

# CS 764: Topics in Database Management Systems Lecture 7: Distribution Query Optimization

Xiangyao Yu 9/29/2021

# Today's Paper: Query Optimization

DISTRIBUTED QUERY PROCESSING IN A RELATIONAL DATA BASE SYSTEM

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ABSTRACT: In this paper we present a new algorithm for retrieving and updating data from a distributed relational data base. Within such a data base, any number of relations can be distributed over any number of sites. Moreover, a user supplied distribution criteria can optionally be used to specify what site

The algorithm is an efficient way to process any query by "breaking" the qualification into separate "pieces" using a few simple heuristics. The cost criteria considered are minimum response time and minimum communications traffic. In addition, the algorithm can optimize separately for two models of a communication network representing respectively ARPANET and ETHERNET like networks. This algorithm is being implemented as part of the INGRES data base system.

KEYWORDS AND PHRASES: Distributed databases, relational model, distributed decomposition, communication networks, distribution criteria.

#### I Introduction

In this paper we are concerned with algorithms. algorithms for processing data base commands that involve data from multiple section II we formalize the problem by machines in a distributed data base indicating our view of a distributed data environment. These algorithms are being base and the interactions to be solved. implemented as part of our work in extend- Then, in section III we discuss our model ing INGRES [HELD75, STON76] to manage a for the computer network. In section IV a distributed data base. As such, we are detailed algorithm is presented for hanconcerned with processing interactions in dling the decomposition of queries in a the data sublanguage, QUEL. The specific distributed environment. There are a few data model that we use is discussed in Sec- complications concerning updates and aggretion II. Some of our initial thoughts on gates in a distributed data base which are these subjects have been presented else- covered in sections V and VI. Lastly, in where [STON77, WONG77].

We are not concerned here with control of concurrent updates or multiple copies [THOM75, LAMP76, ROTH77, CHU76]. Rather we assume that these are handled by a separate

mechanism or can be integrated into our

This paper is organized as follows: In section VII we draw some conclusions.

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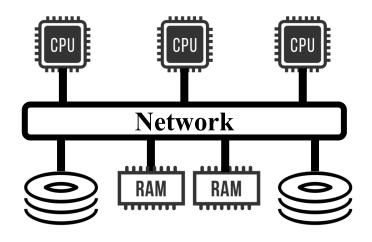
# Agenda

Distributed database architecture

Data partitioning

Parallel operators

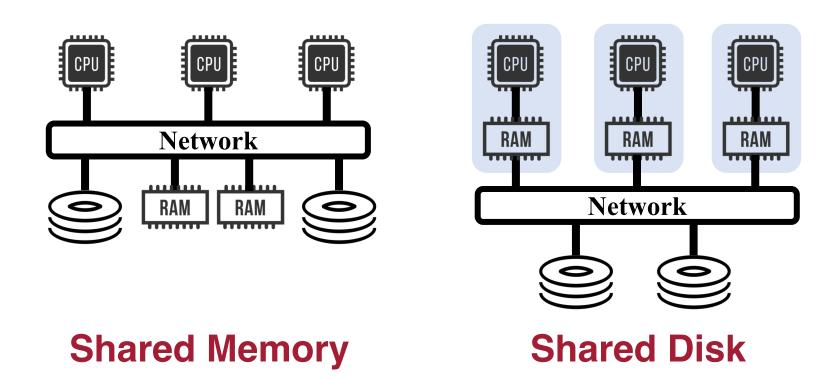
Distributed query optimization



### **Shared Memory**

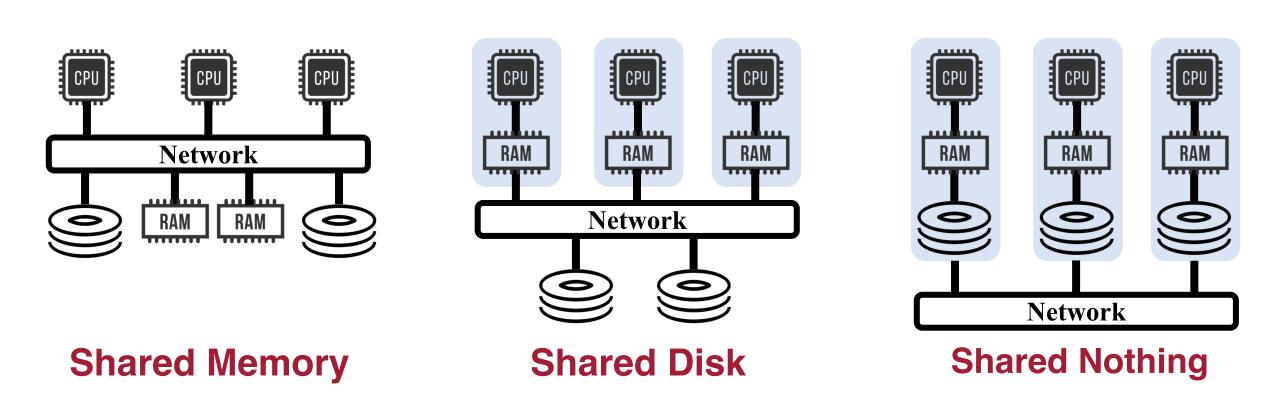
Example: Multicore shared-memory machine

Scale: Single machine



Example: Network attached storage (NAS) and storage area network (SAN), some Oracle and IBM database systems

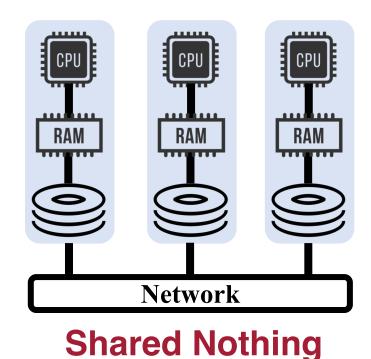
Scale: Cluster with tens of machines



Example: Modern massively parallel databases including Google Spanner, Redshift, CosmosDB, etc.

Scale: Any number of machines

# Shared-Nothing Architecture

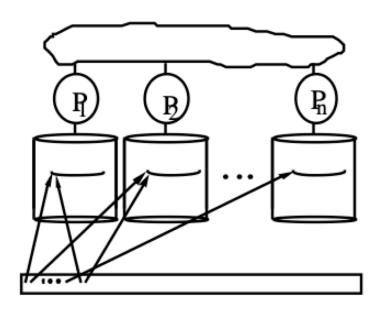


### Advantages:

- High scalability
- High availability
- Good data locality

### Challenges:

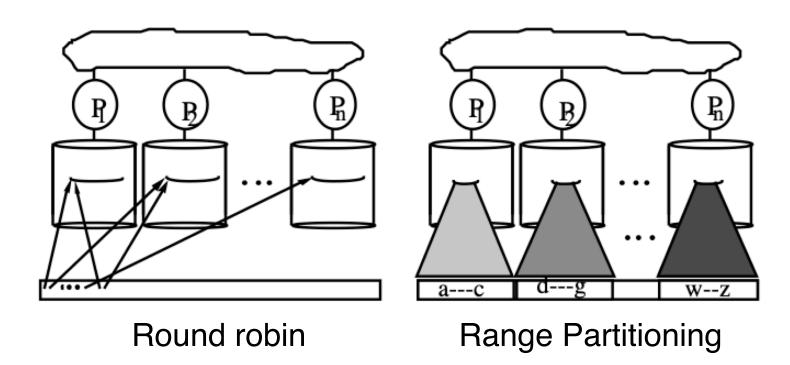
- Must partition data
- Network overhead



Round robin

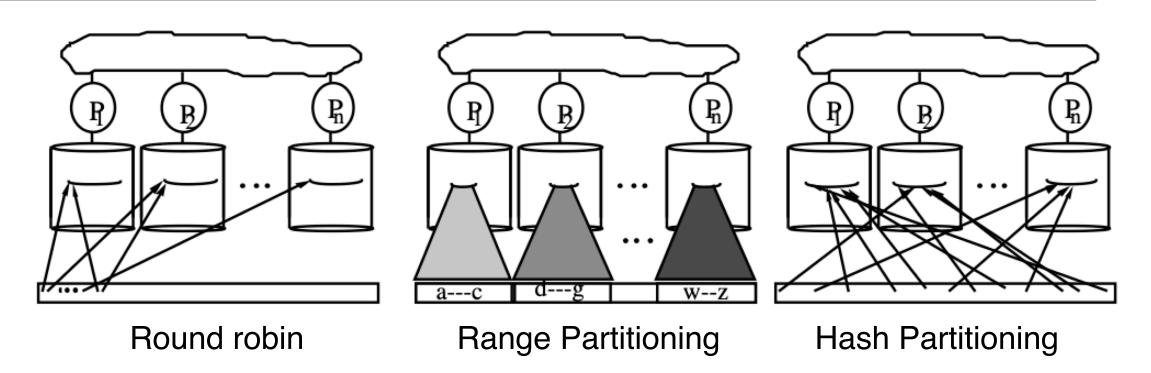
### Map tuple *i* to disk (*i mode n*)

- Advantage: Simplicity, good load balancing
- Disadvantage: Hard to identify the partition of a particular record



Map contiguous attribute ranges to partitions

- Advantage: Good locality due to clustering
- Disadvantage: May suffer from skewness

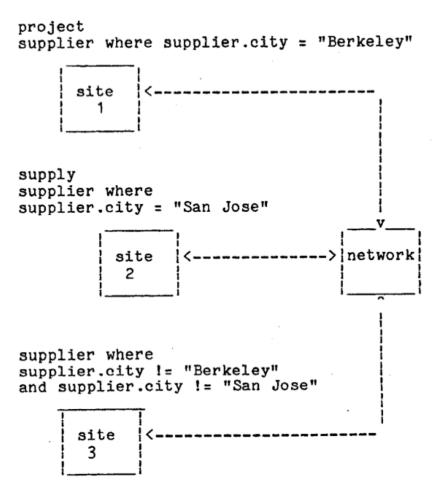


Map based on the hash value of tuple attributes

- Advantage: Good load balance, low skewness
- Disadvantage: Bad locality

Generally, the partitioning function (distribution criteria) can be any function that maps tuples to partition ID

For example:



# Algorithm Flowchart

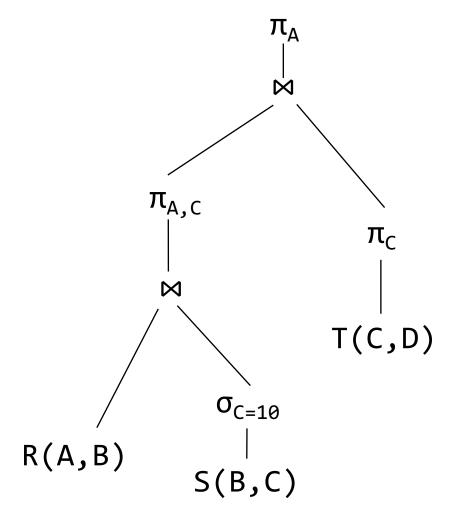
```
Perform all single-table operations

While (exists next piece of query) {

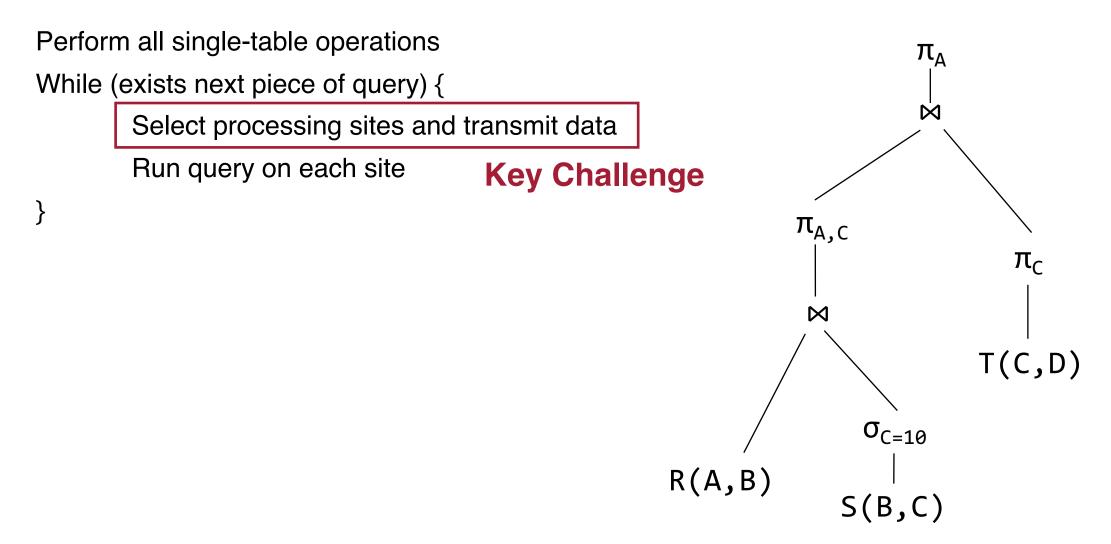
Select processing sites and transmit data

Run query on each site

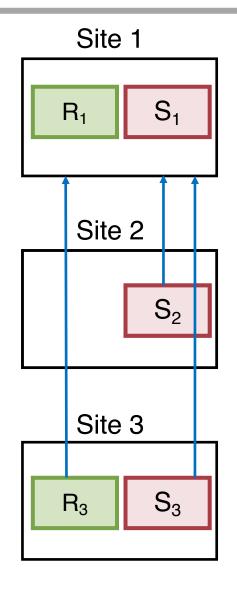
}
```



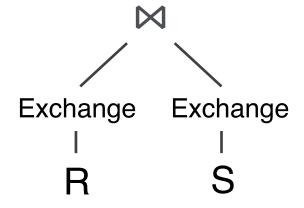
# Algorithm Flowchart



# Join - Single-Site



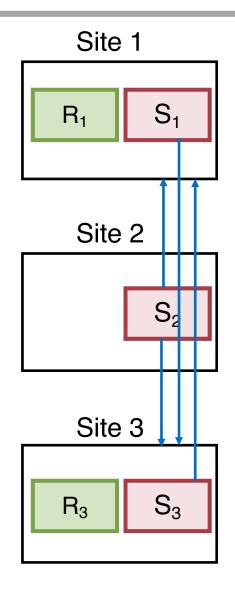
Query plan



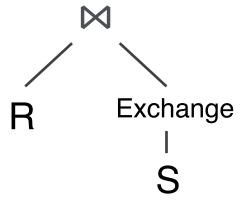
Solution 1: send all the involved tables to a single site

- Advantage: Single-site query execution is a solved problem
- Disadvantage: (1) Single site execution can be slow (2) Data may not fit in single site's memory or disk

### Join – Broadcast



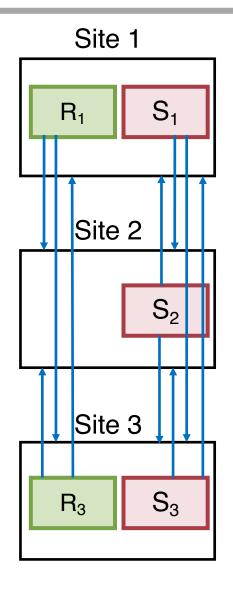
Query plan



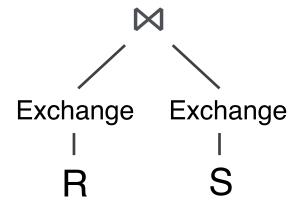
Solution 2: Keep one relation partitioned and broadcast the other relation to all sites

- Advantage: One relation does not need to move
- Disadvantage: Still need to broadcast the other relation to all sites

# Join - Co-partition



Query plan



Solution 3: Partition both relations using the join key

- Advantage: Each site has less data to process
- Disadvantage: Both relations are shuffled (if not already partitioned based on join key)

### Distributed Join

### Single-site

Preferred when both relations are small

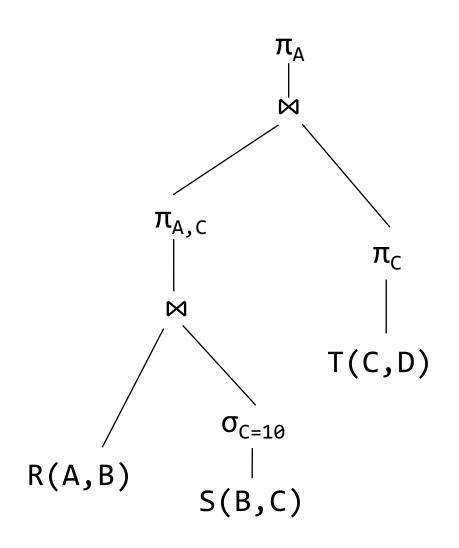
#### **Broadcast**

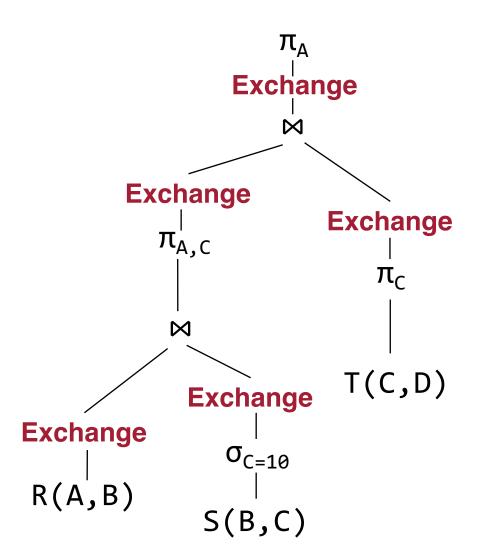
Preferred when one relation is small

### Co-partition

Preferred when both relations are large

# Distributed Query Optimization

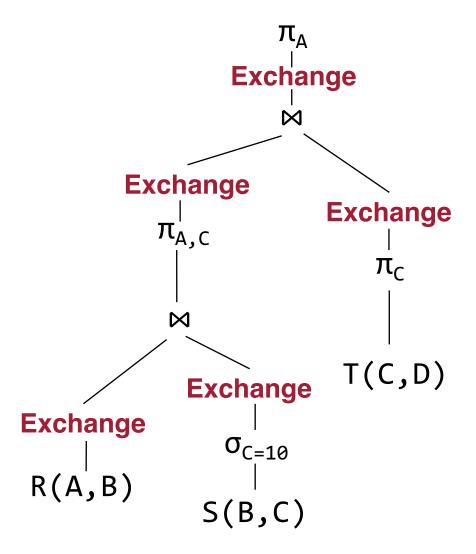




# Distributed Query Optimization

### Extra design complexity

- Which exchange operator to use?
- Which nodes to use to run the operator?



#### Site 1

project (200 tuples)
supplier (50 tuples)

#### Join order

- (project ⋈ supply) ⋈ supplier
- project ⋈ (supply ⋈ supplier)

#### Site 2

supply (400 tuples)
supplier (50 tuples)

#### Site 1

project (200 tuples)
supplier (50 tuples)

#### Site 2

supply (400 tuples)
supplier (50 tuples)

#### Join order

- (project ⋈ supply) ⋈ supplier
- project ⋈ (supply ⋈ supplier)

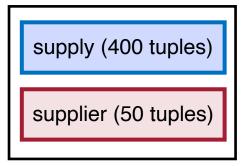
### Plan 1: Send everything to Site 2

Network traffic 250 tuples

#### Site 1

project (200 tuples)
supplier (50 tuples)

#### Site 2



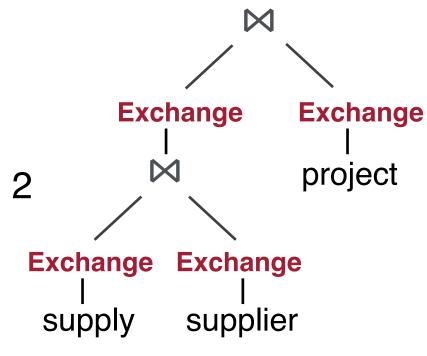
### Join order

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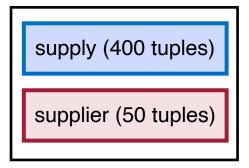
Plan 2:



#### Site 1

project (200 tuples)
supplier (50 tuples)

#### Site 2



#### Join order

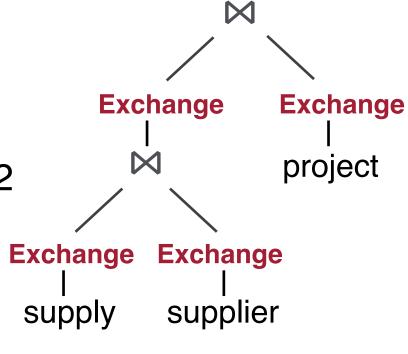
- (project ⋈ supply) ⋈ supplier
- project ⋈ (supply ⋈ supplier)

### Plan 1: Send everything to Site 2

- Network traffic 250 tuples

#### Plan 2:

- 1st join: 50 tuples network traffic (on site 2)
- 2<sup>nd</sup> join: depends result of 1<sup>st</sup> join

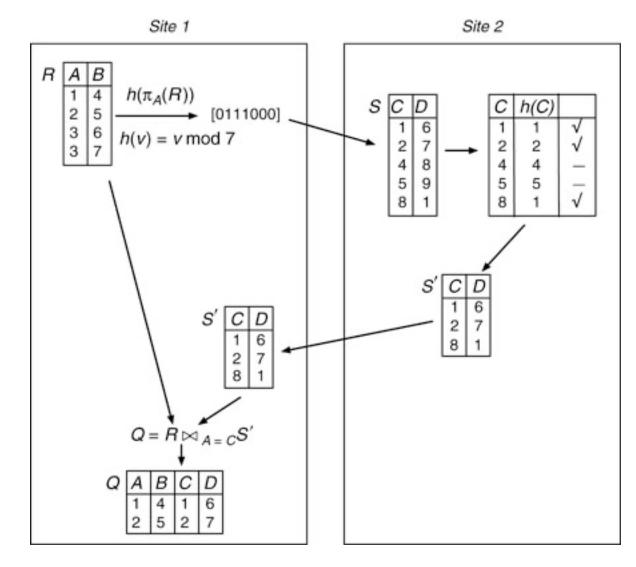


# Specialized Parallel Operators

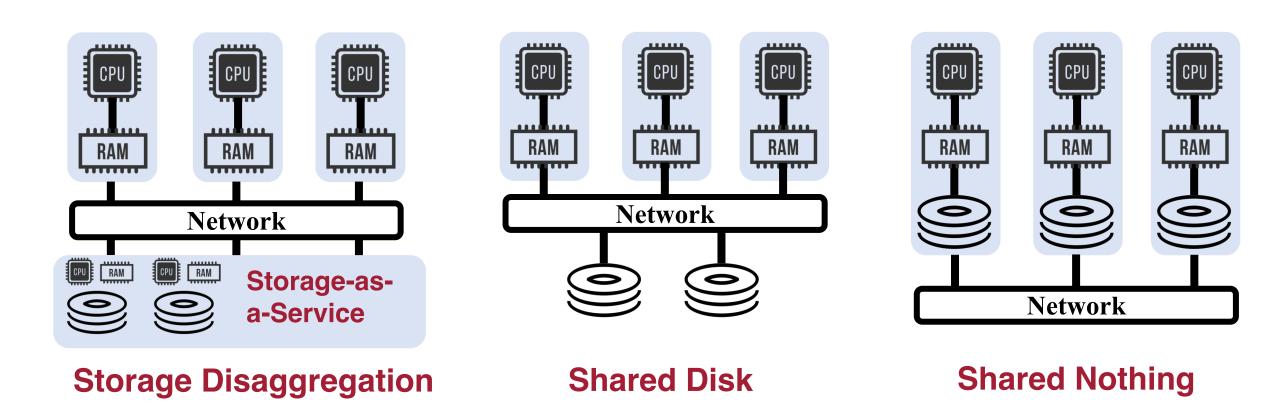
### Semi-join

• Example:

```
SELECT *
FROM T1, T2
WHERE T1.A = T2.C
```



<sup>\*</sup> Source: Sattler KU. (2009) Semijoin. Encyclopedia of Database Systems.



Storage-disaggregation architecture is popular in cloud-native databases

- Features: (1) in-storage computation, (2) high-availability, (3) shared access to storage
- More on this topic in last few lectures

# Q/A – Distributed Query Optimization

Details of reduction algorithm?

How is the paper related to modern distributed databases?

How does master-slave failover work?

Fragments being a project of the relation?

Distribute data without distribution logic?

More on updates? (will cover in next few lectures)

Which is the bottleneck, network or compute?

### Before Next Lecture

Submit review for

Jim Gray, et al., <u>Granularity of Locks and Degrees of</u> <u>Consistency in a Shared Data Base</u>. Modelling in Data Base Management Systems, 1976