Minimizing Communication Cost in Distributed Multi-query Processing

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Outline

- Motivation & Problem Formulation
- Summary of Our Results
- Multiple Queries
 - NP-Hardness
 - Polynomial time algorithm for tree communication networks
 - Approximation algorithms
- Experiments
- Future Work

Motivation

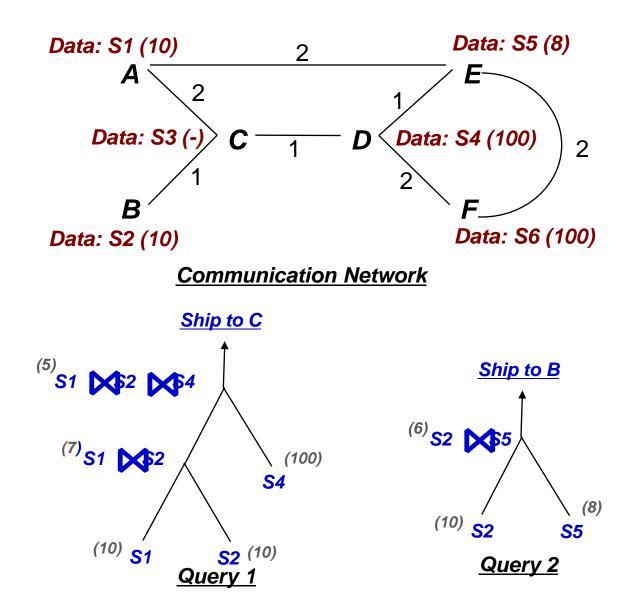
- Emergence of large-scale distributed query processing
 - Scientific federations like SkyServer, GridDB
 - Publish-subscribe systems and content delivery networks
 - Distributed data streams and web sources
 - Sensor networks
 - Large scale data analytics (MapReduce, Hadoop)

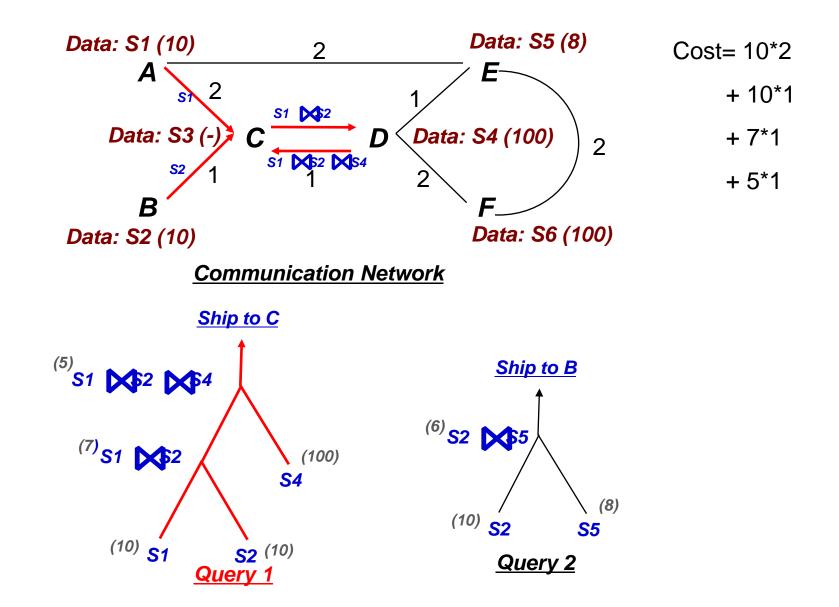
Motivation

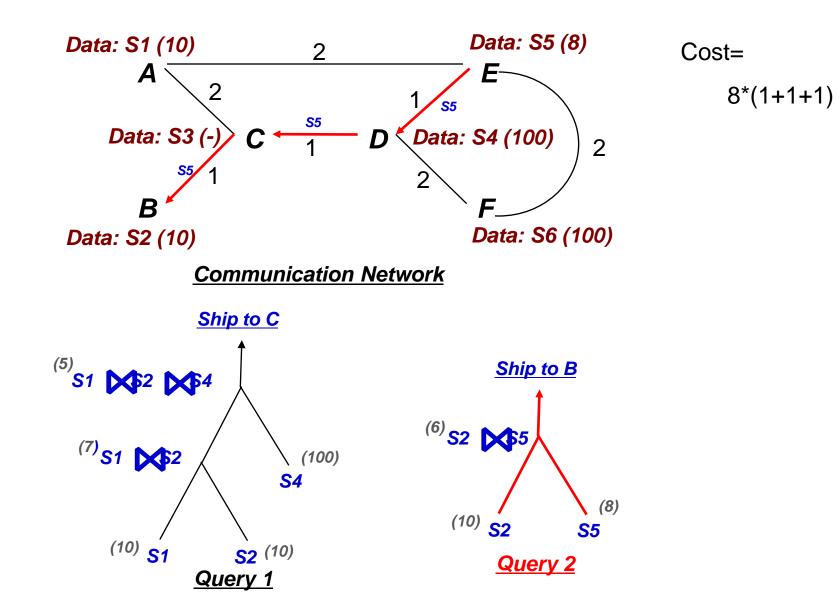
- Need to support:
 - Very large datasets and/or
 - Large numbers of users and queries
- Minimization of communication cost often a key problem
 - Network utilization in Internet-scale systems
 - Energy consumed in sensor networks

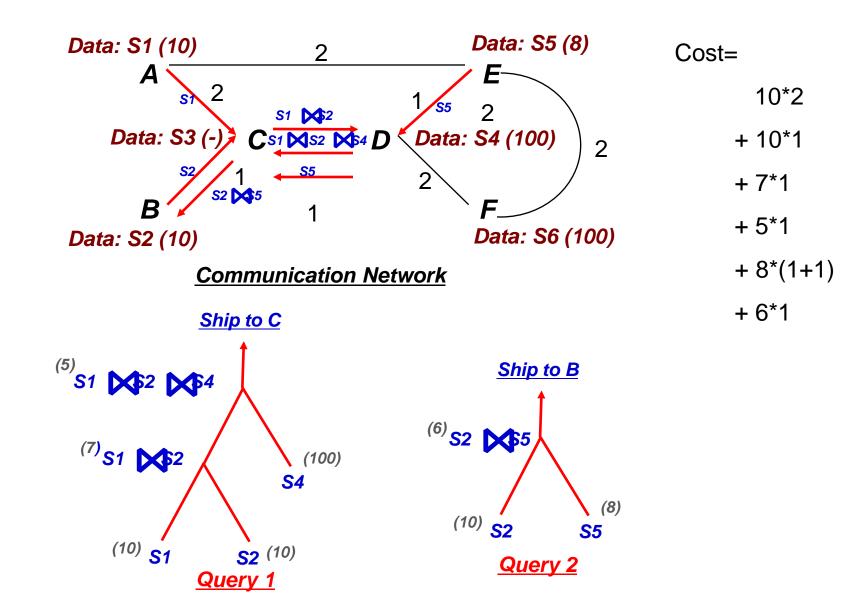
Challenges:

- How to choose query plan
- How to ship data across the network to implement these plans



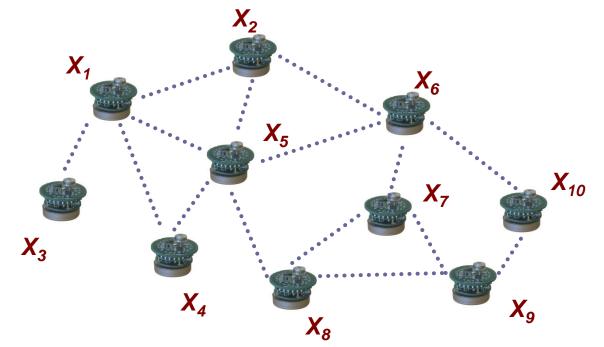






Example: Sensor Network Aggregates

Silberstein and Yang, 2007] Many-to-many Aggregation



Q1 (issued by X_7): $2X_1 + 3X_2 + X_4$ **Q2** (issued by X_8): $X_1 + X_2 + X_3 + X_4$ **Q3** (issued by X_6): $2X_2 + 3X_3 + X_5$

Problem Formulation

Input:

- Communication Network G(V,E)
 - Edge weights indicate the communication costs
- **Data sources:** S_1, \dots, S_n
- A set of queries: Q_1, Q_2, \ldots
- □ For each query Q, a query plan (tree) is given
 - No join order optimization
- Goal:
 - Minimize the communication cost of executing the queries

Our Results

Single Query

Polynomial time solvable (by standard dynamic programming)

Multiple Queries

- NP-Hard on general communication networks
- Polynomial time solvable on tree communication networks
- O(logn)-approximation for general communication networks
- O(1)-approximation for some special cases

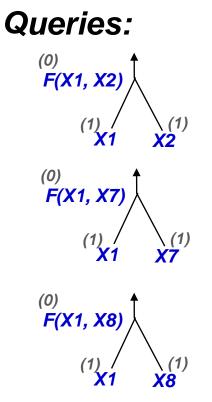
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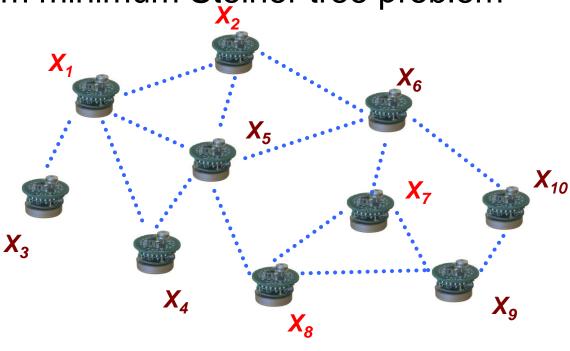
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Complexity

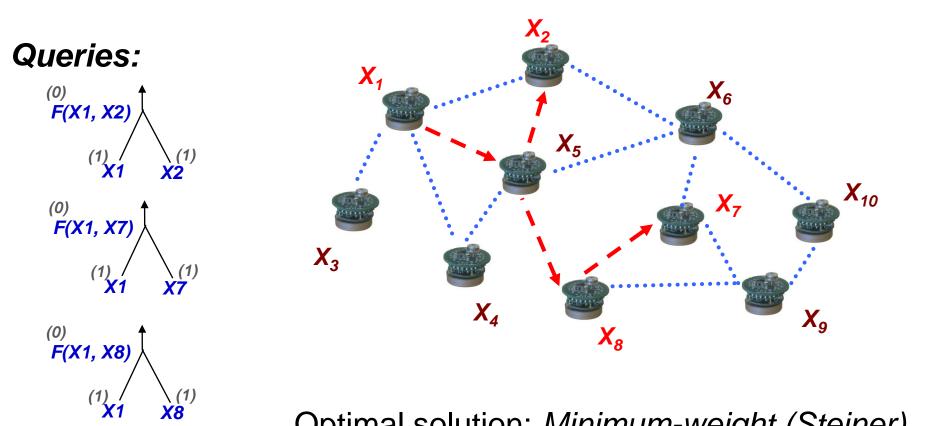
NP-Hard for general communication networks

Reduction from minimum Steiner tree problem





ComplexityNP-Hard for general communication networks



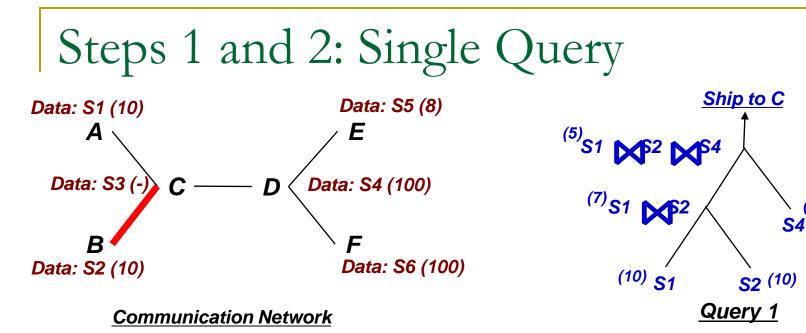
Optimal solution: *Minimum-weight (Steiner)* tree connecting X1, X2, X7, X8

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High-level Overview of Our Approach

- 1. Combine all the query plans into a single hypergraph
 - That explicitly captures the data movement sharing opportunities
- 2. For each edge, decide which data are communicated along that edge
 - By solving a *hypergraph min-cut* problem
- 3. Combine the local solutions into a single global solution

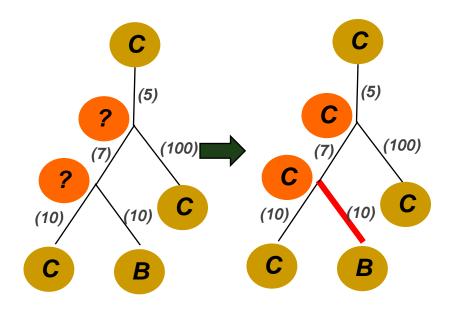


Solving for edge (B, C)

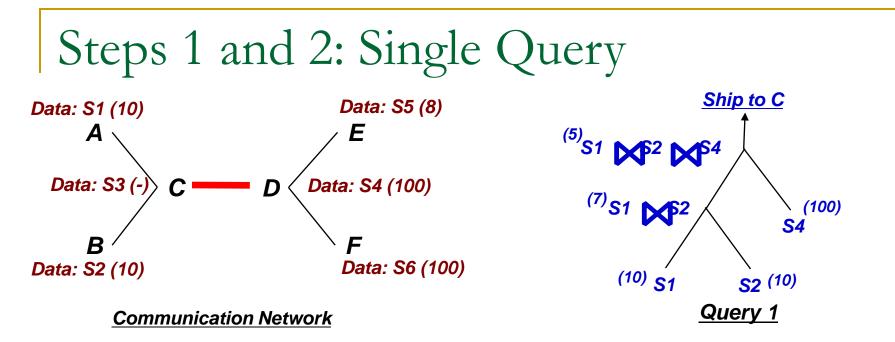
- 1.Label the nodes as B or C if possible, ? Otherwise
- 2. Solve a partition problem to resolve ?'s
- 3. "Cut" edges indicate the data movement

Solution

S2 moves across edge (B, C)



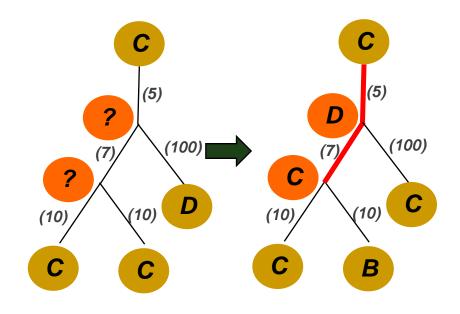
(100)

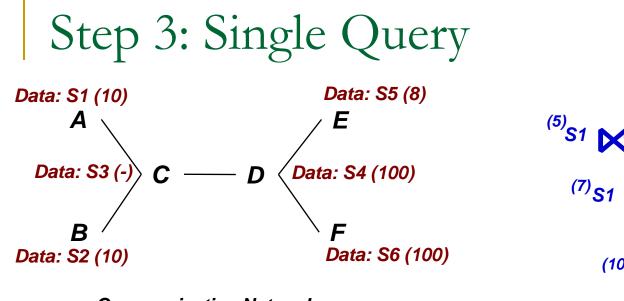


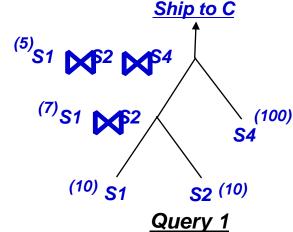
Solving for edge (C, D)

Solution

| S1S2 | moves from C to D |
|--------|-------------------|
| S1S2S4 | moves from D to C |







Communication Network

Solution for (B, C)

S2 moves from B to C Solution for (C, D) S1S2 moves from C to D S1S2S4 moves from D to C Solution for (A, C)

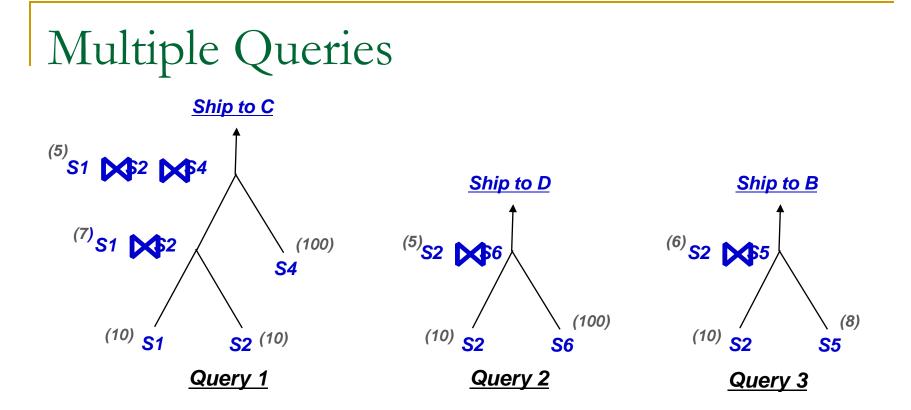
Key Question:

Are these movements consistent with each other ?

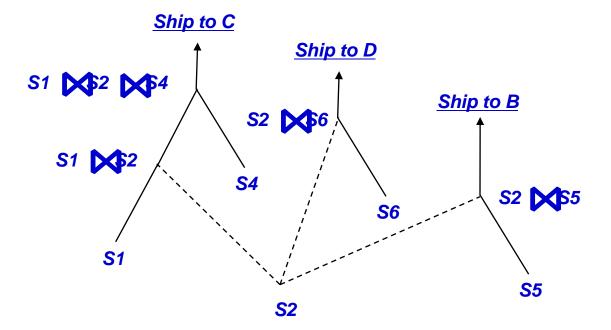
Answer:

Yes, given unique min-cut solutions.

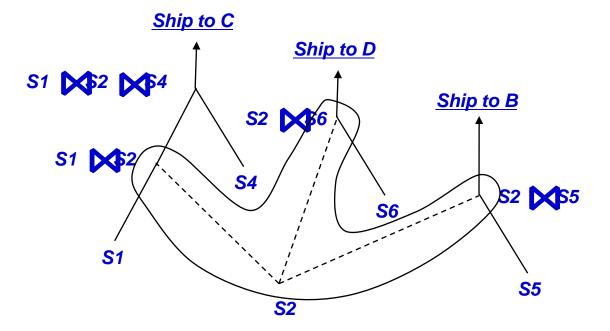
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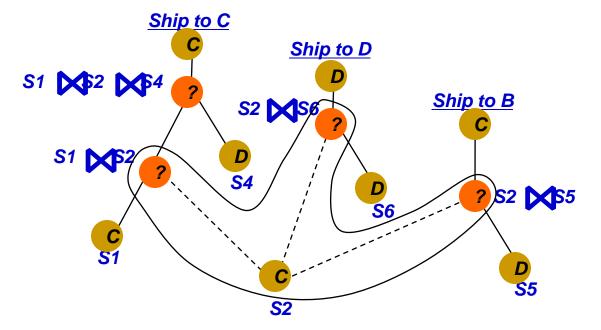
We create hyperedges



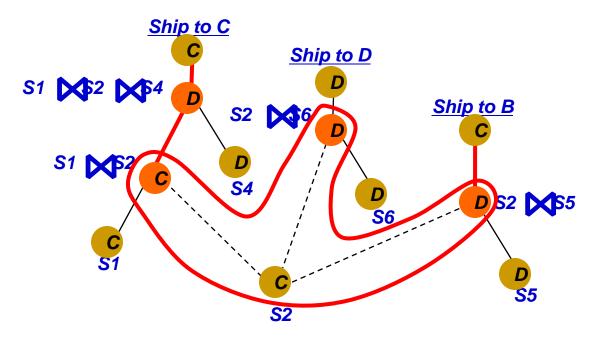
We create hyperedges



Solve for edge (C,D)



Solution for edge (C,D) : Hypergraph Partition



Why hyperedges ? So we don't count data movements multiple times (e.g. Data item S2 above)

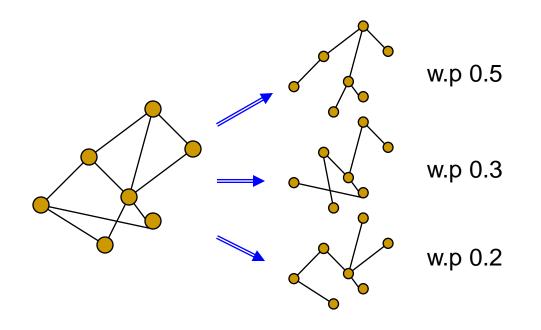
- Add hyperedges corresponding to shared data sources
- For each edge, solve a hypergraph partition problem, (which can be solved by min-cut algorithm)
- Again we can prove the consistency of these local movements
- Complexity: *m* max-flow min-cut computations where *m* is #edges in the tree

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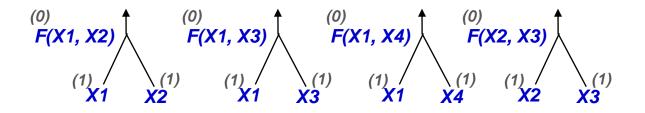
O(logn)-approximation for General Networks

- 1. Construct a distribution of trees base on the communication network by using metric embedding [Fakcharoenphol/Rao/Talwar 06]
- 2. Randomly pick a tree and solve the problem on the tree optimally
- 3. Map the solution back to the original network



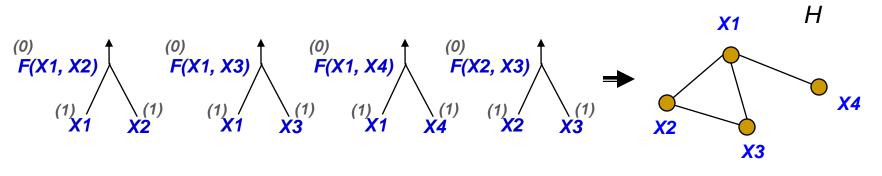
[FRT 06] Any metric can be embedded into a distribution of tree metrics with an O(log n)-distortion. O(1)-approximations for some special cases

 "Pairs Problem": Each query has only two data sources. The size of the result is zero.



O(1)-approximations for some special cases

"Pairs Problem": Each query has only two data sources. The size of the result is zero.



We can capture the queries by a graph H

- □ *H* is a tree : *2p*
- □ *H* is planar : *6p*
- $\Box Deg(H) <= D : D$

Where *p* is the approixmation ratio for minimum Steiner tree problem

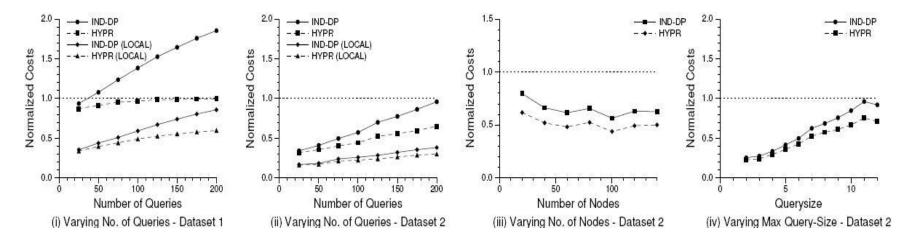
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 - On trees, Polynomial time Algorithm
 - Approximations
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Experiments

IND-DP: optimize each query separately

HYPR: the hypergraph min-cut approach



Communication network: a spanning tree over a set of point randomly distributed in a 2-d plane

Datasets1: the sizes of sources are identical. Datesets2: the sizes of sources are randomly chosen from a skewed distribution. Workload: Each query is over a randomly chosen subset of sources. LOCAL: all queries are chosen to be geometrically co-located sources

Future Directions

- Constant approximations for general communication networks
- Sharing intermediate results generated during query execution
- Online algorithms for handling new queries

Thanks