Parallel Query Processing

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Concepts and goals

Main problem: I/O bottleneck

transferring the data from disk to main memory

Solution

increase the I/O bandwidth through parallelism



Ideal advantages of parallel systems

High performance

short response time or total time good load balancing among processors

High availability

handling failures of hardware elements redundancy and consistency

Extensibility

more processing and storage power can be added

Speedup and scaleup

Speedup

Scaleup

twice as much hardware results in half elapsed time

twice as much hardware can perform twice as large a task in the same elapsed time

Bottlenecks

Start-up many processors \Rightarrow long start-up time

Interference more processors \Rightarrow more communications

Data skew it makes data distribution difficult

Multiprocessor architectures

Shared-memory architectures

- **Shared-disk architectures**
- **Shared-nothing architectures**

Hybrid architectures

Shared-memory architectures

More processors - one memory



Characteristics

any CPU has access to any memory module or disk unit

Advantages

simplicity load balancing **Disadvantages**

cost little extensibility low availability

Shared-disk architectures

More processors and memory units



Characteristics

any CPU has access to any disk unit but exclusive access to its main memory

Advantages

cost extensibility load balancing availability easy migration

Disadvantages

higher complexity potential performance problems

Shared-nothing architectures

Exclusive access to memory and disk



Characteristics

any CPU has only exclusive access to its main memory and disk

Advantages

cost extensibility availability

Disadvantages

higher complexity load balancing

Hybrid architectures

Goals:

combine the advantages of different architectures

use different processing elements in the system

The system is basically a shared-nothing architecture where each node can be a multicomputer of any architecture.



Parallel relational operators

Data partitioning

Parallelisation of relational operators

Join

Parallelisation of join

Pipelined hash-join

Data partitioning

- Partitioning = distributing the tuples of a relation over several disks
- Goal: allowing parallel databases to exploit the I/O bandwidth of multiple disks by reading and writing them in parallel

Relations are declustered (partitioned horizontally) based on a function:

round-robin

range index

hash function

Round-robin partitioning

It maps the i'th tuple to disk i mod n



Range index partitioning

It clusters tuples with similar attributes together in the same partition



Hash partitioning

It maps each tuple to a location based on a hash function. Hashing tends to randomise data rather than cluster it.



Advantages of the partitioning strategies

Round-robin

Sequential scan of all tuples in each query

Range index

Sequential scan of all tuples in each query

Associative search for data

Clustering data

Hash

Sequential scan of all tuples in each query

Associative search for data

Problems with data partitioning

data skew A query may work mainly on one partition because of the actual data placement ⇒ unbalanced load on the processors

Dynamic reorganisation of relations

With a given partitioning the criteria used for data placement can change to the extent that load balancing degrades significantly \Rightarrow the relation should be reorganised. This should be done

on-line

efficiently

transparently to the users and compiled queries

Parallel relational operators

Relational algebra allows parallel processing due to its properties

set-oriented processing

simple operations

limited number of operations

simple improvements have substantial effect (large volume of data)

relatively independent of hardware architecture

Parallel relational operators

Basic idea

use parallel data streams instead of writing new parallel operators \Rightarrow use existing sequential relational operators in parallel

Solution

each relational operator has a set of *input ports* on which input tuples arrive and an *output port* to which the operator's output stream is sent

The parallel dataflow works by partitioning and merging data streams into these sequential ports

From sequential to parallel execution



Extending relational algebra for parallel query processing

Relational operations (select, project, cp, join, union, diff, inters)

Splitting of results over multiple output streams

Operands consisting of multiple input streams

Explicit allocation of processes to processors



Splitting is consistent if

$$\mathbf{R}_i \bowtie \mathbf{S}_j = \emptyset$$
 if $i \neq j$

Pipelined hash-join



as possible

Parallel query processing

=

automatic translation of a query into an efficient execution plan

+

its parallel execution

The translation must be correct

the execution of the plan produces the expected result

The execution plan must be optimal

in that it minimises a cost function that captures resource consumption

Parallel query processing



JOQR: Join Ordering and Query Rewrite

Parallel query processing

1. translation

of the relational algebra expression to a query tree

2. optimisation

reordering of join operations in the query tree and choose among different join algorithms to minimise the cost of the execution

3. parallelisation

transforming the query tree to a physical operator tree and loading the plan to the processors

4. execution

running the concurrent transactions

Translation to query tree

Relational query \Rightarrow (rewrite) \Rightarrow query tree





Optimisation

Reordering of the join operations and choosing an evaluation algorithm for each join operation



Parallelisation of the query

1. Extract parallelism by macro-expansion of the annotated query tree to an operator tree

The operator tree identifies

atomic code pieces (operators)

timing constraints between the operators

2. Schedule the operator tree on the parallel machine

Goal: allocate machine resources so as to exploit the available parallelism while respecting timing and data placement constraints.

Parallelisation of the query

s1=sel S.thema

pj1=proj S.class pj2=proj P.id, P.name pj3=proj R.id pj4=proj P.name

b1, b2: build hash-table

p1, p2: probe hash-table

Physical operator tree

Constraints on available parallelism

Precedence constraint

timing constraints between operators e.g. hash table must be built completely before probing

Parallel constraint

piping the output of an operator to the input of another one \Rightarrow eliminating the need of buffering intermediate results

Data placement constraint

in shared-nothing systems the scan operations must be localised to specific processors that can access the relation

Parallel execution control strategies

Control flow

a single control node controls all processes related to a query it starts all processes and synchronise them scheduling is performed by the control node

Data flow

no centralised control

processes on different nodes trigger each other with data messages

data driven execution: if enough input is available the process starts automatically

Parallel execution control strategies

Control flow is the traditional approach

Advantages of data flow control

reduces the control messages transferred trough the network reduces the workload of a particular node (the control node) provides pipelining naturally

Disadvantages of data flow control

it means more asynchronous activity more competition for a processor higher protocol complexity providing data allocation information to all nodes is difficult

Parallel Database Systems

Research systems and projects

Gamma

HC16-186

XPRS project

Bubba

Prisma/DB Volcano project

Commercial systems

Oracle Parallel Server

Informix Online XPS

NCR Teradata DBS

Sybase MPP IBM DB2 Parallel Edition Tandem NonStop SQL

Gamma

Bubba

University of Wisconsin (David DeWitt)

hypercube, 32 nodes

horizontal partition

hash-based parallel algorithms for operations

data flow query execution (no pipelining)

hybrid hash-join

ported to an Intel iPSC/2 hypercube with 32 nodes MCC, Austin

never completely implemented

OLTP and DSS load

horizontal data partition over all disks

compiled PFAD, a parallel, setoriented execution model

data flow query execution (no pipelining)

operations take place where data is

PRISMA/DB

HC16-186

University of Twente (Peter M.G. Apers)

main-memory parallel database system

shared-nothing architecture

data flow query execution (with pipelining)

simple, grace and hybrid hash-join

Trondheim, Norway

hypercube intrerconnect, 16 nodes

not a fullfledge DBMS

horizontal partition of the data over *all* disks

redistribution of data

XPRS

M. Stonebraker

Shared-memory system

- Based on the Postgres next-generation DBMS
- intra- and inter-operator parallelism

high performance and availability

Volcano project

Götz Graefe

Extensible and parallel query evaluation system

supports shared-memory, sharednothing and hybrid architectures

provides a rich environment for research and education

data flow query execution

Commercial parallel database systems (brief)

Architecture

SMP, Symmetric Multiprocessing (shared memory)

Clustered SMP

MPP, Massively Parallel Processors (shared nothing)

Oracle Parallel Server

shared-disk

parallel cache management

Sybase MPP

shared-nothing (MPP) partitioned database

Commercial parallel database systems (brief)

Informix Online XPS

combination of shared-memory and shared-nothing architecture

partitioned database

Tandem NonStop SQL

primarily designed for OLTP transactions

shared-nothing architecture

NCR Teradata DBS

first commercial system

shared-nothing architecture

installed on systems with more than 100 processors

IBM DB2 Parallel Edition

shared-nothing architecture