

# CSE 4125: Distributed Database Systems

## Chapter – 4 (Part – D)

### Distributed Database Design

# **The Design of Vertical Fragmentation**

# Vertical Fragmentation

- ✓ Partitioning the attributes of a relation into a set of smaller relations.

So that many of the applications will run on only one fragment.

- ✓ Vertical Fragmentation can be done in the following ways:

- *Clustering*: sets can be overlapped

- *Partitioning*: sets must be disjoint.

## Approaches:

**Grouping (clustering):** Progressively assigning each attribute to constitute clusters.

A	B

A	B	C

A	B	C	D

**Splitting (partitioning):** Progressively splitting global relations into fragment.

A	B	C	D

A	B	C

A	B

A	B

A	B

# Bond Energy Algorithm (BEA)

Steps:

1. Attribute Usage Matrix
2. Attribute Affinity Matrix
3. Clustered Affinity Matrix
4. Partitioning

# Bond Energy Algorithm (BEA)

## Example

### PROJ

<b>PNO</b>	<b>PNAME</b>	<b>BUDGET</b>	<b>LOC</b>
P1	Instrumental	150,000	Montreal
P2	Database Dev	135,000	New York
P3	CAM/CAD	250,000	New York
P4	Maintenance	310,000	Orlando

Consider the following 4 queries for relation PROJ, where PNO is the primary key column of the table.

$q_1$ :    **SELECT**    BUDGET  
           **FROM**     PROJ  
           **WHERE**    PNO=Value

$q_2$ : **SELECT**    PNAME,BUDGET  
           **FROM**     PROJ

$q_3$ :    **SELECT**    PNAME  
           **FROM**     PROJ  
           **WHERE**    LOC=Value

$q_4$ : **SELECT**    SUM(BUDGET)  
           **FROM**     PROJ  
           **WHERE**    LOC=Value

Assume that, **PROJ** relation is located in three different sites. The access frequency of each query for each site is stated below –

	<b>S1</b>	<b>S2</b>	<b>S3</b>
<b>q1</b>	15	20	10
<b>q2</b>	5	0	0
<b>q3</b>	25	25	25
<b>q4</b>	3	0	0

FM

Using the Bond Energy Algorithm, group the columns of the table and after that split the columns vertically at required position with the help of goal function.

Note: You should show Attribute Affinity Matrix, Clustered Affinity Matrix, and the calculation for each of the ordering. Take the first two columns as the starting bonding state.

# Step 1: Attribute Usage Matrix

Let, A1 = PNO, A2 = PNAME, A3 = BUDGET, A4 = LOC

1

	A1	A2	A3	A4
Q1	1	0	1	0
Q2	0	1	1	0
Q3	0	1	0	1
Q4	0	0	1	1

2

$q_1$ : **SELECT** BUDGET  
**FROM** PROJ  
**WHERE** PNO=Value

$q_2$ : **SELECT** PNAME,BUDGET  
**FROM** PROJ

$q_3$ : **SELECT** PNAME  
**FROM** PROJ  
**WHERE** LOC=Value

$q_4$ : **SELECT** SUM(BUDGET)  
**FROM** PROJ  
**WHERE** LOC=Value



# Step 2: Attribute Affinity Matrix

	A1	A2	A3	A4
A1	45	0	45	0
A2	0	80	5	75
A3	45	5	53	3
A4	0	75	3	78

3

1

	S1	S2	S3	SUM
Q1	15	20	10	45
Q2	5	0	0	5
Q3	25	25	25	75
Q4	3	0	0	3

2

	A1	A2	A3	A4
Q1	1	0	1	0
Q2	0	1	1	0
Q3	0	1	0	1
Q4	0	0	1	1

# Step 3: Clustered Affinity Matrix

Consider the following *AA* matrix and the corresponding *CA* matrix where  $A_1$  and  $A_2$  have been placed. **Place  $A_3$ :**

AA =

	A1	A2	A3	A4
A1	45	0	45	0
A2	0	80	5	75
A3	45	5	53	3
A4	0	75	3	78

CA =

Starting Bonding State				
	A1	A2		
A1	45	0		
A2	0	80		
A3	45	5		
A4	0	75		

New attribute always  
will be in middle

# Global Affinity Measure

$$cont(A_i, A_k, A_j) = 2bond(A_i, A_k) + 2bond(A_k, A_j) - 2bond(A_i, A_j)$$

Contribution  
Function

$$bond(A_x, A_y) = \text{SUMMATION}(\text{For all rows } (A_x * A_y))$$

Example –

$$cont(A_1, A_4, A_2) = 2bond(A_1, A_4) + 2bond(A_4, A_2) - 2bond(A_1, A_2)$$

$$bond(A_1, A_4) = 45*0 + 0*75 + 45*3 + 0*78 = 135$$

$$bond(A_4, A_2) = 11865$$

$$bond(A_1, A_2) = 225$$

	A1	A2	A3	A4
A1	45	0	45	0
A2	0	80	5	75
A3	45	5	53	3
A4	0	75	3	78

$$cont(A_1, A_4, A_2) = 2*135 + 2*11865 - 2*225 = 23550$$

# Step 3: Clustered Affinity Matrix

Consider the following AA matrix and the corresponding CA matrix where  $A_1$  and  $A_2$  have been placed. **Place  $A_3$ :**

AA =

	A1	A2	A3	A4
A1	45	0	45	0
A2	0	80	5	75
A3	45	5	53	3
A4	0	75	3	78

CA =

	A1	A2		
A1	45	0		
A2	0	80		
A3	45	5		
A4	0	75		

Starting Bonding State

Ordering (0-3-1) :

$$\begin{aligned} cont(A_0, A_3, A_1) &= 2bond(A_0, A_3) + 2bond(A_3, A_1) - 2bond(A_0, A_1) \\ &= 2 * 0 + 2 * 4410 - 2 * 0 = 8820 \end{aligned}$$

Ordering (1-3-2) :

$$\begin{aligned} cont(A_1, A_3, A_2) &= 2bond(A_1, A_3) + 2bond(A_3, A_2) - 2bond(A_1, A_2) \\ &= 2 * 4410 + 2 * 890 - 2 * 225 = 10150 \end{aligned}$$

Ordering (2-3-4) :

$$\begin{aligned} cont(A_2, A_3, A_4) &= 2bond(A_2, A_3) + 2bond(A_3, A_4) - 2bond(A_2, A_4) \\ &= 2 * 890 + 2 * 0 - 2 * 0 = 1780 \end{aligned}$$

## Step 3: Continued

Therefore, the CA matrix has to form

	<b>A1</b>	<b>A3</b>	<b>A2</b>	
<b>A1</b>	45	45	0	
<b>A2</b>	0	5	80	
<b>A3</b>	45	53	5	
<b>A4</b>	0	3	75	

Similarly, Now for placing A4 do the calculations. You must have to show the calculation in the exam.

Place A4: All possible orderings will be (0-4-1), (1-4-3), (3-4-2), (2-4-5)

(2-4-5) ordering will be the highest value, means A4 is after A2.

## Step 3: Continued

When  $A_4$  is placed, the final form of the  $CA$  matrix (**after column organization**) is

	<b>A1</b>	<b>A3</b>	<b>A2</b>	<b>A4</b>
<b>A1</b>	45	45	0	0
<b>A2</b>	0	5	80	75
<b>A3</b>	45	53	5	3
<b>A4</b>	0	3	75	78

The final form of the  $CA$  matrix (**after row organization**) is

	<b>A1</b>	<b>A3</b>	<b>A2</b>	<b>A4</b>
<b>A1</b>	45	45	0	0
<b>A3</b>	45	53	5	3
<b>A2</b>	0	5	80	75
<b>A4</b>	0	3	75	78

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

## Clustered Affinity Matrix (CA)

Cluster 1:  $A_1$  &  $A_3$

Cluster 2:  $A_2$  &  $A_4$

What if the clustered affinity matrix looks like this? →

	A1	A3	A2	A4
A1	45	0	75	45
A3	0	53	5	0
A2	75	5	3	0
A4	45	0	0	78

	A1	A3	A2	A4
A1	45	0	75	45
A3	0	53	5	0
A2	75	5	3	0
A4	45	0	0	78

First Left Rotate:

Column

1

	A3	A2	A4	A1
A1	0	75	45	45
A3	53	5	0	0
A2	5	3	0	75
A4	0	0	78	45

Row

2

	A3	A2	A4	A1
A3	53	5	0	0
A2	5	3	0	75
A4	0	0	78	45
A1	0	75	45	45



# Clustering Summary (Steps 1,2,3)

- We need AUM that reflects the query-attribute relationship
- AUM and FM are used to make AA
- Global Affinity Measure is used to establish the clusters of attributes
- Stronger affinities attributes and weaker ones are grouped in CA

# Step 4: Partitioning

We define -

**TQ** = set of applications that access only TA

**BQ** = set of applications that access only BA

**OQ** = set of applications that access both TA and BA

	A1	A3	A2	A4
A1	TA			
A3				
A2			BA	
A4				

# Step 4: Partitioning

We define -

**CTQ** = number of accesses to attributes by applications that access only TA

**CBQ** = number of accesses to attributes by applications that access only BA

**COQ** = number of accesses to attributes by applications that access both TA and BA

# Step 4: Partitioning

Goal Function –

Find the point  $z$  along the diagonal that maximizes

$$z = (CTQ * CBQ) - (COQ * COQ)$$

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

# Step 4: Partitioning

Goal Function –

Find the point  $z$  along the diagonal that maximizes

$$z = (CTQ * CBQ) - (COQ * COQ)$$

## Setting 1

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

## Setting 2

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

## Setting 3

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

# Step 4: Partitioning

## Setting 1

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

## AUM

	A1	A2	A3	A4
Q1	1	0	1	0
Q2	0	1	1	0
Q3	0	1	0	1
Q4	0	0	1	1

## FM

	S1	S2	S3
q1	15	20	10
q2	5	0	0
q3	25	25	25
q4	3	0	0

$$TQ = \{\}$$

$$CTQ = 0$$

$$BQ = \{q2, q3, q4\}$$

$$CBQ = 5+0+0+25+25+25+3+0+0 = 83$$

$$OQ = \{q1\}$$

$$COQ = 15 + 20 + 10 = 45$$

$$Z1 = (CTQ * CBQ) - (COQ * COQ) = (0 * 83) - (45 * 45) = -2025$$

# Step 4: Partitioning

## Setting 2

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

## AUM

	A1	A2	A3	A4
Q1	1	0	1	0
Q2	0	1	1	0
Q3	0	1	0	1
Q4	0	0	1	1

## FM

	S1	S2	S3
q1	15	20	10
q2	5	0	0
q3	25	25	25
q4	3	0	0

$$TQ = \{q1\}$$

$$CTQ = 15+20+10 = 45$$

$$BQ = \{q3\}$$

$$CBQ = 25 + 25 + 25 = 75$$

$$OQ = \{q2, q4\}$$

$$COQ = 5+0+0+3+0+0 = 8$$

$$Z2 = (CTQ * CBQ) - (COQ * COQ) = (45 * 75) - (8 * 8) = 3311$$

# Step 4: Partitioning

## Setting 3

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

## AUM

	A1	A2	A3	A4
Q1	1	0	1	0
Q2	0	1	1	0
Q3	0	1	0	1
Q4	0	0	1	1

## FM

	S1	S2	S3
q1	15	20	10
q2	5	0	0
q3	25	25	25
q4	3	0	0

$$TQ = \{q1, q2\}$$

$$CTQ = 15+20+10+5+0+0 = 50$$

$$BQ = \{\}$$

$$CBQ = 0$$

$$OQ = \{q3, q4\}$$

$$COQ = 25+25+25+3+0+0 = 78$$

$$Z3 = (CTQ * CBQ) - (COQ * COQ) = (50 * 0) - (78 * 78) = -6084$$



# Step 4: Partitioning

Goal Function z2 is maximum with setting 2.

## Setting 2

	A1	A3	A2	A4
A1	45	45	0	0
A3	45	53	5	3
A2	0	5	80	75
A4	0	3	75	78

Two Fragments:

PROJ1 = {A1, A3}  
= {PNO, BUDGET}

PROJ2 = {A1, A2, A4}  
= {PNO, PNAME, LOC}

PNO is the primary key

# Vertical Fragmentation

Introduces *replication*.

- Tuple identifier.

Convenient for read-only application.

- Why?

Not convenient for update application.

- Why?

## Answer to why:

Let us consider what happens when two fragments  $R_1$  and  $R_2$  are overlapping ;i.e., there exists a set of attributes  $I$  which belong to both  $R_1$  and  $R_2$ . Assume that  $R_1$  and  $R_2$  are at **sites 1 and 2**.

Then **read applications** at site 1, using attributes of  $I$  together with other attributes of  $R_1$ , are local to site 1; likewise, read applications at site 2, using attributes of  $I$  together with other attributes of  $R_2$ , are local to site 2.

However, update applications which-change the value of attributes of  $I$  must reference them at both sites.

# Exercise

Consider the applications “AP1”, “AP2”, “AP3” and “AP4” as shown below. These applications work on the **EMP** relation defined as **EMP(EmpID, Name, Loc, Dept)**, where **EmpID** is the primary key column of the table.

AP1: SELECT **EmpID** FROM **EMP** WHERE **Dept** = “Payroll”

AP2: SELECT **Dept** FROM **EMP**

AP3: UPDATE **EMP** SET **Loc** = “Chittagong” WHERE **Name** = “Kalam”

AP4: UPDATE **EMP** SET **Loc** = “Comilla” WHERE **EmpID** = 109288

Assume that there is only **one site** and the access frequency of AP1, AP2, AP3 and AP4 is 3, 7, 4, 3 respectively.

Using the Bond Energy Algorithm, group the columns of the table and after that split the columns vertically at required position with the help of goal function.

Note: You should show Attribute Affinity Matrix, Clustered Affinity Matrix, and the calculation for each of the ordering, goal function. Take the first two columns as the starting bonding state.