

## **QUESTION 1**

### **What is UBER mode ?**

When a particular Hadoop job is invoked from client machine, RM will create separate container for that application. Uber configuration, will allow running that job in the same process as the Application Master (AM).

### **Uber jobs:**

Uber jobs are jobs that are executed within the ApplicationMaster. It do not need to communicate with RM to create the containers on different worker nodes. As the job is given is evaluated small so the AM runs tasks within its own process and avoided the overhead of launching and communicate with remote containers located in different worker nodes can be considered as advantage in certain cases.

### **Why?**

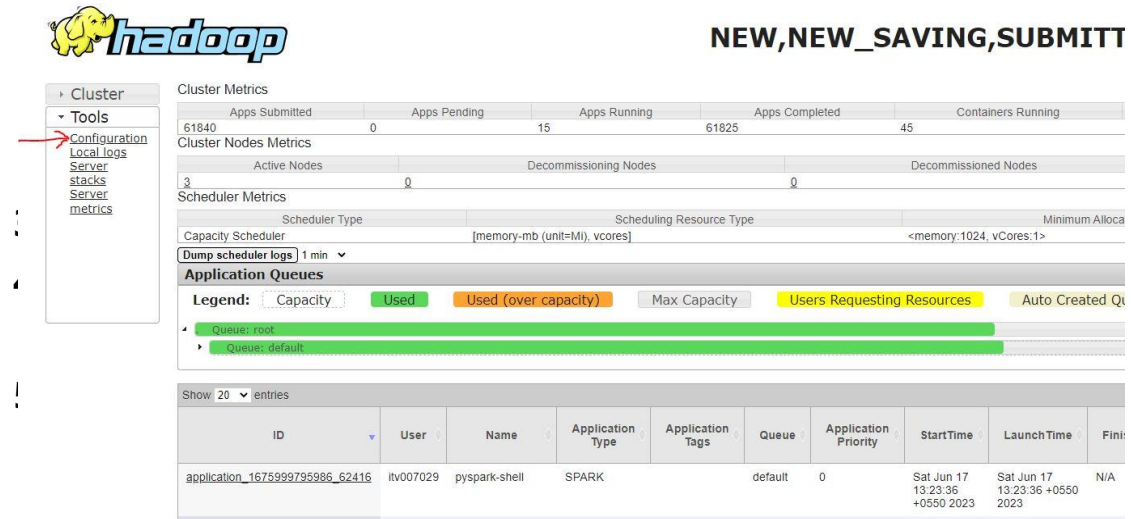
If you have a small dataset, Uber configuration will help you out, by reducing additional time required to complete the certain Hadoop job.

### **Configuration options for Uber Jobs**

Property	Description
mapreduce.job.ubertask.enable	Whether to enable the small-jobs "ubertask" optimization, which runs "sufficiently small" jobs sequentially within a single JVM. "Small" is defined by the following maxmaps, maxreduces, and maxbytes settings. Users can override this value.  Default = false
mapreduce.job.ubertask.maxmaps	The threshold for the number of maps beyond which a job is considered too large for the ubertasking optimization. Users can override this value, but only downward.  Default = 9
mapreduce.job.ubertask.maxreduces	The threshold for the number of reduces beyond which a job is considered too large for the uber tasking optimization. CURRENTLY THE CODE CANNOT SUPPORT MORE THAN ONE REDUCE and will ignore larger values (zero is a valid maximum value, however). Users can override this value, but only downward.  Default = 1
mapreduce.job.ubertask.maxbytes	The threshold for the number of input bytes beyond which a job is considered too large for the uber tasking optimization. If no value is specified, dfs.block.size is used as the default. Be sure to specify a default value in mapred-site.xml if the underlying file system is not HDFS. Users can override this value, but only downward.  Default = HDFS Block Size

## SOP for accessing UBER mode in Resource Manager –

1. Go to <http://172.16.1.104:19088/> by enabling foxy proxy server.
2. Click on Configuration option under Tool tab on left side of screen.



**NEW,NEW\_SAVING,SUBMITT**

**Cluster Metrics**

Apps Submitted	Apps Pending	Apps Running	Apps Completed	Containers Running
61840	0	15	61825	45

**Cluster Nodes Metrics**

Active Nodes	Decommissioning Nodes	Decommissioned Nodes
3	0	0

**Scheduler Metrics**

Scheduler Type	Scheduling Resource Type	Minimum Allocated
Capacity Scheduler	[memory-mb (unit=Mi), vcores]	<memory:1024, vCores:1>

**Application Queues**

Legend: Capacity Used Used (over capacity) Max Capacity Users Requesting Resources Auto Created Queue

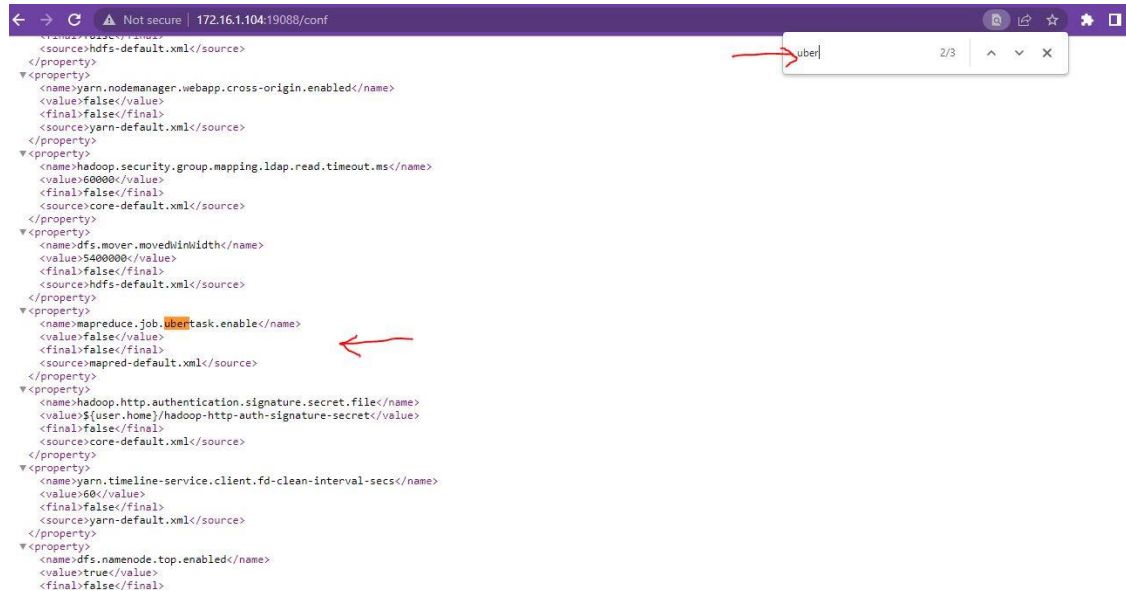
Queue: root

Queue: default

Show: 20 entries

ID	User	Name	Application Type	Application Tags	Queue	Application Priority	StartTime	LaunchTime	Final
application_1675999795986_62416	litv007029	pyspark-shell	SPARK		default	0	Sat Jun 17 13:23:36 +0550 2023	Sat Jun 17 13:23:36 +0550 2023	N/A

3. Find in page for the keyword "Uber". Setting will appear change as mentioned.




Not secure | 172.16.1.104:19088/conf

uber 2/3

`<name>mapreduce.job.ubertask.enable</name>  
<value>false</value>  
<final>false</final>  
<source>mapred-default.xml</source>`

## QUESTION 2

**ABOUT** - Gives the information about the cluster and it's metrics eg – memory used, total memory, v cores etc. Other information like Cluster Node Metrics and Scheduler Metrics (min and max vcore allocation info). We can also find certain details related to commissioning of Resource Manager.



Logged in as: drwho

### About the Cluster

Cluster

About

Nodes

Node Labels

Applications

NEW

NEW SAVING

SUBMITTED

ACCEPTED

RUNNING

FINISHED

FAILED

KILLED

Scheduler

Tools

Cluster Metrics

Apps Submitted	Apps Pending	Apps Running	Apps Completed	Containers Running	Memory Used	Memory Total	Memory Reserved	VCores Used	VCores Total	VCores Reserved
61881	0	14	61867	42	70 GB	151.00 GB	0 B	42	90	0

Cluster Nodes Metrics

Active Nodes	Decommissioning Nodes	Decommissioned Nodes	Lost Nodes	Unhealthy Nodes	Rebooted Nodes	Shutdown Nodes
3	0	0	1	0	0	0

Scheduler Metrics

Scheduler Type	Scheduling Resource Type	Minimum Allocation	Maximum Allocation	Maximum Cluster Application Priority
Capacity Scheduler	[memory-mb (unit=M), vcores]	<memory:1024, vCores:1>	<memory:8192, vCores:4>	0

Cluster overview

Cluster ID: 167599795986

ResourceManager state: STARTED

ResourceManager HA state: active

ResourceManager HA zookeeper connection state: Could not find leader elector. Verify both HA and automatic failover are enabled.


ResourceManager RMStateStore: org.apache.hadoop.yarn.server.resourcemanager.recovery.NullRMStateStore

ResourceManager started on: Thu Feb 09 22:29:55 -0500 2023

ResourceManager version: 3.3.0 from aa96f1871bf0858f9bac59cf2a81ec470da649af by brahma source checksum e0a276649f889c15d0e8f08eccd0c10 on 2020-07-06T18:58Z

Hadoop version: 3.3.0 from aa96f1871bf0858f9bac59cf2a81ec470da649af by brahma source checksum 5dc29b802d6ccd77b262ef9d04d19c4 on 2020-07-06T18:44Z

**Nodes** – Shows information about 3 worker nodes present on this cluster.



Logged in as: drwho

### Nodes of the cluster

Cluster

About

Nodes

Node Labels

Applications

NEW

NEW SAVING

SUBMITTED

ACCEPTED

RUNNING

FINISHED

FAILED

KILLED

Scheduler

Tools

Cluster Metrics

Apps Submitted	Apps Pending	Apps Running	Apps Completed	Containers Running	Memory Used	Memory Total	Memory Reserved	VCores Used	VCores Total	VCores Reserved
61888	0	15	61873	45	75 GB	151.00 GB	0 B	45	90	0

Cluster Nodes Metrics

Active Nodes	Decommissioning Nodes	Decommissioned Nodes	Lost Nodes	Unhealthy Nodes	Rebooted Nodes	Shutdown Nodes
3	0	0	1	0	0	0

Scheduler Metrics

Scheduler Type	Scheduling Resource Type	Minimum Allocation	Maximum Allocation	Maximum Cluster Application Priority
Capacity Scheduler	[memory-mb (unit=M), vcores]	<memory:1024, vCores:1>	<memory:8192, vCores:4>	0


Show: 20 entries

Node Labels	Rack	Node State	Node Address	Node HTTP Address	Last health update	Health report	Containers	Allocation Tags	Mem Used	Mem Avail	VCores Used	VCores Avail	Version
/default-rack		RUNNING	w01.iversity.com:35127	w01.iversity.com:8042	Sat Jun 17 05:15:41 -0400 2023		10		15 GB	35.33 GB	10	20	3.3.0
/default-rack		RUNNING	w03.iversity.com:41791	w03.iversity.com:8042	Sat Jun 17 05:14:55 -0400 2023		20		34 GB	16.33 GB	20	10	3.3.0
/default-rack		RUNNING	w02.iversity.com:46669	w02.iversity.com:8042	Sat Jun 17 05:16:29 -0400 2023		15		26 GB	24.33 GB	15	15	3.3.0

Showing 1 to 3 of 3 entries

First Previous 1 Next Last

By clicking on link provided under Node HTTP Address of particular worker node we can get info about that particular node and as well **node manager** information.



ResourceManager

NodeManager

Node Information

List of Applications

List of Containers

Tools

Total Vmem allocated for Containers

251.66 GB

Vmem enforcement enabled

true

Total Pmem allocated for Container

50.33 GB

Pmem enforcement enabled

true

Total VCores allocated for Containers

30

Resource types

memory-mb (unit=M), vcores

NodeHealthyStatus

true

LastNodeHealthTime

Sat Jun 17 05:35:41 EDT 2023

NodeHealthReport

NodeManager started on

Wed Feb 15 08:19:38 EST 2023


NodeManager Version:

3.3.0 from aa96f1871bf0858f9bac59cf2a81ec470da649af by brahma source checksum e0a276649f889c15d0e8f08eccd0c10 on 2020-07-06T18:58Z

Hadoop Version:

3.3.0 from aa96f1871bf0858f9bac59cf2a81ec470da649af by brahma source checksum 5dc29b802d6ccd77b262ef9d04d19c4 on 2020-07-06T18:44Z

**Applications** – Shows all the links and key information about all the application present on this cluster, where different tab under applications can show particular job which are running, finished or killed etc.



**All Applications**

Cluster Metrics

Apps Submitted	Apps Pending	Apps Running	Apps Completed	Containers Running	Memory Used	Memory Total	Memory Reserved	V-Cores Used	V-Cores Total	V-Cores Reserved
1	0	16	61890	62	108 GB	181.00 GB	0 B	62	90	0

Scheduler Metrics

Active Nodes	Decommissioning Nodes	Decommissioned Nodes	Lost Nodes	Unhealthy Nodes	Rebooted Nodes	Shutdown Nodes
0	0	0	0	0	0	0

Capacity Scheduler

Schedule Type	Scheduling Resource Type	Minimum Allocation	Maximum Allocation	Maximum Cluster Application Priority
memory=10 (unit=MB), vCores	memory=1024 vCores=16	0	0	0

Show 20 entries

ID	User	Name	Application Type	Application Tag	Queue	Application Priority	StartTime	LaunchedTime	FinishTime	State	FinalStatus	Running Containers	Allocated CPU V-Cores	Allocated Memory MB	Reserved CPU V-Cores	Reserved Memory MB	% of Queue	% of Cluster	Progress	Tracking UI	Blacklist Nodes
application_1675999795986_02416	itv00709	itv00709-grouping-aggregation	SPARK		default	0	Sat Jun 17 18:21:18 +0500 2023	Sat Jun 17 18:21:17 +0500 2023	N/A	RUNNING	UNDEFINED	3	3	8120	0	0	3.3	3.3	<div></div>	ApplicationMaster	0
application_1675999795986_02411	itv00265	supern-sitel	SPARK		default	0	Sat Jun 17 18:21:18 +0500 2023	Sat Jun 17 18:21:17 +0500 2023	N/A	RUNNING	UNDEFINED	10	10	19466	0	0	12.6	12.6	<div></div>	ApplicationMaster	0
application_1675999795986_02410	itv04946	supern-sitel	SPARK		default	0	Sat Jun 17 18:17:43 +0500 2023	Sat Jun 17 18:17:52 +0500 2023	N/A	RUNNING	UNDEFINED	3	3	8120	0	0	3.3	3.3	<div></div>	ApplicationMaster	0
application_1675999795986_02415	itv00721	supern-sitel	SPARK		default	0	Sat Jun 17 18:16:43 +0500 2023	Sat Jun 17 18:16:44 +0500 2023	N/A	RUNNING	UNDEFINED	3	3	8120	0	0	3.3	3.3	<div></div>	ApplicationMaster	0
application_1675999795986_02413	itv00081	supern-sitel	SPARK		default	0	Sat Jun 17 18:16:43 +0500 2023	Sat Jun 17 18:16:48 +0500 2023	N/A	RUNNING	UNDEFINED	3	3	8120	0	0	3.3	3.3	<div></div>	ApplicationMaster	0
application_1675999795986_02417	itv00081	supern-sitel	SPARK		default	0	Sat Jun 17 18:16:24 +0500 2023	Sat Jun 17 18:16:24 +0500 2023	N/A	RUNNING	UNDEFINED	3	3	8120	0	0	3.3	3.3	<div></div>	ApplicationMaster	0
application_1675999795986_02414	itv00265	supern-sitel	SPARK		default	0	Sat Jun 17 18:16:17 +0500 2023	Sat Jun 17 18:16:17 +0500 2023	N/A	FINISHED	SUCCEEDED	N/A	N/A	N/A	N/A	N/A	0.0	0.0	<div></div>	History	0

Link under the ID of particular running application will lead to that particular application information in detailed view.



**Application application\_1675999795986**

Kill Application

User: itv00709  
 Name: itv00709-grouping-aggregation  
 Application Type: SPARK  
 Application Tags:  
 Application Priority: 0 (Higher integer value indicates higher priority)  
 YarnApplicationState: RUNNING: AM has registered with RM and started running.  
 Queue: default  
 FinalStatus Reported by AM: Application has not completed yet.  
 Started: Sat Jun 17 05:51:15 -0400 2023  
 Launched: Sat Jun 17 05:51:17 -0400 2023  
 Finished: N/A  
 Elapsed: 5mins, 33sec  
 Tracking URL: ApplicationMaster  
 Log Aggregation Status: NOT\_START  
 Application Timeout (Remaining Time): Unlimited  
 Diagnostics:  
 Unmanaged Application: false  
 Application Node Label expression: <Not set>  
 AM container Node Label expression: <DEFAULT\_PARTITION>


Total Resource Preempted: <memory=0, vCores=0>  
 Total Number of Non-AM Containers Preempted: 0  
 Total Number of AM Containers Preempted: 0  
 Resource Preempted from Current Attempt: <memory=0, vCores=0>  
 Number of Non-AM Containers Preempted from Current Attempt: 0  
 Aggregate Resource Allocation: 1666750 MB-seconds, 607 vcore-seconds  
 Aggregate Preempted Resource Allocation: 0 MB-seconds, 0 vcore-seconds

Show 20 entries

Attempt ID	Started	Node	Logs	Nodes blacklisted by the app
attempts_1675999795986_000001	Sat Jun 17 18:21:15 +0500 2023	http://w01.lhvacslv.com:8042	Logs	0

Showing 1 to 1 of 1 entries

By clicking on link under Tracking URL named as Application Master will lead us to **Spark UI** of that particular job and we can navigate through various information about the job in that particular application.



**Spark Jobs**

User: itv00709  
 Total Uptime: 9.1 min  
 Scheduling Mode: FIFO  
 Completed Jobs: 7

Event Timeline

Executors

- Added
- Removed

Jobs

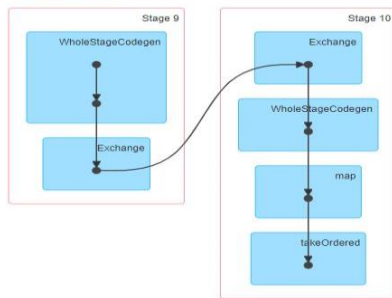
- Succeeded
- Failed
- Running

Completed Jobs (7)

Job id	Description	Submitted	Duration	Stages: Succeeded/Total	Tasks (for all stages): Succeeded/Total
6	showString at NativeMethodAccessorImpl.java:0 showString at NativeMethodAccessorImpl.java:0	2023/06/17 05:54:51	0.8 s	2/2	202/202

**Link under Description** section will lead us to DAG Visualization and stages information. **Stages tab** show completed stages. Storage tab show the cached results. **Environment tab** gives us information about spark properties, system properties etc. **Executors tab** show the information about information about driver and executors involved for this application in order to complete a job. When clicked on **SQL tab** shows the completed queries and when clicked in description details of the query with physical, optimized, analyzed and parsed logical plan.

#### ▼ DAG Visualization



#### ▼ Completed Stages (2)

Stage Id	Description	Submitted	Duration	Tasks: Succeeded/Total	Input	Output	Shuffle Read	Shuffle Write
10	showString at NativeMethodAccessorImpl.java:0	+details 2023/06/17 05:54:51	0.4 s	200/200			597.6 KB	
9	showString at NativeMethodAccessorImpl.java:0	+details 2023/06/17 05:54:51	0.4 s	2/2	44.0 MB			597.6 KB

## Executors

► Show Additional Metrics

### Summary

	RDD Blocks	Storage Memory	Disk Used	Cores	Active Tasks	Failed Tasks	Complete Tasks	Total Tasks	Task Time (GC Time)	Input	Shuffle Read	Shuffle Write	Blacklisted
Active(3)	0	113.4 KiB / 1.1 GiB	0.0 B	2	0	0	1	1	1 s (25.0 ms)	128 KiB	0.0 B	0.0 B	0
Dead(0)	0	0.0 B / 0.0 B	0.0 B	0	0	0	0	0	0.0 ms (0.0 ms)	0.0 B	0.0 B	0.0 B	0
Total(3)	0	113.4 KiB / 1.1 GiB	0.0 B	2	0	0	1	1	1 s (25.0 ms)	128 KiB	0.0 B	0.0 B	0

### Executors

Show 20 entries

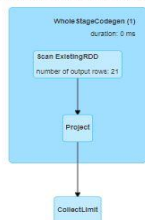
Search:

Executor ID	Address	Status	RDD Blocks	Storage Memory	Disk Used	Cores	Active Tasks	Failed Tasks	Complete Tasks	Total Tasks	Task Time (GC Time)	Input	Shuffle Read	Shuffle Write	Logs	Thread Dump
driver	g02.itversity.com:43861	Active	0	56.7 KiB / 397.5 MiB	0.0 B	0	0	0	0	0	0.0 ms (0.0 ms)	0.0 B	0.0 B	0.0 B		Thread Dump
1	w02.itversity.com:44731	Active	0	56.7 KiB / 366.3 MiB	0.0 B	1	0	0	1	1	1 s (25.0 ms)	128 KiB	0.0 B	0.0 B	stdout stderr	Thread Dump
2	w01.itversity.com:38963	Active	0	0.0 B / 366.3 MiB	0.0 B	1	0	0	0	0	0.0 ms (0.0 ms)	0.0 B	0.0 B	0.0 B	stdout stderr	Thread Dump

Duration: 1 s

Succeeded Jobs: 3

☐ Show the Stage ID and Task ID that corresponds to the max metric



#### ▼ Details

```

-- Parsed Logical Plan --
GlobalLimit 21
+ LocalLimit 21
+ Project [cast(_$B331 as string) AS _$B443, cast(_$B432 as string) AS _$B444, cast(_$B433 as string) AS _$B445, cast(_$B434 as string) AS _$B446]
+ LogicalRDD [_$B431, _$B432, _$B433, _$B434], false

-- Analyzed Logical Plan --
_1: string, _2: string, _3: string, _4: string
GlobalLimit 21
+ LocalLimit 21
+ Project [cast(_$B431 as string) AS _$B443, cast(_$B432 as string) AS _$B444, cast(_$B433 as string) AS _$B445, cast(_$B434 as string) AS _$B446]
+ LogicalRDD [_$B431, _$B432, _$B433, _$B434], false

-- Optimized Logical Plan --
GlobalLimit 21
+ LocalLimit 21
+ Project [cast(_$B431 as string) AS _$B443, _$B432, cast(_$B433 as string) AS _$B445, _$B434]
+ LogicalRDD [_$B431, _$B432, _$B433, _$B434], false

-- Physical Plan --
CollectLimit 21
+ *(1) Project [cast(_$B431 as string) AS _$B443, _$B432, cast(_$B433 as string) AS _$B445, _$B434]
+ *(1) Scan ExistingRDD[_$B431, _$B432, _$B433, _$B434]
  
```



**TOOLS** – under tools section we can open configuration tab in order to edit predefined configurations and property related to resource manager e.g. – changing uber mode configuration.



**NEW,NEW\_SAVING,SUBMITT**

Cluster
Tools
Configuration
Local logs
Server
Stacks
Server metrics

### Cluster Metrics

Apps Submitted	Apps Pending	Apps Running	Apps Completed	Containers Running
61840	0	15	61825	45

### Cluster Nodes Metrics

Active Nodes	Decommissioning Nodes	Decommissioned Nodes
3	0	0

### Scheduler Metrics

Scheduler Type	Scheduling Resource Type	Minimum Allocation
Capacity Scheduler	[memory-mb (unit=Mi), vcores]	<memory:1024, vCores:1>

**Dump scheduler logs** 1 min

### Application Queues

**Legend:** Capacity Used Used (over capacity) Max Capacity Users Requesting Resources Auto Created Q

- Queue: root
- Queue: default

ID	User	Name	Application Type	Application Tags	Queue	Application Priority	StartTime	LaunchTime	Final
application_1675999795986_62416	itv007029	pyspark-shell	SPARK		default	0	Sat Jun 17 13:23:36 +0550 2023	Sat Jun 17 13:23:36 +0550 2023	N/A

Not secure | 172.16.1.104:19088/conf

```

<source>hdfs-default.xml</source>
</property>
<property>
<name>yarn.nodemanager.webapp.cross-origin.enabled</name>
<value>false</value>
</property>
</property>
<source>yarn-default.xml</source>
</property>
<property>
<name>hadoop.security.group.mapping.ldap.read.timeout.ms</name>
<value>60000</value>
</property>
</property>
<source>core-default.xml</source>
</property>
<property>
<name>dfs.mover.movedwintwidth</name>
<value>540000</value>
</property>
</property>
<source>hdfs-default.xml</source>
</property>
<property>
<name>mapreduce.job.ubertask.enable</name>
<value>false</value>
</property>
</property>
<source>mapred-default.xml</source>
</property>
<property>
<name>hadoop.http.authentication.signature.secret.file</name>
<value>${user.home}/hadoop-http-auth-signature-secret</value>
</property>
</property>
<source>core-default.xml</source>
</property>
<property>
<name>yarn.timeline-service.client.fd-clean-interval-secs</name>
<value>60</value>
</property>
</property>
<source>yarn-default.xml</source>
</property>
<property>
<name>dfs.namenode.top.enabled</name>
<value>true</value>
</property>
</property>

```

uber

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## QUESTION 3

```
[1]: from pyspark.sql import SparkSession
import getpass
username = getpass.getuser()
spark = SparkSession. \
builder. \
appName("Ultron"). \
config('spark.ui.port','0'). \
config("spark.sql.warehouse.dir", f"/user/itv005937/warehouse"). \
enableHiveSupport(). \
master('yarn'). \
getOrCreate()

[2]: groceries_schema = "order_id integer, location string, item string, order_date string, quantity integer"

[3]: from pyspark.sql import *
from pyspark.sql.functions import *

[4]: groceries_new_df = spark.read \
.format("csv") \
.schema(groceries_schema) \
.option("header", "true") \
.load("/user/itv005937/data/groceries_new.csv")

[5]: groceries_df = groceries_new_df.withColumn("order_date", to_date("order_date", "dd/mm/yyyy"))

[6]: groceries_df.createOrReplaceTempView("groceries_sql")
```

### Aggregate function

```
[7]: groceries_df.select(count(*).alias("No_of_Records"),
                        countDistinct("location").alias("No_of_location"),
                        sum("quantity").alias("Total_Quantity")).show()
```

No_of_Records	No_of_location	Total_Quantity
21	7	273

```
[8]: groceries_df.selectExpr("count(*) as No_of_Records",
                             "count(distinct(location)) as No_of_location",
                             "sum(quantity) as Total_Quantity").show()
```

No_of_Records	No_of_location	Total_Quantity
21	7	273

---

```
count(distinct(location)) as No_of_Location",
"sum(quantity) as Total_Quantity").show()
```

No_of_Records	No_of_Location	Total_Quantity
21	7	273

```
[9]: spark.sql("""select count(*) as No_of_Records,
count(distinct(location)) as No_of_Location,
sum(quantity) as Total_Quantity from groceries_sql""").show()
```

No_of_Records	No_of_Location	Total_Quantity
21	7	273

## Grouping Aggregation

```
[10]: GA1 = groceries_df.groupBy("location","item") \
.agg(sum("quantity").alias("Total_quantity_per_group"), sum(expr("(quantity/273)*100")) \
.alias("percentage_of_total_quantity")).sort("location")
```

```
[11]: GA1.show()
```

location	item	Total_quantity_per_group	percentage_of_total_quantity
Bellevue	Flowers	10	3.6630036630036633
Bellevue	Bread	125	45.78754578754578
Issaquah	Meat	14	5.128205128205129
Issaquah	Onion	12	4.395604395604395
Issaquah	Tomato	6	2.197802197802198
Kent	Apples	20	7.326007326007327
Redmond	Cheese	15	5.4945054945054945
Redmond	Meat	40	14.652014652014653
Redmond	Bread	5	1.8315018315018317
Renton	Bread	5	1.8315018315018317
Sammanish	Bread	5	1.8315018315018317
Seattle	Bananas	7	2.564102564102564
Seattle	Potatoes	9	3.296703296703297

```
[12]: GA2 = groceries_df.groupBy("location","item") \
.agg(expr("sum(quantity) as Total_quantity_per_group"), expr("sum((quantity/273)*100 as percentage_of_total_quantity")) \
.sort("location")
```

```
[13]: GA2.show()
```

```
[12]: GA2 = groceries_df.groupBy("location","item") \
.agg(expr("sum(quantity) as Total_quantity_per_group"), expr("sum((quantity/273)*100 as percentage_of_total_quantity")) \
.sort("location")
```

```
[13]: GA2.show()
```

location	item	Total_quantity_per_group	percentage_of_total_quantity
Bellevue	Flowers	10	3.6630036630036633
Bellevue	Bread	125	45.78754578754578
Issaquah	Meat	14	5.128205128205129
Issaquah	Onion	12	4.395604395604395
Issaquah	Tomato	6	2.197802197802198
Kent	Apples	20	7.326007326007327
Redmond	Cheese	15	5.4945054945054945
Redmond	Bread	5	1.8315018315018317
Redmond	Meat	40	14.652014652014653
Renton	Bread	5	1.8315018315018317
Sammanish	Bread	5	1.8315018315018317
Seattle	Bananas	7	2.564102564102564
Seattle	Potatoes	9	3.296703296703297

```
[15]: spark.sql("""select location, item, sum(quantity) as Total_quantity_per_group,
sum((quantity/273)*100) as percentage_of_total_quantity
from groceries_sql group by location,item order by location""").show()
```

location	item	Total_quantity_per_group	percentage_of_total_quantity
Bellevue	Flowers	10	3.6630036630036633
Bellevue	Bread	125	45.78754578754578
Issaquah	Meat	14	5.128205128205129
Issaquah	Onion	12	4.395604395604395
Issaquah	Tomato	6	2.197802197802198
Kent	Apples	20	7.326007326007327
Redmond	Bread	5	1.8315018315018317
Redmond	Cheese	15	5.4945054945054945
Redmond	Meat	40	14.652014652014653
Renton	Bread	5	1.8315018315018317
Sammanish	Bread	5	1.8315018315018317
Seattle	Bananas	7	2.564102564102564
Seattle	Potatoes	9	3.296703296703297



## Window Functions - Running Total, Rank, Dense Rank, Row Num, Lead, Lag

```
[9]: window_open_1 = Window.partitionBy("location", "item") \
    .orderBy("order_date") \
    .rowsBetween(Window.unboundedPreceding, Window.currentRow)
```

```
[10]: RT_df = groceries_df.withColumn("Running_Total", sum("quantity").over(window_open_1))
```

```
[11]: RT_df.show()
```

order_id	location	item	order_date	quantity	Running_Total
2	Kent	Apples	2017-01-02	20	20
13	Sammamish	Bread	2017-01-07	5	5
15	Issaquah	Meat	2017-01-08	3	3
16	Issaquah	Meat	2017-01-09	5	8
17	Issaquah	Meat	2017-01-10	6	14
11	Renton	Bread	2017-01-05	5	5
1	Seattle	Bananas	2017-01-01	7	7
5	Seattle	Potatoes	2017-01-04	9	9
9	Redmond	Cheese	2017-01-05	15	15
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	8
12	Issaquah	Onion	2017-01-07	4	12
7	Redmond	Bread	2017-01-05	5	5
4	Redmond	Meat	2017-01-03	40	40
3	Bellevue	Flowers	2017-01-02	10	10
14	Issaquah	Tomato	2017-01-07	6	6
6	Bellevue	Bread	2017-01-04	5	5
18	Bellevue	Bread	2017-01-11	7	12
19	Bellevue	Bread	2017-01-12	54	66
20	Bellevue	Bread	2017-01-13	34	100

only showing top 20 rows

```
[12]: window_open_2 = Window.partitionBy("location") \
    .orderBy("order_date") \
```

```
[13]: Rank_df = groceries_df.withColumn("Rank", rank().over(window_open_2))
```

```
[14]: Rank_df.show()
```

order_id	location	item	order_date	quantity	Rank
8	Issaquah	Onion	2017-01-05	4	1
10	Issaquah	Onion	2017-01-06	4	2
12	Issaquah	Onion	2017-01-07	4	3
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
17	Issaquah	Meat	2017-01-10	6	7
13	Sammamish	Bread	2017-01-07	5	1
4	Redmond	Meat	2017-01-03	40	1
7	Redmond	Bread	2017-01-05	5	2
9	Redmond	Cheese	2017-01-05	15	2
1	Seattle	Bananas	2017-01-01	7	1
5	Seattle	Potatoes	2017-01-04	9	2
2	Kent	Apples	2017-01-02	20	1
3	Bellevue	Flowers	2017-01-02	10	1
6	Bellevue	Bread	2017-01-04	5	2
18	Bellevue	Bread	2017-01-11	7	3
19	Bellevue	Bread	2017-01-12	54	4
20	Bellevue	Bread	2017-01-13	34	5
21	Bellevue	Bread	2017-01-14	25	6

only showing top 20 rows

```
[15]: Dense_Rank_df = groceries_df.withColumn("Dense_Rank", dense_rank().over(window_open_2))
```

```
[16]: Dense_Rank_df.show()
```

order_id	location	item	order_date	quantity	Dense_Rank
8	Issaquah	Onion	2017-01-05	4	1
10	Issaquah	Onion	2017-01-06	4	2
12	Issaquah	Onion	2017-01-07	4	3
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	4
16	Issaquah	Meat	2017-01-09	5	5
17	Issaquah	Meat	2017-01-10	6	6
13	Sammamish	Bread	2017-01-07	5	1
4	Redmond	Meat	2017-01-03	40	1
7	Redmond	Bread	2017-01-05	5	2
9	Redmond	Cheese	2017-01-05	15	2
1	Seattle	Bananas	2017-01-01	7	1
5	Seattle	Potatoes	2017-01-04	9	2
2	Kent	Apples	2017-01-02	20	1
3	Bellevue	Flowers	2017-01-02	10	1
6	Bellevue	Bread	2017-01-04	5	2
18	Bellevue	Bread	2017-01-11	7	3
19	Bellevue	Bread	2017-01-12	54	4
20	Bellevue	Bread	2017-01-13	34	5
21	Bellevue	Bread	2017-01-14	25	6

only showing top 20 rows

```
[17]: Row_Num_df = groceries_df.withColumn("Row_Num", row_number().over(window_open_2))
```

```
[18]: Row_Num_df.show()
```

order_id	location	item	order_date	quantity	Row_Num
8	Issaquah	Onion	2017-01-05	4	1
10	Issaquah	Onion	2017-01-06	4	2
12	Issaquah	Onion	2017-01-07	4	3
14	Issaquah	Tomato	2017-01-07	6	4
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
17	Issaquah	Meat	2017-01-10	6	7
13	Sammamish	Bread	2017-01-07	5	1
4	Redmond	Meat	2017-01-03	40	1
7	Redmond	Bread	2017-01-05	5	2
9	Redmond	Cheese	2017-01-05	15	3
1	Seattle	Bananas	2017-01-01	7	1
5	Seattle	Potatoes	2017-01-04	9	2
2	Kent	Apples	2017-01-02	20	1
3	Bellevue	Flowers	2017-01-02	10	1
6	Bellevue	Bread	2017-01-04	5	2
18	Bellevue	Bread	2017-01-11	7	3
19	Bellevue	Bread	2017-01-12	54	4
20	Bellevue	Bread	2017-01-13	34	5
21	Bellevue	Bread	2017-01-14	25	6

only showing top 20 rows

```
[19]: Lead_df = groceries_df.withColumn("Upcoming_quantity", lead("quantity").over(window_open_2))
```

```
[20]: Lead_df.show()
```

order_id	location	item	order_Date	quantity	Upcoming_quantity
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	6
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
17	Issaquah	Meat	2017-01-10	6	null
13	Sammamish	Bread	2017-01-07	5	null
4	Redmond	Meat	2017-01-03	40	5
7	Redmond	Bread	2017-01-05	5	15
9	Redmond	Cheese	2017-01-05	15	null
1	Seattle	Bananas	2017-01-01	7	9
5	Seattle	Potatoes	2017-01-04	9	null
2	Kent	Apples	2017-01-02	20	null
3	Bellevue	Flowers	2017-01-02	10	5
6	Bellevue	Bread	2017-01-04	5	7
18	Bellevue	Bread	2017-01-11	7	54
19	Bellevue	Bread	2017-01-12	54	34
20	Bellevue	Bread	2017-01-13	34	25
21	Bellevue	Bread	2017-01-14	25	null

only showing top 20 rows

```
[21]: Lag_df = groceries_df.withColumn("Previous_quantity", lag("quantity").over(window_open_2))
```

```
[22]: Lag_df.show()
```

order_id	location	item	order_Date	quantity	Previous_quantity
8	Issaquah	Onion	2017-01-05	4	null
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	4
14	Issaquah	Tomato	2017-01-07	6	4
15	Issaquah	Meat	2017-01-08	3	6
16	Issaquah	Meat	2017-01-09	5	3
17	Issaquah	Meat	2017-01-10	6	5
13	Sammamish	Bread	2017-01-07	5	null
4	Redmond	Meat	2017-01-03	40	null
7	Redmond	Bread	2017-01-05	5	40
9	Redmond	Cheese	2017-01-05	15	5
1	Seattle	Bananas	2017-01-01	7	null
5	Seattle	Potatoes	2017-01-04	9	7
2	Kent	Apples	2017-01-02	20	null
3	Bellevue	Flowers	2017-01-02	10	null
6	Bellevue	Bread	2017-01-04	5	10
18	Bellevue	Bread	2017-01-11	7	5
19	Bellevue	Bread	2017-01-12	54	7

## Pivot

```
[51]: spark.sql("select item, location from groceries_sql").groupBy("item").pivot("location").count().show()
```

item	Bellevue	Issaquah	Kent	Redmond	Renton	Sammamish	Seattle
Potatoes	null	null	null	null	null	null	1
Cheese	null	null	null	1	null	null	null
Meat	null	3	null	1	null	null	null
Apples	null	null	1	null	null	null	null
Onion	null	3	null	null	null	null	null
Bread	5	null	null	1	1	1	null
Flowers	1	null	null	null	null	null	null
Bananas	null	null	null	null	null	null	1
Tomato	null	1	null	null	null	null	null

```
[ ]:
```



## QUESTION 4

### Ways to handle Null Values :-

1. **Filtering** – With combination of filter() and isNotNull() function we can filter out the rows containing Null values.

```
[33]: Lead_df = groceries_df.withColumn("Upcoming_quantity", lead("quantity").over(window_open_2))
```

```
[34]: Lead_df.show()
```

order_id	location	item	order_Date	quantity	Upcoming_quantity
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	6
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
17	Issaquah	Meat	2017-01-10	6	null
13	Sammanish	Bread	2017-01-07	5	null
4	Redmond	Meat	2017-01-03	40	5
7	Redmond	Bread	2017-01-05	5	15
9	Redmond	Cheese	2017-01-05	15	null
1	Seattle	Bananas	2017-01-01	7	9
5	Seattle	Potatoes	2017-01-04	9	null
2	Kent	Apples	2017-01-02	20	null
3	Bellevue	Flowers	2017-01-02	10	5
6	Bellevue	Bread	2017-01-04	5	7
18	Bellevue	Bread	2017-01-11	7	54
19	Bellevue	Bread	2017-01-12	54	34
20	Bellevue	Bread	2017-01-13	34	25
21	Bellevue	Bread	2017-01-14	25	null

only showing top 20 rows

```
[45]: filter_df = Lead_df.filter(Lead_df['Upcoming_quantity'].isNotNull())
```

```
[46]: filter_df.show()
```

order_id	location	item	order_Date	quantity	Upcoming_quantity
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	6
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
4	Redmond	Meat	2017-01-03	40	5
7	Redmond	Bread	2017-01-05	5	15
1	Seattle	Bananas	2017-01-01	7	9
3	Bellevue	Flowers	2017-01-02	10	5
6	Bellevue	Bread	2017-01-04	5	7
18	Bellevue	Bread	2017-01-11	7	54
19	Bellevue	Bread	2017-01-12	54	34
20	Bellevue	Bread	2017-01-13	34	25

2. **Dropping Null values** – The function drop() is used to remove rows containing Null values from your dataframes. This will create a new DataFrame without Null values.

**DF\_without\_Nulls = df.dropna()**

3. **Filling Null values** - fillna() is used to fill a value in place of null values present in particular column of a dataframe. We can use this to fill value in single column as well as all the columns containing nulls.

**Single Column –** filled\_df = df.fillna(replacement\_value, subset = ['column name'])

**All column –** df.fillna(replacement\_value)

```
[33]: Lead_df = groceries_df.withColumn("Upcoming_quantity", lead("quantity").over(window_open_2))
[34]: Lead_df.show()
```

order_id	location	item	order_date	quantity	Upcoming_quantity
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	6
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
17	Issaquah	Meat	2017-01-10	6	null
13	Sammanish	Bread	2017-01-07	5	null
4	Redmond	Meat	2017-01-03	40	5
7	Redmond	Bread	2017-01-05	5	15
9	Redmond	Cheese	2017-01-05	15	null
1	Seattle	Bananas	2017-01-01	7	9
5	Seattle	Potatoes	2017-01-04	9	null
2	Kent	Apples	2017-01-02	20	null
3	Bellevue	Flowers	2017-01-02	10	5
6	Bellevue	Bread	2017-01-04	5	7
18	Bellevue	Bread	2017-01-11	7	54
19	Bellevue	Bread	2017-01-12	54	34
20	Bellevue	Bread	2017-01-13	34	25
21	Bellevue	Bread	2017-01-14	25	null

only showing top 20 rows

```
[42]: filled_df = Lead_df.fillna(0, subset = ["Upcoming_quantity"])
[43]: filled_df.show()
```

order_id	location	item	order_date	quantity	Upcoming_quantity
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	6
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
17	Issaquah	Meat	2017-01-10	6	0
13	Sammanish	Bread	2017-01-07	5	0
4	Redmond	Meat	2017-01-03	40	5
7	Redmond	Bread	2017-01-05	5	15
9	Redmond	Cheese	2017-01-05	15	0
1	Seattle	Bananas	2017-01-01	7	9
5	Seattle	Potatoes	2017-01-04	9	0
2	Kent	Apples	2017-01-02	20	0
3	Bellevue	Flowers	2017-01-02	10	5
6	Bellevue	Bread	2017-01-04	5	7
18	Bellevue	Bread	2017-01-11	7	54
19	Bellevue	Bread	2017-01-12	54	34
20	Bellevue	Bread	2017-01-13	34	25
21	Bellevue	Bread	2017-01-14	25	0

only showing top 20 rows

```
[36]: pivot_df = spark.sql("select item, location from groceries_sql").groupBy("item").pivot("location").count()
[37]: pivot_df.show()
```

item	Bellevue	Issaquah	Kent	Redmond	Renton	Sammanish	Seattle
Potatoes	null	null	null	null	null	null	1
Cheese	null	null	null	1	null	null	null
Meat	null	3	null	1	null	null	null
Apples	null	null	1	null	null	null	null
Onion	null	3	null	null	null	null	null
Bread	5	null	null	1	1	1	null
Flowers	1	null	null	null	null	null	null
Bananas	null	null	null	null	null	null	1
Tomato	null	1	null	null	null	null	null

```
[40]: filled_df = pivot_df.fillna(0)
[41]: filled_df.show()
```

item	Bellevue	Issaquah	Kent	Redmond	Renton	Sammanish	Seattle
Potatoes	0	0	0	0	0	0	1
Cheese	0	0	0	1	0	0	0
Meat	0	3	0	1	0	0	0
Apples	0	0	1	0	0	0	0
Onion	0	3	0	0	0	0	0
Bread	5	0	0	1	1	1	0
Flowers	1	0	0	0	0	0	0
Bananas	0	0	0	0	0	0	1
Tomato	0	1	0	0	0	0	0



4. **Nulls in Mathematical operations** – Use the na property to handle nulls in mathematical operations. For example - null values can be replaced by a value when performing arithmetic operation.

**Math\_df = df.na.fill(replacement\_value)**

5. **Spark.sql query filter** - We can write custom spark.sql query to filter out null values using where column\_name is not null.

```
[25]: Lead_df = groceries_df.withColumn("Upcoming_quantity", lead("quantity").over(window_open_2))
```

```
[26]: Lead_df.show()
```

order_id	location	item	order_Date	quantity	Upcoming_quantity
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	6
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
17	Issaquah	Meat	2017-01-10	6	null
13	Sammanish	Bread	2017-01-07	5	null
4	Redmond	Meat	2017-01-03	40	5
7	Redmond	Bread	2017-01-05	5	15
9	Redmond	Cheese	2017-01-05	15	null
1	Seattle	Bananas	2017-01-01	7	9
5	Seattle	Potatoes	2017-01-04	9	null
2	Kent	Apples	2017-01-02	20	null
3	Bellevue	Flowers	2017-01-02	10	5
6	Bellevue	Bread	2017-01-04	5	7
18	Bellevue	Bread	2017-01-11	7	54
19	Bellevue	Bread	2017-01-12	54	34
20	Bellevue	Bread	2017-01-13	34	25
21	Bellevue	Bread	2017-01-14	25	null

only showing top 20 rows

```
[46]: Lead_df.createOrReplaceTempView("Lead_sql")
```

```
[49]: spark.sql("select * from Lead_sql where Upcoming_quantity is not null").show()
```

order_id	location	item	order_Date	quantity	Upcoming_quantity
8	Issaquah	Onion	2017-01-05	4	4
10	Issaquah	Onion	2017-01-06	4	4
12	Issaquah	Onion	2017-01-07	4	6
14	Issaquah	Tomato	2017-01-07	6	3
15	Issaquah	Meat	2017-01-08	3	5
16	Issaquah	Meat	2017-01-09	5	6
4	Redmond	Meat	2017-01-03	40	5
7	Redmond	Bread	2017-01-05	5	15
1	Seattle	Bananas	2017-01-01	7	9
3	Bellevue	Flowers	2017-01-02	10	5
6	Bellevue	Bread	2017-01-04	5	7
18	Bellevue	Bread	2017-01-11	7	54
19	Bellevue	Bread	2017-01-12	54	34
20	Bellevue	Bread	2017-01-13	34	25