{subscribers to } c_1 \text{ \& } c_2} \right) * (\text{number of subscribers})$

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Subscriptions

- Optimal Subscription Problem:
 - Find set of topics and clusters to subscribe, that cover all interests and minimizes cost
 - NP complete
- Greedy Subscription Algorithm
 - Treat each individual request for adding/removing a topic heuristically

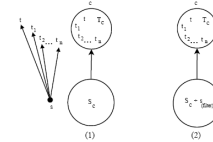
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Greedy Subscription Algorithm

- User s wishes to subscribe to topic t
 - If t belongs to a cluster s is subscribed to, update the filter
 - Otherwise two choices:
 - Subscribe s to t directly
 - Subscribe s to new cluster that includes t

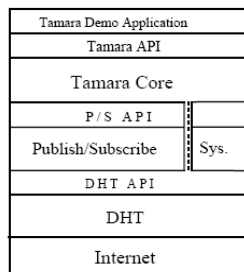


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Implementation



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Experiments on Synthetic Data 1/4

- Events Frequency Models

Frequency	% of topics
0.6-0.6	15
0.2-0.4	5
0.1-0.2	25
0.02 - 0.1	45
0.002 - 0.02	10

(a) Average

Frequency	% of topics
1-3	5
0.5-1	20
0.1-0.5	50
0.01-0.1	20
0.001-0.01	5

(b) Frequent

Frequency	% of topics
0.1-0.2	10
0.01-0.1	40
0.001-0.01	50

(c) Intrequent

- Quality of Clustering

Events	Algo.	Structure maint.	Event diss.	OC
Average	Tamara	-78.75%	+45.45%	-63.73 %
	K-means	-53.75%	+9.1%	-46.15%
Frequent	Tamara	-73.75%	+15.62%	-48.21%
	K-means	-65%	+9.3%	-43.75%
Intrequent	Tamara	-88.75%	+150%	-77.38%
	K-means	-65%	0	-61.9%

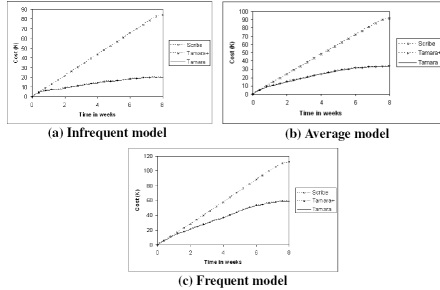
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Experiments on Synthetic Data 2/4

Operation Cost Over Time



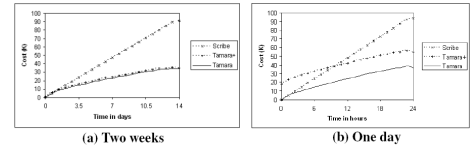
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Experiments on Synthetic Data 3/4

Effect of User Subscription Rate



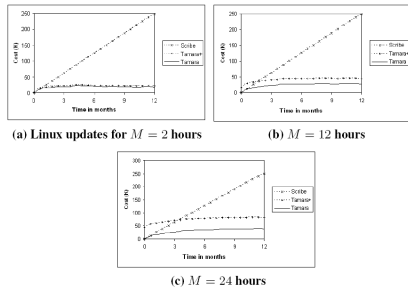
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Experiments on Synthetic Data 4/4

Effect of Maintenance Interval



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Discussion

- Stability of Dynamic Algorithm is not addressed
- Functions used in Cost Model are too generic
 - Do not take network distance into consideration
- Adjustment cost is ignored
- Excessive use of heuristics
 - Triggering benefit estimation process
 - Greedy subscription algorithm

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