UNIT - III

Processing data with Hadoop

- A large file to be processed is divided into chunks and is processed parallelly
- using Map reduce
- 2 tasks (sorting & shuffling) keys
  i) Map task - process each every chunk of data & produce intermediate result
  ii) reduce task - O/P of map task is i/p to reduce task & combines the o/p of all map task

HDFS & map reduce is present in same node making data to be available increasing performance

Processing using Map reduce

2 trackers

jobtracker . task tracker

- Present in name node.
- gives task
- rescheduling
- check for failure

- present in every data node

Mapreduce phases & daemons

Phase

Map: doesmapping & produce o/p as key/value pair

reduce: combine o/p from different maps & gives result

daemons

jobtracker
- master
- schedule task

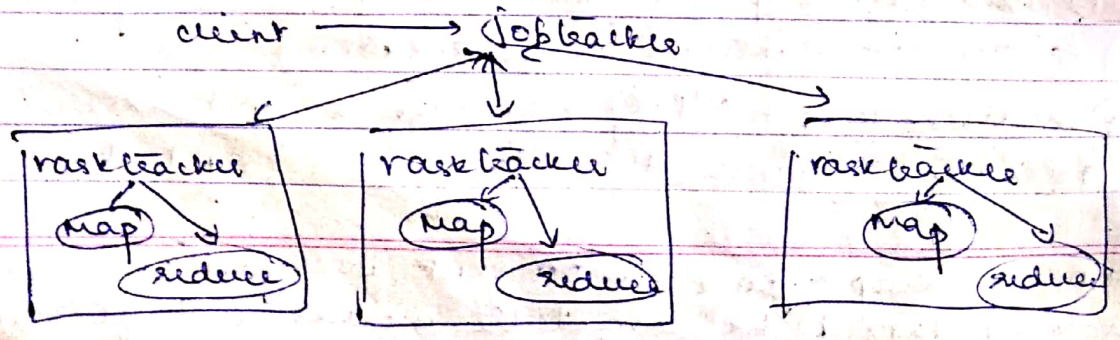
tasktracker
- slave
- exe. task

Daemons

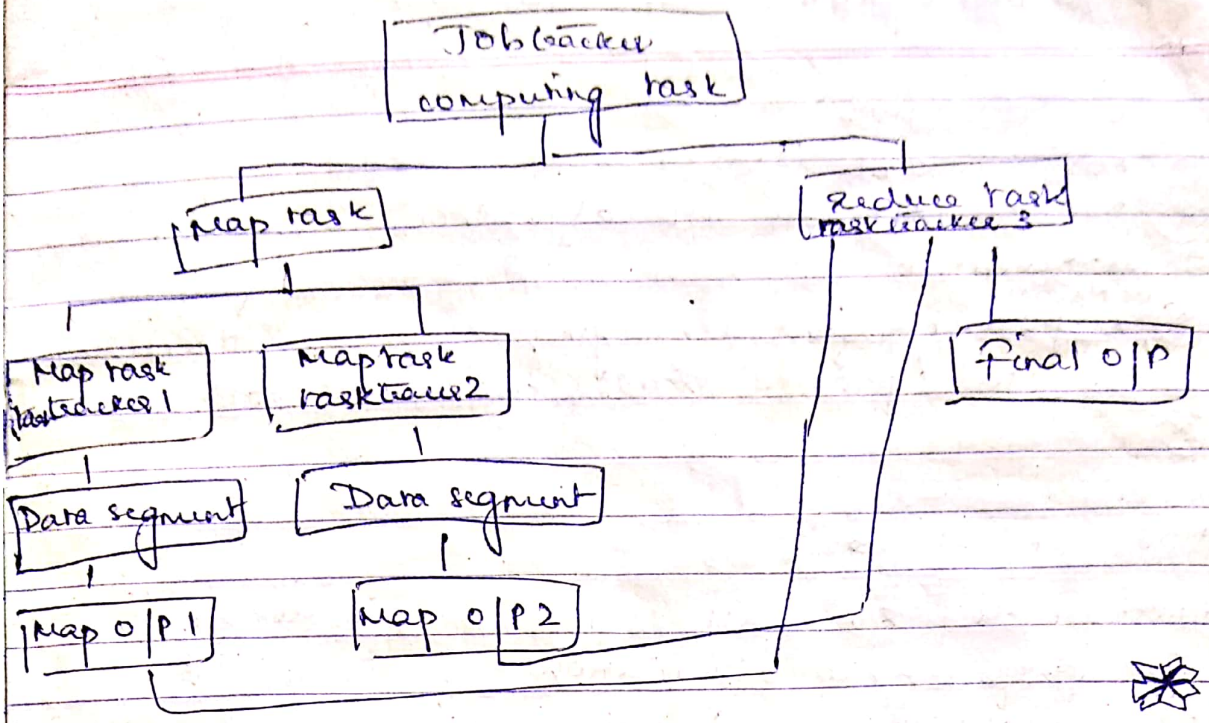
i) Jobtracker : - provides connectivity b/w application & hadoop
- figures out how to assign task to task tracker
- incase of failure it tries again & again & task is given to other node

ii) Task tracker - execute task assigned by jobtracker
- each node has one task tracker & multiple VMs
- Sends heartbeat message to jobtracker to show exists
- & incase of failure the task is rescheduled to another node.

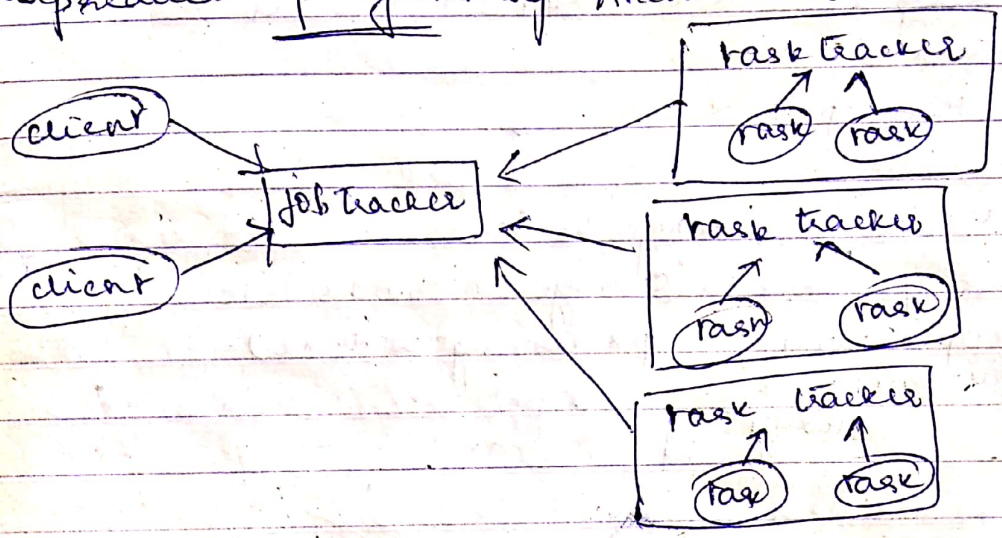
Jobtracker & tasktracker interaction



Mapreduce programming workflow

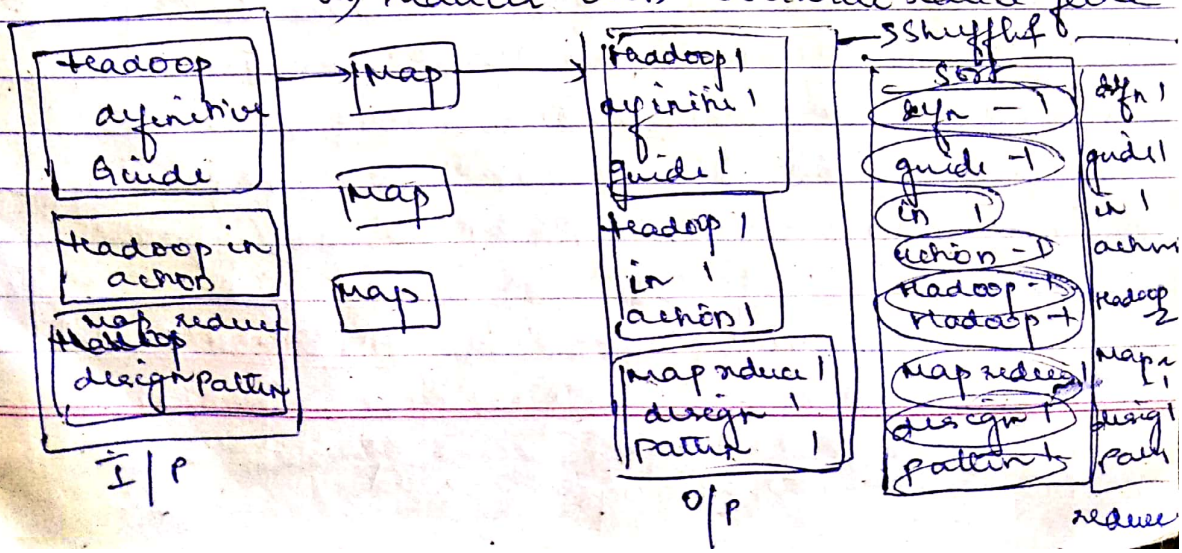


Mapreduce programming Architecture.



Eg. word count

Support class i) driver class - info of job config
ii) mapper class - overwrite map func
iii) reducer class - overwrite reduce func



Limitations of Hadoop 1.0

1. single NameNode - all the responsibility.
2. Processing is restricted that is carried - batch oriented jobs
3. not suitable for interactive analysis processing of data
4. not suitable for ML algorithms & graph, menu application
5. MapReduce responsible for managing resources & processing of data (produce resource utilization problems)
6.

HDFS limitations

1. NameNode stores all data in main Memory leading to performance degradation

This is overcome using Hadoop 2.0

On Hadoop 2.0 we use YARN - another resource negotiator

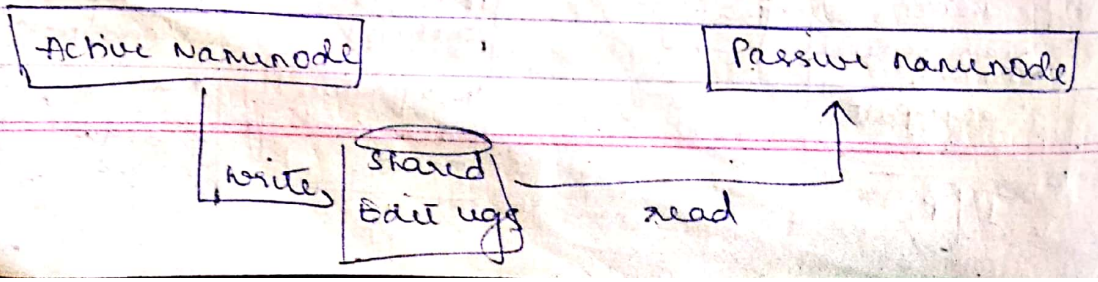
Hadoop 2.0

- Namespace (handles diff file operation) create | update | delete | modify
- block storage service (storage in blocks & low processing is done)

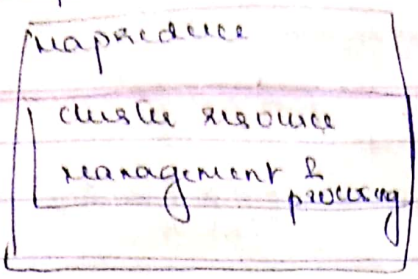
HDFS-2 features

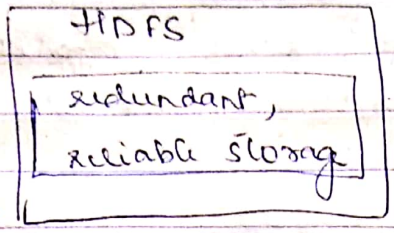
1. Horizontal scalability & stability
2. highly available.

1. multiple name nodes with data node & register itself with NameNode. No direct interaction within NameNode: one NameNode is active & other NameNodes will be passive, in case of failure, passive becomes active NameNode

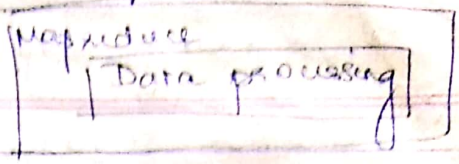


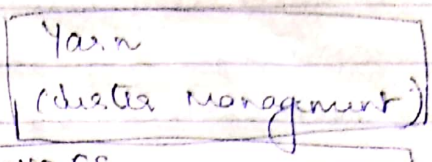
Hadoop 1.0

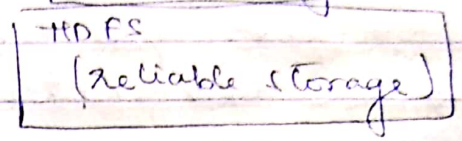




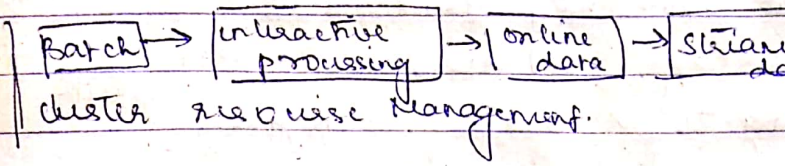
Hadoop 2.0







Hadoop Yarn



major components of yarn

- Resource Manager — in Master node
- Node Manager — in Slave node
- Application Master — in Slave node

- Responsibility
  i) Managing resources (optimal)
  ii) Schedule & apply Manager -
  - co-ordination
  - scheduling exe of job by RM
  - allocating

Responsibility of RM
i) Managing resources
ii) Schedule & application Manager

Scheduler:
- Schedules job execution as requested by RM
- allocating resource to application submitted to cluster

- communicating with AM, keep track of resources of running application

iii) Application Manager
- co-ordinate with scheduler to keep track of running application
- accepting the job submission from the client
- negotiating first container for executing application specific task with suitable application master on slave node.

(2) Node Manager - responsibility
i) managing & executing containers
ii) monitoring usage of resource - Memory, CPU and reporting to resource manager
iii) Sending heart beat messages regarding status update to resource manager

(3) Application Master - responsibility
i) per application specific library which works with node manager to execute task
ii) if multiple jobs are submitted on cluster, more than one instances of application master on slave node
iii) negotiating resource containers
iv) working with one or more node managers

Diff
Differences b/w yarn & map reduce

yarn	Map reduce
i) Support variety of processing engines & appln	i) support its own appln
ii) separates its duties access multiple component	ii) consolidates most of task by single component.
iii) dynamically allocate ports of resources to appln	iii) Static allocation of resources for designated task.

Needs of yarn

- opened up new usecases for Apache HBase & Apache Hive
- offers scalability - any amt of data
- resource utilization &
- high availability
- performance is improved wrt to mapreduce
- suitable for real time processor since it separates HDFS from map reduce, real time processing & appln can't wait for batch jobs to finish
- manages resources in cluster envt.
- in case of scarcity of resources - communicates with compute resources & assign resources.

Application of Map-Reduce — Softcopy.