

**INDEXING STRATEGY FOR BIG DATA PROCESSING: A
CASE STUDY OF PINGER**



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Abstrak

Dengan adanya jumlah data yang besar yang dikumpul secara berterusan dan dikongsi oleh organisasi, maka telah menjadi keperluan untuk memenuhi pemprosesan yang baru muncul dan mendapatkan semula keperluan yang berkaitan dengan jumlah data yang besar ini. Hal ini boleh dicapai melalui mengindeks set data dan mengurangkan pengiraan berat biasa kepada strategi pengindeksan yang terkini semasa pemprosesan jumlah set data yang sangat besar. Kajian ini mencadangkan satu strategi Pengindeksan yang dinamakan sebagai Big Data INDEXing Strategy (BIND), yang menggunakan konsep pengkomputeran selari berprestasi tinggi. BIND menyokong pengedaran data selari dan menjalankan pemprosesan dalam bentuk MapReduce. Bagi membina strategi BIND, konsep penjadualan tugas lan foster untuk pemprosesan selari digunakan. Cadangan strategi pengindeksan yang pertama kali diuji ke atas persekitaran kelompok 2-nod mempunyai pelbagai saiz set data digunakan untuk mengambil perhatian samada prestasinya bertambah baik atau merosot mengikut pertambahan saiznya. Selepas itu, ia telah diuji pada kelompok 3-nod untuk diambil perhatian prestasinya apabila bilangan sumber pengiraan ditambah. Keputusan menunjukkan BIND mengurangkan pemprosesan dan query time berbanding dengan strategi semasa. Hasil kajian mempunyai implikasi yang signifikan dalam keberkesanannya menguruskan Big Data dan memudahkan penyimpanan data serta dapatan semula maklumat untuk pengguna dan organisasi yang menguruskan Big Data ini.

Kata kunci: Big Data; Pemprosesan; Pengindeksan; MapReduce; Dapatan Semula Maklumat

Abstract

With the huge amount of data continuously accumulated and shared by individuals and organizations, it has become necessary to meet the emerging processing and retrieval requirements associated with these large volumes of complex data. This could be achieved by indexing the data sets and reducing heavy computational overhead accustomed to most current indexing strategies during processing of very large amount of data sets. This study proposed a novel Indexing strategy called Big Data INDEXing Strategy (BIND), using a concept of high performance parallel computing. BIND supports parallel distribution of data and performs processing in a MapReduce fashion. To develop BIND strategy, Ian foster's task-scheduling concept for parallel processing is applied. The proposed indexing strategy was first tested on a 2-node cluster environment where varying sizes of datasets were used to note if the performance improves or declines as the size of the data increases. Subsequently, it was tested on a 3-node cluster to note the performance when the number of computation resources are increased. The results demonstrate that BIND minimizes the processing and query time as compared to the current strategy. The findings have significant implication in efficiently managing Big Data and facilitating data storage and information retrieval for users and organizations that manage Big Data.

Keywords: Big Data; Processing; Indexing; MapReduce; Information Retrieval

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In the name of ALLAH, Most Gracious, Most Merciful:

“Work; so Allah will see your work and (so will) His Messenger and the believers;”

(The Holy Quran - AtTawbah 9:105)

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Dedication

To my family.



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List of Abbreviations

BIND	-	Big Data INDEXing
GPS	-	Global Positioning System
RFID	-	Radio Frequency Identification System
IT	-	Information Technology
MB	-	Megabyte
ZB	-	Zettabyte
TB	-	Terabyte
EX	-	Exabyte
SQL	-	Structured Query Language
RDBMS	-	Relational Database Management System
IDC	-	International Data Corporation
PingER	-	Ping End-to-end Reporting
ABDC	-	Analytics and Big Data Committee
SDAV	-	Scalable Data Management, Analysis and Visualization Institute
MA	-	Measurement Agents
PC	-	Personal Computer
IEPM	-	Internet End-to-end Performance Measurement
SLAC	-	Standard Linear Accelerator Center
HDFS	-	Hadoop Distributed File System
CSV	-	Comma Separated Values
JT	-	JobTracker
RM	-	ResourceManager
TT	-	TaskTracker
NM	-	NodeManager
DN	-	DataNode
NN	-	NameNode
SNN	-	SecondaryNameNode
RAM	-	Random Access Memory
CPU	-	Central Processing Unit

NNS	- Nearest Neighbor Search
AI	- Artificial Intelligence
NAI	- Non-Artificial Intelligence
LSI	- Latent Semantic Indexing
HMM	Hidden Markov Model
SVD	Singular Value Decomposition
RDF	Resource Description Framework
GiST	Generalized Search Tree
GIN	Generalized Inverted Index
PT	Posting Tree
ET	Entries Tree
PL	Posting List
RTT	Round Trip Time
ICMP	Internet Control Message Protocol
QoS	Quality of Service
AIS	Asynchronous Index Strategy
ETL	Extract, Transform, Load
LFS	Local File System
K-V	Key-Value
K-L(V)	Key-List (Value)
UI	User Interface
SSH	Secure Shell
JAR	Java Archive
IDE	Integrated Development Environment
HQL	Hive Query Language
IS	Indexing Strategy
TTL	Time-To-Leave
IR	Information Retrieval

CHAPTER ONE

INTRODUCTION

1.1 OVERVIEW

Big Data is a term used to describe very large data sets that are of different forms or structure (complex), generated at a very high speed, and cannot be managed by traditional database management systems [6]. This definition explains the three (3) main characteristics associated with Big Data: volume, variety and velocity (3Vs), and the value that can be extracted from it is seen as a fourth characteristic (4V's) [7]. Big Data is sourced from so many end devices such as Personal Computers (PC), smart phones, Global Positioning System (GPS) devices [8], sensors, Radio Frequency Identification (RFID) devices, etc. Also, online applications such as social networks and applications that involve video streaming are great sources that generate Big Data.

According to Zhou et al. in [9], the total size of data generated will surpass 7.9 Zettabytes (ZB) by the end of 2015, and predicted to reach 35ZB in 2020. Significant interest have been taken in this area lately, because of the ubiquitous and pervasive nature of data, and the emergence of mobile and cloud computing which results in the generation of huge amounts of data. The application of Big Data analytic has been of great value in terms of decision making in areas concerned with development [10, 11], security [12], and health care [13, 14, 15]. Big Data has also been used in the prediction of natural disaster, energy consumption, economic productivity, social network (for customer or consumer interest), Information Technology (IT), business[16], and other domains. Cisco related that organizations such as Facebook, Yahoo, Google, Twitter, etc., deal with Big Data on a daily basis, and have realized the need to analyze historical data for decision making [17]. The use of so many wired and wireless devices has generated voluminous data that are to be collected, transferred, processed,

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