Evolution of Database Emerging to Sybase Adaptive Server Enterprise and Ensuring Better Server Performance Tuning and Query Optimization

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ABSTRACT
Evolution of the database is emerging to a robust Sybase Adaptive Server Enterprise Database System. Key issue in the Sybase Database System is to obtain better server performance by tuning the server and optimizing the query processing. The server performance tuning and query optimization are two main issues to gain high throughput and less or fast response time. In this paper we have considered well two issues and explained how to gain better server performance by tuning the server and optimizing the query processing. We are also illustrating solutions to some situations normally occurs related to this two issues in this paper. This paper will introduce the reader to the most basic concepts of server performance tuning, query processing and optimization of query processing. We hope this paper will provide an adequate support to Database Administrator(s) (DBAs) who are working on Sybase and researchers who are working on Database Server performance tuning and query optimization. We have discussed in brief in the introduction section of this paper about the evolution of the database emerging to SYBASE ASE version under different operating system environments.

General Terms
Database Server Performance Tuning and Query Optimization

Keywords
Database, Database Management System, Relational Database Management System, Performance Tuning, Query Processing, Query Optimization.

1. INTRODUCTION
Keeping in mind the progress in communication and database technologies (concurrency, consistency and reliability) has increased the data processing potential. Various protocols and algorithms are proposed and implemented for network reliability, concurrency, atomicity, consistency, recovery, replication, query processing and query optimization. Any application or information system is mainly built from four components and that are user interface that handles information on screen, application program logic that handles the processing and information, integrity logic that governs how information is processed and kept accurate within an organization and finally access that is the method of storing and retrieving data from a physical device. We now want an optimal performance of Database Server and fast query processing in any Information System proposed and that will access to the database. We want an efficient and quick retrieval of information for any kind of decision supports. For this, server performance in general and optimization in particular are two main issues. We have considered Sybase as a robust database system in this paper. Recent versions of Sybase ASE have brought a great deal of change in the realm of optimization. Sybase ASE is known for its reliability, robustness, and availability, supports large applications with hundreds and thousands of concurrent users, and well-suited for customized transactional applications. This paper has addressed two issues (Server Performance Tuning and Query Optimization) well however this two issues are big. We can also say poor SQL statement performance and high CPU usage on our Sybase ASE server almost always indicates the need for performance tuning. This paper has an introduction section including evolution of database in brief and an introduction to Sybase SQL Server and Sybase ASE including Sybase ASE overview, performance tuning section and query processing and optimization of query processing section. We are also illustrating some solutions to situations normally occurs during server run time related to the server performance tuning and query optimization in a separate section, and finally we conclude the paper.

1.1 Evolution of Database
We first discuss few about evolution of database. We first explain about Data Hierarchy. A bit binary digits, A byte represents a character (basic building unit of data), A field (a logical grouping of characters), A record (a logical grouping of related fields), A file (a logical grouping of related records), A database (a logical grouping of related files).

In general database is a collection of organized and related records. The Database Management System (DBMS) provides a systematic and organized way of storing data, managing data and retrieving data from a collection of logically related information stored in a database. Database Management System (DBMS) did not come into industries until 1960’s. First DBMS appeared right before 1970. All database systems are evolved from a file system.
Originally, databases were flat. This means that the information was stored in one long text file, called a tab delimited file. Each entry in the tab delimited file is separated by a special character, such as a vertical bar (|). Each entry contains multiple pieces of data in the form of information about a particular object or person grouped together as a record. The text file makes it difficult to search for specific information or to create reports that include only certain fields from each record.

The File Management System also called as FMS in short is one in which all data is stored on a single large file.

Researchers in database field, however, found the data has its value, and models based only on data be introduced to improve the reliability, security, efficiency of the access and to come out from the drawbacks of the file based systems and that leads to the introduction of the Hierarchical Data Model and the Hierarchical Database Management System.

Data models provide a way in which the stored data is organized as specified structure or relation for quick access and efficient management. Data models formally define data elements and relationships among data elements for a domain of interest.

A data model instance may be one of three kinds according to ANSI/SPARC (American National Standards Institute, Standards Planning and Requirements Committee) in 1975. 1) Logical Schema/Conceptual Schema/Conceptual Level, 2) Physical Schema/Basic Level and 3) External Schema/External Level.

We discuss in brief about different generation data models and database management systems.

1.1.1. 1st Generation Data Model and Database Management System (DBMS)
Hierarchical and Network Data Models are considered under 1st Generation Data Model.

Hierarchical Data Model and Hierarchical Database Management System existed from mid 1960- early 1980. The previous system FMS drawback of accessing records and sorting records which took a long time was removed in this by the introduction of parent-child relationship between records in the database. Each parent item can have multiple children, but each chilled item can have one and only one parent. Thus relationships in the hierarchical database either one-to-one or one-to-many. Many-to-Many relationships are not allowed. The origin of the data is called the root from which several branches have data at different levels and the last level is called the leaf. Hierarchical databases are generally large databases with large amounts of data. The hierarchical data model organizes data in a tree structure or as some call it, an "inverted" tree. There is a hierarchy of parent and child data segments. Prominent hierarchical database model was IBM’s first DBMS called IMS. In mid 1960s Rockwell partner with IBM created information Management System (IMS), IMS DB/DC leads to the mainframe database market in 70’s and early 80’s.

In order to avoid all drawbacks of the Hierarchical Data Model, a next data model and database management system took its origin which is called as the Network Model and Network Database Management System. Network Data Model and Network Database Management System was introduced in 1969 almost the exact opposite of the Hierarchical Data Model. In this the main concept of many-many relationships got introduced. Network Model eliminates redundancy but at the expense of more complicated relationships. This model can be better than hierarchical model for some kinds of data storage tasks, but worse for others. Neither one is consistently superior to other.

1.1.2. 2nd Generation Data Model (Relational Model) and Relational Database Management System (RDBMS) – 1970
Relational Data Model is considered under 2nd Generation Data Model.

In order to overcome all the drawbacks of the previous data models and database management systems the Relational Data Model and the Relational Database Management System got introduced in 1970, in which data get organized as tables and each record forms a row with many fields or attributes in it. Relationships between tables are also formed in this system. A tuple or row contains all the data of a single instance of the table. In the relational model, every tuple must have a unique identification or key based on the data that uniquely identifies each tuple in the relation. Often, keys are used to join data from two or more relations based on matching identification. The relational model also includes concepts such as foreign keys, which are primary keys in one relation that are kept in another relation to allow for the joining of data.

In 1970 Dr. Edgar F. Codd at IBM published the relational model. A relational database is a collection of relations or tables. By definition, a relation becomes a set of tuples having the same attributes. Operations, which can be performed on the relations are select, project and join. The join operation combines relations, the select queries are used for data retrieval and the project operation identifies attributes. Similar to other database models, even relational databases support the insert, delete and update operations.

Basically, relational databases are based on relational set theory. Normalization is a vital component of relational model of the databases. Relational operations which are supported by the relational databases work best with normalized tables. A relational database supports relational algebra, relational calculus, consequently supporting the relational operations of the set theory. Apart from mathematical set operations namely, union, intersection, difference and Cartesian product, relational databases also support select, project, relational join and division operations. These operations are unique to relational databases.

Relational databases support an important concept of dynamic views. In a relational database, a view is not a part of the physical schema, it is dynamic. Hence changing the data in a table alters the data depicted by the view. Views can subset data, join and simplify multiple relations, dynamically hide the complexity in the data and reduce the data storage requirements.

Relational databases use Structured Query Language (SQL) which is invented by IBM in 1970 and SQL is a declarative language, which is an easy and human-readable language. SQL instructions are in the form of plain instructions, which can be put to the database for implementation. Most of the database vendors support the SQL standard.

Relational databases have an excellent security. A relational database supports access permissions, which allow the database
administrator to implement need-based permissions to the access of the data in database tables. Relational databases support the concept of users and user rights, thus meeting the security needs of databases. Relations are associated with privileges like create privilege, grant privilege, select, insert and delete privileges, which authorize different users for corresponding operations on the database.

The other important advantages of relational databases include their performance, power, and support to new hardware technologies as also flexibility and a capacity to meet all types of data needs. Relational databases are scalable and provide support for the implementation of distributed systems.

Owing to their advantages and application in the operations of data storage and retrieval of the modern times, relational databases have revolutionized database management systems. SYBASE is an example of relational database. However, SYABASE ASE has also objected oriented features.

In relational Model, values are atomic (value as one that cannot be decomposed into smaller pieces by the DBMS such as Bank Account No., Employee Code etc. or Columns in a relational table are not repeating group or arrays), each row is unique, column values are of the same kind, the sequence of columns is insignificant (ordering of the columns in the relational table has no meaning). Columns can be retrieved in any order and in various sequences. The benefit of this property is that it enables many users to share the same table without concern of how the table is organized. It also permits the physical structure of the database to change without affecting the relational tables), the sequence of rows is insignificant (the main benefit is that the rows of a relational table can be retrieved in different order and sequences. Adding information to a relational table is simplified and does not affect existing queries), each column has unique name, certain fields may be designed as keys which means that searches for specific values of that field will use indexing to speed up them.

The RDBMS allows for data independence, this helps to provide a sharp and clear boundary between the logical and physical aspects of database management.

The RDBMS provides simplicity, this provides a more simple structure than those that were being before it. A simple structure that is easy to communicate to users and programmers and a wide variety of users in an enterprise can interact with a simple model.

The RDBMS has good theoretical background, this means that it provides a theoretical background for database management field.

1.1.3. 3rd Generation Data Model and Database Management System (DBMS)

ER-Model and Semantic Data Model are considered under 3rd Generation Data Model.

In 1976, six year after Dr. Codd published the relational Model, Dr. Peter Chen published a paper in the ACM Transaction on Database Systems, introducing the Entity Relationship Model (ER Model). An ER Model is intended as a description of real-world entities. The ER data model views the real world as a set of basic objects (entities) and relationships among these objects (Entities, relationships, and attributes). It is intended primarily for the DB design process by allowing the specification of an enterprise scheme. This represents the overall logical structure of the database. Although, it is constructed in such a way as to allow easy translation to the relational model. The ER diagram represents the conceptual level of database design. A relational schema is at the logical level of database design. ER model is not supported directly by any DBMS. It needs to be translated into a model that is supported by DBMSs. The entity-relationship model (or ER model) is a way of graphically representing the logical relationships of entities (or objects) in order to create a database. Any ER diagram has an equivalent relational table, and any relational table has an equivalent ER diagram.

Semantic Data Model was developed by M. Hammer and D. McLeod in 1981. The Semantic Data Model (SDM), like other data models, is a way of structuring data to represent it in a logical way. SDM differs from other data models, however, in that it focuses on providing more meaning of the data itself, rather than solely or primarily on the relationships and attributes of the data.

SDM provides a high-level understanding of the data by abstracting it further away from the physical aspects of data storage.

The semantic data model (SDM) has been designed as a natural application modeling mechanism that can capture and express the structure of an application environment. It can serve as a formal specification and documentation mechanism for a data base, can support a variety of powerful user interface facilities, and can be used as a tool in the data base design process.

Object-Oriented Data Model (OODM) and Object Relational Data Model (ORDM) are considered under Semantic Data Model.

Object-Oriented Data Model and Object-Oriented Database Management System (OODBMS) - (early 1980s and 1990s). Research in the field of databases has resulted in the emergence of new approaches such as the object-oriented data model and Object-Oriented Database Management System, which overcome the limitation of earlier models.

Object-oriented model has adopted many features that were developed for object oriented programming languages. These include objects, inheritance (the inheritance feature allows new classes to be derived from existing ones. The derived classes inherit the attributes and methods of the parent class. They may also refine the methods of the parent class and add new methods. The parent class is a generalization of the child classes and the child classes are specializations of the parent class), polymorphism (a mechanism associating one interface with multiple code implementations) and encapsulation (the association of data with the code that manipulates that data, by storing both components together in the database).

Object-Oriented features mainly are Complex objects, object identity, encapsulation, classes, inheritance, overriding, overloading, late binding, computational completeness, extensibility.

Object Database features are mainly persistence, performance, concurrency, reliability, and declarative queries.

We do not discuss in details here about OODM.
Object-Relational Data Model and Object Relational Database Management System (ORDBMS). The object-relational model is designed to provide a relational database management that allows developers to integrate databases with their data types and methods. It is essentially a relational model that allows users to integrate object-oriented features into it. The primary function of this new object-relational model is to more power, greater flexibility, better performance, and greater data integrity then those that came before it.

The ORDBMS provides an addition of new and extensive object storage capabilities to the relational models at the center of the more modern information systems of today. These services assimilate the management of conventional fielded data, more complex objects such as a time-series or more detailed geospatial data (such as imagery, maps, and vector data) and varied dualistic media (such as audio, video, images, and applets). This can be done due to the model working to summarize methods with data structures, the ORDBMS server can implement complex analytical data and data management operations to explore and change multimedia and other more complex objects.

It can be said that the object relational model is an evolutionary technology, this approach has taken on the robust transaction and performance management aspects of its predecessors.

Database developers can now work with somewhat familiar tabular structures and data definition but with more power and capabilities.

Object-relational models allow users to define data types, function, and also operators. As a direct result of this the functionality and performance of this model are optimized. The massive scalability of the object-relational model is its most notable advantage, and it can be seen at work in many of today’s vendor programs.

In ORDBMS, SQL-3 is supported. SQL3 is a superset of SQL/92. SQL3 supports all of the constructs supported by that standard, as well as adding new ones of its own such as Extended Base Types, Row Types, User-Defined Types, User-Defined Routines, Sub-Types and Super-Types, Sub-Tables and Super-Tables, Reference Types and Object Identity, Collection Types.

**1.1.4. 4th Generation DBMS**

4th Generation DBMS is also based on Object-Oriented Data Model has come into an existence. VERSANT Database is an example of 4th Generation Database.

*Each of these models modeled the data and the relationship between the data in different ways. Each of the models encountered some limitations in being able to represent the data which resulted in other models to compensate for the limitations.*

**1.2 Sybase (10/11) SQL Server and Sybase Adaptive Server Enterprise (ASE)**

Sybase first began to design its relational database management system in 1984. Sybase Adaptive Server Enterprise (i.e., the product formerly known as the Sybase SQL Server) became the leading RDBMS product for high performance transaction processing supporting large numbers of client application connections on relatively small, inexpensive hardware. To be successful in this demanding area, Sybase focused heavily to optimize and streamline all the layers of database engine for transaction processing workloads. While this focus turned out to be the asset that catapulted Sybase to a dominant position in high end OLTP applications during the 1990’s, it increasingly became a liability in some environments as application workloads have changed and hardware performance has increased [12].

Sybase SQL Server is unique among the most popular Relational Database Management System (RDBMS) products in that it was planned and designed to operate a client-server architecture. Sybase Inc., built SQL Server with the network in mind. Each client process establishes a connection to SQL Server over the network. Information is sent from client to server and back again over the network, using standard Application Programming Interfaces (APIs) [1].

The architecture of SYBASE SQL Server enables developers to create robust, production-level client-server applications. A basic understanding of the components of that architecture will aid developers to create optimally performing applications.

Nowadays the Sybase Data Server is known as Adaptive Server Enterprise or ASE (to distinguish it from other Sybase database products like Sybase IQ). Previously it was known as Sybase SQL Server. The basic unit of storage in ASE is data page. A data page is the minimum amount of data transferred to and from disk to the cache. The supported page sizes for ASE are 2K, 4K, 8K and 16K [6].

Version 15.5 of Sybase ASE provides in-memory database (IMDB) capabilities designed to deliver low response time and high throughput for mission-critical system. More specifically, IMDB tends to deliver better performance for write-intensive workloads [32].

**1.2.1 Classification of DBMS Software Architecture and Sybase Architecture**

**Single Process**

A product is designed to use a single-process architecture single thread requests through the DBMS. This can produce bottlenecks in a multi-user environment. Single process architectures are typical for single-user workstation DBMS products.

**Multi-Process**

A product is designed to use a multi-process architecture creates new processes to handle each new request. Quick exhaustion of system resources (such as memory) can occur as more processes are initiated. Many multi-user DBMS products are multi-process architecture.

**Single Process, Multi-thread Architecture (SYBASE)**

Single Process, multi-thread architecture, employing multiple threads within a single process to service multiple users. Only one process will ever run on the server regardless of how many clients are connected [1].
Sybase ASE runs as an application on top of an operating system and depends solely on the services exported by the operating system to function. It uses operating system services for process creation and manipulation; device and file processing; shared memory for interprocess communication. ASE uses its own thread package for creating, managing, and scheduling tasks inside Sybase ASE engines for running SQL queries and for system tasks like checkpoint, network handler, housekeeper etc. [29]. Figure - 1 illustrates the architecture of the SYBASE ASE. We do not go in details on this issue here. We refer to SYBASE manual.

1.2.2. Sybase ASE Overview
We discuss few about Sybase ASE overview [5]. For details please refer to Sybase Manual.

1.2.2.1 The Sybase Server
A Sybase database server consists of A) two processes, data server and backup server. Besides there are Monitor Server, XP Server and Adaptive Server Utilities. Data Server or Adaptive Server maintains overall operations of the system databases and user databases. Backup Server is an open server-based application that manages all databases backup (dump) and restore (load) operations for Adaptive Server. Monitor Server is an open server application that obtains performance statistics on Adaptive Server makes those statistics available to monitors Monitor Historical Servers and application build with Monitor Client Library. XP Server is an open server application that manages and executes extended stored procedures from within Adaptive Server. B) devices which house the databases; one database (master) contains system and configuration data. When we install Sybase, default databases are [master (6MB), tempdb (3 MB), sybsystemdb (2 MB), model (2 MB)], and sybsystemprocs (120 MB). Master database contains the operation of Adaptive Server as whole and stores information about all users, user databases, devices, objects, and system table entries as we have discussed earlier too. The master database is contained entirely on the master device and cannot be expanded onto any other device. Model database provides a template for new user databases. The model database contains required system table, which are copied into a new user database with the create database command. Tempdb is a work area for Adaptive Server, each time Adaptive Server is started, the tempdb database is created and built from the model database. Sybsystemdb stores information about transaction in progress, and which is also used during recovery. Sybsystemprocs database contains most of the Sybase-supplied System Stored Procedures. Sybase Utilities are collection of SQL statements and flow-of-control statements that perform system tasks. Optional databases are sybsecurity for the auditing. The pubs2 and pubs3 databases are sample databases provided as learning tool for Adaptive Server. Sybsyntax database contains syntax help for Transact-SQL commands, Sybase system procedures, dbccdb database stores the result of dbcc when dbcc checkstorage or dbcc check verifying are used.
interpubs database contains French and German data, inetpubs contains Japanese data. Adaptive Server utilities and open client routines. C) a configuration file which contains the server attributes. We do not discuss in details about these here. Please refer to Sybase Manual.

1.2.2.2. Memory Model
The Sybase memory model consists of A) the program area, which is where the data server executable is stored; B) the data cache, stores recently fetched pages from the database device; C) the stored procedure cache, which contains optimized SQL calls. The Sybase data server runs as a single process within the operating system as we have discussed earlier; when multiple users are connected to the database, only one process is managed by the OS. Each Sybase database connection requires 40-60k of memory. The "total memory" configuration parameter determines the amount of memory allocated to the server. This memory is taken immediately upon startup, and does not increase.

1.2.2.3. Transaction Processing
Transactions are written to the data cache, where they advance to the transaction log, and database device. When a rollback occurs, pages are discarded from the data cache. The transaction logs are used to restore data in event of a hardware failure. A checkpoint operation flushes all updated (committed) memory pages to their respective tables. Transaction logging is required for all databases; only image (blob) fields may be exempt.

During an update transaction, the data page(s) containing the row(s) are locked. This will cause contention if the transaction is not efficiently written. Record locking can be turned on in certain cases, but this requires sizing the table structure with respect to the page size.

The transaction logging subsystem is one of the most critical components of a database server. To be able to accomplish the goal of providing recoverability of databases, transactions write log records to persistent storage. Since a number of such log record writes are directly dependent on the number of executing transactions, the logging system can potentially become a bottleneck in high throughput OLTP environments. All the users working on a particular database share the log; thus, to guarantee high performance of the application, it is essential for the DBA to monitor and configure the log to provide for best throughput and response time of the application. Out-of-the-box, Adaptive Server Enterprise (ASE) already provides a high performance logging subsystem that scales to thousands of users and very large database (VLDB) environments. ASE also provides options for the DBA to customize the logging subsystem to satisfy their unique environments for best throughput and response times [14].

1.2.2.4. Backup Procedures
A "dump database" operation can be performed when the database is on-line or offline. Subsequent "dump transaction" commands need to be issued during the day, to ensure acceptable recovery windows.

1.2.2.5. Recovery Procedures
A "load database" command loads the designated database with the named dump file. Subsequent "load transaction" commands can then be issued to load multiple transaction dump files.

1.2.2.6. Security and Account Setup
The initial login shipped with Sybase is "sa" (System Administrator). This login has the role "sa_role" which is the super-user, in Sybase terms. User logins are added at the server level, and then granted access to each database, as needed. Within each database, access to tables can be granted per application requirements. A user can also be aliased as "dbo", which automatically grants them all rights within a database.

1.2.2.7. Database Creation
User databases are initialized with the "create database" command. In practical Sybase can maintain 100 different databases in one box. Tables are created within each database; users refer to tables by using databasesname.ownernname.tablename. When we first create a database, Adaptive Server creates three segments in the database (System Segment, Log Segment and Default Segment). A typical Sybase database will consist of six segments spread across various devices (non-SAN environment). Maximum database size may be 8 Tera Bytes. Maximum size of the database devices may be 32 Giga Bytes. We can create maximum number of database devices per server is 256. We can create maximum number of segments per database is 31.

1.2.2.8. Data Types
Supported data types include integer, decimal, float, money, char, varchar, datetime, image, and text data types.

1.2.2.9. Storage Concepts
Tables are stored in segments; a segment is an area within a device, with a name and a size, that is allocated for a database. The transaction log is stored in its own segment, usually on a separate device.

1.2.2.10. Transact-SQL
SQL is a relational calculus, and when we submit SQL query it is decomposed into a relational algebra. SQL includes commands not only for querying (retrieving data from) a database, but also for creating new databases and database objects, adding new data, modifying existing data, and other functions. Sybase provides Transact-SQL (T-SQL) is a robust programming language in which stored procedures can be written. The procedures are stored in a compiled format, which allows for faster execution of code. Cursors are supported for row by row processing. Temporary tables are supported, which allows customized, private work tables to be created for complex processes. Any number of result sets can be returned to calling applications via SELECT statements.

2. PERFORMANCE TUNING SYBASE
Sybase Adaptive Server Enterprise and Sybase SQL Server provide extensive performance and tuning features. Performance and tuning is an art, not a science. As just discussed, there are many environmental factors that can have an impact on performance. There are many tuning strategies to choose from, and scores of configuration parameters that can be set. In the face of this complexity, all we can do is use our training and experience to make informed judgments, about which configurations might work, try each, then measure performance and compare the results [9]. See details in [15, 16, 17, 18, 19].
Database server performance tuning means adjustment and balancing the server. A systems administrator (SA) is responsible, at the server level, for maintaining a secure and stable operating environment; for forecasting capacity requirements of the CPU; and for planning and executing future expansion. Further, an SA is responsible for the overall performance at the server level. Database administration is the act of migrating a logical model into a physical design and performing all tasks essential to consistent, secure, and prompt database access. A Database Administrator (DBA) is responsible for access, consistency, and performance within the scope of the database. The Systems Administrator of a Sybase ASE (also referred to as SA, Database Administrator, or sometimes DBA) is responsible for all aspects of creating, maintaining, and monitoring the server and database environment [4].

Performance of any Database Server is the measure of efficiency of an application or multiple applications running in the same environment. Performance is usually measured in response time and throughput. After all, performance can be expressed in simple and clear figures, such as "1 second response time", or "100 transactions per second". Response time is the time that a single task takes to complete. Throughput refers to the volume of work completed in a fixed time period. Tuning is optimizing performance. Sybase Adaptive Server and its environment and applications can be broken into components, or tuning layers. In many cases, two or more layers must be tuned so that they work optimally together.

SQL Server performance issues includes configuration options, database design, locking, query structure, stored procedure usage, hardware and network issues, and remote processing considerations. Anyone seeking to improve system performance is eager to find the one magic parameter they can set to suddenly improve performance by an order of magnitude. Unfortunately, there are no magic bullets for SQL server. Rather, the practitioner must follow some basic configuration guidelines and tailor the environment to the application’s needs and then design the application carefully. SQL server can only be expected to deliver proper performance if we design the application with performance in mind from the outset. Over 80% of the things that an SQL server user can do to improve performance relate to application and database design [3]. According to Sybase, 80% of Sybase database performance problems can be solved by properly created index and carefully design of SQL queries.

Performance is determined by all these factors:

1) The client application itself; How efficiently is it written? We look at application tuning 2) The client-side library; What facilities does it make available to the application? How easy are they to use? 3) The network; How efficiently is it used by the client/server connection? 4) The DBMS; How effectively can it use the hardware? What facilities does it supply to help build efficient fast applications? 5) The size of the database; How long does it take to dump the database? How long to recreate it after a media failure? [8].

We broadly classify the tuning layers in Sybase Adaptive Server as follows [2]:


2.1.1. Application Layer
Most performance gains come from query tuning, based on good database design. Most of this guide is devoted to an Adaptive Server internals and query processing techniques and tools. Most of our efforts in maintaining high Adaptive Server will involve tuning the queries on our Server. Decision support (DSS) and online transaction processing (OLTP) require different performance strategies. Transaction design can reduce concurrency, since long-running transaction hold locks, and reduce the access of other users to data. Referential integrity requires joins for data modification. Indexing to support selects increases time to modify data. Auditing for security purposes can limit performance. Issues at the Network Layers are using remote or replicated processing to move decision support off the OLTP machine, using stored procedures to reduce compilation time and network usage and using minimum locking level that meets our application needs.

2.1.2. Database Layer
Applications share resources at the database layer, including disk, the transaction log, and data and procedure cache. Issues at the database layer are developing a backup and recovery scheme, distributing data across devices, auditing affect performance; audit only what we need, schedule maintenance activities that can slow performance and lock users out of table. Options to address these issues include using transaction log thresholds to automate log dumps and avoid running out of space, using thresholds for space monitoring in data segments, using partition to speed loading data, placing objects on devices to avoid disk contention or to take advantage of I/O parallelism and caching for high availability of critical tables and indexes.

2.1.3. Server Layer
At the server layer there are many shared resources, including the data and procedure caches, locks, and CPUs. Issues at the Adaptive Server layer are the application types to be supported: OLTP, DSS, or a mix, the number of users to be supported can affect tuning decisions — as the number of users increases, contention for resource can shift, network loads, replication server or other distributed processing can be option when the number of users and transaction rate reach high level. Options to address these issues include tuning memory and other parameters, deciding on client vs. server processing — can some processing take place at the client side?, configuring cache sizes and I/O sizes, adding CPUs to match workload, Configuring the housekeeper task to improve CPU utilization (following multiprocessor application design guideline to reduce contention) and configuring multiple data caches.

2.1.4. Device Layer
The disk and controllers that stores data.

2.1.5. Network Layer
The network or networks that connect users to Adaptive Server.

2.1.6. Hardware Layer
The CPU(s) available

2.1.7. Operating System Layer
Adaptive server a major application shares CPU, memory, and other resources with the operating system, and other Sybase software such as Backup Server and Monitor Server. At the operating system layer, the major issues are the file system
available to Adaptive Server, Memory Management – accurately estimating operating system overhead and other program use, CPU availability, network interface, choosing between files and raw partition, increasing the memory size, moving client operations and batch processing to other machines and multiple CPU utilization for Adaptive Server. Following two system procedures are very useful for server monitoring and performance tuning:

sp_sysmon  a system procedure that monitors Adaptive Server performance and provides statistical output describing the behavior of our Adaptive Server system.

sp_monitor  a system procedure displays statistics about Adaptive Server.

2.2. Sybase ASE Configuration Issues
Sybase ASE has many configuration parameters that we can set as per our requirements and to obtain better server performance. That can be set by using system procedure sp_configure. We do not discuss in details about this here.

2.3. How Memory Affects Performance
Having ample memory reduces disk I/O, which improves performance, since memory access is much faster than disk access. When a user issues a query, the data and index pages must be in memory, or read into memory, in order to examine the values on them. If the pages already reside in memory, Adaptive Server does not need to perform disk I/O [10].

Adding more memory is cheap and easy, but developing around memory problems is expensive. Give Adaptive Server as much memory as possible. Memory conditions that can cause poor performance are:

Total data cache size is too small. Procedure cache size is too small. Only the default cache is configured on an SMP system with several active CPUs, leading to contention for the data cache. User-configured data cache sizes are not appropriate for specific user applications. Configured I/O sizes are not appropriate for specific queries. Audit queue size is not appropriate if auditing feature is installed [27].

2.4. How Much Memory to Configure
Memory is the most important consideration when we are configuring Adaptive Server. Memory is consumed by various configuration parameters, procedure cache and data caches. Setting the values of the various configuration parameters and the caches correctly is critical to good system performance.

Total memory allocated during boot-time is the sum of memory required for all the configuration needs of Adaptive Server. This value can be obtained from the read-only configuration parameter 'total logical memory'. This value is calculated by Adaptive Server. The configuration parameter 'max memory' must be greater than or equal to 'total logical memory'. 'max memory' indicates the amount of memory we will allow for Adaptive Server needs.

During boot-time, by default, Adaptive Server allocates memory based on the value of 'total logical memory'. However, if the configuration parameter 'allocate max shared memory' has been set, then the memory allocated will be based on the value of 'max memory'. The configuration parameter 'allocate max shared memory' will enable a system administrator to allocate, the maximum memory that is allowed to be used by Adaptive Server, during boot-time.

The key points for memory configuration are:

1) The system administrator should determine the size of shared memory available to Adaptive Server and set 'max memory' to this value. 2) The configuration parameter 'allocate max shared memory' can be turned on during boot-time and run-time to allocate all the shared memory up to 'max memory' with the least number of shared memory segments. Large number of shared memory segments has the disadvantage of some performance degradation on certain platforms. Please check our operating system documentation to determine the optimal number of shared memory segments. Note that once a shared memory segment is allocated, it cannot be released until the next server reboot. 3) Configure the different configuration parameters, if the defaults are not sufficient. 4) Now the difference between 'max memory' and 'total logical memory' is additional memory available for procedure, data caches or for other configuration parameters.

The amount of memory to be allocated by Adaptive Server during boot-time, is determined by either 'total logical memory' or 'max memory'. If this value too high:

1) Adaptive Server may not start, if the physical resources on our machine does is not sufficient. 2) If it does start, the operating system page fault rates may rise significantly and the operating system may need to be configured to compensate.

The System Administration Guide provides a thorough discussion of:

1) How to configure the total amount of memory used by Adaptive Server, 2) Configurable parameters that use memory, which affects the amount of memory left for processing queries, 3) Handling wider character literals requires Adaptive Server to allocate memory for string user data. Also, rather than statically allocating buffers of the maximum possible size, Adaptive Server allocates memory dynamically. That is, it allocates memory for local buffers as it needs it, always allocating the maximum size for these buffers, even if large buffers are unnecessary. These memory management requests may cause Adaptive Server to have a marginal loss in performance when handling wide-character data, 4) If we require Adaptive Server to handle more than 1000 columns from a single table, or process over 10000 arguments to stored procedures, the server must set up and allocate memory for various internal data structures for these objects. An increase in the number of small tasks that are performed repeatedly may cause performance degradation for queries that deal with larger numbers of such items. This performance hit increases as the number of columns and stored procedure arguments increases, 5) Memory that is allocated dynamically (as opposed to rebooting Adaptive Server to allocate the memory) slightly degrades the server’s performance 6) When Adaptive Server uses larger logical page sizes, all disk I/Os are done in terms of the larger logical page sizes. For example, if Adaptive Server uses an 8K logical page size, it retrieves data from the disk in 8K blocks. This should result in an increased I/O throughput, although the amount of throughput is eventually limited by the controller’s I/O bandwidth.
What remains after all other memory needs have been met is available for the procedure cache and the data cache. Figure-2 shows how memory is divided [27].

<table>
<thead>
<tr>
<th>OS and other programs</th>
</tr>
</thead>
<tbody>
<tr>
<td>Adaptive Server Executable</td>
</tr>
<tr>
<td>Static Overhead</td>
</tr>
<tr>
<td>Kernel and Server Structures</td>
</tr>
<tr>
<td>Procedure Cache</td>
</tr>
<tr>
<td>Data Cache Overhead</td>
</tr>
<tr>
<td>Data Cache</td>
</tr>
<tr>
<td>Total Physical Memory</td>
</tr>
</tbody>
</table>

Figure-2 How Adaptive Server uses memory

2.5. Server Performance Gains Through Query Optimization [27,28]

It is very important in this paper to discuss well about table, index and page as these are very important for better Server Performance Tuning and Query Optimization.

The Adaptive Server optimizer attempts to find the most efficient access path to our data for each table in the query, by estimating the cost of the physical I/O needed to access the data, and the number of times each page needs to be read while in the data cache.

In most database applications, there are many tables in the database, and each table has one or more indexes. Depending on whether we have created indexes, and what kind of indexes we have created, the optimizer’s access method options include:

- A table scan – reading all the table’s data pages, sometimes hundreds or thousands of pages. Index access – using the index to find only the data pages needed, sometimes as few as three or four page reads in all. Index covering – using only a non clustered index to return data, without reading the actual data rows, requiring only a fraction of the page reads required for a table scan.

Having the proper set of indexes on our tables should allow most of our queries to access the data they need with a minimum number of page reads.

2.5.1. Query Processing and Page Reads

Most of a query’s execution time is spent reading data pages from disk. Therefore, most of our performance improvement — more than 80%, according to many performance and tuning experts — comes from reducing the number of disk reads needed for each query.

2.5.2. Adaptive Server pages

The basic unit of storage for Adaptive Server is a page. Page sizes can be 2K, 4K, 8K to 16K. The server’s page size is established when we first build the source. Once the server is built the value cannot be changed. These types of pages store database objects:

- Data pages – store the data rows for a table, Index pages – store the index rows for all levels of an index and Large object (LOB) pages – store the data for text and image columns, and for Java off-row columns.

2.5.3. Extents

Adaptive Server pages are always allocated to a table, index, or LOB structure. A block of 8 pages is called an extent. The size of an extent depends on the page size the server uses. The extent size on a 2K server is 16K where on an 8K it is 64K, etc. The smallest amount of space that a table or index can occupy is 1 extent, or 8 pages. Extents are deallocated only when all the pages in an extent are empty.

The use of extents in Adaptive Server is transparent to the user except when examining reports on space usage.

2.5.4. Allocation Pages

When we create a database or add space to a database, the space is divided into allocation units of 256 data pages. The first page in each allocation unit is the allocation page. Page 0 and all pages that are multiples of 256 are allocation pages.

The allocation page tracks space in each extent on the allocation unit by recording the object ID and index ID for the object that is stored on the extent, and the number of used and free pages. The allocation page also stores the page ID for the table or index’s OAM page.
2.5.5. Object Allocation Map Pages
Each table, index, and text chain has one or more Object Allocation Map (OAM) pages stored on pages allocated to the table or index. If a table has more than one OAM page, the pages are linked in a chain. OAM pages store pointers to the allocation units that contain pages for the object.

The first page in the chain stores allocation hints, indicating which OAM page in the chain has information about allocation units with free space. This provides a fast way to allocate additional space for an object and to keep the new space close to pages already used by the object.

2.5.6. How Indexes Work
We discuss here Adaptive Server stores indexes and how it uses indexes to speed data retrieval for select, update, delete, and insert operations.

Indexes are the most important physical design element in improving database performance and mainly are 1) Indexes help prevent table scans. Instead of reading hundreds of data pages, a few index pages and data pages can satisfy many queries, 2) For some queries, data can be retrieved from a nonclustered index without ever accessing the data rows, 3) Clustered indexes can randomize data inserts, avoiding insert “hot spots” on the last page of a table and 4) Indexes can help avoid sorts, if the index order matches the order of columns in an order by clause.

In addition to their performance benefits, indexes can enforce the uniqueness of data.

Indexes are database objects that can be created for a table to speed direct access to specific data rows. Indexes store the values of the key(s) that were named when the index was created, and logical pointers to the data pages or to other index pages.

Although indexes speed data retrieval, they can slow down data modifications, since most changes to the data also require updating the indexes. Optimal indexing demands are 1) An understanding of the behavior of queries that access unindexed heap tables, tables with clustered indexes, and tables with nonclustered indexes, 2) An understanding of the mix of queries that run on our server and 3) An understanding of the Adaptive Server optimizer.

2.5.7. How Indexes Affect Performance
Carefully considered indexes, built on top of a good database design, are the foundation of a high-performance Adaptive Server installation. However, adding indexes without proper analysis can reduce the overall performance of our system. Insert, update, and delete operations can take longer when a large number of indexes need to be updated.

Analyze our application workload and create indexes as necessary to improve the performance of the most critical processes.

The Adaptive Server query optimizer uses a probabilistic costing model. It analyzes the costs of possible query plans and chooses the plan that has the lowest estimated cost. Since much of the cost of executing a query consists of disk I/O, creating the correct indexes for our applications means that the optimizer can use indexes to 1) Avoid table scans when accessing data, 2) Target specific data pages that contain specific values in a point query, 3) Establish upper and lower bounds for reading data in a range query, 4) Avoid data page access completely, when an index covers a query and 5) Use ordered data to avoid sorts or to favor merge joins over nested-loop joins.

In addition, we can create indexes to enforce the uniqueness of data and to randomize the storage location of inserts.

2.5.8. Types of Indexes
Adaptive Server provides two types of indexes; 1) Clustered indexes, where the table data is physically stored in the order of the keys on the index such as for allpages-locked tables, rows are stored in key order on pages, and pages are linked in key order and for data-only-locked tables, indexes are used to direct the storage of data on rows and pages, but strict key ordering is not maintained and 2) Nonclustered indexes, where the storage order of data in the table is not related to index keys.

We can create only one clustered index on a table because there is only one possible physical ordering of the data rows. We can create up to 249 nonclustered indexes per table.

A table that has no clustered index is called a heap. The rows in the table are in no particular order, and all new rows are added to the end of the table.

2.5.9. Index Pages
Index entries are stored as rows on index pages in a format similar to the format used for data rows on data pages. Index entries store the key values and pointers to lower levels of the index, to the data pages, or to individual data rows.

Adaptive Server uses B-tree indexing, so each node in the index structure can have multiple children.

Index entries are usually much smaller than a data row in a data page, and index pages are much more densely populated than data pages. If a data row has 200 bytes (including row overhead), there are 10 rows per page.

An index on a 15-byte field has about 100 rows per index page (the pointers require 4–9 bytes per row, depending on the type of index and the index level).

Indexes can have multiple levels 1) Root level, 2) Leaf level and 3) Intermediate level.

Root level: The root level is the highest level of the index. There is only one root page. If an allpages-locked table is very small, so that the entire index fits on a single page, there are no intermediate or leaf levels, and the root page stores pointers to the data pages. Data-only-locked tables always have a leaf level between the root page and the data pages. For larger tables, the root page stores pointers to the intermediate level index pages or to leaf-level pages.

Leaf level: The lowest level of the index is the leaf level. At the leaf level, the index contains a key value for each row in the table, and the rows are stored in sorted order by the index key. For clustered indexes on allpages-locked tables, the leaf level is the data. No other level of the index contains one index row for each data row. For nonclustered indexes and clustered indexes on data-only-locked tables, the leaf level contains the index key value for each row, a pointer to the page where the row is stored, and a pointer to the rows on the data page. The leaf level is the
level just above the data; it contains one index row for each data row. Index rows on the index page are stored in key value order.

Intermediate level: All levels between the root and leaf levels are intermediate levels. An index on a large table or an index using long keys may have many intermediate levels. A very small allpages-locked table may not have an intermediate level at all; the root pages point directly to the leaf level.

Please see manual in details about Clustered indexes and select operations, Clustered indexes and insert operations, Clustered indexes and delete operations, Nonclustered indexes and select operations, Nonclustered indexes and insert operations, Nonclustered indexes and delete operations.

2.5.10. New Optimization Techniques and Query Execution Operator Supports in ASE

The query optimizer provides speed and efficiency for online transaction processing (OLTP) and operational decision-support systems (DSS) environments. We can choose an optimization strategy that best suits our query environment. Query optimizer uses a number of algorithms and formulas. The information needed by the optimizer is supplied by the statistics.

The query optimizer is self-tuning, and requires fewer interventions than versions of Adaptive Server Enterprise earlier than 15.0. It relies infrequently on worktables for materialization between steps of operations; however, the query optimizer may use more worktables when it determines that hash and merge operations are more effective.

Some of the key features in the release 15.0 query optimizer include support for 1) New optimization techniques and query execution operator supports that enhance query performance, such as a) On-the-fly grouping and ordering operator support using in-memory sorting and hashing for queries with group by and order by clauses b) hash and merge join operator support for efficient join operations and c) index union and index intersection strategies for queries with predicates on different indexes, 2) Improved index selection, especially for joins with or clauses, and joins with and search arguments (SARGs) with mismatched but compatible datatypes, 3) Improved costing that employs join histograms to prevent inaccuracies that might otherwise arise due to data skews in joining columns, 4) New cost-based pruning and timeout mechanisms in join ordering and plan strategies for large, multiway joins, and for star and snowflake schema joins, 5) New optimization techniques to support data and index partitioning (building blocks for parallelism) that are especially beneficial for very large data sets, 6) Improved query optimization techniques for vertical and horizontal parallelism and 7) Improved problem diagnosis and resolution through a) Searchable XML format trace outputs, b) Detailed diagnostic output from new set commands.

List of optimization techniques and operator support provided in Adaptive Server Enterprise are hash join, hash union distinct, merge join, merge union all, merge union distinct, nested-loop-join, append union all, distinct hashing, distinct sorted, group-sorted, distinct sorting, group hashing, multi table store ind, opportunistic distinct view, index intersection. Please see manuals for these techniques and operators in details. Please see details work on join in [30].
We can also say query optimization is the process of choosing the efficient execution strategy for execution a query. In systematic query optimization, the system estimates the cost of every plan and then chooses the best one. The best cost plan is not always universal since it depends on the constraints put on data. The cost considered in systematic query optimization includes access cost to secondary storage, storage cost, computation cost for intermediate relations and communication cost.

In ASE 15 the query processing engine has been enhanced to include very efficient techniques. ASE 15 incorporates into the optimizer component of the query processing engine proven and well tested ‘state-of-the-art’ technology. ASE 15 performs very well out of the box by automatically analyzing and selecting high performance query plan using much wider array of options and methods than were available in earlier versions [20].

Optimization Goals are provided in ASE 15 in order to allow query optimization behavior to fit the needs of our application. The optimization goals are groupings of pre-set optimization criteria that in combination affect the overall behavior of the optimizer component of the query processing engine. Each of the goals is designed to direct the optimizer to use features and functionality that will allow it to find the most efficient query plan.

There are three optimization goals that can be set. The first is designed to allow the query processing engine to use all the techniques available, including the new features and functionality to find and execute the most efficient query plans. This goal optimizes the complete result set and balances the needs of both OLTP and DSS style queries; it is on by default.

The second optimizer goal will allow the query processing engine to use those techniques most suitable to finding the most efficient query plan for purely OLTP query.

The third optimization goal is designed to generate query plans that will return the first few rows of the result set as quickly as possible. This goal is very efficient in cursor-based and web-based applications.

These optimization goals are designed to be set at the server-wide level. However, they can also set at the session or query level for testing purposes. Once set there should be no further need to tune the query processing engine’s behavior for the environment we have chosen.

To start with query processing and optimization, first we execute:

```
set showplan on
go
set statistics io on
```

then we run our query. We check if the query plan chooses the indexes we expected. We check the logical I/Os for each table and look for unexpected (i.e. high) values.

Steps to tune our query involved our database is normalized, our SQL code is reasonably well designed, the ASE resources have been allocated appropriately for (a) the system and (b) our database, the indices on the table(s) in the query are correct and the table(s) in each query are implemented correctly (physical attributes).

SET FORCEPLAN ON instructs the Optimizer to use the order we have listed in the FROM clause.

SET STATISTICS IO ON, SET STATISTICS TIME ON is important when tuning.
3.2. Abstract Query Plans (APs)

ASE introduces a new and powerful feature, Abstract Query Plans (APs) [11]. This is a very large feature which cannot be fully covered here. Please also see [12,13,23]. APs are persistent human-readable and editable copy of the query plan created by the optimizer. APs can be captured when a query is run. Once captured, they can then be associated to their originating query and used to execute that query whenever it is run again. This features allow us to run a query, capture and save the query plan created by the query optimizer and then reuse it whenever the same query is run again. We can edit an Abstract Plan, thus telling the optimizer exactly what to do, or simply use it as is. This is a very powerful and flexible feature.

We can say abstract plan is a description of a query plan which is kept in persistent storage, can be read and edited, and is associated with a specific query. When executing that query, the optimizer’s plan will be based on the information in the abstract plan. We cannot generate abstract plans on pre ASE 12.0 server. We turn on the AP capture mode and run the slow-performing query(s) to capture APs. Then we examine the AP to identify any possible problem. We edit the plan, creating a partial or full AP to resolve the issue. We make sure the AP is used whenever we run the query.

Before we go into APs in detail, let’s back up a bit and define query plans and their role in executing a query. The optimizer’s job is to determine the most efficient way (method) to access the data and pass this information on for execution of the query. The query plan is the information on how to execute the query, which is produced by the optimizer. Basically, the query plan contains a series of specific instructions on how to most efficiently retrieve the required data. Prior to ASE 12.0, the only option has been to view it via showplan or dbcc traceon 310 output (as FINAL PLAN). In most cases, changes to the optimizer’s cost model result in more efficient query plans. In some cases, of course, there will be no change; and in some there may be performance regression. APs can be captured before applying the upgrade, and then, once the upgrade has been completed, queries can be rerun to see if there is any negative change in performance. If a performance regression is identified after the upgrade, the AP from previous version can then be used in the newer version to run the query and we can contact Sybase regarding the problem. In the case of possible optimization bug, we may write an AP to workaround the problem while Sybase investigates it.

Once execution is complete, the query plan is gone. In ASE 12.0, however, APs make each query plan fully available to us. They can be captured, associated to their original query, and reused over and over, bypassing the optimizer fully or partially. They can even be edited and included in a query using the new T-SQL PLAN statement as we have stated earlier. Abstract Query Plan was first introduced in ASE 12.0 version.

When ASE creates an AP, it contains all the access methods specified by the optimizer such as how to read the table (table or index scan), which index to use if an index scan is specified, which join type to use if a join is performed, what degree of parallelism, what I/O size, and whether the LRU or MRU strategy is to be utilized.

Let’s take a quick look at a couple of simple APs.

select * from t1 where c=0
The simple search argument query generates the AP below:

(i_scan c_index t1) 
(prop t1 (parallel 1)(prefetch 16)(lru))

This AP says access table t1 using index c_index. It also says that table t1 will be accessed using no parallelism, 16K I/O prefetch, and the LRU strategy.

select * from t1, t2 
where t1.c = t2.c and t1.c = 0

The simple join above results in the AP below:

(nl_g_join t1.c = t2.c and t1.c = 0 (i_scan i1 t1) (i_scan i2 t2) 
(prop t1 (parallel 1)(prefetch 2)(lru) 
(prop t2 (parallel 1)(prefetch 16)(lru))

This AP says to perform a nested loop join, using table t1 as the outer table, and to access it using an index i1. Use table t2 as the inner table; access it using index i2. For both tables, use no parallelism and use the LRU strategy. For table t1, use 2K prefetch and for table t2, use 16K prefetch. As we can see, with a little practice APs are easy to read and understand.

APs are captured when we turn on the capture mode.

set plan dump on

ASE configuration value:

sp_configure "abstract plan dump", 1

The optimizer will optimize the query as usual, but it will also save a copy of its chosen query plan in the form of an AP. Keep in mind that if we are capturing an AP for a compiled object, such as stored procedure, we will need to ensure that it is recompiled. We can do this by executing it with recompile or by dropping and recreating it and then executing again.

As APs are captured, they are written to the new system table sysqueryplans and placed in a capture group.

When an AP is captured, it is stored along with a unique AP ID number, a hash key value, the query text (trimmed to remove white space), the user’s ID, and the ID of the current AP group. The hash key is a computed number used later to aid when associating the query to an AP, created using the trimmed query text. In general, there are at least one million possible hash key values for every AP, thus making conflicts unlikely.

APs can also be created manually without conflict by using the new create plan T-SQL statement, writing the AP text along with the SQL statement. When we save an AP using the create plan command, the query will not be executed. It’s advisable to run the query as soon as possible using the AP to ensure that it performs the way we expect it to. Let’s take a look:

create plan

select c1, c2 from tableA
where c1 = 10
and c2 > 100
"(i_scan tableA_index tableA)"

The AP in this example will access tableA using index tableA_index.

There are many more related to APs, please see the Sybase manual.

3.3. Optimization of Query Processing

Query optimization is the process of analyzing a query to determine what resources it requires and how the execute the query with the least possible query cost. To understand the optimization a query, we need to understand how the query accesses database objects, the sizes of the objects, and the indexes on the tables in order to determine whether it is possible to improve the query’s performance [2]. Please also see [12].

Symptoms of optimization problems are a query runs more slowly than we expect, based on indexes and table size, a query runs more slowly than similar query, a query suddenly starts running more slowly than usual, a query processed within a stored procedure takes longer than when it is processed as an adhoc statement and the query plan shows the use of a table scan when we expect it to use an index.

Source of optimization problems are statistics have not been updated recently so the actual data distribution does not match the values used by Adaptive Server to optimize queries, the rows to be referenced by a given transaction do not fit the pattern reflected by the index statistics, an index is being used to access a large portion of the table, where clauses are written in unoptimized form, no appropriate index exists for a critical query and a stored procedure was compiled before significant changes to the underlying tables were performed.

ASE 15 introduces completely new Optimizer and Query Processing engines, incorporating many fundamental changes. With these improvements comes the requirement that we give the optimizer as much accurate information about our data and databases as we possibly can. All that the optimizer needs to know is found in the ‘statistics’. ASE 15’s optimizer incorporates state of the art technology. A great deal of highly skilled and dedicated work went into the huge project of redesigning and rewriting it. It will produce amazing performance levels as long as it gets the most accurate information about our data as possible. The statistics contain everything the optimizer needs to do its job quickly and efficiently [7].

Before we continue let’s quickly review exactly what statistics are available to the optimizer. There are two types of statistics – column level and object level. The column level statistics describe the distribution of values in the column; they consist of the column’s histogram and density values and are updated when an index is created or an ‘update statistics’ command is run. The object level statistics describe a table and its indexes and include values such as number of rows and pages in the table and/or index (es), the number of empty pages, and the cluster ratios among others. Some of the object level statistics are updated automatically by ASE, others when ‘update statistics’ is run.

3.3.1. Factors analyzed in optimizing queries

Query plans consist of retrieval tactics and an ordered set of execution steps, which retrieve the data needed by the query. In developing query plans, the query optimizer examines:

1) The size of each table in the query, both in rows and data pages, and the number of OAM and allocation pages to be read,
2) The indexes that exist on the tables and columns used in the
query, the type of index, and the height, number of leaf pages, and cluster ratios for each index. 3) The index coverage of the query; that is, whether the query can be satisfied by retrieving data from the index leaf pages without accessing the data pages. Adaptive Server can use indexes that cover queries, even if no where clauses are included in the query. 4) The density and distribution of keys in the indexes, 5) The size of the available data cache or caches, the size of I/O supported by the caches, and the cache strategy to be used, 6) The cost of physical and logical reads; that is, reads of physical I/O pages from the disk, and of logical I/O reads from main memory, 7) join clauses, with the best join order and join type, considering the costs and number of scans required for each join and the usefulness of indexes in limiting the I/O, 8) Whether building a worktable (an internal, temporary table) with an index on the join columns is faster than repeated table scans if there are no useful indexes for the inner table in a join, 9) Whether the query contains a max or min aggregate that can use an index to find the value without scanning the table and, 10) Whether data or index pages must be used repeatedly, to satisfy a query such as a join, or whether a fetch-and-discard strategy should be employed to avoid flushing of the buffer cache of useful pages of other tables, since the pages of this table need to be scanned only once.

For each plan, the query optimizer determines the total cost by computing the costs of logical and physical I/Os, and CPU processing. If there are proxy tables, additional network related costs are evaluated as well. The query optimizer then selects the cheapest plan.

Statements in a stored procedure or trigger are optimized when the respective statements are first executed, and the query plan is stored in the procedure cache. If a respective statement is not executed, then it will not be optimized until a later execution of the stored procedure in which the statement is executed. If other users execute the same procedure while an unused instance of a stored procedure resides in the cache, then that instance is used, and previous executed statements in that stored procedure are not recompiled [35].

3.3.2. Why Creating and Maintaining Accurate Statistics is Important to ASE 15

The optimizer has always been dependant on the statistics because it is ‘cost based’. That is, it makes its decisions about which query plan to use based on the estimated resource costs of executing it. Without them it’s flying blind and can only guess at which is the most efficient query plan.

In fact, the statistics have become even more critical to good performance. Why is this? Because many of the optimizer’s new features and functionality can be very I/O intensive, especially if an inefficient query plan is used. Some of the new features and functionality include new methods for handling groupings, unions and all query level sorting operations.

New and improved join processing that is sensitive to the accuracy of the statistics has also been added. Hash joins are new to ASE and Sort-Merge joins have been improved and are turned on by default in ASE 15 [28]. If there are no useful indexes or if the statistics tell the optimizer that an index would not be efficient to use, then a worktable has to be created for the join. Joining values are moved into the worktable where they are sorted into the order required by the join and then merged with rows from the other joining table. All this, especially the sorting of the values in the worktable requires a great deal of I/O. Since both of these join methods can include sorting steps it is imperative that efficient query plans are chosen by the optimizer. The bottom-line is that large, unnecessary sorts are the bane of good query performance. It can’t be over emphasized how important it is to keep accurate statistics available for the optimizer to take advantage of when estimating the costs of joins. Even though ASE 15 is designed to avoid worktables that were often used in earlier versions, inaccurate statistics can lead to query plans that revert to using them.

One new piece of functionality added to ASE 15 in order to deal directly with a long-standing join performance issue is Join Histograms.

3.3.3. Join Histograms

The Problem - Prior to ASE 15 the optimizer could only use a column’s density value, if it was available, to estimate the cost of joining one column to another. The density value is a ‘weighted average’ of the distribution of values in the column. If there was no density value, the optimizer used a preset selectivity value, a sort of ‘magic number’ that was based on the join operator. Since the most common join is an equi-join, the ‘magic number’ was 0.10 (the optimizer believed that 10% of the rows in the column qualified for the join). As we might imagine most joins don’t qualify exactly 10% of either column.

When the column contained a fairly even distribution of values, the density value was accurate. However, when the column contained any degree of data skew (many values occupying a small number of rows each and a few values occupying many rows each) the density value was not accurate. When the optimizer used a density value of a skewed column it would lead the optimizer to believe that a join of the skewed table would be more efficient than it actually was. This in turn resulted in some very poorly performing joins.

The ASE 15 Solution - The join histograms of ASE 15 always give the optimizer an accurate view of the table. For example if the where clause of a join statement disqualifies a highly duplicated value it is not included in the join histogram; why estimate the cost of retrieving the skewed values if they aren’t needed for the join? How do join histograms work? Very simply – If there are statistics on a column that is being joined and there is a search argument (SARG) in the query then a histogram will be built on the fly containing only the values that the SARG has qualified.

A quick example: customerID table contains 100 distinct customer ids, orderID table contains all the orders placed by the 100 customers, 100K rows. Let’s say that of the 100 customers 3 have placed 50% of all the orders and of the 3 all have placed close to the same number of orders; the orderID table will contain data skew while the customerID table will be evenly distributed.

In pre-ASE 15, the density value would be larger due to the highly duplicated values for the three big customers. There is a good chance that the optimizer would estimate that an index on the custID column in the orders table would be expensive to use and call for a table scan. In fact, the index would be very selective for the query. In ASE 15, the join histogram would not include the highly duplicated values thus accurately making the index access look cheaper than a table scan.
Of course, for join histograms to be accurate and useful, accurate statistics need to be in place for the joining columns.

Here’s another situation that missing or inaccurate statistics can cause – The optimizer having to examine more query plans than is necessary [13].

3.3.4. Timeout

The Problem – In ASE 15 the new functionality in the optimizer can lead to the optimizer having MANY more query plans to examine than in earlier versions. Add to this the large and complex queries common today and the optimizer is working much harder than ever before. The more query plans to estimate costs for, the more time it takes to find the best plan to execute. In fact, there is a possibility that the optimizer can take more time optimizing the query than it takes to actually execute it and return the results.

The ASE 15 Solution - It is for this reason that ‘timeout’ was included in the ASE 15 optimizer. Put simply, once the optimizer has reached the timeout limit it will choose the cheapest (most efficient) query plan it has examined up to that point. This plan will be used to execute the query whether it’s the best or not. Accurate statistics can result in the optimizer having to examine far fewer query plans and finding the most efficient plan before the timeout limit is reached. This in turn will result in less procedure cache usage for the optimizer’s search. When there are no statistics or the statistics are inaccurate the optimizer can overestimate the cost of what is actually the best query plan and go on to examine many more [13].

3.3.5. How to Create and Maintain Accurate Statistics

Statistics are kept up to date by regularly running the update statistics command, or by dropping and recreating indexes, which can be very time and resource consuming. In the past, there were a number of ‘rules of thumb’ floating around among DBAs about when to update statistics. Some of the more popular were 1) when 5-10% of the data has changed, 2) once a week, 3) every day, 4) only when queries start performing badly, 4) only when we have time to do it and 5) We never run update statistics, etc.

The honest answer to the question was, is, and will always be - It all depends on our data and our queries [13].

3.4.6. Locking and Concurrency

The Optimizer decides on Lock Type and Granularity. Decisions on lock type (share, exclusive, or update) and granularity (page or table) are made during optimization so make sure our updates and deletes don't scan the table. Exclusive Locks are only released upon Commit or Rollback. Lock Contention can have a large impact on both throughput and response time if not considered both in the application and database design. Keep transactions as small and short as possible to minimize blocking. Consider alternatives to "mass" updates and deletes such as a v10.0 cursor in a stored procedure which frequently commits. Never include any "user interaction" in the middle of transactions. Shared Locks generally released after page is read. Share locks "roll" through result set for concurrency. Only "HOLDLOCK" or "Isolation Level 3" retains share locks until commit or rollback. Remember also that HOLDLOCK is for read-consistency. It doesn't block other readers. Use optimistic locking techniques such as timestamps and the tsequal() function to check for updates to a row since it was read (rather than holdlock)[8].

3.4.7. Tuning Transact-SQL Queries

Poor Performance SQL problem can occur at any stage of the development lifecycle, but catching them early is always much less expensive that letting them reach a production environment.

Often times, the main bottleneck is a SQL statement taking up too many resources then we examine the following 1) Are the table and index statistics up to date? 2) Are there any missing indexes? 3) Do any of the columns need extended histogram or frequency statistics? 4) Are all the unique and not-null constraints correctly defined on the columns and tables? 5) Is the database choosing the right access path? [21].

One of the largest factors determining performance is TSQL. Test not only for efficient plans but also semantic correctness.

3.4.8. Normalized -vs- Denormalized

Always start with a completely normalized database. Denormalization should be an optimization taken as a result of a performance problem. Benefits of a normalized database include; reduce data redundancy, accelerates searching, sorting, and index creation since tables are narrower, allows more clustered indexes and hence more flexibility in tuning queries, since there are more tables, accelerates index searching since indexes tend to be narrower and perhaps shorter, allows better use of segments to control physical placement of tables, fewer indexes per table, helping UPDATE, INSERT, and DELETE performance, fewer NULLs and less redundant data, increasing compactness of the database, accelerates trigger execution by minimizing the extra integrity work of maintaining redundant data, joins are generally very fast provided proper indexes are available, cost of a logical I/O (get page from cache) only 1-2 milliseconds. We can say benefits of normalization mainly are 1) More rows per page (less logical I/O), 2) More rows per I/O (more efficient), and 3) More rows fit in cache (less physical I/O). There are some good reasons to denormalize; all queries require access to the "full" set of joined data, majority of applications scan entire tables doing joins, computational complexity of derived columns require storage for SELECTs [8]. We have also discussed well in section 4.13. For details please see [15].

3.4.9. Index Selection

Without a clustered index, all INSERTs and "out-of-place" UPDATEs go to the last page. The lock contention in high transaction environments would be prohibitive. This is also true for INSERTs to a clustered index on a monotonically increasing key.

High INSERT environments should always cluster on a key which provides the most "randomness" (to minimize lock / device contention) that is usable in many queries. Note this is generally not our primary key!

Prime candidates for clustered index (in addition to the above) include: Columns Accessed by a Range, Columns Used with Order By, Group By, or Joins

Indexes Help SELECTs and Hurt INSERTs

Too many indexes can significantly hurt performance of INSERTs and "out-of-place" UPDATEs.
Prime candidates for non-clustered indexes include; Columns Used in Queries requiring Index Coverage, Columns Used to Access Less than 20% (rule of thumb) of the Data, Unique indexes should be defined as UNIQUE to help the optimizer, minimize index page splits with fill factor (helps concurrency and minimizes deadlocks), Keep the Size of the Key as Small as Possible, Accelerates index scans and tree traversals

Use small datatypes whenever possible. Numerics should also be used whenever possible as they compare faster than strings. For details please see [27].

4. ILLUSTRATING SOLUTIONS TO SOME SITUATIONS NORMALLY OCCURS DURING SERVER RUN TIME RELATED TO THE SERVER PERFORMANCE TUNING AND QUERY OPTIMIZATION

This section is illustrating solutions to some situations normally occurs during server run time to obtain better server performance and query optimization. Following two system procedures that we always use during server monitoring and performance tuning:

sp_sysmon a system procedure that monitors Adaptive Server performance and provides statistical output describing the behavior of our Adaptive Server system. Mainly we use sp_sysmon to look buffer hit ratio and spinlock contention. Buffer hit ratio reports the % times a requested datapage is found in the data caches. A buffer hit ratio of less than 97% could indicate memory starvation. If sufficient amount of memory is not allocated to the data cache the first thing to do would be to increase memory. Spinlock Contention reports the number of times an engine encountered spinlock contention on the cache, and had to wait to acquire memory. Splitting the data caches into cachelets will reduce spinlock contention in case of SMP environments.

sp_monitor a system procedure displays statistics about Adaptive Server. It is a well known stored procedure available for the longest time on ASE to capture CPU busy and IO busy statistics since the last run of the same procedure. This for me provides a first insight at what is going on in the data server as far as IO and CPU utilization is concerned. Besides this stored procedure, there are some other features that can help us for performance troubleshooting.

We have discussed about sp_sysmon and sp_monitor in section 2.1.7 also. For details please refer to Sybase manual.

4.1. We normally use following tools to evaluate and tune query

We use the "set showplan on" command to see the plan chosen as "most efficient" by optimizer. We run all queries through during development and testing to ensure accurate access model and known performance. Information comes through the Error Handler of a DB-Library application. We can also say set showplan on is used for look for table scans in the results. We can also say showplan enables us to find out which plan the optimizer has chosen, what indexes it intends to use, and which tables will be accessed most heavily. Given this knowledge, we could alter our queries, create indexes, or alter our table design, caching etc to improve application performance. We use set noexec on, so that the showplan is returned, but the server does not actually return the data. We show here an example.

```
select * into #dennis from sysobjects
```

```
go
set showplan on
set noexec on
```

```
go
select name from #dennis where type = 'U' order by name
```

We use the "dbcc traceon(3604, 302, 310)" command to see each alternative plan evaluated by the optimizer. Generally, this is only necessary to understand why the optimizer won't give us the plan we want. If we think the problem is with an index, here is how to see they are chosen using dbcc traceon(302) and dbcc traceon(310).

We use the "set statistics io on" command to see the number of logical and physical i/o's for a query. Scrutinize those queries with high logical i/o's and it usually takes a lot of time.

We use the "set statistics time on" command to see the amount of time (elapsed, execution, parse and compile) a query takes to run or we can say it gathers time statistics so we can see where the time is spent.

We use "set forceplan on" command to change join order to be the order of the tables in the FROM clause.

If the optimizer refuses to select the proper index for a table, we force it by adding the index id in parentheses after the table name in the FROM clause.

```
SELECT * FROM orders(2), order_detail(1) WHERE ...
```

We use set statistics subquerycache on to display the number of cache hits and misses and the number of rows in the cache for each subquery.

We also use optdiag utility command to display statistics for tables, indexes, and columns. This utility allows us to read, write and simulate the statistics used by the optimizer to determine the most efficient method to retrieve the required data. Please see details in [11,13].

We also use set table count that increases the number of tables that the optimizer considers at one time while determining join order.

Adaptive Server's cost-based optimizer uses statistics about the tables, indexes, and columns named in a query to estimate query costs. It chooses the access method that the optimizer determines has the least cost. But this cost estimate cannot be accurate if statistics are not accurate. For details please also see [12]. We have also discussed well in section 3.3.1 and 3.3.2.

Some statistics, such as the number of pages or rows in a table, are updated during query processing. Other statistics, such as the histograms on columns, are only updated when we run the update statistics command or when indexes are created.

We can use the optdiag command to see the time update statistics was last run.

The update statistics commands update the column-related statistics such as histograms and densities. So statistics need to
be updated on those columns where the distribution of keys in
the index changes in ways that affect the use of indexes for our
queries.

Running the update statistics commands requires system
resources. Like other maintenance tasks, it should be scheduled
at times when load on the server is light. In particular, update
statistics requires table scans or leaf-level scans of indexes, may
increase I/O contention, may use the CPU to perform sorts, and
uses the data and procedure caches. Use of these resources can
adversely affect queries running on the server if we run update
statistics at times when usage is high. In addition, some update
statistics commands require shared locks, which can block
updates.

4.2. To start with query processing and optimization we first execute

set showplan on

exec

then we run our query. We check if the query plan chooses the
indexes we expected. We check the logical I/Os for each table
and look for unexpected (i.e. high) values.

4.3. Checking how much time query taking

Suppose we have a batch of queries, which we want to optimize.
First we need to identify which query is creating the problem
and then to identify where the problem lies in that query. Query
that is taking maximum time is the first we want to consider for
the optimization. To check which query is taking maximum time
to execute we place getdate() both before and after the query.
We identify the type of the query whether it is a data look-up
query or data modification query.

We use ‘set showplan on’ to see what query plan is executed by
the SQL Server? If query is taking a lot of time to execute it
is not possible to execute it again and again. We use ‘set noexec
on’ to just build the query plan and not to execute it. When these
two options are set SQL Server will parse the query, optimize it
and develop a query plan, but will not submit it to execute.

As SQL server uses cost based query optimizer, it’s always
better to turn on the following options to check the statistics of
the query:

set statistics io on gives actual logical and physical page reads
incurred by the query.

set statistics time on gives total CPU and elapsed time to execute
a query and the time to parse and compile the query.

4.4. Server was running fine when starts and if we find slow we can think of just few

If query plans are used by query optimizers to execute a
Query/Stored Procedure. Over time the query plans become
outdated and are no longer efficient then we should run update
statistics <table name> and sp_recycle <table name>.

If the query optimizer does a table scans when it cannot resolve
a badly written query or Stored Procedure. Here query
optimization is the only solution for this. We try to find out such
queries by running them during off-peak time.

4.5. Increase RAM size and number of concurrent user connections

Every user will consume resources on the Sybase Server, which
reduces the amount of memory that Sybase has for itself. For
this we should increase RAM on our server and increase the
amount of memory dedicated to Sybase server. Increase the no.
of concurrent user connections.

4.6. Rebuild Reorg command improve Performance

We always clean-up tables that have been updated frequently by
using command rebuild reorg.

4.7. Checking whether query is using an index

We look at the query plan that is generated by the optimizer. We
check whether the optimizer is picking up proper indexes for the
SARG, OR clauses and Join operations. We see whether it is
using indexes to execute the query or not, and if yes which
indexes are used. If there is an index on the table and optimizer
is not using that index to resolve the query, we try to figure out
why optimizer is not using the index [40]. We have figured out
this issue in section 4.1, 4.2 and 4.8.

4.8. Checking whether statistics is up-to-date

As a general advice, once the table gets 20% updated, we have
to run the update statistics for the table. Update statistics for a
table updates the distribution page [server keeps distribution
information for each index on a separate page in the database]
(so distribution statistics) for the indexes used in the table. We
use ‘update statistics’ command to update the statistics for all
indexes on a table, using following commands:

To update the statistics of all the indexes related to the table:

update statistics Sales.SalesOrderDetail where Sales is a
database and SalesOrderDetailes is a table.

To update the statistics of a particular index on the table:

update statistics Sales.SalesOrderDetail
AK_SalesOrderDetail_rowguid

where

AK_SalesOrderDetail_rowguid is an index under table
SalesOrderDetail

We run dbcc traceon 302 for getting the meta-information to
know how the query optimizer is handling the index selection
and OR clauses. We run dbcc traceon 310 for getting the meta-
information to know how query optimizer is picking up the join
order [40].

Again we run that query and see if it using the indexes or not. If
the query is still not using any indexes, and we think that an
index can be used to execute the query faster, we can force the
optimizer to use the index by specifying the index ID number
after the table name in the query. SQL Server would use that
index to satisfy the query. We can always use ‘set statistics io’
option to verify whether there was an I/O savings over the
optimizers choice or not.
We identify which indexes to use to resolve the query: clustered or non-clustered. Clustered index is better for range queries, if data is in sorted order, column has number of duplicate values, and column is frequently referenced in the order by clause. Non-clustered index is better for single row lookups, queries containing joins, queries with small range retrieval.

We keep following things in mind for the indexes 1) unless the OR strategy is applied, SQL server uses only one index per table to satisfy the query, 2) sometimes a table scan may be cheaper than a candidate index in terms of total I/O. e.g. large range retrievals on a non-clustered-index column and 3) a non-clustered index that covers a query will be faster than a similarly defined clustered index.

4.9. If we are using Join and Subquery
If the query contains any joins, we look what is the join order selected by the optimizer. Joins are performed as a set of nested loops. Generally Join queries consume more memory than subquery. Subquery involves the table creation so affects performance, because of I/O fetches need for processing. Subquery is always better than join if we have hardware bottleneck (memory constraints). We check if we can replace join with the subquery. But if we can scale up our hardware (nowadays it is not a big deal) then we go with join.

Order in which the join is performed has effect on the total number of I/O performed. By knowing the meaning of data we can select a particular join order and we check if it improves performance. We can force join order with the ‘set forceplan’ option:

```
set forceplan {on / off}
```

It might be possible that optimizer was using different indexes before forcing the join order, as certain indexes may not be useful for certain join orders. As the number of table increases, query optimizer takes more time to determine the join order. If the optimizer is picking the proper join order but taking a lot of time in determining the join order, we can force that particular join order and save time. One another reason query not picking up proper indexes may be “set table count value” is set to lower number but the number of tables involved in the join operations is relatively higher [40]. We have also discussed in section 4.15.

4.10. We can avoid transaction logging under some circumstances
Another factor while optimizing the query is to avoid the transaction log in data modification queries. In the following cases we can avoid the transaction to be logged; 1) we use fast BCP instead of slow BCP to avoid the data to be logged, 2) we use truncate table instead of delete and 3) we use select into instead of creating temporary table to avoid the transaction log [40].

4.11. We should prefer dropping and rebuilding index under some circumstances
A table having indexes and triggers takes long time to perform data modification queries. Depending on how important the query is, we can think of dropping the indexes during the daytime and rebuilding the indexes at end of the day. For these types of queries, in many cases index usage is discouraged, because the update statements.

If we need to perform large data modification its better to drop indexes and rebuild indexes after the data modification operation is performed.

Another factor we want to consider is whether query is using any worktable/temporary table to resolve the query. Creating the worktable to solve the query always take much time to process. Queries containing ‘order by’, ‘group by’, ‘distinct’ always uses a worktable. We check if we can satisfy the query by avoiding these clauses. Probable solutions to avoid these clauses are 1) if column has a unique/primary key, then there is no need to specify the ‘distinct’ clause, 2) if we have more than four tables to join, it will definitely use a worktable to resolve the query. We try to keep the number of tables to join to minimum. We try to create subquery to avoid join of more than four tables, 3) if there is an index on a column, which keeps the column sorted, we don’t need to specify ‘order by’ clause in the query, 4) if we know that data is unique, then there is no need to specify the ‘group by’ clause in the query.

We check for the following scenarios in the where clause of the query 1) negative Logic (!=, <>, <>); negative logic always results in the table scan unless index covering is applied, 2) calculated comparison: if right side of an expression needs to be calculated first, it will not be evaluated until it reaches the execution stage and 3) if a query is frequently using computed expressions, insert a column of the computed expression in the table. We need to write a trigger to maintain that column. e.g. suppose a query always calculates discounted price from a table, its better to insert a column of discounted price in the table [40].

4.12. We should avoid trigger and adopt stored procedure under some circumstances
A trigger gets fired for every transaction. It is another overhead in the query. Depending on how important the query is we can drop triggers and write a store procedure for that. And at end of the day we can run that stored procedure to perform the operations performed by the trigger. e.g. During the day we need to handle a lot of queries for OLTP, at that time we can consider to drop the trigger and run stored procedure at end of the day [40].

4.13. We should denormalize tables under some circumstances
Depending upon the type of queries we can think of denormalizing the tables. Denormalizing can improve performance by 1) Minimizing the need of join, 2) Reducing the No. of foreign keys on tables, 3) Reducing the No. of indexes, saving storage space, and reducing data modification time, 4) Pre computing aggregate values, that is computing them at data modification time rather than at select time and 5) Reducing the No. of tables (in some cases). Following are the options to denormalize the data depending upon the importance of the queries 1) Inserting calculated columns: it incurs another overhead to maintain the calculated column value. When a data row is inserted, we need to calculate the value of that column. Depending upon the importance of the query we can insert a calculated column, 2) Vertical merge of tables: If a query always performs the join of two or more tables. It is better to merge the tables and build a new table to satisfy that query faster. Again it depends on how much important the query is and how we are going to manage the duplicate data. Duplicating the data
improves the query to run faster but it incurs another overhead to maintain the duplicate data, 3) Vertical split of tables: If query always select certain columns from the table, its better to separate the columns from the table and perform the query to improve the performance, 4) Horizontal merge of tables: Merge the data horizontally, e.g. If a query shows the transaction done by the customer in the current month only and if the user wants to see the record for certain months, we can horizontally combine data from old transactions table and recent transactions table to give the result and 5) Horizontal split of tables: Split the data horizontally. e.g. in a online banking scenario, split the data in most recent transactions and old transactions. We can just show the most recent transactions, instead of showing the whole transactions [40].

4.14. Query plan optimization

ASE 15 Set Plan Commands allow us to set an optimization goal which fits our database environment.

In Decision Support System and Reporting Environments if we give the optimizer more time to compile a query plan through set plan command is beneficial for optimization.

set plan opttimeoutlimit 999

where current optimization timeout limit for query optimization is 999.

The range of values for opttimeoutlimit has been changed to 0 – 4000, with 0 indicating no optimization limit.

There are three new optimization goals in Sybase ASE and Optimization goal can be chosen at the Server, Session and Query level.

allrows_mix – the default goal, as well as the most useful goal in a mixed query environment. It balances the needs of OLTP and DSS query environments (merge joins; A merge join can use multiple worker processes to perform: The scan that selects rows into a worktable, note a merge join may requires a sort + allrows.oltp)

allrows.oltp – the most useful goal for purely OLTP queries (nested loop join; Nested-loop joins provide efficient access when tables are indexed on join columns).

allrows.dss – the most useful goal for operational DSS queries of medium-to-high complexity (hash joins + allrows_mix). The hash join algorithm builds an in-memory hash table of the smaller of its targets.

We use session level syntax as follows:

set plan optgoal allrows_mix

This query depicts an equijoin

set plan optgoal allrows.oltp

Another example in Sybase ASE

set plan optgoal allrows.dss

SELECT * FROM employee JOIN department ON employee.DepartmentID = department.DepartmentID;

4.15. Various Joins operation

We have discussed about joins operation in ASE 15 in section 2.5.10. Here again we discuss few about various joins operation in various flavors of Sybase and this will guide in query optimization.

4.15.1. Equijoin

Joins based on equality (=) are called equijoins. Equijoins compare the values in the columns being joined for equality and then include all the columns in the tables being joined in the results.

Example in Sybase ASE

select * from publisher pub, author au where pub.id = au.id

This query depicts an equijoin

Another example in Sybase ASE

SELECT * FROM employee JOIN department ON employee.DepartmentID = department.DepartmentID;

4.15.2. Natural Join

Natural join is basically a form of equijoin where one of the join fields is projected out. i.e. it avoids repetition of the join column.

Example in Sybase ASE

select pub.id, pub.name, au.* from publisher pub, author au where pub.id = au.id

This query depicts a natural join

Example other than Sybase ASE

SELECT * FROM employee NATURAL JOIN department;

4.15.3. Outer Join

Joins that include all rows, regardless of whether there is a matching row, are called outer joins. Adaptive Server supports both left and right outer joins. For example, the following query joins the titles and the titleauthor tables on their title_id column:

Example in Sybase ASE

select * from titles, titleauthor where titles.title_id *= titleauthor.title_id

Transact-SQL outer joins (shown above) use the *= command to indicate a left outer join and the //= command to indicate a right outer join.

The left outer join, *=, selects all rows from the first table that meet the statement's restrictions. The second table generates values if there is a match on the join condition. Otherwise, the second table generates null values.

For example, the following left outer join lists all authors and finds the publishers (if any) in their city:

Example in Sybase ASE

select au_fname, au_lname, pub_name from authors, publishers where authors.city *= publishers.city

SELECT * FROM employee, department WHERE employee.DepartmentID *= department.DepartmentID

Another example in Sybase ASE
The result of a left outer join (or simply left join) for table A and B always contains all records of the "left" table (A), even if the join-condition does not find any matching record in the "right" table (B). This means that if the ON clause matches 0 (zero) records in B, the join will still return a row in the result—but with NULL in each column from B. This means that a left outer join returns all the values from the left table, plus matched values from the right table (or NULL in case of no matching join predicate). If the right table returns one row and the left table returns more than one matching row for it, the values in the right table will be repeated for each distinct row on the left table.

The right outer join, =*, selects all rows from the second table that meet the statement's restrictions. The first table generates values if there is a match on the join condition. Otherwise, the first table generates null values.

**Example in Sybase ASE**

```sql
SELECT * FROM employee RIGHT OUTER JOIN department ON employee.DepartmentID = department.DepartmentID;
```

Another example in Sybase ASE

```sql
SELECT * FROM employee, department WHERE employee.DepartmentID = department.DepartmentID
```

A right outer join (or right join) closely resembles a left outer join, except with the treatment of the tables reversed. Every row from the "right" table (B) will appear in the joined table at least once. If no matching row from the "left" table (A) exists, NULL will appear in columns from A for those records that have no match in B. A right outer join returns all the values from the right table and matched values from the left table (NULL in case of no matching join predicate).

A table is either an inner or an outer member of an outer join. If the join operator is *=, the second table is the inner table; if the join operator is =*, the first table is the inner table.

**4.15.4. Full Outer Join**

Full outer join Conceptually, a full outer join combines the effect of applying both left and right outer joins. Where records in the FULL OUTER JOINed tables do not match, the result set will have NULL values for every column of the table that lacks a matching row. For those records that do match, a single row will be produced in the result set (containing fields populated from both tables).

For example, this allows us to see each employee who is in a department and each department that has an employee, but also see each employee who is not part of a department and each department which doesn't have an employee.

**Example other than Sybase ASE:**

```sql
SELECT * FROM employee FULL OUTER JOIN department ON employee.DepartmentID = department.DepartmentID;
```

Some database systems (like MySQL) do not support this functionality directly, but they can emulate it through the use of left and right outer joins and unions. The same example can appear as follows in SYBASE ASE and MySQL:

```sql
SELECT * FROM employee LEFT JOIN department ON employee.DepartmentID = department.DepartmentID UNION ALL
SELECT * FROM employee RIGHT JOIN department ON employee.DepartmentID = department.DepartmentID;
```

Another Example in Sybase ASE

```sql
SELECT employee.*, department.* FROM employee LEFT JOIN department ON employee.DepartmentID = department.DepartmentID;
```

This is based on cartesian product or cross join and is inefficient. Hash join and sort-merge join is better to this.

**4.15.5. Inner Join**

Inner joins, in which the joined table includes only the rows of the inner and outer tables that meet the conditions of the ON clause. We can also say that the query compares each row of A with each row of B to find all pairs of rows which satisfy the join-predicate. When the join-predicate is satisfied, column values for each matched pair of rows of A and B are combined into a result row.

**Example in Sybase ASE**

```sql
SELECT * FROM employee INNER JOIN department ON employee.DepartmentID = department.DepartmentID
```

This join is used for comparing values within a column of a table. Since this operation involves a join of a table within itself, we need to give the table two temporary names, or correlation names which then are used to qualify the column names in the rest of the query.

**Example in Sybase ASE**

```sql
update t1 set t1.c1 = t1.c1 + 1 FROM t1 a, t1 b where a.c1 = b.c2
```

and

```sql
delete t1 FROM t1 a, t1 b WHERE a.c1 = b.c2
```

For example, we can use a self-join to find out which authors in Oakland, California, live in the same postal code area. Since this query involves a join of the authors table with itself, the authors table appears in two roles. To distinguish these roles, we can temporarily and arbitrarily give the authors table two different correlation names—such as au1 and au2—in the from clause. These correlation names qualify the column names in the rest of the query. The self-join statement looks like this:

**Example in Sybase ASE**

```sql
select au1.au_fname, au1.au_lname,
au2.au_fname, au2.au_lname
from authors au1, authors au2
where au1.city = "Oakland" and au2.city = "Oakland"
```
and au1.state = "CA" and au2.state = "CA"
and au1.postalcode = au2.postalcode
another example
select * from table1, table1

14.15.7. Cross Join (Cartesian product)
Sybase 12 to 15 does not support Cross Joins.
CROSS JOIN returns the Cartesian product of rows from tables in the join. In other words, it will produce rows which combine each row from the first table with each row from the second table.
Example of an explicit cross join other than Sybase ASE:
SELECT * FROM employee CROSS JOIN department
Example of an implicit cross join in Sybase ASE:
SELECT * FROM employee, department
Another example in Sybase ASE
Select * from table1, table2
If table1 has 6 rows and table 2 has six rows then result is 6 * 6 = 36 rows.
Example of an explicit cross join other than Sybase ASE:
SELECT * FROM employee CROSS JOIN department;
Example of an implicit cross join in Sybase ASE:
SELECT * FROM employee, department;

4.15.8. Nested loop join
Nested Loop Join was the only Join in Sybase 12.5 and lower. Nested Loop Join works well in the condition where number of rows to be returned or there is a SARG condition (filter condition) in the query.
This physical operator supports the nested-loop-join algorithm.
In this join left child forming the outer data stream and the right child forming the inner data stream. For every row from the outer data stream, the inner data stream is opened. Often, the right child is a scan operator.
Here is a simple pseudo-code listing for joining the two relations r and s utilizing the nested for loop [36].
for each tuple tr in r
for each tuple ts in s
if join condition is true for (tr,ts)
add tr+ts to the result
or
We illustrate nested loop join and pseudo code again as follows:
select table1.*, from table1, table2 where table1.C1 = table2.C2
where columns C1 and C2 are non-clustered indexes.
pseudo-code is as follows:
for each row R1 in table1
for each row R2 in table2
if (C1=C2)
return (table1.*)
We can also say that one of the relations is counted as the inner relation, while the other one is counted as the outer one. Then, all the tuple from the inner relation are compared with tuple from the outer relation. Once a matching occurred, both tuples are placed in the outer buffer [37].
Please see [36,37,38,39] for details.
We can also say in the nested loop join the system will pick an outer table. System will point to the first row in the outer table. Now the system will scan the inner table, will examine each row and will check to see if it matches. Once the system scanned the inner table, will move to the next row in the outer table and will scan the inner row again.

4.15.9. Merge join is considered to be better than the nested loop join in performance
Merge join is considered better than nested loop join in term of performance. Merge Join is executed if it satisfies the following condition; 1) Only under Equijoin condition and 2) Data on the columns to be joined should be sorted. If the optimizer decides that it is effective to sort the data then we do the join then it will create a work table in the tempdb for the columns to be joined, sorts it and then joins the tables. Merge joins can perform better than hash joins if the row sources already sorted and a sort operation does not have to be done. If merge join involves choosing lower access method (an index scan as opposed to a full table scan), then the benefit of using soft merged might be lost. Merge joins are useful when the join condition between two tables is an inequality condition (but not a nonequality) like <, <=, >, >=.
This physical operator supports the merge join algorithm, which relies on ordered input. merge join is most valuable when input is ordered on the merge key, for example, from an index scan. merge join is less valuable if sort operators are required to order input.
The left and right children are the outer and inner data streams, respectively. Both data stream must be sorted on the MERGE JOIN’s key values.
First, a row from the outer stream is fetched. This initializes the MERGE JOIN’s key values. Then, rows from the inner stream are fetched until a row with key values that match or are greater than (less than if key column is descending) is encountered. If the join key matches, the qualifying row is passed on for additional processing, and a subsequent next call to the MERGE JOIN operator continues fetching from the currently active stream.
If the new values are greater than the current comparison key, these values are used as the new comparison join key while fetching rows from the outer stream. This process continues until one of the data streams is exhausted.
Generally, the MERGE JOIN strategy is effective when a scan of the data streams requires that most of the rows must be processed, and that, if any of the input streams are large, they are already sorted on the join keys.
We illustrate merge join and pseudo code as follows:
select table1.*, from table1, table2 where table1.C1 = table2.C2
where columns C1 and C2 are non-clustered indexes
Here is a simple pseudo-code listing for joining the two relations table1 and table2 utilizing the Merge Join.

get first row C1 from table1
get first row C2 from table2
while not at the end of table1 for table2
begin
  if (C1=C2)
  begin
    return (table1.*)
    get next C2 value from table2
  end
  else if (C1 < C2)
  get next row C1 value from table1
  else
  get next C2 value from table2
end

In merge join the both tables (indexes) are read only once and hence has very less logical and physical I/O.

Enabling and disabling merge joins

By default, merge joins are enabled, at the server level, for allrows mix and for allrows_dss optgoal, and are disabled at the server level for other optgoals, including allrows_oltp. When merge joins are disabled, the server only costs other join types that are not disabled. To enable merge joins server-wide, set enable merge join to 1. The pre-version 15.0 configuration enable sort-merge joins and JTC does not affect the new query processor.

The command set merge_join on overrides the server level to allow use of merge joins in a session or stored procedure.

To enable merge joins, we use:
set merge_join on
To disable merge joins, we use:
set merge_join off

Please see [36,37,38,39] for details.

4.15.10. Hash join improves query performance

Hashing is a Kind of Indexing technique which allows random access to a Hashed Column value. Using the "hash" join operator introduced in ASE 15, Hash Joins have been shown to improve queries performance by as much as 500%. Hash Join will be applied only in Equi Join condition. Hash joins are used when the joining large tables. The optimizer uses smaller of the two tables to build a hash table in memory and the scans the large tables and compares the hash value (of rows from large table) with this hash table to find the joined rows.

We illustrate hash join and pseudo code as follows:
select table1.*, table2 where table1.C1 = table2.C2
where columns C1 and C2 are non-clustered indexes
The algorithm of hash join is divided in two parts; 1) Build an in-memory hash table on smaller on smaller of the two tables, and 2) Probe this hash table with hash value for each row second table [38].

Build Phase: (We can also say optimizer takes the smaller of the two tables to be joined (based on the statistics) and read all the rows and constructs an in-Memory Hash table:
for each row R1 in Table1
begin
  calculate hash value C1
  insert R1 into the appropriate hash bucket
end

Probe Phase: (We can also say in this phase reads all rows from the second table (often called right probe input), hashes these rows on the same equijoin keys, and looks or probes for matching rows in the hash table)
for each row R2 in the table2
begin
  calculate hash value on C2
  for each row C1 in the corresponding hash bucket
  if(C1=C2)
  return table1.*
  end
end

This physical operator supports the hash join algorithm. hash join may consume more runtime resources, but is valuable when the joining columns do not have useful indexes or when a relatively large number of rows satisfy the join condition, compared to the product of the number of rows in the joined tables.

The left child generates the build input stream. The right child generates the probe input stream. The build set is generated by completely draining the build input stream when the first row is requested from the HASH JOIN operator. Every row is read from the input stream and hashed into an appropriate bucket using the hash key.

Each row from the probe set is hashed. A lookup is done in the corresponding build bucket to check for rows matching hash keys. This occurs if the build set’s bucket is memory resident.

If there is not enough memory to hold the entire build set, then a portion of it spills to disk. This portion is referred to as a hash partition and should not be confused with table partitions.

Below is an example of forced hash join.
select t1.invoice_id, t1.total, t1.status_cd, t1.fiscal_qtr, t1.margin
from invoice_master t1, client_master d
where d.region_id = 2001
  and t1.invoice_id = d.invoice_id
  and t1.fiscal_qtr between 20031 and 20044
order by t1.invoice_id, t1.fiscal_qtr
plan " (h_join (scan t1) (scan d))"
go [42]

Enabling and disabling hash joins

By default, hash joins are enabled only at allrows_dss optgoal. To override the server level to allow use of hash join in a session or stored procedure, use set hash_join on.

To enable hash joins, we use:
set hash_join on
To disable hash joins, we use:
4.20. Improve performance at network layer
We can take advantage of several techniques that will improve network performance and are 1) Using small packets for most database activity, 2) Using large packet size for task that perform large data transfer, 3) Using stored procedure to reduce overall traffic, 4) Filtering data to avoid large transfers, 5) Isolating heavy network users from ordinary users and 6) Using client control mechanism for special case.

4.21. Tuning I/O System
With more data available to query, we now has to understand and tune the I/O system in order to provide adequate response time for retrieving and writing data. Please see details in [31].

4.22. Forcing an index with abstract plans
We show an example to scan the lineitem table without an index.

```sql
select count(*) from orders, lineitem where o_orderkey = l_orderkey
```

This method may not create the best available query plan, and may run faster if we use the l_idx1 index on lineitem. We try to rewriting the query to force the index.

```sql
select count(*) from orders, lineitem (index l_idx1) where o_orderkey = l_orderkey
```

Although using force parameter often solves query plan issues, it requires that we change the application code. Even if changing the code is not a problem for us, this can take much longer than using abstract query plans.

First we enable abstract plan.

```sql
set option show_abstract_plan on
```

Adaptive server generates the abstract plans, which we edit and then force to use an index.

```sql
select count(*) from orders, lineitem where o_orderkey = l_orderkey
```

The abstract plan of the final query execution plan is

```
(nl_join (t_scan orders) (t_scan lineitem))(prop orders (parallel 1)(prefetch 2)(lru))(prop lineitem (parallel 1)(prefetch 2)(lru))
```

4.23. Move hot files and tables/indexes to a faster storage device
Once all the tuning has been done and performance still does not meet the fast enough, alternative measures must be investigated. In this situation we can take one measure to move hot files and tables/indexes to a faster storage device. The fastest storage device available is a solid-state file cache. Hot files are 1) Transaction Logs, 2) tempdb and 3) Highly accessed tables/indexes. Solid-state file cache can retrieve a random data block with no mechanical delays, access to the data is virtually instantaneous.
Solid-state file caching systems have no moving parts, so they experience no mechanical delays when accessing data. They can support random I/O rates measured in the thousands per second. This compares to rotating disk products that can support I/O rates on the order of one hundred per second.

We can increase overall system performance by 200%, 400%, 800% or more by placing the most I/O intensive data segments on a Solid Data solid-state file cache [33].


For high performance applications, we typically configure Sybase applications with data segments on raw disk partitions. In contrast to a raw partition, a typical UNIX file system uses a read-ahead buffer cache, which can actually slow down the I/O performance if the database has its own buffer caches.

There is also a resiliency benefit for using raw partitions with Sybase devices. A normal UNIX file system uses a buffer cache for disk I/O. Since the Sybase database does a write-ahead buffering scheme, it assumes that the write has been committed when there is a possibility that data is still in the buffer waiting to be written to disk. In this scenario, if there were a system/disk failure before the buffer cache is written to disk, there would be no way to recover or rollback the data.

By using raw partitions, the server can then manage the disk I/O without any buffering from a file system so that if there were any system / disk failures, the system would know what part of the transaction completed and could recover or rollback the data [33].

4.25. Smart Partitions

ASE 15 makes large databases easy to manage and more efficient by allowing to divide tables into smaller partitions that can be individually managed. We can run maintenance tasks on selected partitions to avoid slowing overall performance and queries run faster because ASE 15’s smart query optimizer bypass partitions that don’t contain relevant data [34].

5. CONCLUSIONS

On the basis of our survey of existing works, findings, technical merits, Sybase Reference Manuals, and illustrating solutions to some situations normally occurs during server run time for better server performance, we have discussed systematically about the performance tuning of database server and query processing and optimization of query processing taking into consideration a Sybase Adaptive Server Enterprise as a robust Database system at present. This paper will provide an adequate support to a person who is using Sybase and cater knowledge how to achieve a challenging goal of high throughput and less or fast response time. This paper will also provide good support to researchers who are working on Database Server Performance Tuning and Query Optimization. In future we shall address this issue in other databases.

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