

## Practical No. 1

### Aim: Implementation of Data partitioning through Range and List partitioning

**Objectives:** Understand partitioning concept with Range and List partitioning method

#### Theory:

- What is Partitioning?

Partitioning enables tables and indexes to be subdivided into individual smaller pieces. Each piece of the database object is called a partition. A partition has its own name, and may optionally have its own storage characteristics

- Key features and benefits of partitioning

Partitioning allows tables, indexes, and index-organized tables to be subdivided into smaller pieces, enabling these database objects to be managed and accessed at a finer level of granularity.

- Types of Partitioning: -

- Range Partitioning: The data is distributed based on a range of values of the partitioning key.
- List Partitioning: The data distribution is defined by a discrete list of values of the partitioning key

#### Range Partitioning:

- Create table sales\_range with fields salesman\_id, salesman\_name, sales\_amount, sales\_date. Partition the table into 4 partitions using range partitioning.

```
SQL> CREATE TABLE sales_range
  2  (salesman_id NUMBER(5),
  3   salesman_name VARCHAR2(30),
  4   sales_amount NUMBER(10),
  5   sales_date DATE)
  6 PARTITION BY RANGE(sales_date)
  7  (
  8   PARTITION sales_jan2000 VALUES LESS THAN(TO_DATE('01/02/2000','DD/MM/YYYY')),
  9   PARTITION sales_feb2000 VALUES LESS THAN(TO_DATE('01/03/2000','DD/MM/YYYY')),
 10  PARTITION sales_mar2000 VALUES LESS THAN(TO_DATE('01/04/2000','DD/MM/YYYY')),
 11  PARTITION sales_apr2000 VALUES LESS THAN(TO_DATE('01/05/2000','DD/MM/YYYY'))
 12 )
 13 ;
```

Table created.

- Check to See If The Partitions Are Correctly Built

```
SQL> SELECT TABLE_NAME,PARTITION_NAME FROM USER_TAB_PARTITIONS WHERE
  2  TABLESPACE_NAME='USERS';
```

TABLE_NAME	PARTITION_NAME
SALES_RANGE	SALES_JAN2000
SALES_RANGE	SALES_FEB2000
SALES_RANGE	SALES_MAR2000
SALES_RANGE	SALES_APR2000

- Insert data in sales\_range table

```
SQL> insert into sales_range values(1,'Janaki Joshi',5000,TO_DATE('23/02/2000','DD/MM/YYYY'));
1 row created.

SQL> insert into sales_range values(1,'Anand Sharma',15000,TO_DATE('12/03/2000','DD/MM/YYYY'));
1 row created.

SQL> insert into sales_range values(2,'Sonali Shah',35000,TO_DATE('17/04/2000','DD/MM/YYYY'));
1 row created.

SQL> insert into sales_range values(3,'Swati Chaudhary',14000,TO_DATE('24/01/2000','DD/MM/YYYY'));
1 row created.

SQL> insert into sales_range values(4,'Saniya Sane',4000,TO_DATE('16/08/1998','DD/MM/YYYY'));
1 row created.
```

- Display data from sales\_range table

```
SQL> select * from sales_range;
```

SALESMAN_ID	SALESMAN_NAME	SALES_AMOUNT	SALES_DATE
3	Swati Chaudhary	14000	24-JAN-00
4	Saniya Sane	4000	16-AUG-98
1	Janaki Joshi	5000	23-FEB-00
1	Anand Sharma	15000	12-MAR-00
2	Sonali Shah	35000	17-APR-00

- Display data from sales\_jan2000 partition of sales\_range table

```
SQL> select * from sales_range partition(sales_jan2000);
```

SALESMAN_ID	SALESMAN_NAME	SALES_AMOUNT	SALES_DATE
3	Swati Chaudhary	14000	24-JAN-00
4	Saniya Sane	4000	16-AUG-98

#### List Partitioning:

- Create table sales\_list with fields saleman\_id, salesman\_name, sales\_state, sales\_amount, sales\_date. Apply list partitioning on the table.

```
SQL> CREATE TABLE sales_list
  2  (salesman_id NUMBER(5),
  3   salesman_name VARCHAR2(30),
  4   sales_state VARCHAR2(20),
  5   sales_amount NUMBER(10),
  6   sales_date DATE)
  7 PARTITION BY LIST(sales_state)
  8 (
  9   PARTITION sales_west VALUES('California', 'Hawaii'),
 10  PARTITION sales_east VALUES ('New York', 'Virginia', 'Florida'),
 11  PARTITION sales_central VALUES('Texas', 'Illinois'),
 12  PARTITION sales_other VALUES(DEFAULT)
 13 )
 14 enable row movement
 15 ;
```

Table created.

(NOTE: enable row movement: This clause either enables or disables the migration of a row to a new partition if its key is updated. The default is **DISABLE ROW MOVEMENT**. In existing row if we try to update partition col value which causes change of partition is only possible if this option is enable)

#### ➤ Check to See If the Partitions Are Correctly Built

```
SQL> SELECT TABLE_NAME, PARTITION_NAME FROM USER_TAB_PARTITIONS WHERE
  2  TABLESPACE_NAME='USERS' ;
```

TABLE_NAME	PARTITION_NAME
SALES_LIST	SALES_CENTRAL
SALES_LIST	SALES_OTHER

#### ➤ Insert data in sales\_list table

```
SQL> insert into sales_list values(1,'Jayesh Patel','New York',2300,TO_DATE('23/02/2020','DD/MM/YYYY')) ;
1 row created.

SQL> insert into sales_list values(2,'Yash Pal','California',12300,TO_DATE('17/10/2020','DD/MM/YYYY')) ;
1 row created.

SQL> insert into sales_list values(3,'Sumit Rao','Texas',60000,TO_DATE('05/12/2021','DD/MM/YYYY')) ;
1 row created.

SQL> insert into sales_list values(4,'Sayli Joshi','Florida',18000,TO_DATE('09/09/2022','DD/MM/YYYY')) ;
1 row created.

SQL> insert into sales_list values(5,'Mina Shetty','Banglore',24000,TO_DATE('19/07/2022','DD/MM/YYYY')) ;
1 row created.
```

#### ➤ Display data from sales\_list table

```
SQL> select * from sales_list;
```

SALESMAN_ID	SALESMAN_NAME	SALES_STATE	SALES_AMOUNT
2	Yash Pal	California	12300
1	Jayesh Patel	New York	2300
4	Sayli Joshi	Florida	18000
3	Sumit Rao	Texas	60000
5	Mina Shetty	Banglore	24000

#### ➤ Display data from sales\_east partition of sales\_list table

```
SQL> select * from sales_list partition(sales_east);
```

SALESMAN_ID	SALESMAN_NAME	SALES_STATE	SALES_AMOUNT
1	Jayesh Patel	New York	2300
4	Sayli Joshi	Florida	18000

#### ➤ Display data from sales\_west partition of sales\_list table

```
SQL> select * from sales_list partition(sales_west);
```

SALESMAN_ID	SALESMAN_NAME	SALES_STATE	SALES_AMOUNT
2	Yash Pal	California	12300

## Practical No. 2

**Aim:** Implementation of Analytical queries like ROLL\_UP, CUBE, RANK, DENSE\_RANK, LEAD, LAG, FIRST and LAST.

**Objective:** Understand different types of Analytical functions

### Theory:

**Analytical functions:-** Analytic functions compute an aggregate value based on a group of rows. They differ from aggregate functions in that they return multiple rows for each group.

### Syntax:-

```
analytic_function([arguments])OVER(analytic_clause)
)
```

The analytic\_clause breaks down into the following optional elements.

[query\_partition\_clause][order\_by\_clause]

**ROLL UP:-** The ROLLUP is an extension of the GROUP BY clause. The ROLLUP calculates multiple levels of subtotals across a group of columns (or dimensions) along with the grand total. the ROLLUP extension produces group subtotals from right to left and a grand total. If "n" is the number of columns listed in the ROLLUP, there will be  $n+1$  levels of subtotals.

**CUBE:-** In addition to the subtotals generated by the ROLLUP extension, the CUBE extension will generate subtotals for all combinations of the dimensions specified. If "n" is the number of columns listed in the CUBE, there will be  $2^n$  subtotal combinations.

**ROW NUMBER:-** It is an analytical function and unlike NTILE this function assigns a unique sequential number to each row of the result set.

**RANK():-** The RANK() function is an analytic function that calculates the rank of a value in a set of values. The RANK() function returns the same rank for the rows with the same values.

**DENSE RANK():-** The DENSE\_RANK() function is an analytic function that calculates the rank of a value in a set of values.

The DENSE\_RANK() function returns the same rank for the rows with the same values. DENSE\_RANK() does not have any gap in rankings

**LEAD() and LAG():-** The LEAD function is used to access data from SUBSEQUENT rows along with data from the current row. The LAG function is used to access data from PREVIOUS rows along with data from the current row

**Syntax:-** LEAD(expression,[offset],[default]) over ([query\_partition\_clause] order\_by\_clause)

**FIRST and LAST :-** FIRST is an analytic function as the name suggests is used to provide the value of the first row in an ordered set of rows. LAST is also an analytical function which is used to get the value of the last row in an ordered set of rows.

### Implementation:-

- Create table employee with fields  
emp\_no,dep\_no,emp\_name,dob,salary,comm,job.

SQL> CREATE TABLE employee

```
2 (emp_no NUMBER(10),
3 dep_no NUMBER(10),
4 emp_name VARCHAR(25),
5 dob DATE,
6 salary NUMBER(10),
7 comm NUMBER(10),
8 job VARCHAR(25)
9 );
```

Table created.

```
SQL> CREATE TABLE Employee
2 (EMP_NO NUMBER(5),
3 DEP_NO NUMBER(5),
4 EMPLOYEE_NAME VARCHAR2(10),
5 BDATE DATE,
6 SALARY NUMBER(10),
7 COMM NUMBER(10),
8 JOB VARCHAR2(15));
```

Table created.

- Insert 10 records
- SQL> insert into employee  
VALUES(101,10,'TEJAS',TO\_DATE('12/01/82','DD/MM/YYYY'),22000,1000,'CLERK');
- 1 row created.

- Display all the records.

```
SQL> select * from employee;
```

	EMP_NO	DEP_NO	EMP_NAME	DOB	SALARY	COMM
JOB						
CLERK	101	10	TEJAS	12-JAN-82	22000	1000
CLERK	102	10	MANISH	12-FEB-83	52000	2000
CLERK	103	10	SACHIN	12-MAR-84	25000	4000
JOB						
manager	104	20	VANCHITA	12-APR-85	35000	5000
manager	105	20	SWARUPA	12-MAY-86	45000	6000
manager	106	20	ASHAWINI	12-JUN-87	25000	2000
JOB						
clerk	107	10	SUDIP	12-JUL-88	55000	6000
clerk	108	20	AKSHAY	12-AUG-89	35000	6000

8 rows selected.

### Roll up

- Display dep\_no,job,job count, sum of salary group them up using a roll up function in the order of dep\_no,job.

```
SQL> select dep_no,job,count(*),sum(salary)
```

- from employee

- group by rollup(dep\_no,job);

DEP_NO	JOB	COUNT(*)	SUM(SALARY)
10	CLERK	3	99000
10	manager	1	55000
10		4	154000
20	clerk	1	35000
20	manager	3	105000
20		4	140000
		8	294000

7 rows selected.

### Cube:

- Display dep\_no,job,job count, sum of salary group them up using a cube function in the order of dep\_no,job.

```
SQL> select dep_no,job,count(*),sum(salary)
```

- from employee

- group by cube(dep\_no,job);

DEP_NO	JOB	COUNT(*)	SUM(SALARY)
10	CLERK	8	294000
10	clerk	3	99000
10	manager	1	35000
10		4	160000
20	CLERK	4	154000
20	manager	3	99000
20		1	55000
20	clerk	4	140000
20	manager	1	35000
		3	105000

10 rows selected.

- Display emp\_no,emp\_name and salary from employee table and give numbers to each row.

```
SQL> select rownum,emp_no,emp_name,salary from employee;
```

ROWNUM	EMP_NO	EMP_NAME	SALARY
1	101	TEJAS	22000
2	102	MANISH	52000
3	103	SACHIN	25000
4	104	VANCHITA	35000
5	105	SWARUPA	45000
6	106	ASHAWINI	25000
7	107	SUDIP	55000
8	108	AKSHAY	35000

8 rows selected.

```
SQL> select rownum,emp_no,emp_name,salary from employee order by salary ;
```

ROWNUM	EMP_NO	EMP_NAME	SALARY
1	101	TEJAS	22000
6	106	ASHAWINI	25000
3	103	SACHIN	25000
8	108	AKSHAY	35000
4	104	VANCHITA	35000
5	105	SWARUPA	45000
2	102	MANISH	52000
7	107	SUDIP	55000

8 rows selected.

- Display emp\_no, emp\_name and descending order of salary from employee table and give numbers to each row.

```
SQL> select rownum,emp_no,emp_name,salary from employee order by salary desc;
```

ROWNUM	EMP_NO	EMP_NAME	SALARY
7	107	SUDIP	55000
2	102	MANISH	52000
5	105	SWARUPA	45000
8	108	AKSHAY	35000
4	104	VANCHITA	35000
6	106	ASHAWINI	25000
3	103	SACHIN	25000
1	101	TEJAS	22000

### Row number():

- Use Row\_number() analytical function to give numbering according to salary.

```
SQL>select row_number() over(order by salary),emp_no,emp_name,salary from employee order by salary desc;
```

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SQL> select row_number()over(order by salary),emp_no,emp_name,salary from employee order by salary desc;				
ROW_NUMBER()OVER(ORDERBYSALARY)	EMP_NO	EMP_NAME	SALARY	
8	107	SUDIP	55000	
7	102	MANISH	52000	
6	105	SWARUPA	45000	
4	108	AKSHAY	35000	
5	104	VANCHITA	35000	
2	106	ASHAWINI	25000	
3	103	SACHIN	25000	
1	101	TEJAS	22000	

### Rank():

- Display eno, ename and salary from employee table and rank them according to ascending order of salary.

SQL>select emp\_no,emp\_name,salary,rank() over(order by salary) from employee order by salary;

SQL> select emp_no,emp_name,salary,rank() over(order by salary) from employee order by salary;				
EMP_NO	EMP_NAME	SALARY	RANK()OVER(ORDERBYSALARY)	
101	TEJAS	22000	1	
106	ASHAWINI	25000	2	
103	SACHIN	25000	2	
108	AKSHAY	35000	4	
104	VANCHITA	35000	4	
105	SWARUPA	45000	6	
102	MANISH	52000	7	
107	SUDIP	55000	8	

8 rows selected.

### Dense rank() :

- Display eno, name, salary from employee table and rank them according to ascending order of salary using dense\_rank()

SQL> select emp_no,emp_name,salary,dense_rank() over(order by salary) from employee order by salary;				
EMP_NO	EMP_NAME	SALARY	DENSE_RANK()OVER(ORDERBYSALARY)	
101	TEJAS	22000	1	
106	ASHAWINI	25000	2	
103	SACHIN	25000	2	
108	AKSHAY	35000	3	
104	VANCHITA	35000	3	
105	SWARUPA	45000	4	
102	MANISH	52000	5	
107	SUDIP	55000	6	

8 rows selected.

- Display three highest salaried person.

SQL> select * from(select dense_rank() over(order by salary desc)top,emp_name,emp_no,salary from employee) where top <= 3;				
TOP	EMP_NAME	EMP_NO	SALARY	
1	SUDIP	107	55000	
2	MANISH	102	52000	
3	SWARUPA	105	45000	

- Partition by department\_number

SQL> select emp_no,dep_no,salary,comm,rank() over(partition by dep_no order by salary)as Rank from employee;				
EMP_NO	DEP_NO	SALARY	COMM	RANK
101	10	22000	1000	1
103	10	25000	4000	2
102	10	52000	2000	3
107	10	55000	6000	4
106	20	25000	2000	1
108	20	35000	6000	2
104	20	35000	5000	2
105	20	45000	6000	4

### Update employee salary

SQL> update Employee

2 set salary=33000

3 where emp\_no=101;

1 row updated.

SQL> update Employee

2 set salary=44000

3 where emp\_no=102;

1 row updated.

SQL> select * from employee;					
EMP_NO	DEP_NO	EMP_NAME	DOB	SALARY	COMM
101	10	TEJAS	12-JAN-82	33000	1000
102	10	MANISH	12-FEB-83	44000	2000
103	10	SACHIN	12-MAR-84	25000	4000
EMP_NO	DEP_NO	EMP_NAME	DOB	SALARY	COMM
104	20	VANCHITA	12-APR-85	35000	5000
105	20	SWARUPA	12-MAY-86	45000	6000
106	20	ASHAWINI	12-JUN-87	25000	2000
EMP_NO	DEP_NO	EMP_NAME	DOB	SALARY	COMM
107	10	SUDIP	12-JUL-88	55000	6000
108	20	AKSHAY	12-AUG-89	35000	6000

- Inserting null value in commission

SQL>insert into employee values(101,10,'TEJAS',TO\_DATE('12/01/82','DD/MM/YYYY'),22000,null,'CLERK');

- Replace null values of commission by 1000

SQL>select emp\_no,emp\_name,salary,nvl(comm,1000)new\_comm from employee order by comm desc;

SQL> Select emp\_no, emp\_name,salary,nvl(comm,1000) new\_comm from employee order by comm desc;

EMP_NO	EMP_NAME	SALARY	NEW_COMM
101	TEJAS	22000	1000
105	SWARUPA	45000	6000
108	AKSHAY	35000	6000
107	SUDIP	55000	6000
104	VANCHITA	35000	5000
103	SACHIN	25000	4000
106	ASHAWINI	25000	2000
102	MANISH	44000	2000
101	TEJAS	33000	1000

- Display details of employee and give ranking only for employee in dept\_no 10

SQL> select dense\_rank()over(partition by dep\_no order by salary)Rank,dep\_no,emp\_name,salary from employee where dep\_no=10;

EMP_NO	DEP_NO	SALARY	COMM	RANK
101	10	22000	1000	1
103	10	25000	4000	2
102	10	52000	2000	3
107	10	55000	6000	4
EMP_NO	DEP_NO	EMP_NAME	SALARY	
101	10	TEJAS	22000	
102	10	SACHIN	25000	
103	10	TEJAS	33000	
104	10	MANISH	44000	
105	10	SUDIP	55000	

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- Display name, job & salary and Rank the salary job wise

SQL> select dense\_rank()over(partition by job order by salary)Rank,job,emp\_name,salary from employee order by job;

RANK	JOB	EMP_NAME	SALARY
1	CLERK	TEJAS	22000
2	CLERK	SACHIN	25000
3	CLERK	TEJAS	33000
4	CLERK	MANISH	44000
1	clerk	AKSHAY	35000
1	manager	ASHWINI	25000
2	manager	VANCHITA	35000
3	manager	SWARUPA	45000
4	manager	SUDIP	55000

9 rows selected.

- Display information of employee & rank them for employee working as manager

SQL> select dense\_rank()over(partition by job order by salary)Rank,job,emp\_name,salary from employee where job='manager';

RANK	JOB	EMP_NAME	SALARY
1	manager	ASHWINI	25000
2	manager	VANCHITA	35000
3	manager	SWARUPA	45000
4	manager	SUDIP	55000

- Display first five records of employee in descending order of salary

SQL>select\*from(select emp\_no, emp\_name, salary, dense\_rank()over(order by salary desc)rank from employee) where rank<=5;

EMP_NO	EMP_NAME	SALARY	RANK
107	SUDIP	55000	1
105	SWARUPA	45000	2
102	MANISH	44000	3
104	VANCHITA	35000	4
108	AKSHAY	35000	4
101	TEJAS	33000	5

### LEAD()

- Display Employee details using Lead() analytical function.

SQL> select emp\_no,dob,lead(dob,1)over(order by dob)as "next" from employee;

EMP_NO	DOB	next
101	12-JAN-82	12-JAN-82
101	12-JAN-82	12-FEB-83
102	12-FEB-83	12-MAR-84
103	12-MAR-84	12-APR-85
104	12-APR-85	12-MAY-86
105	12-MAY-86	12-JUN-87
106	12-JUN-87	12-JUL-88
107	12-JUL-88	12-AUG-89
108	12-AUG-89	

### LAG():

- Display Employee details using Lag() analytical function.

SQL> select emp\_no,dob,lag(dob,1)over(order by dob)as "previous" from employee;

EMP_NO	DOB	previous
101	12-JAN-82	
101	12-JAN-82	12-JAN-82
102	12-FEB-83	12-JAN-82
103	12-MAR-84	12-FEB-83
104	12-APR-85	12-MAR-84
105	12-MAY-86	12-APR-85
106	12-JUN-87	12-MAY-86
107	12-JUL-88	12-JUN-87
108	12-AUG-89	12-JUL-88

### **First:**

SQL>select dep\_no, salary, max(salary) keep(DENSE\_RANK FIRST ORDER BY salary desc)over(PARTITION BY dep\_no)"max" from employee;

DEP_NO	SALARY	max
10	25000	55000
10	55000	55000
10	44000	55000
10	33000	55000
10	22000	55000
20	35000	45000
20	25000	45000
20	35000	45000
20	45000	45000

### **Last:**

SQL>select dep\_no,salary,min(salary)keep(DENSE\_RANK LAST ORDER BY salary desc)over(PARTITION BY dep\_no)"min" from employee;

DEP_NO	SALARY	MIN
10	25000	22000
10	55000	22000
10	44000	22000
10	33000	22000
10	22000	22000
20	35000	25000
20	25000	25000
20	35000	25000
20	45000	25000

9 rows selected.

### Practical No.3

**Aim:** Implementation of ORDBMS using ADT (Abstract Data Types), References, etc.

**Objectives:** Understand implementation of ORDBMS concept using ADT(Abstract Data Types) and References.

**Theory:** An object relational database management system (ORDBMS) is a database management system with that is similar to a relational database, except that it has an object-oriented database model. This system supports objects, classes and inheritance in database schemas and query language.

Abstract Data type (ADT) is a type (or class) for objects whose behavior is defined by a set of values and a set of operations.

#### Implementation:

➤ Create ADT for name and address  
create type type\_name13 As object(fname varchar(20), mname varchar(20), lname varchar(20));  
/

Type created

```
SQL> create type type_name13 As object
  2  (
  3  fname varchar(20),mname varchar(20),lname varchar(20)
  4  );
  5 /
```

Type created.

```
create type type_address13 As object(street
varchar(20), city varchar(20), pincode
number(10));
/
```

Type created

```
SQL> create type type_address13 As object
  2  (
  3  street varchar(20),
  4  city varchar(20),
  5  pincode number(10)
  6  );
  7 /
```

Type created.

➤ Create table customer with fields cid(number), cname(type\_name), cadd(type\_add), cphone(number)

Creating table :

```
create table customer013
```

```
(c_id number(5) primary key,c_name
type_name,
c_add type_address,c_phno number(10)
);
```

Table created.

```
SQL> create table customer013
  2  (
  3  c_id number(5) primary key,
  4  c_name type_name13,
  5  c_add type_address13,
  6  c_phno number(10)
  7  );
```

Table created.

Insert data:

```
insert into customer013
values(1,type_name('varsha','s','atul'),
type_address('Sainagar','Mumbai',400042),1
23456789);
```

```

SQL> insert into customer013
  2  values(2,type_name13('simran','p','gupta'),
  3  type_address13('Sainagar','Mumbai',480042),9434945924);

1 row created.

SQL> insert into customer013
  2  values(3,type_name13('sailee','s','kadam'),
  3  type_address13('virar','Mumbai',407042),1212346789);

1 row created.

SQL> insert into customer013
  2  values(4,type_name13('sumit','v','gawde'),
  3  type_address13('kandiwali','Mumbai',407042),1212346789);

1 row created.

SQL> insert into customer013
  2  values(5,type_name13('pratik','a','ahirao'),
  3  type_address13('borivali','Mumbai',407042),1212346789);

1 row created.

```

1 row created.

Selecting data from table:

`select * from customer013;`

```

SQL> select * from customer013;

C_ID
-----
C_NAME(FNAME, MNAME, LNAME)
C_ADD(STREET, CITY, PINCODE)
C_PHNO
-----
1
TYPE_NAME13('aadarsh', 'p', 'choudhari')
TYPE_ADDRESS13('virar', 'Mumbai', 400042)
123456789

C_ID
-----
C_NAME(FNAME, MNAME, LNAME)
C_ADD(STREET, CITY, PINCODE)
C_PHNO
-----
2
TYPE_NAME13('simran', 'p', 'gupta')
TYPE_ADDRESS13('Sainagar', 'Mumbai', 480042)
9434945924

C_ID
-----
C_NAME(FNAME, MNAME, LNAME)
C_ADD(STREET, CITY, PINCODE)
C_PHNO
-----
3
TYPE_NAME13('sailee', 's', 'kadam')
TYPE_ADDRESS13('virar', 'Mumbai', 407042)
1212346789

```

➤ Select street value of customer whose cid=1

`SQL> select c.c_add.street from customer1 c where c_id=1;`

```

SQL> select c.c_add.street from customer013 c where c_id=1;

C_ADD.STREET
-----
virar

```

➤ Select fname of customer whose cid=1  
`SQL> select c.c_name.fname from customer1 c where c_id=1;`

```

SQL> select c.c_name.fname from customer013 c where c_id=1;

C_NAME.FNAME
-----
aadarsh

```

➤ Select lname of customer whose cid=1

```

SQL> select c_id,c.c_name.lname from customer013 c;

C_ID C_NAME.LNAME
-----
1 choudhari
2 gupta
3 kadam
4 gawde
5 ahirao

```

➤ Select cname of customers.

`SQL> select c_name from customer1;`

```

SQL> select c_name from customer013;

C_NAME(FNAME, MNAME, LNAME)
-----
TYPE_NAME13('aadarsh', 'p', 'choudhari')
TYPE_NAME13('simran', 'p', 'gupta')
TYPE_NAME13('sailee', 's', 'kadam')
TYPE_NAME13('sumit', 'v', 'gawde')
TYPE_NAME13('pratik', 'a', 'ahirao')

```

➤ Select full name of customers.

`SQL> select c.c_name.fname||' '||c.c_name.mname||' '||c.c_name.lname from customer1 c;`

```

SQL> select c.c_name.fname||' '||c.c_name.mname||' '||c.c_name.lname
  2  from customer013 c;

C.C_NAME.FNAME||'||C.C_NAME.MNAME||'||C.C_NAME.LNAME
-----
aadarsh p choudhari
simran p gupta
sailee s kadam
sumit v gawde
pratik a ahirao

```

**REF and DREF function:**

## Creating Object tables

```
SQL> create or replace type  
ANIMAL_TY12 as object(Breed  
varchar2(25),  
Name varchar2(25),BirthDate DATE);  
/  
Type created.
```

```
SQL> create or replace type ANIMAL_TY12 as object (Breed varchar2(25),  
2 Name varchar2(25), BirthDate DATE);  
3 /  
Type created.
```

```
SQL> create table ANIMAL12 of  
ANIMAL_TY12;  
Table created.
```

```
SQL> create table ANIMAL12 of ANIMAL_TY12;  
Table created.
```

## Inserting Rows into Object Tables

```
SQL> insert into ANIMAL12 values(  
ANIMAL_TY12('MULE','FRANCES','01-  
APR-02'));  
insert into ANIMAL12 values(  
ANIMAL_TY12('DOG','BENJI','03-SEP-  
01'));  
1 row created.
```

```
SQL> insert into ANIMAL12 values( ANIMAL_TY12('MULE','FRANCES','01-APR-02'));  
1 row created.  
SQL> insert into ANIMAL12 values( ANIMAL_TY12('DOG','BENJI','03-SEP-01'));  
1 row created.
```

## The REF Function

```
SQL> select REF(A) from ANIMAL12 A;
```

```
SQL> select REF(A) from ANIMAL12 A;  
REF(A)  
-----  
0000280209E1F34FF0BDE64B8BA6317510F72BD625BF03A2BED4A04A6880D2F0B3D3874139010002  
4D0000  
0000280209D827A5504ABE4AFBBFB98B476DF07686BF03A2BED4A04A6880D2F0B3D3874139010002  
4D0001
```

## Using the DEREF Function

```
SQL> create table KEEPER12  
(KeeperName varchar2(25), AnimalKept  
REF ANIMAL_TY12);
```

Table created.

```
SQL> create table KEEPER12 (KeeperName varchar2(25), AnimalKept REF ANIMAL_TY12);  
Table created.
```

```
SQL> describe KEEPER12
```

Name	Null?	Type
KEEPERNAME		VARCHAR2(25)
ANIMALKEPT		REF OF ANIMAL_TY12

```
SQL> insert into KEEPER12 select  
'CATHERINE', REF(A) from ANIMAL12  
A where Name='BENJI';  
1 row created.
```

```
SQL> insert into KEEPER12 select 'CATHERINE', REF(A) from ANIMAL12 A where Name='BENJI';  
1 row created.
```

```
SQL> select * from KEEPER12;
```

```
SQL> select * from KEEPER12;  
KEEPERNAME  
-----  
ANIMALKEPT  
-----  
CATHERINE  
0000220208D827A5504ABE4AFBBFB98B476DF07686BF03A2BED4A04A6880D2F0B3D3874139
```

```
SQL> select KeeperName,  
DEREF(K.AnimalKept)from KEEPER12 K;
```

```
SQL> select KeeperName, DEREF(K.AnimalKept) from KEEPER12 K;  
KEEPERNAME  
-----  
DEREF(K.ANIMALKEPT)(BREED, NAME, BIRTHDATE)  
-----  
CATHERINE  
ANIMAL_TY12('DOG', 'BENJI', '03-SEP-01')
```

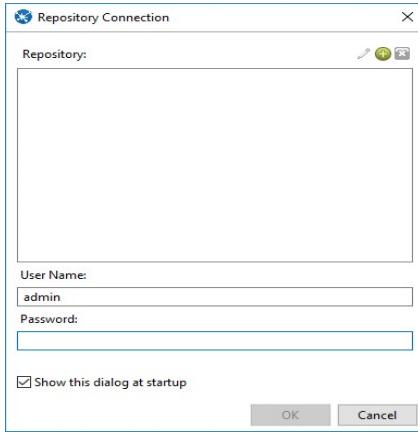
## ETL Implementation through Pentaho

**Aim :** Perform Transformation on source Table and store the data to Output table in SQL.  
Starting with Pentaho

**Step 1. Open Data Integration folder (C:\data-integration)**

**Step 2. Double click on spoon(Windows Batch file)**

**Step 3. Click on cancel.**

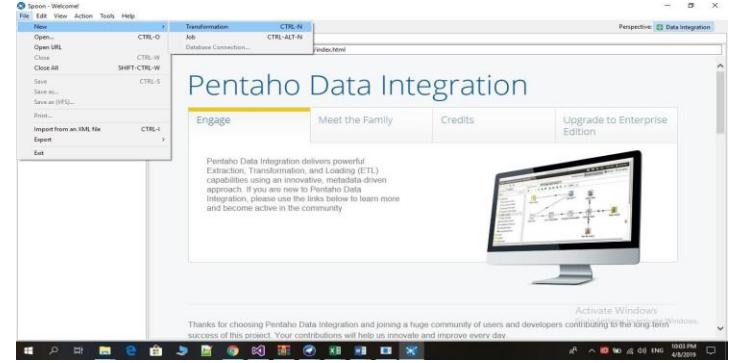


1. Transforming Source Table and storing to Output Table in SQL

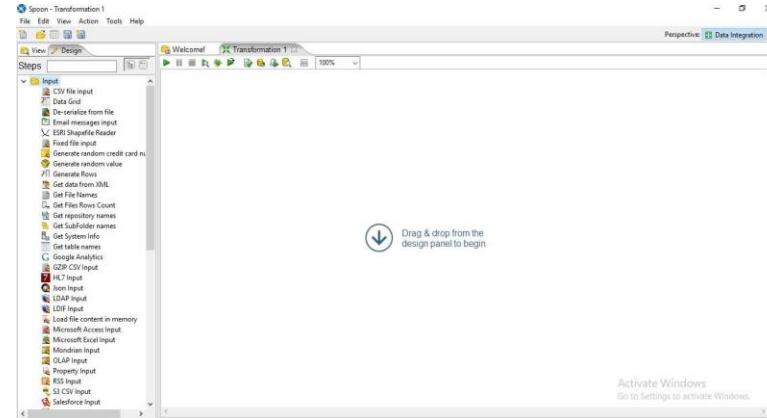
**Step1: Go to Pentaho**



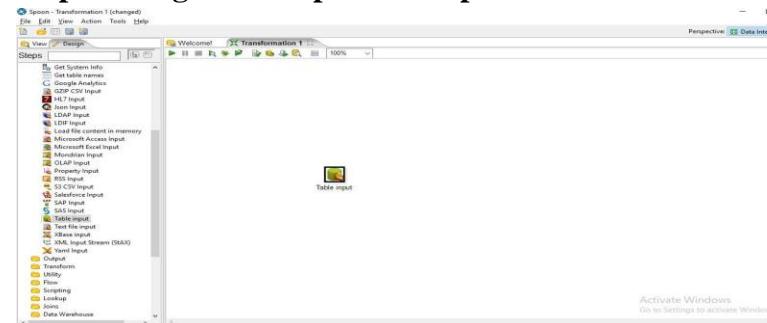
**Step2: Go to file->New->Transformation.**



**Step 3: Click on Input**

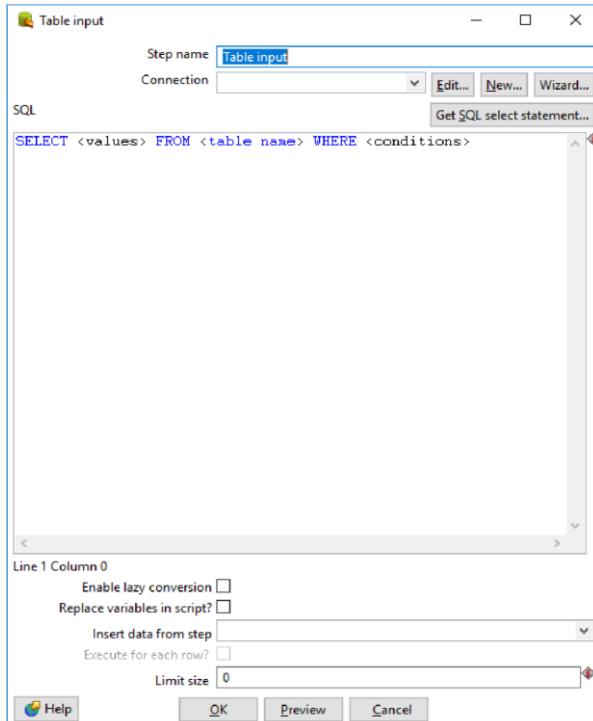


**Step 4: Drag Table Input on the panel**

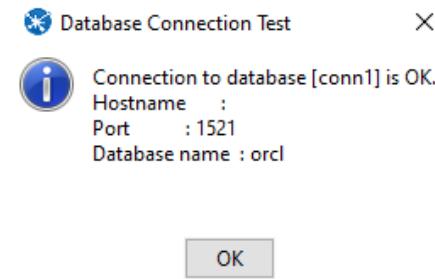


**Step 5: Double Click on Table Input and the following tab will appear.**

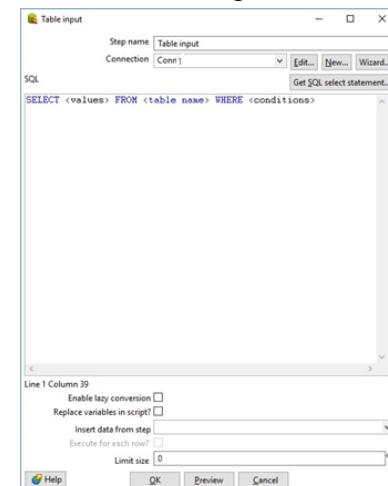
a. Click On New



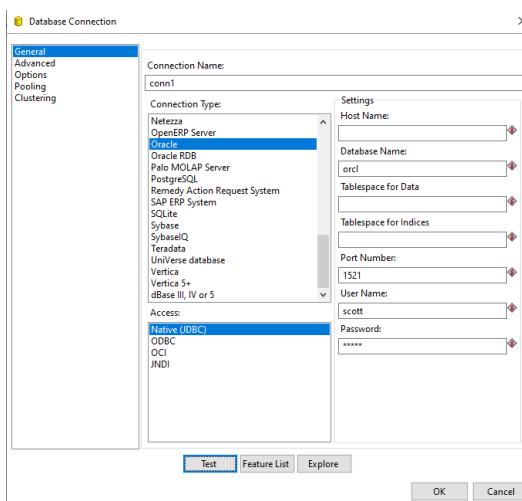
Connection Successful



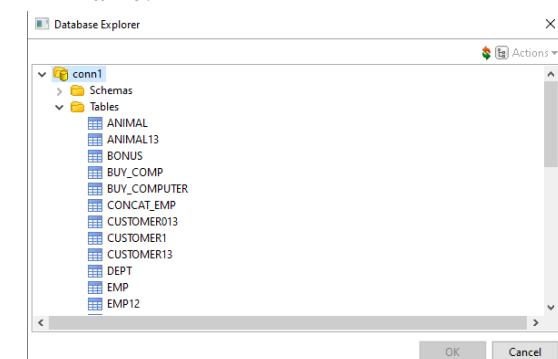
c. Click on get SQL select statement



- b. Select Oracle in Connection Type and enter Connection Name, Database Name, User Name and Password and click on Test.



d. Click on database -> table and select the table name.



e. Click on get Preview.

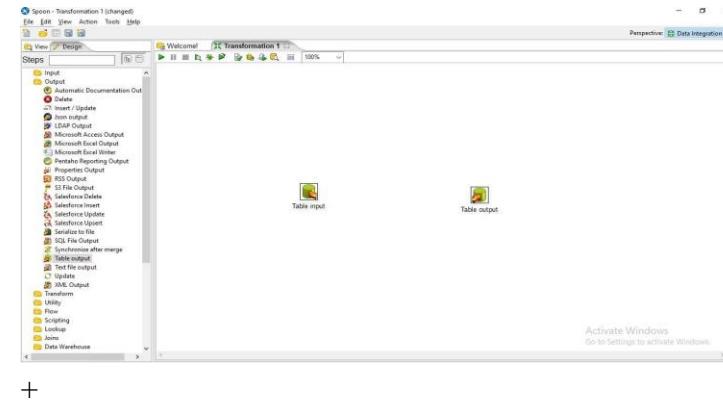


### Examine preview data

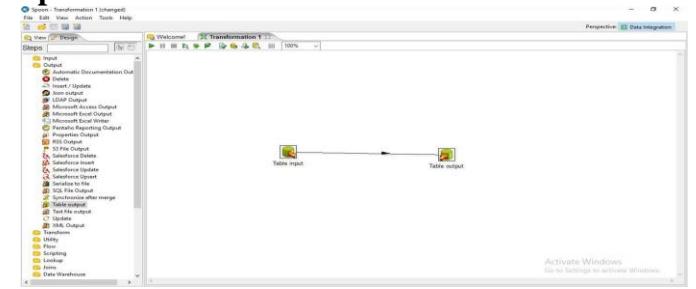
Rows of step: Table input (5 rows)

#	EMP_NO	FNAME	LNAME
1	1	Sonam	Singh
2	2	Pradnya	Suryavanshi
3	3	Komal	Malviya
4	4	Nidhi	Rai
5	5	Riya	Sharma

**Step 6: Click on Output and drag table Output.**

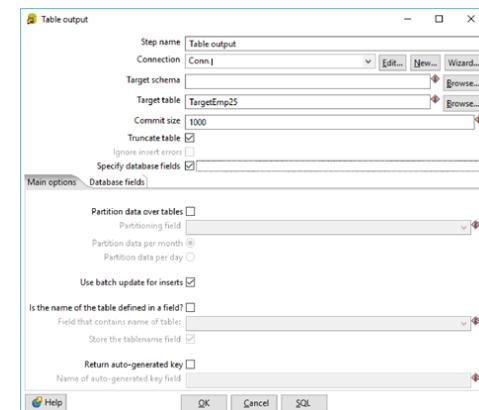


**Step 7: Hold the mouse Pointer on table input and select and drag the Output connector to the Table output.**

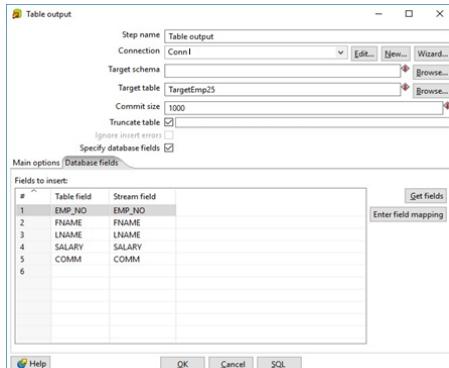


### **Step 8: Double Click on target table**

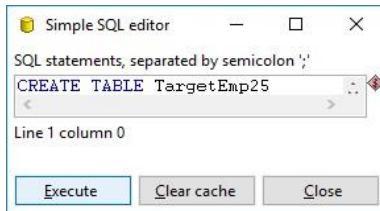
- Enter the Target table name and check the truncate table and Specify database fields Check Box.



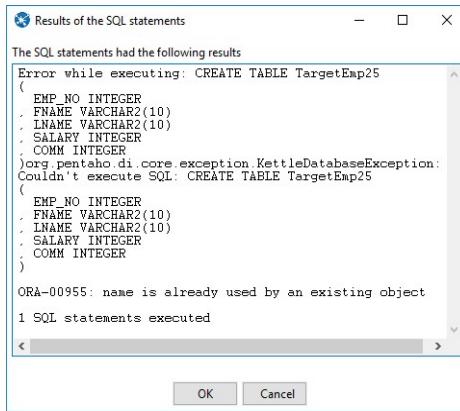
- Click On Database fields and click on Get fields.



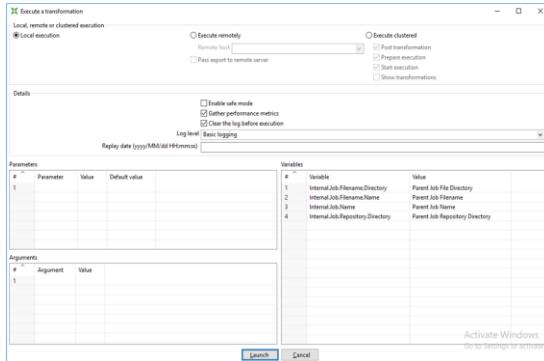
c. Click on SQL and click on Execute.



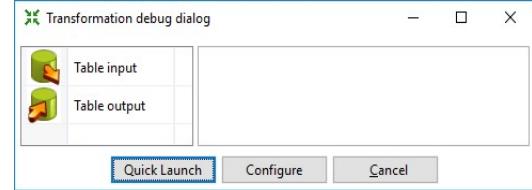
d. Click Ok



## Step 9: Click on Run Transformation and Click on Launch

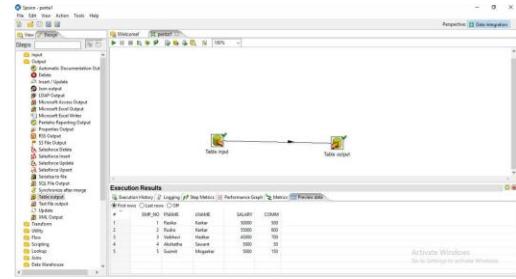


## Step 10: Click on Debug Transformation(Spider) and Click on Quick Launch



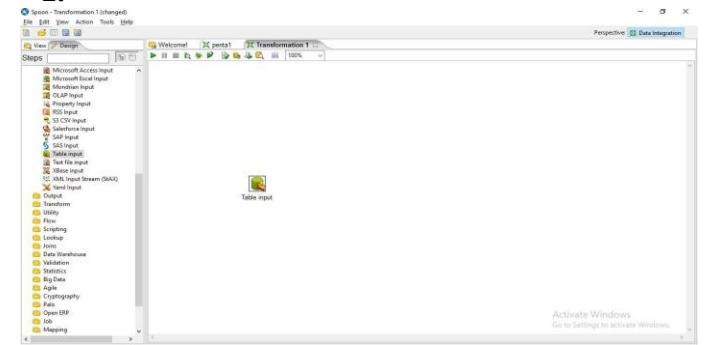
### Output:

The Green ticks on the table input and table output shows successful transformation.

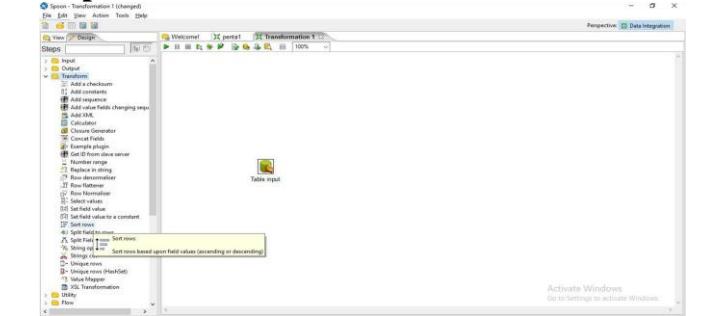


## 2. Sorting Operation and adding Sequence to Output Table.

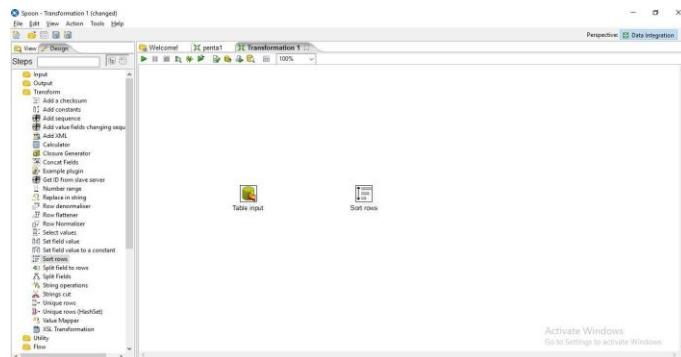
### Step 1: Perform the first 5 steps same as practical 1.



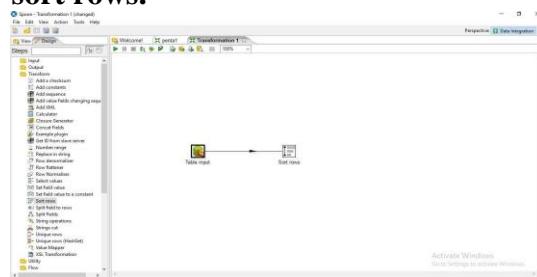
### Step 2: Click on Transform in the left



### Step 3: Drag sort rows on the Panel

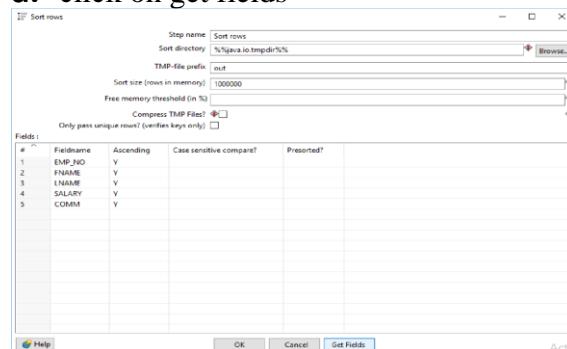


**Step 4:** Hold the mouse pointer on the Table Input and then drag the output connector to the sort rows.

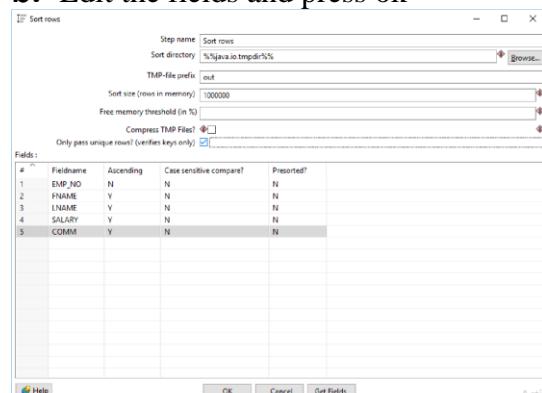


**Step 5: Double click on the sort rows**

- a. click on get fields



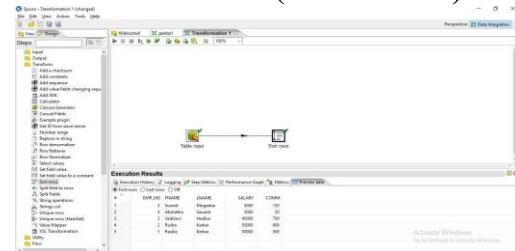
- b. Edit the fields and press ok



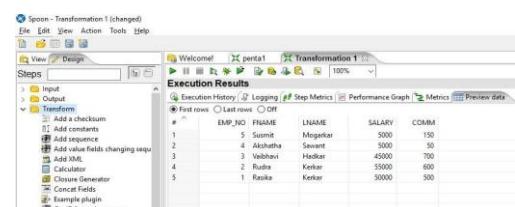
**Step 6: Click on Debug the Transformation and Click on Quick Launch.**



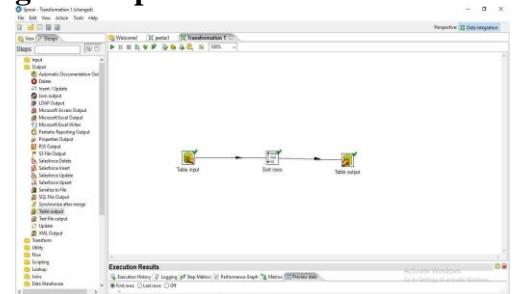
**Successful connections(Green Ticks)**



**Sorted Data**

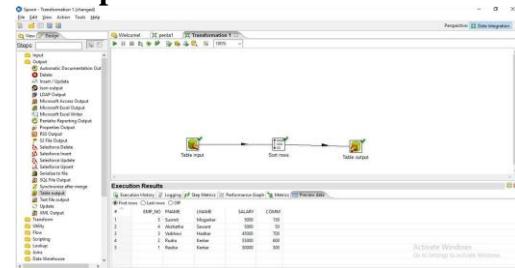


**Step 7: hold the cursor on the sort row and then drag the output connector to the table output.**



**Step 8: now perform step no 8 to step 10 from practical 1.**

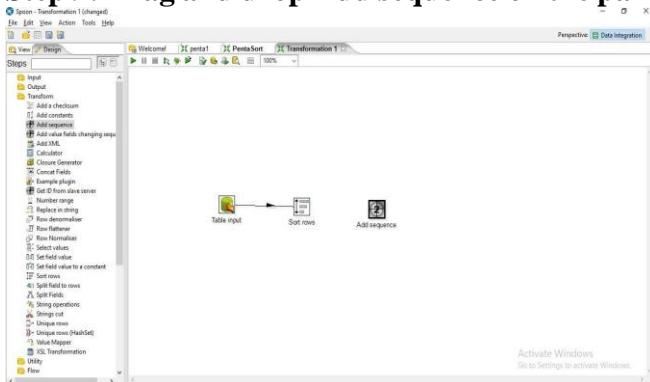
**Final Output:**



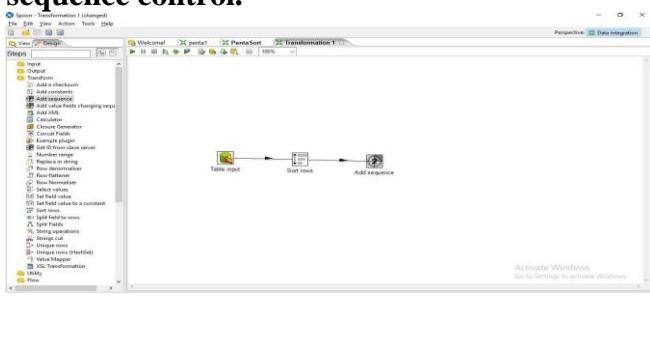
**SQL> select \* from emppp13;**

SQL> select * from emp_target_concat;				
EMP_NO	FNAME	LNAME	SALARY	COMM
-----				
1	Sonam	Singh	25000	500
2	Pradnya	Suryavanshi	20000	700
3	Komal	Malviya	17000	1000
4	Nidhi	Rai	30000	900
5	Riya	Sharma	15000	1100

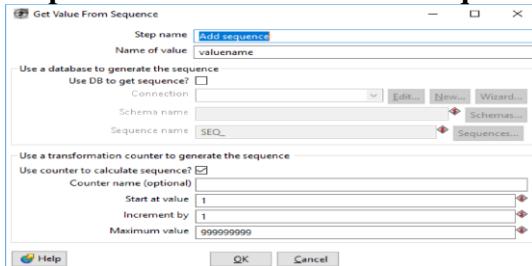
### Step 9: Drag and drop Add Sequence on the panel



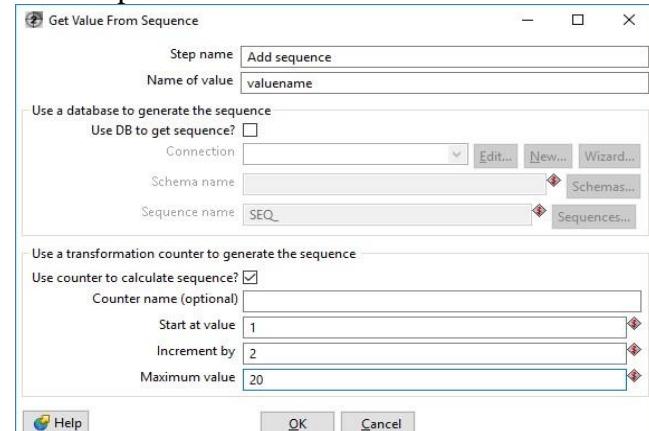
**Step 10: Hold the mouse pointer on the sort rows and drag the output connector to the Add sequence control.**



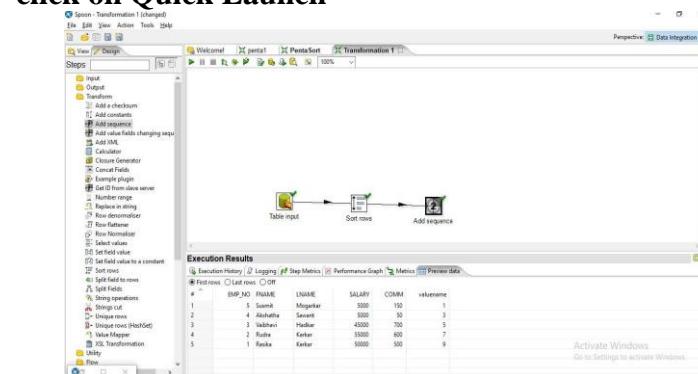
### Step 11: Double Click on Add Sequence Control



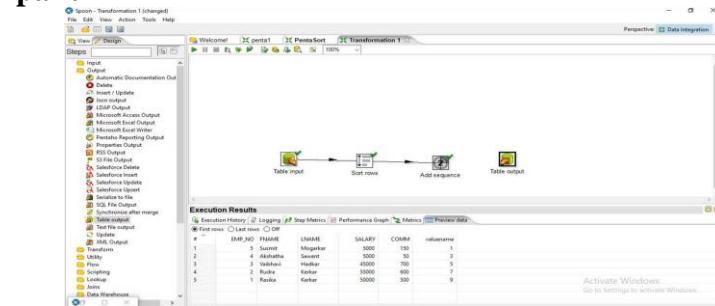
- Enter the values for Start at Values, Increment by and Maximum Values and press ok.



### Step 12: Click on Debug the Transformation and click on Quick Launch



**Step 13: Drag and drop the table output on the panel**



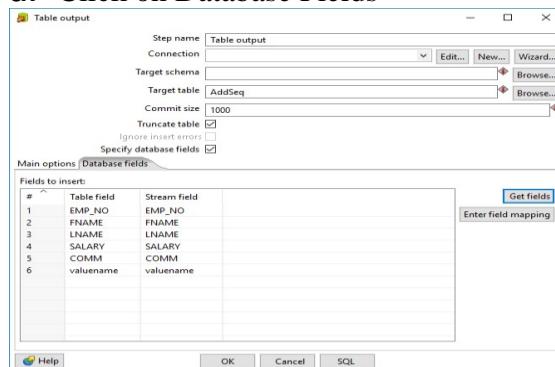
**Step 14: Hold the mouse pointer on the add sequence control and drag the output connector to the table output.**

**Execution Results**

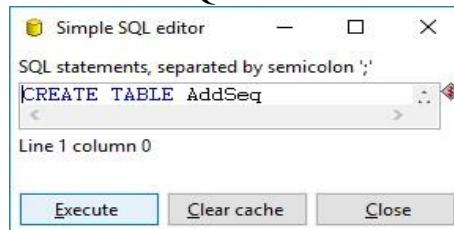
#	EMP_NO	FNAME	LNAME	SALARY	COMM	valuename
1	5	Sumeet	Mogarkar	5000	150	1
2	4	Akashtha	Savant	5000	30	3
3	3	Vishal	Hedkar	45000	700	5
4	2	Rutu	Kekre	55000	600	7
5	1	Raksha	Kekre	50000	500	9

### Step 15: Double Click on the table output.

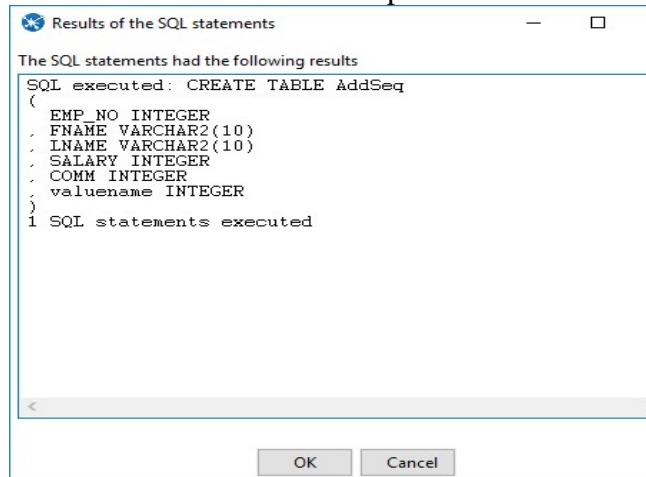
#### a. Click on Database Fields



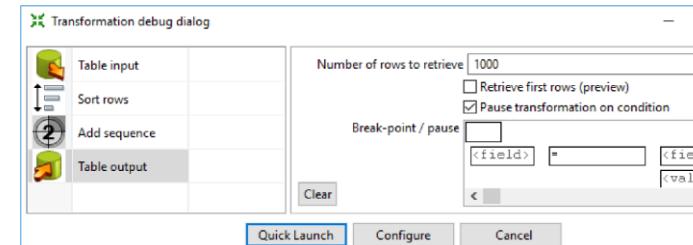
#### b. Click on SQL



#### c. Click on Execute an then press ok



### Step 16: Click on debug Transformation and then quick Launch



### Output



SQL> select \* from empp13;

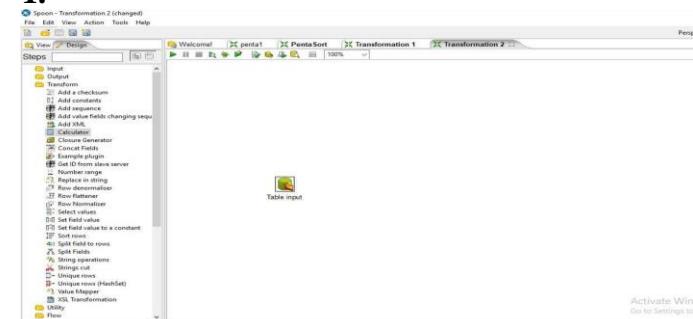
SQL> select \* from empp13;

ENO	FNAME	LNAME	SALARY	COMM	VALUENAME
5	smit	bole	160000	12000	1
4	prateek	ahirrao	150000	15000	3
2	aadarsh	choudhari	200000	74000	5
1	simran	gupta	100000	44000	7

### 3. Calculator Operation

#### Step 1: Perform first 5 steps same as practical

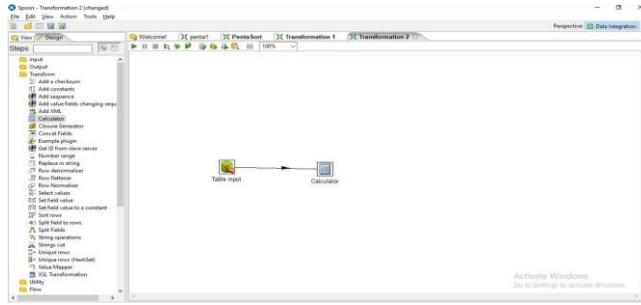
#### 1.



#### Step 2: Drag and Drop Calculator from transformation on the panel.

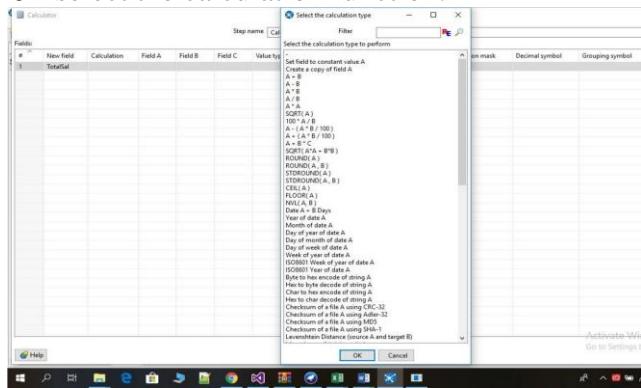


**Step 3: Hold the mouse pointer on the Table input and drag the output connector to the calculator control.**

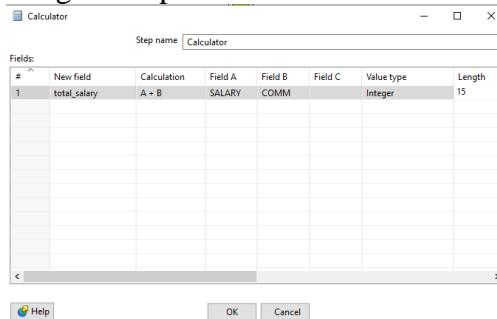


**Step 4: Double Click on the Calculator**

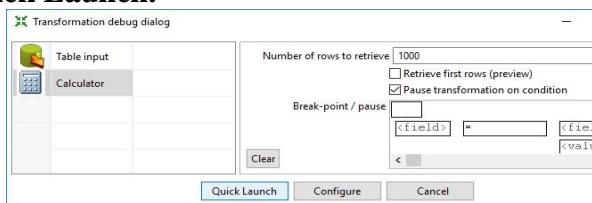
- specify the new field name
- select the calculation function.



- Enter Field A, Field B, Value type and Length and press ok.



**Step 5: Click on Debug transformation and click on Quick Launch.**



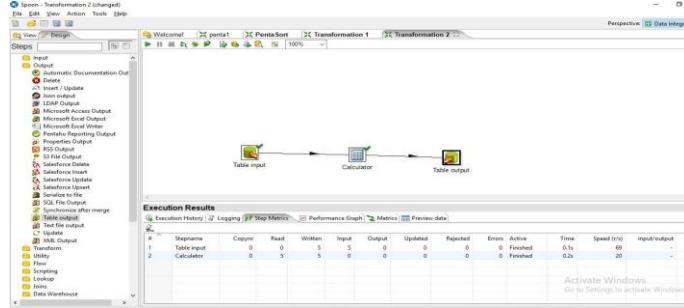
**Preview**

Examine preview data

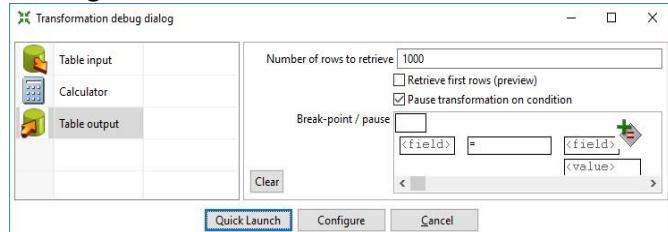
Rows of step: Calculator (4 rows)

#	ENO	FNAME	LNAME	SALARY	COMM	total_salary
1	5	smit	bole	160000	12000	00172000
2	4	prateek	ahirrao	150000	15000	00165000
3	2	aadarsh	choudhari	200000	74000	00274000
4	1	simran	gupta	100000	44000	00144000

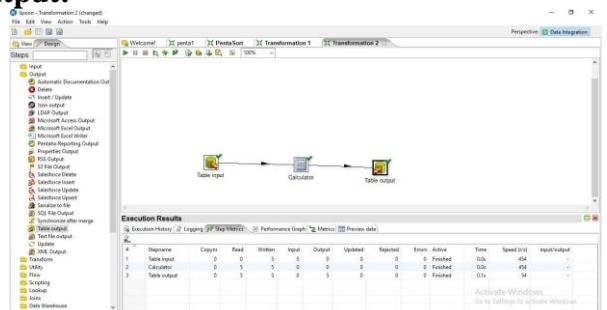
**Step 6: Now Perform the steps 6-10 same as practical 1.**



**Step 7: Click on Debug Transformation and then click on Quick Launch**



**Output:**



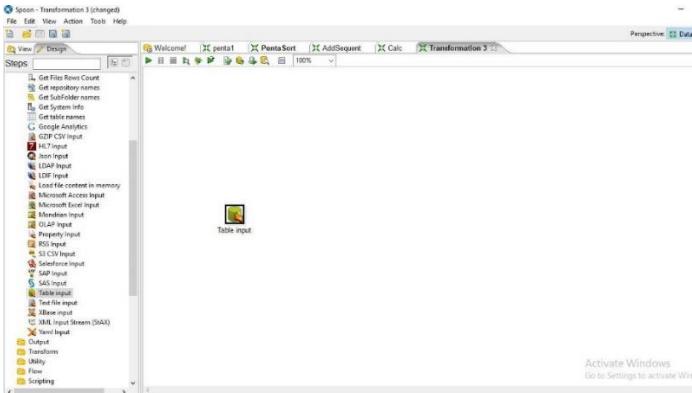
SQL> select \* from emppp13;

SQL> select \* from emppp13;

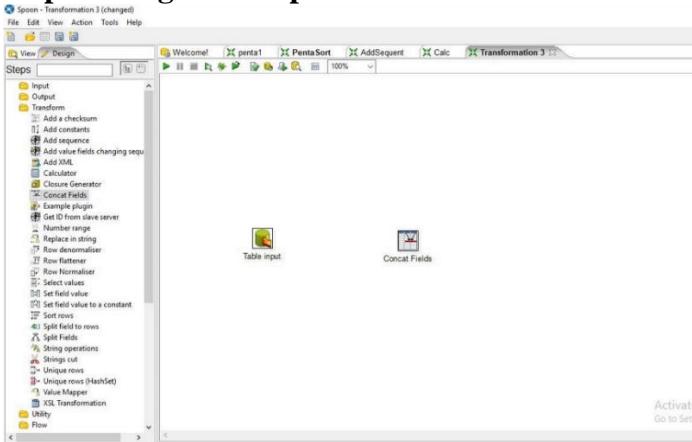
ENO	FNAME	LNAME	SALARY	COMM	TOTAL_SALARY
1	simran	gupta	100000	44000	144000
2	aadarsh	choudhari	200000	74000	274000
3	prateek	ahirrao	150000	15000	165000
4	smit	bole	160000	12000	172000

**d. Concatenation Operation**

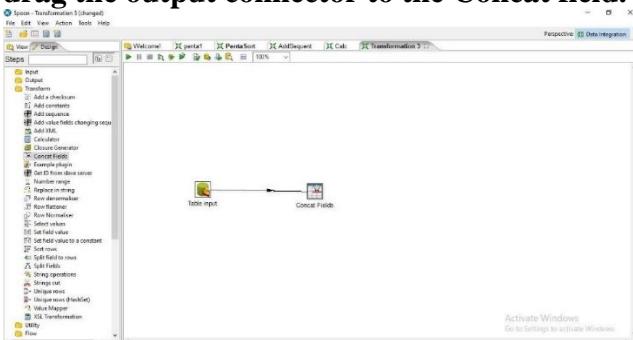
**Step 1: Perform first 6 steps same as practical 1.**



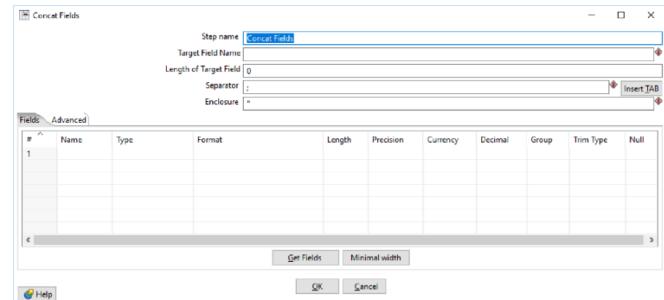
## Step 2: Drag and drop Concat Field



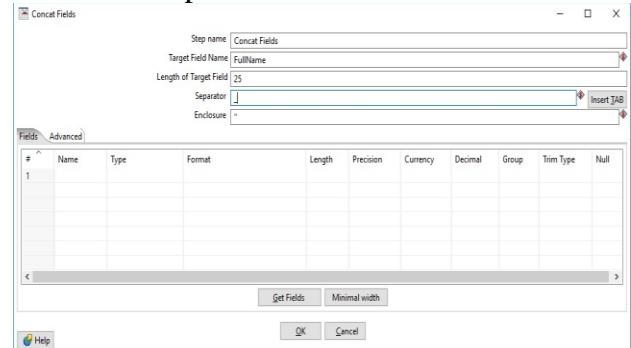
## Step 4: Hold mouse pointer on input table and drag the output connector to the Concat field.



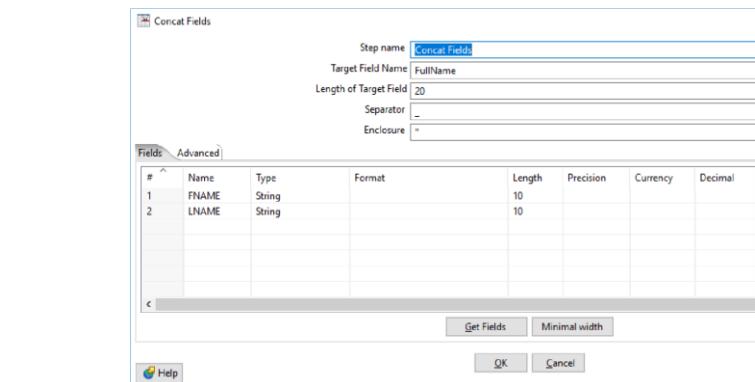
## Step 5: Double Click on Cncat Field



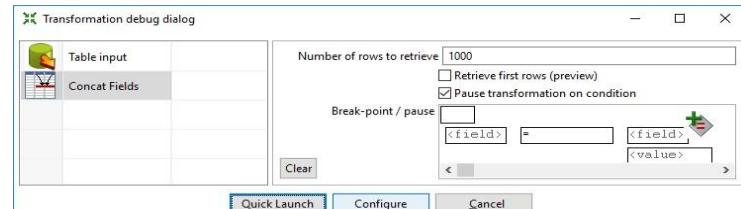
- a. Enter the values for Target Field Name, Length of the Target Field and the Seperator and press on Get Fields



- b. Select the columns that you want to concat



## Step 6: Click on Debug transformation and click on Quick Launch.



**Examine preview data**

Rows of step: Concat Fields (5 rows)

#	EMP_NO	FNAME	LNAME	SALARY	COMM	FullName
1	5	Susmit	Mogarkar	5000	150	Susmit _Mogarkar
2	4	Akshatha	Sawant	5000	50	Akshatha _Sawant
3	3	Vaibhavi	Hadkar	45000	700	Vaibhavi _Hadkar
4	2	Rudra	Kerkar	55000	600	Rudra _Kerkar
5	1	Rasika	Kerkar	50000	500	Rasika _Kerkar

**Close**

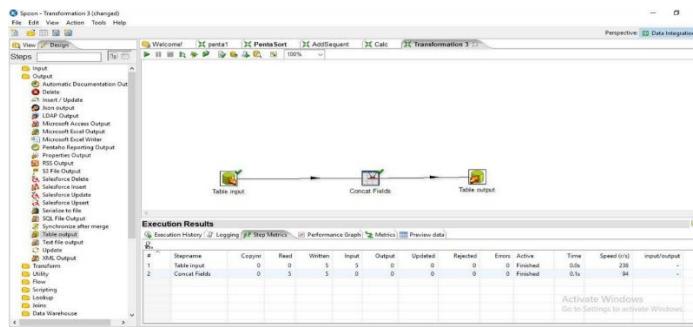
**Examine preview data**

Rows of step: Concat Fields (5 rows)

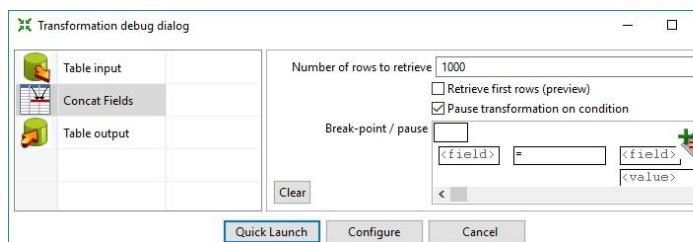
#	EMP_NO	FNAME	LNAME	SALARY	COMM	FullName
1	5	Susmit	Mogarkar	5000	150	Susmit _Mogarkar
2	4	Akshatha	Sawant	5000	50	Akshatha _Sawant
3	3	Vaibhavi	Hadkar	45000	700	Vaibhavi _Hadkar
4	2	Rudra	Kerkar	55000	600	Rudra _Kerkar
5	1	Rasika	Kerkar	50000	500	Rasika _Kerkar

**Close**

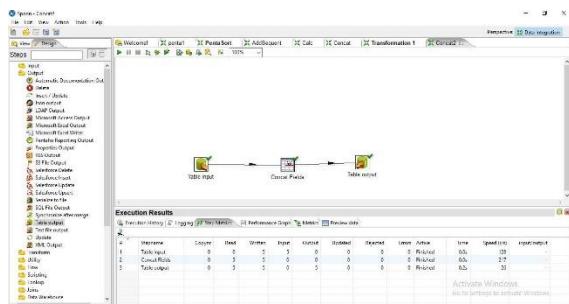
**Step 7:** Now Perform the steps 6-10 same as practical 1.



**Step 8:** Click on debug Transformation and click on Quick Launch



**Output:**

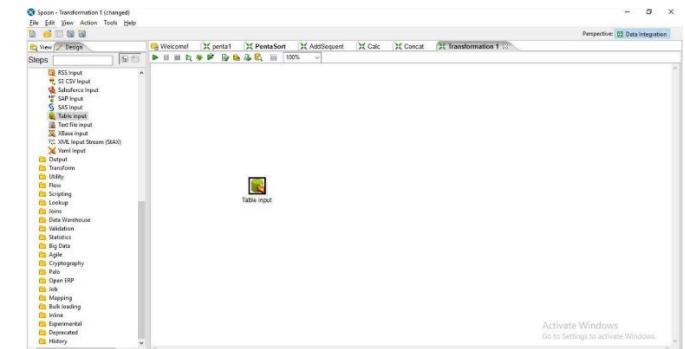


**SQL> select \* from Concat;**

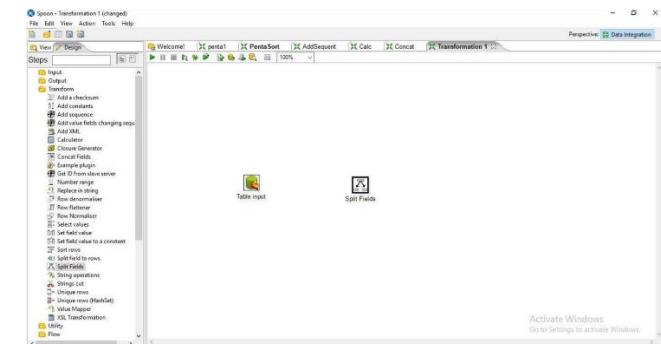
SQL> select \* from emp\_target\_concat;

EMP_NO	FNAME	LNAME
SALARY	COMM	FULLNAME
1 Sonam	500	Sonam _Singh
25000		Singh
2 Pradnya	700	Pradnya _Suryavanshi
20000		Suryavanshi
3 Komal	1000	Komal _Malviya
17000		Malviya
EMP_NO	FNAME	LNAME
SALARY	COMM	FULLNAME
4 Nidhi	900	Nidhi _Rai
30000		Rai
5 Riya	1100	Riya _Sharma
15000		Sharma

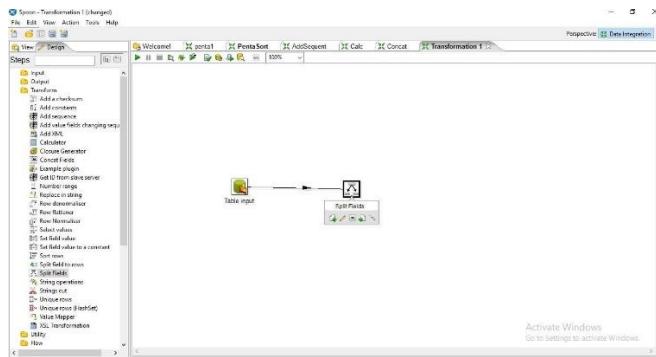
**Step 1:** Perform first 6 steps same as practical 1.



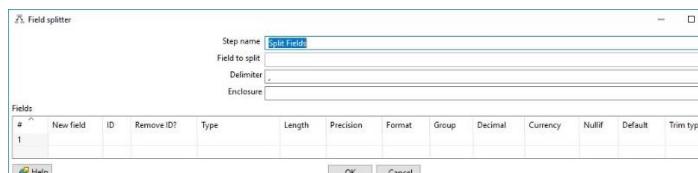
**Step 2:** Drag and drop Split Field on the Panel



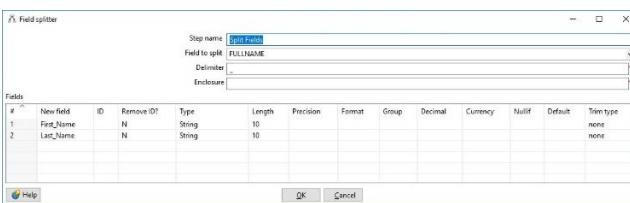
**Step 3:** Hold the mouse pointer over table input and drag it to the Split Field.



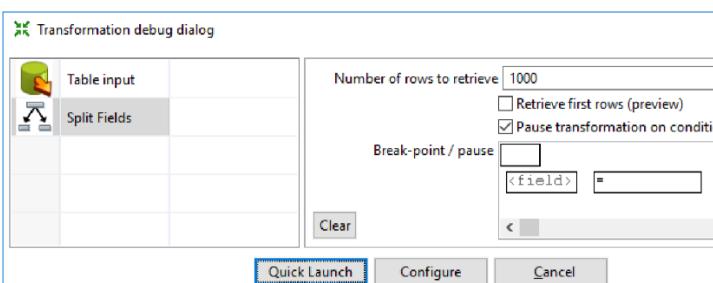
#### Step 4: Double Click On Split Field.



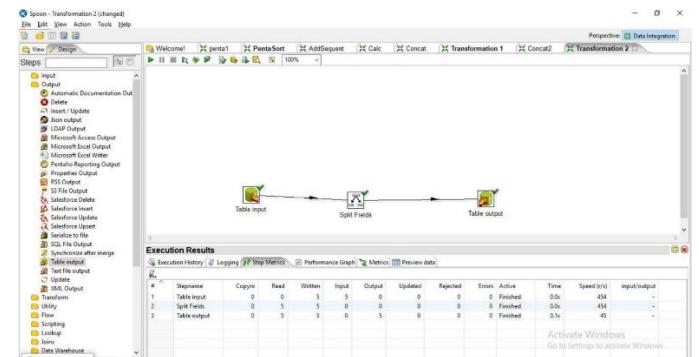
- Enter the values For Field to Split, Delimiter and the field names and press ok



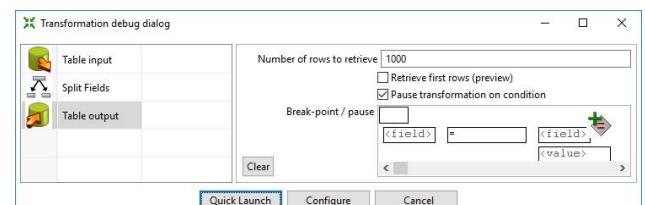
#### Step 6: Click on Debug transformation and Click on Quick Launch.



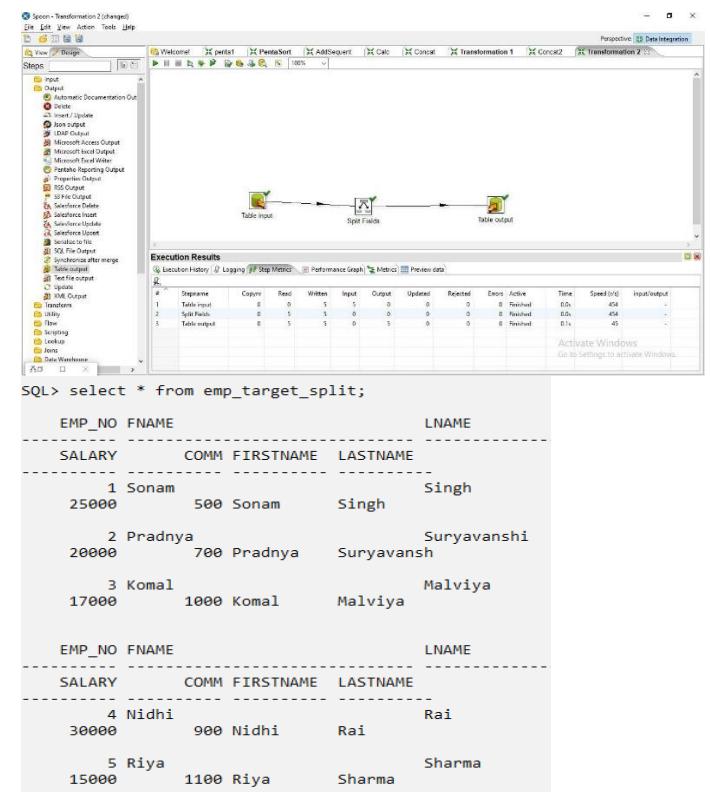
#### Step 7: Perform steps 6-10 same as practical 1.



#### Step 8: Click on Debug transformation and Click on Quick Launch.

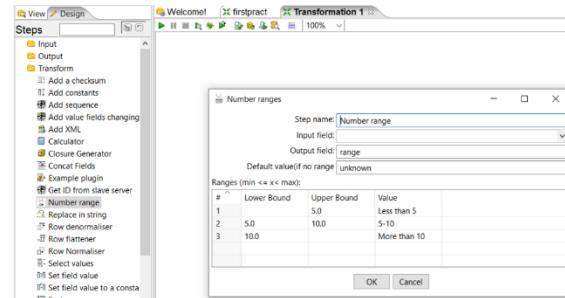


#### Output:

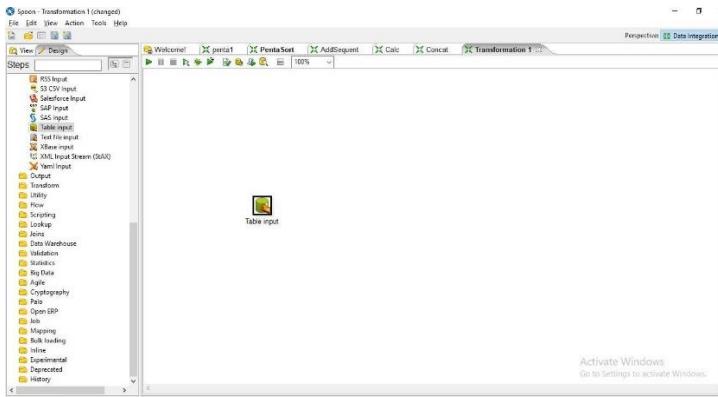


```
SQL> select * from student13;
```

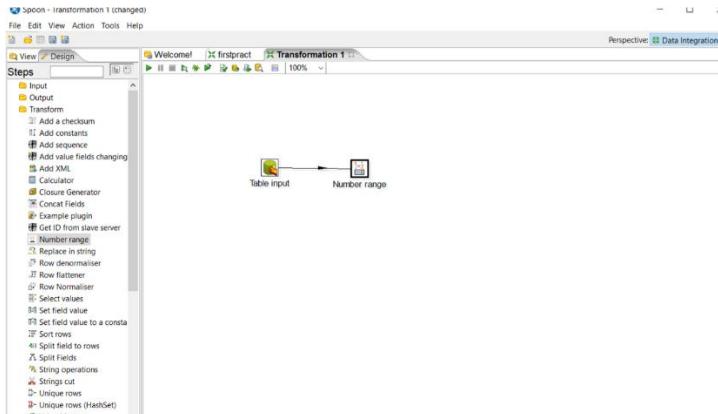
ROLL_NO	PERCENTAGE
101	55
102	59
103	68
104	75
105	83
106	35
107	88



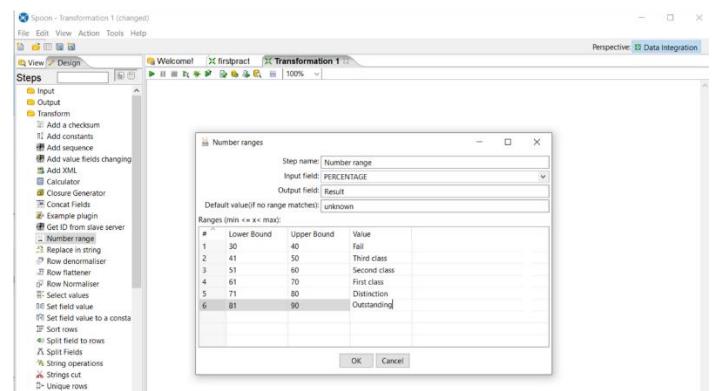
## Step 1: Perform first 6 steps same as practical 1.



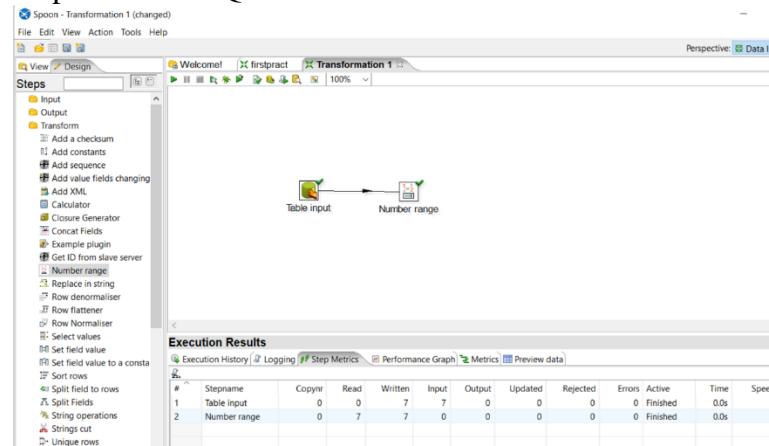
Step2: Drag and drop Number Range transformation and double click on it.



Step3: Set values for Input field, Output field, Lower bound, Upper bound and value according to the values specified in percentage column.



## Step4: Click on Quick launch



## Step 5: Click on Preview data

**Execution Results**

#	RDXL_NO	PERCENTAGE	Result
1	101	55	Second class
2	102	59	First class
3	103	68	Distinction
4	104	75	Outstanding
5	105	83	Outstanding
6	106	87	Outstanding
7	107	88	Outstanding

## 7. String Operation: Required SQL table:

```
SQL> select * from newemp13;
```

EMPN	ENAME	DEPT
E001	JOHN	D002
MANAGER		50000
E002	SMIT	D001
HR		40000
E003	STEVEKING	D002
		30000
EMPN	ENAME	DEPT
JOB		SALARY
E004	LEX HEAN	D003
MANAGER		50000
E005	BRUCE	D003
MANAGER		50000
E006	KEVINNASH	D002
		50000

### Step 1: Perform first 6 steps same as practical 1.

Step 2: Drag and drop String operation transformation, make connection between table input and string operation and double click on it.

### Step3: Set values for the parameters

### Step 4: Launch the transformation

#	EMPNO	ENAME	DEPT_NO	JOB	SALARY
1	E001	JOHN	D002	***Manager	50000
2	E002	SMIT	D001	*****Hr	40000
3	E003	STEVENKING	D002	<null>	30000
4	E004	LEX HEAN	D003	***Manager	14000
5	E005	BRUCE	D003	***Manager	17000
6	E006	KEVINNASH	D002	****Sales	18000

### Step 5: Drag and drop output table

### Step 6: Launch the transformation

### Step 7: In SQL output table

```
SQL> select * from stringgg;
EMPN      ENAME          DEPT
-----      -----
JOB          SALARY
-----      -----
E001      JOHN           D002
***Manager          50000

E002      SMIT            D001
*****Hr           40000

E003      STEVEKING        D002
                      30000

EMPN      ENAME          DEPT
-----      -----
JOB          SALARY
-----      -----
E004      LEX HEAN         D003
***Manager          50000

E005      BRUCE            D003
***Manager          50000

E006      KEVINNASH         D002
*****Sales          50000
```

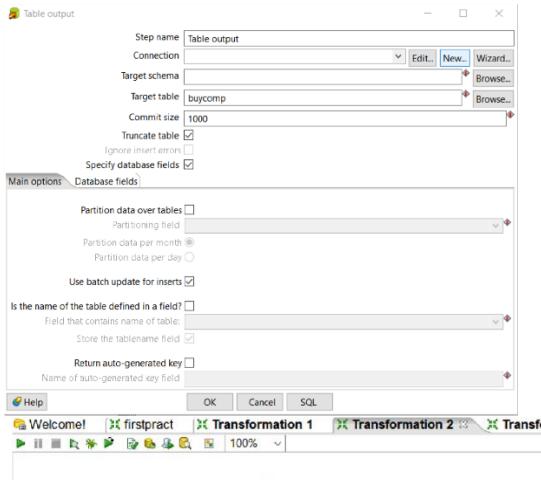
6 rows selected.

8. Copy contents of.CSV file to the target table
- Step1:Drag and drop CSV file input on a panel

Step2: Double click on it and open respective CSV file, Lazy conversation and Header row present? Options should be checked, click on Getfield and then ok.

In this step we can change data type and format of CSV files after doing setting click on ok  
 Step3: Drag and drop Table output and connect it with CSV file.

Step4: Double click on Table output and create new connection and follow same stapes like before and launch the transformation.



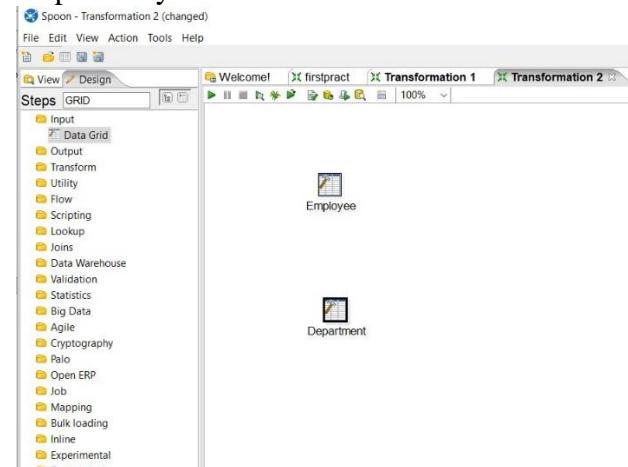
Execution Results						
<a href="#">Execution History</a>   <a href="#">Logging</a>   <a href="#">Step Metrics</a>   <a href="#">Performance Graph</a>   <a href="#">Metrics</a>   <a href="#">Preview data</a>						
<input checked="" type="radio"/> First rows   <input type="radio"/> Last rows   <input type="radio"/> Off						
<hr/>						
#	age	income	student	credit_rating	buys_computer	
1	<=30	high	no	fair	no	
2	<=30	high	no	excellent	no	
3	31..40	high	no	fair	yes	
4	>40	medium	no	fair	yes	
5	>40	low	yes	fair	yes	
6	>40	low	yes	excellent	no	
7	31..40	low	yes	excellent	yes	
8	<=30	medium	no	fair	no	
9	<=30	low	yes	fair	yes	
10	>40	medium	yes	fair	yes	
11	<=30	medium	yes	excellent	yes	
12	31..40	medium	no	excellent	yes	
13	31..40	high	yes	fair	yes	
14	..	..	..	..	..	

## Step 5: Output table created in SQL

```
SQL> select * from csv13;
+-----+-----+-----+-----+-----+
| AGE  | INCOME | STU | CREDIT_RA | BUY |
+-----+-----+-----+-----+-----+
| <=30 | high   | no  | fair    | no   |
| <=31 | high   | no  | excellent| no   |
| 31>40| high   | no  | fair    | yes  |
| >40  | medium | no  | fair    | yes  |
| >40  | low    | yes | fair    | yes  |
| >40  | low    | yes | excellent| no   |
| 31>40| low    | yes | excellent| yes  |
| <=30 | medium | no  | fair    | no   |
| <=30 | low    | yes | fair    | yes  |
| >40  | medium | yes | fair    | yes  |
| <=30 | medium | yes | excellent| yes  |
+-----+-----+-----+-----+-----+
14 rows selected.
```

## 9. Implement the merge join transformation on

**tables Step1:** Drag two Data grid on panel rename them with Employee and Department respectively.



## Step2: Insert records into respective grids.

Add constant rows							
Step name: Employee							
Meta	Data	#	Name	Type	Format	Length	Precision
		1	Empid	Integer			
		2	Empname	String			
		3	Age	Integer			
		4	Depno	Integer			

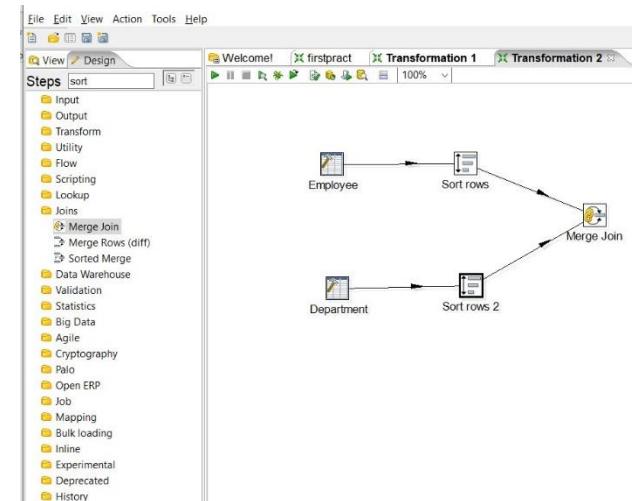
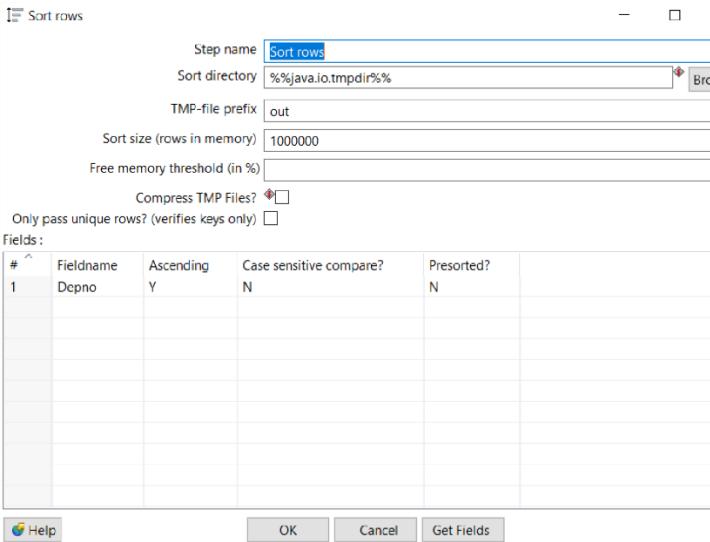
  

Add constant rows						
Step name: Employee						
Meta	Data	#	Empid	Empname	Age	Depno
		1	101	sandip	38	10
		2	102	vishwas	39	20
		3	103	vaishu	37	30
		4	104	sheetal	40	40

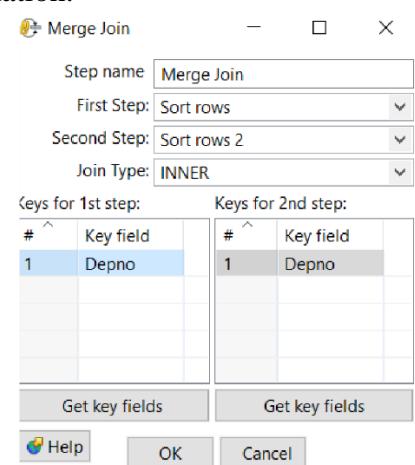
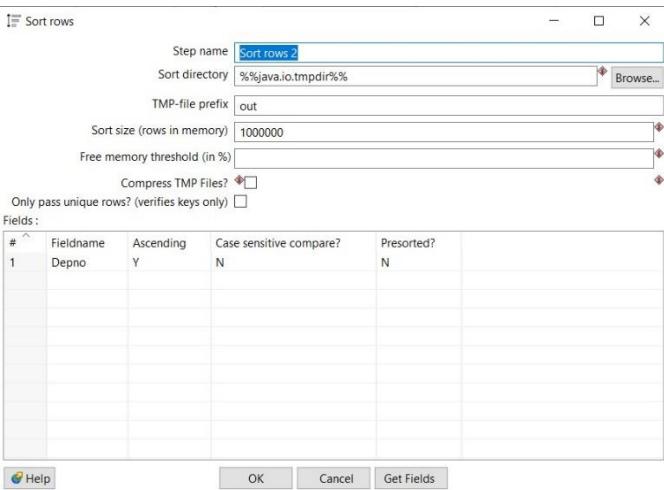
  

Meta Data						
#	Depno	Depname				
1	10	IT				
2	30	Admin				
3	20	Hr				

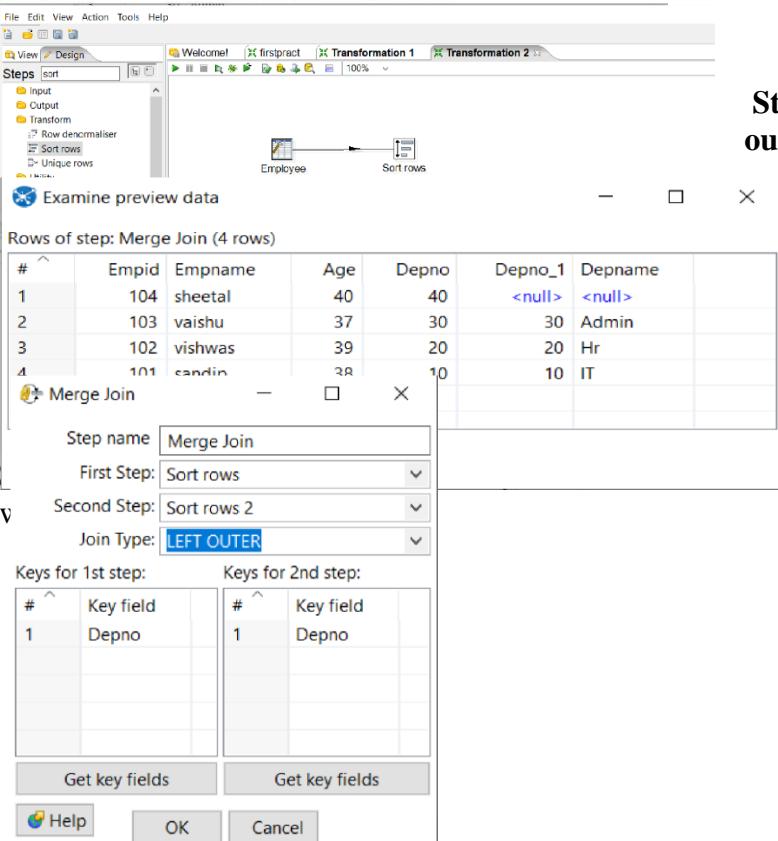
**Step3: Insert sort rows transformation(on Empid) for both Employee and Department.**

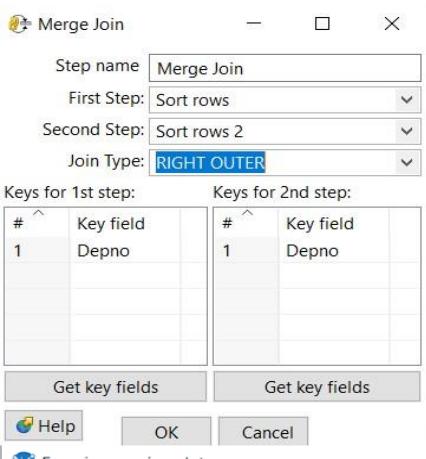


**Step5:** Double click on Merge join and made necessary settings First select First\_step, then Second\_step, select Join type INNER, and from get fields select Depno only, after that quick launch the transformation.



**Step 6: Select Join Type LeftOuter,then Right outer and finally Full outer join**



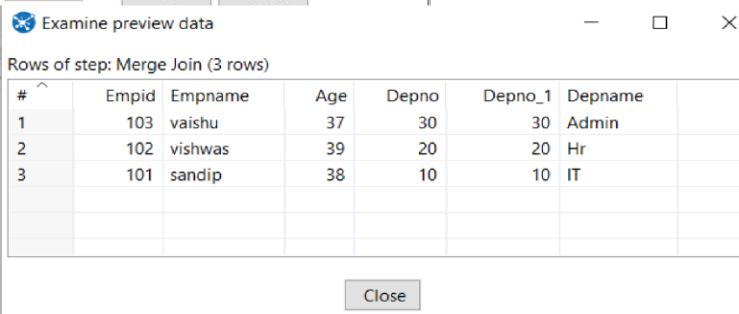


### Examine preview data

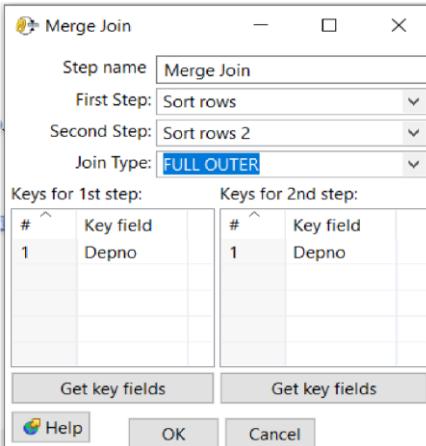
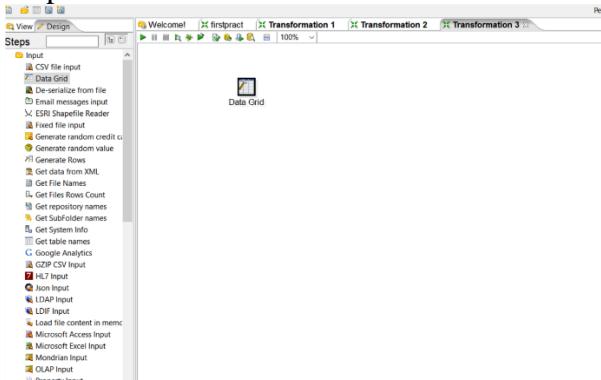
Rows of step: Merge Join (4 rows)

#	Empid	Emplname	Age	Depno	Depno_1
1	104	sheetal	40	40	<null>
2	103	vaishu	37	30	30
3	102	vishwas	39	20	20
4	101	sandip	38	10	10

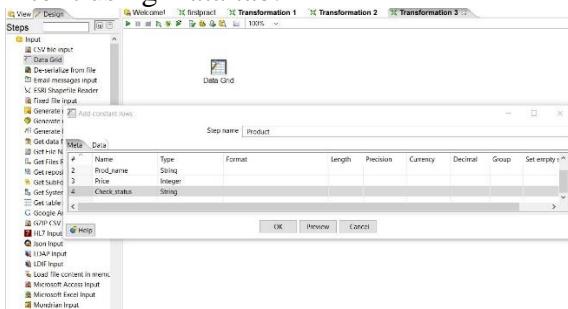
Close

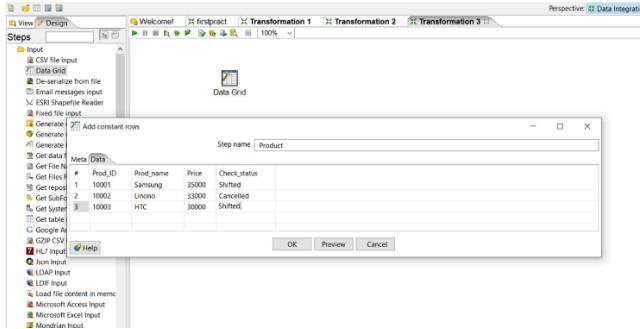


0. Implement different data validations on table. Step1: Drag and drop Data grid on panel.

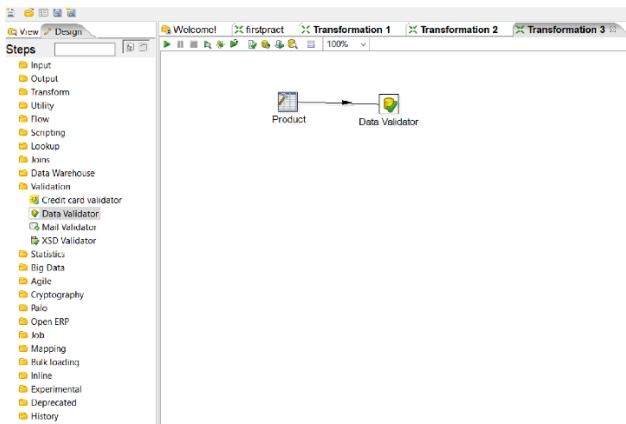


Step2: Double click on Data grid and create product table in Meta tab and enter records into it using Data tab.



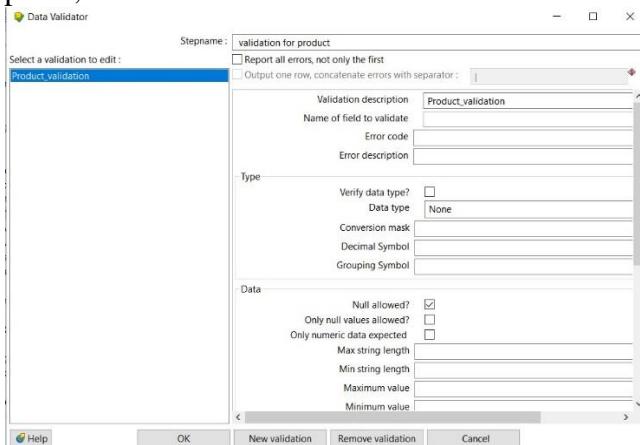


Step3: Go to Validation steps and select, Drag and Drop Data validator on panel and double click on it.

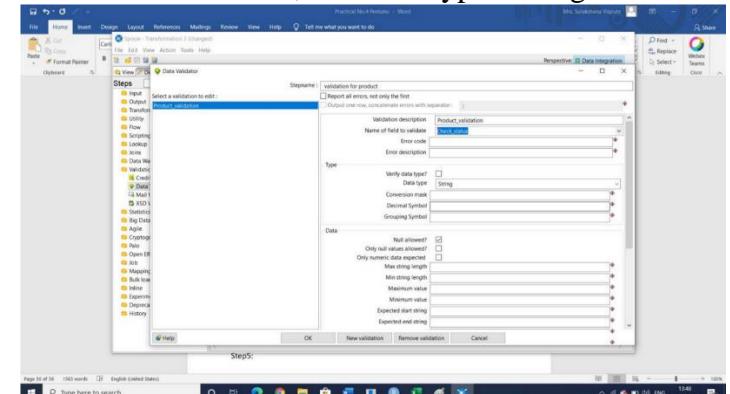


Step4: Click on new validation button and give new name to validation.

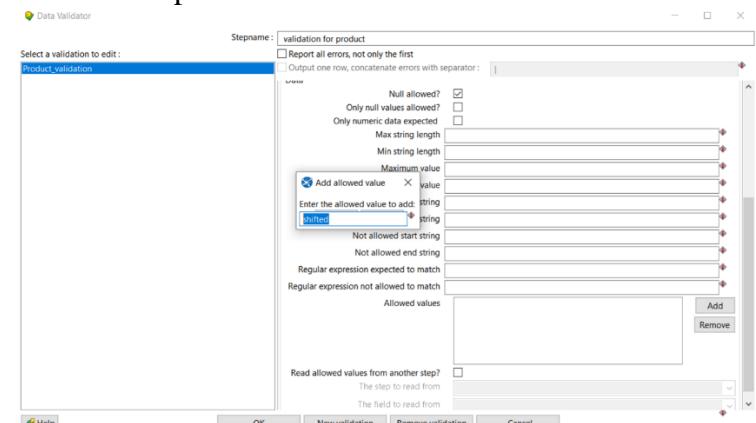
Once new validation is created it will appear on left panel, double click on it to set data validation



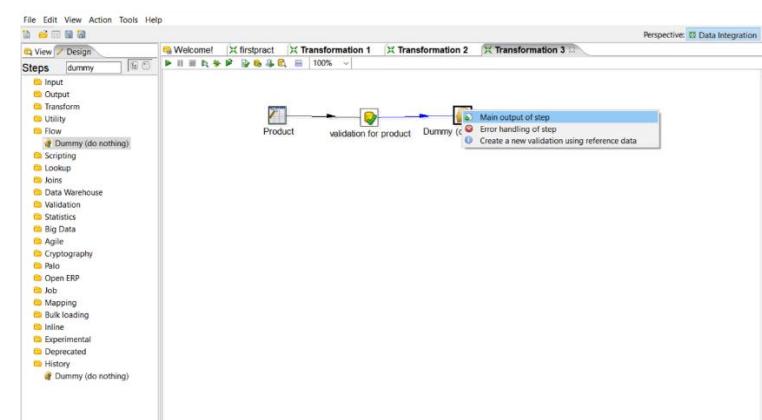
Step5: Set values for Name of field to validate=check status, and Data type =string



Step6: Click on add to set validation, set it to Shifted and press Enter and click on ok.



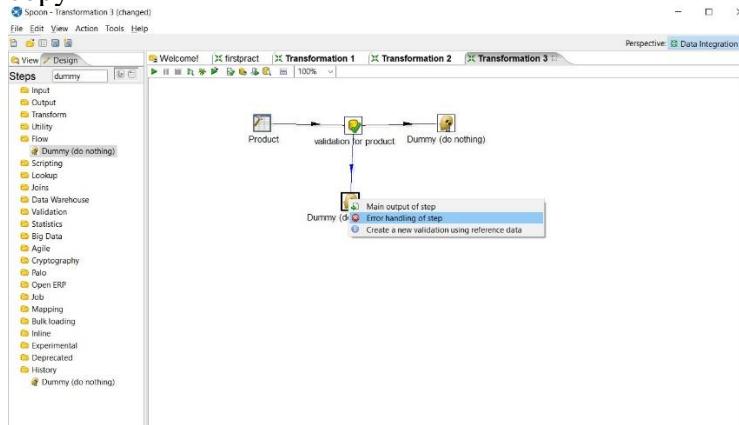
Step7: Select Dummy file from Flow to hold output and connect it with Data validation, while connecting select option “ Main Output of step”



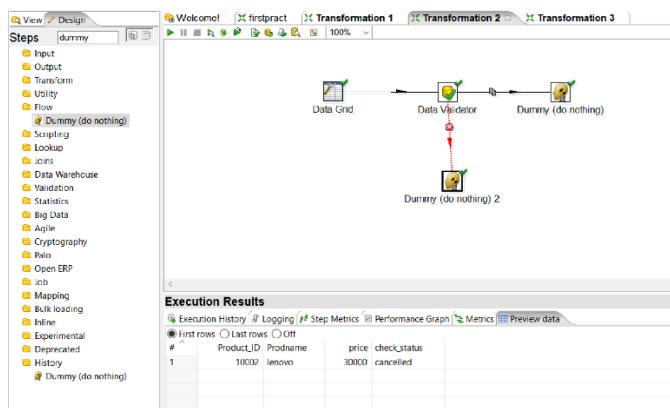
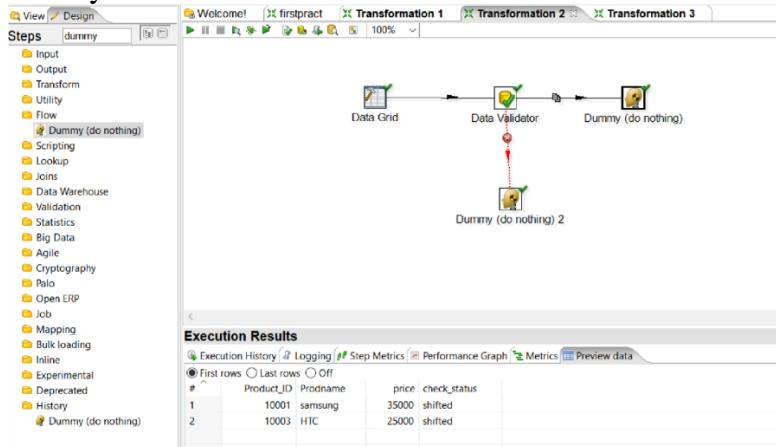
Step8: Select another dummy and connect it to ‘Validation for product’ and while selecting select

'Error handling of step and in next window click on

copy



Step 9: launch transformation by selecting one dummy file each.



## Practical No. 5

**Aim:** - Introduction to R programming and Data acquisition.

**Objective:** - To understand the concept of R Programming. and evaluate different data mining techniques like classification, prediction, clustering and association rule mining in R

### R Commands:-

```
myString <- "Hello, World!"
```

```
print ( myString)
```

```
[1] "Hello, world!"
```

#setwd() - sets working directory.

```
setwd("D:/58")
```

getwd() ##getwd() - gets current working directory.

```
[1] "D:/58"
```

dir() ## dir() - lists the contents of current working directory.

```
[1] "data_apriori.csv" "missing_col1.csv" "missing_col1.xlsx" "na_data.csv"      "PRACT5.R"
> |
```

ls() ##ls() - lists names of objects in R environment

```
[1] "accuracy"          "clustercut"        "clusters"          "df"
[5] "df2"              "fit"                "index"            "iris"
[9] "iris_norm"         "iris_target_category" "iris_rest"         "iris_test_category"
[13] "iris_train"        "iriscluster"       "lmodel1"          "model"
[17] "my_data"           "mystring"          "nor"              "pr"
[21] "preds"             "preds_new"         "ran"              "tab"
[25] "test"              "testdata"          "train"            "trainData"
[29] "xTest"             "xTrain"            "yTest"            "ytrain"
```

#Creating and assigning to a variable:

```
x<-1
```

```
class(x)
```

```
[1] "numeric"
```

#Printing a variable:

```
#auto-printing
```

```
print(x) #explicit printing
```

```
[1] 1
```

```
x<-'c'
```

```
is.character(x) #check if character
```

```
[1] TRUE
```

```
is.integer(x) #check if integer
```

```
[1] FALSE
```

```
y<-'2.14'
```

```
as.integer(y)
```

```
[1] 2
```

##Creating Vector: contains objects of same class.

```
y<-vector("logical",length=10)
length(x)
[1] 3
y<-c(4,5,6)
5*x
[1] 56.5 137.5 169.0
###Creating Matrix: Two-dimensional array having elements of same class.
m<-matrix(c(11,12,13,55,60,65,66,72,78),
nrow=3,ncol=3)
m
[,1] [,2] [,3]
[1,]    11    55    66
[2,]    12    60    72
[3,]    13    65    78
> |
dim(m)
[1] 3 3
attributes(m)
$dim
[1] 3 3
matrix(c(11,12,13,55,60,65,66,72,78),nrow=
3,ncol=3,byrow = TRUE)
m
-----, , -----
[,1] [,2] [,3]
[1,]    11    12    13
[2,]    55    60    65
[3,]    66    72    78
> m
[,1] [,2] [,3]
[1,]    11    55    66
[2,]    12    60    72
[3,]    13    65    78
> |
x<-c(1,2,3) y<-c(11,12,13)
cbind(x,y)
   x  y
[1,] 11.3 4
[2,] 27.5 5
[3,] 33.8 6
> |
rbind(x,y)
[1] [,2] [,3]
x 11.3 27.5 33.8
y  4.0  5.0  6.0
> |
p<-3*m
p
```

```

[,1] [,2] [,3]
[1,]   33  165  198
[2,]   36  180  216
[3,]   39  195  234
> |
n<-
matrix(c(4,5,6,14,15,16,24,25,26),nrow=3,n
col=3)
q<- m+n
q
[,1] [,2] [,3]
[1,]   15   69   90
[2,]   17   75   97
[3,]   19   81   104
> |
o<-matrix(c(4,5,6,14,15,16),nrow=3,ncol=2)
o
[,1] [,2]
[1,]    4   14
[2,]    5   15
[3,]    6   16
> |
r<-m%*% o
r
[,1] [,2]
[1,]  715 2035
[2,]  780 2220
[3,]  845 2405
> |
mdash<-t(m)
mdash
[,1] [,2] [,3]
[1,]   11   12   13
[2,]   55   60   65
[3,]   66   72   78
> |
s<-matrix(c(4,5,6,14,15,16,24,25,26),
nrow=3,ncol=3,byrow=TRUE)
s_det<-det(s)
s_det
[1] 1.110223e-14
###List: A special type of vector containing
elements of different classes. #####Elements
of list can be accessed by giving
element index or name in [[[]]].

x<-list(1,"p",TRUE,2+4i)
x
[[1]]
[1] 1
[[2]]
[1] "p"
[[3]]
[1] TRUE
[[4]]
[1] 2+4i
###Factor: Represents categorical data. Can
be ordered or unordered.
status<-
c("low","high","medium","high","low")
x<-factor(status,
ordered=TRUE,levels=c("low","medium","h
igh"))
x
[1] low   high   medium high   low
Levels: low < medium < high
> |
###Data frame: Used to store tabular data.
Can contain different classes.
student_id<-c(1,2,3)
student_names<-
c("Ram","Shyam","Laxman")
position<-
c("First","Second","Third")
data<-
data.frame(student_id,student_names,positio
n)
data
  student_id student_names position
1           1          Ram     First
2           2         Shyam    Second
3           3        Laxman   Third
> |
data$student_id
[1] 1 2 3
nrow(data)
[1] 3
ncol(data)
[1] 3
names(data)
[1] "student_id"  "student_names" "position"
smoke <-
matrix(c(51,43,22,92,28,21,68,22,9),ncol=3,
byrow=TRUE)
colnames(smoke)<-
c("High","Low","Middle")
rownames(smoke)<-

```

```
c("current","former","never")
smoke <- as.table(smoke)
smoke
      High Low Middle
current   51  43    22
former    92  28    21
never     68  22     9
> |
Reading and writing data from csv
dataT <- read.table("na_data.csv", sep =",",
header = T)
dataT
      X1      Rick x623.3 x01.01.2012      IT
1 2      Dan 515.20 23/09/2013 operations
2 3 Michelle 611.00 15/11/2014      IT
3 4      Ryan 729.00 11/05/2014      HR
4 NA      Gary 843.25 27/03/2015 Finance
5 6      Nina NA 21/05/2013      IT
6 7 Simon 632.80 30/07/2013 operations
7 8      Guru 722.50 17/06/2014 Finance
8 9      John NA 21/05/2012
9 10 Rock 600.80 30/07/2013      HR
10 11 Brad 1032.80 30/07/2013 operations
11 12 Ryan 729.00 11/05/2014      HR
> |
dim(dataT)
[1] 11 5
head(dataT, 2)
      X1      Rick x623.3 x01.01.2012      IT
1 2      Dan 515.20 23/09/2013 operations
2 3 Michelle 611.0 15/11/2014      IT
> |
tail(dataT, 2)
      X1 Rick x623.3 x01.01.2012      IT
10 11 Brad 1032.8 30/07/2013 operations
11 12 Ryan 729.0 11/05/2014      HR
> |
z <- data.frame(a = 5, b = 10, c = pi)
write.csv(z,file="data.csv")
```

A	B	C	D
a	b	c	
1	5	10	3.141593

Reading and writing data from Excel using XLConnect

```
dataX <- XLConnect::
readWorksheetFromFile("employee_info.xls
",sheet=1)
dataX
```

```
x1      Rick x623.3 x2012.01.01.00.00.00      IT
1 2      Dan 515.20 23/09/2013 operations
2 3 Michelle 611.00 15/11/2014      IT
3 4      Ryan 729.00 2014-11-05 00:00:00      HR
4 NA      Gary 843.25 27/03/2015 Finance
5 6      Nina NA 21/05/2013      IT
6 7 Simon 632.80 30/07/2013 operations
7 8      Guru 722.50 17/06/2014 Finance
8 9      John NA 21/05/2012      <NA>
9 10 Rock 600.80 30/07/2013      HR
10 11 Brad 1032.80 30/07/2013 operations
11 12 Ryan 729.00 2014-11-05 00:00:00      HR
> |
```

```
dataY <- dataX[1:2,]
```

```
dataY
```

```
x1      Rick x623.3 x2012.01.01.00.00.00      IT
1 2      Dan 515.20 23/09/2013 operations
2 3 Michelle 611.0 15/11/2014      IT
> |
```

Reading and writing data from Excel using  
readXL and writeXL  
data <- data.frame(Name=character(),  
Age=numeric())

## Practical No. 6

**Aim: - Implementation of Data preprocessing techniques like,**

1. Naming and Renaming variables, adding a new variable.
2. Dealing with missing data.
3. Dealing with categorical data.
4. Data reduction using sub setting.

### R Commands:-

#my\_data

```
> setwd("D:/46_fymca")
> getwd()
[1] "D:/46_fymca"
> data<-mtcars
> head(data,5)
      mpg cyl disp hp drat wt qsec vs am gear carb
Mazda RX4     21.0   6 160 110 3.90 2.620 16.46  0  1  4   4
Mazda RX4 Wag 21.0   6 160 110 3.90 2.875 17.02  0  1  4   4
Datsun 710    22.8   4 108  93 3.85 2.320 18.61  1  1  4   1
Hornet 4 Drive 21.4   6 258 110 3.08 3.215 19.44  1  0  3   1
Hornet Sportabout 18.7   8 360 175 3.15 3.440 17.02  0  0  3   2
> data1<-data[1:6,1:5]
> data1
      mpg cyl disp hp drat
Mazda RX4     21.0   6 160 110 3.90
Mazda RX4 Wag 21.0   6 160 110 3.90
Datsun 710    22.8   4 108  93 3.85
Hornet 4 Drive 21.4   6 258 110 3.08
Hornet Sportabout 18.7   8 360 175 3.15
Valiant ... 18.1   6 225 105 2.76
```

### ## Renaming columns with dplyr::rename()

```
> require(dplyr)
> install.packages("dplyr")
Error in install.packages : Updating loaded packages
> install.packages("dplyr")
WARNING: Rtools is required to build R packages but is not currently installed.
download and install the appropriate version of Rtools before proceeding

https://cran.rstudio.com/bin/windows/Rtools/
Warning in install.packages :
  package 'dplyr' is in use and will not be installed
> data1<-rename(data1,horse_power~hp)
> data1
      mpg cyl disp horse_power drat
Mazda RX4     21.0   6 160          110 3.90
Mazda RX4 Wag 21.0   6 160          110 3.90
Datsun 710    22.8   4 108          93 3.85
Hornet 4 Drive 21.4   6 258          110 3.08
Hornet Sportabout 18.7   8 360          175 3.15
Valiant ... 18.1   6 225          105 2.76
```

### ## Adding new variable

```
> data1$new_hp1 <- data1$horse_power*0.5
> colnames(data1)
[1] "mpg"       "cyl"        "disp"       "horse_power" "drat"
> data1
      mpg cyl disp horse_power drat new_hp1
Mazda RX4     21.0   6 160          110 3.90      55.0
Mazda RX4 Wag 21.0   6 160          110 3.90      55.0
Datsun 710    22.8   4 108          93 3.85      46.5
Hornet 4 Drive 21.4   6 258          110 3.08      55.0
Hornet Sportabout 18.7   8 360          175 3.15      87.5
Valiant ... 18.1   6 225          105 2.76      52.5
```

#naming variable

Create missing\_col1.csv file with following data.

1	Rick	623.3	01/01/2012	IT
2	Dan	515.2	23/09/2013	Operations
3	Michelle	611	15/11/2014	IT
4	Ryan	729	11/05/2014	HR
	Gary	843.25	27/03/2015	Finance
6	Nina		21/05/2013	IT
7	Simon	632.8	30/07/2013	Operations
8	Guru	722.5	17/06/2014	Finance
9	John		21/05/2012	
10	Rock	600.8	30/07/2013	HR
11	Brad	1032.80	30/07/2013	Operations
12	Ryan	729	11/05/2014	HR

#Reading with read.table() assumes no headers by default. First few lines :

```
> data2 = read.table(file= "D:/46_fymca/missing_col1.csv", sep = ",") 
> data2=read.table(file="D:/46_fymca/na_data.csv",sep = ",") 
> data2
      V1        V2        V3        V4        V5
1  1     Rick  623.30 01/01/2012      IT
2  2      Dan  515.20 23/09/2013 Operations
3  3 Michelle 611.00 15/11/2014      IT
4  4      Ryan  729.00 11/05/2014      HR
5  NA     Gary  843.25 27/03/2015 Finance
6  6      Nina  NA     21/05/2013      IT
7  7     Simon  632.80 30/07/2013 Operations
8  8      Guru  722.50 17/06/2014 Finance
9  9      John  NA     21/05/2012      HR
10 10     Rock  600.80 30/07/2013      HR
11 11     Brad  1032.80 30/07/2013 operations
12 12     Ryan  729.00 11/05/2014      HR
```

#V1, V2, V3.. are given as default names (titles) by R

```
> data2=read.csv(file="D:/46_fymca/na_data.csv",col.names=c("sno","Name","Salary","DateofJoining","Department"))
> data2
      sno      Name  Salary DateofJoining Department
1     2      Dan  515.20 23/09/2013 Operations
2     3 Michelle 611.00 15/11/2014      IT
3     4      Ryan  729.00 11/05/2014      HR
4     NA     Gary  843.25 27/03/2015 Finance
5     6      Nina  NA     21/05/2013      IT
6     7     Simon  632.80 30/07/2013 Operations
7     8      Guru  722.50 17/06/2014 Finance
8     9      John  NA     21/05/2012      HR
9    10     Rock  600.80 30/07/2013      HR
10   11     Brad  1032.80 30/07/2013 operations
11   12     Ryan  729.00 11/05/2014      HR
```

Error Detection and Correction NA: Not Available - Known as missing values

Works as a place holder for something that is 'missing'

Most basic operations (addition, subtraction, multiplication, etc.) in R deal with it without crashing and return NA if one of the inputs is NA

is.na(VALUE) is used to check if the input value is NA or not. Returns a TRUE/FALSE vector Whereas in case of Excel like utilities for numeric computations it's assumed to be 0.

```
# Operation with NA
```

```
NA+4
```

```
[1]NA
```

```
# Create a vector V with 1 NA value
```

```
v <- c(1,2,NA,3)
```

```
# Median with and without NA (remove NA)
```

```
> median(v)
[1] NA
```

```
# On removing NAs
```

```
> median(v,na.rm = T)
[1] 2
```

```
# Apply is.na() to vector
```

```
>is.n(V)
[1] FALSE FALSE TRUE FALSE
```

```
# Removing the NA values by using logical indexing
```

```
> naVals <- is.na(v)
```

```
# Get values that are not NA
```

```
V[!naVals]
[1] 1 2 3
```

```
# Subsetting with complete cases - values that are not NA
```

```
V[complete.cases(V)]
```

```
[1] 1 2 3
```

```
#Create na_data.csv file with following data.
```

1	Rick	623.3	01/01/2012	IT
2	Dan	515.2	23/09/2013	Operations
3	Michelle	611	15/11/2014	IT
4	Ryan	729	11/05/2014	HR
	Gary	843.25	27/03/2015	Finance
6	Nina		21/05/2013	IT
7	Simon	632.8	30/07/2013	Operations
8	Guru	722.5	17/06/2014	Finance
9	John		21/05/2012	
10	Rock	600.8	30/07/2013	HR
11	Brad	1032.8	30/07/2013	Operations
12	Ryan	729	11/05/2014	HR

```
# Subsetting a data frame with complete cases
# Complete Data of Prime Ministers. Notice NAs
```

```
> dataC <- read.csv(file ="D:/46_fymca/na_data.csv", na.strings = "")
> dataC
   X1      Rick  x623.3 x01.01.2012      IT
1  2       Dan  515.20 23/09/2013 operations
2  3  Michelle 611.00 15/11/2014      IT
3  4      Ryan 729.00 11/05/2014      HR
4  NA     Gary 843.25 27/03/2015  Finance
5  6      Nina    NA 21/05/2013      IT
6  7      Simon 632.80 30/07/2013 operations
7  8      Guru 722.50 17/06/2014  Finance
8  9      John    NA 21/05/2012      <NA>
9  10     Rock 600.80 30/07/2013      HR
10 11     Brad 1032.80 30/07/2013 operations
11 12      Ryan 729.00 11/05/2014      HR
```

```
# Subset only the rows without NA
```

```
> dataCompleteCases <- dataC[complete.cases(dataC),]
> dataCompleteCases
   X1      Rick  x623.3 x01.01.2012      IT
1  2       Dan  515.20 23/09/2013 operations
2  3  Michelle 611.00 15/11/2014      IT
3  4      Ryan 729.00 11/05/2014      HR
6  7  simon 632.80 30/07/2013 operations
7  8      Guru 722.50 17/06/2014  Finance
9  10     Rock 600.80 30/07/2013      HR
10 11     Brad 1032.80 30/07/2013 operations
11 12      Ryan 729.00 11/05/2014      HR
```

### Imputation

The process of estimating or deriving missing values There are various methods for imputation

- Imputation of the mean
- Imputation of the median
- Imputation using linear regression models
- Package Hmisc implements many imputation methods, few examples :

```
> install.packages("Hmisc")
WARNING: Rtools is required to build R packages but is not currently installed. Please download and install the appropriate version of Rtools before proceeding.

https://cran.rstudio.com/bin/windows/Rtools/
Installing package into 'C:/Users/ADMIN/Documents/R/win-library/4.1'
(as 'lib' is unspecified)
trying URL 'https://cran.rstudio.com/bin/windows/contrib/4.1/Hmisc_4.7-2.zip'
Content type 'application/zip' length 3302837 bytes (3.1 MB)
downloaded 3.1 MB

package 'Hmisc' successfully unpacked and MD5 sums checked

The downloaded binary packages are in
  C:\Users\ADMIN\AppData\Local\Temp\RtmpCiYL4j\downloaded_packages
> library(Hmisc)
Loading required package: lattice
Loading required package: survival
Loading required package: Formula
Loading required package: ggplot2
Attaching package: 'Hmisc'

The following objects are masked from 'package:base':
  format.pval, units

Warning messages:
1: package 'Hmisc' was built under R version 4.1.3
2: package 'ggplot2' was built under R version 4.1.3
```

## create a vector

```
> x=c(1,2,3,NA,4,4,NA)
```

# mean imputation - from package, mention name of function to be used

```
> x <- impute(x,fun = mean)
> x
  1   2   3   4   5   6   7
1.0 2.0 3.0 2.8* 4.0 4.0 2.8*
```

#median imputation

```
> x <- impute(x,fun = median)
> x
  1   2   3   4   5   6   7
1.0 2.0 3.0 2.8* 4.0 4.0 2.8*
```

\*\* Categorical Data \*\*

## Factors are variables in R which take on a limited number of different values; such variables are often referred to as categorical variables.

#Convert Character into Factor(categorical data)

# Create gender vector

```
> gender_vector <- c("Male", "Female",
> class(gender_vector)
[1] "character"
```

# Convert gender\_vector to a factor

```
> factor_gender_vector <- factor(gende
> class(factor_gender_vector)
[1] "factor"
```

# Create Ordinal categorical vector

```
> day_vector <- c('evening', 'morning', 'afternoon', 'midday', 'midnight', 'evening')
```

# Convert `day\_vector` to a factor with ordered level

```
> factor_day <- factor(day_vector, order = TRUE, levels =c('morning','midday', 'afternoon','evening','midnight'))
```

# Print the new variable

```
> factor_day
[1] evening morning afternoon midday midnight evening
Levels: morning < midday < afternoon < evening < midnigh
```

# Convert Numeric to Factor

# Creating vectors

```
> age <- c(40, 49, 48, 40, 67, 52, 53)
> salary <- c(103200, 106200, 150200, 10606, 10390, 14070, 10220)
> gender <- c("male", "male", "transgender", "female", "male", "female", "transgender")
```

# Creating data frame named employee

```
> employee<-data.frame(age, salary, gender)
> employee
  age salary      gender
1 40 103200     male
2 49 106200     male
3 48 150200 transgender
4 40 10606      female
5 67 10390      male
6 52 14070      female
7 53 10220      transgender
```

# Creating a factor corresponding to age with labels

```
> wfact = cut(employee$age, 3,labels=c('Young',
> table(wfact)
wfact
  Young Medium    Aged
        4       2       1
> |
```

## Practical 7

**Aim:** - Implementation and analysis of Linear regression through graphical methods

**Theory:** - Linear regression analysis is used to predict the value of a variable based on the value of another variable. The variable you want to predict is called the dependent variable. The variable you are using to predict the other variable's value is called the independent variable.

### R Commands:

```
# setwd("D:/53 FYMCA/prac7")
#getwd()
[1] "D:/53 FYMCA/prac7"
# my_data <-mtcars
# names(my_data)
[1] "mpg" "cyl" "disp" "hp"   "drat" "wt"   "q
# dim(my_data)
----- RANDOMIZE -----
# my_data <- my_data[sample(nrow(my_data),
), ]
# head(my_data)
      mpg cyl  disp  hp drat
Honda Civic 30.4 4 75.7 52 4.93
Lotus Europa 30.4 4 95.1 113 3.77
Merc 280 19.2 6 167.6 123 3.92
Merc 230 22.8 4 140.8 95 3.92
Hornet 4 Drive 21.4 6 258.0 110 3.08
AMC Javelin 15.2 8 304.0 150 3.15
...
# TrainData <- my_data[1:20]
# TestData <- my_data[21:32]
-----Linear Model-----
# fit = lm(mpg ~ hp, data=mtcars)
# summary(fit)
Call:
lm(formula = mpg ~ hp, data = mtcars)

Residuals:
    Min      1Q  Median      3Q     Max 
-5.7121 -2.1122 -0.8854  1.5819  8.2360 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept) 30.09886   1.63392 18.421 < 2e-16 ***
hp        -0.06823   0.01012 -6.742 1.79e-07 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.863 on 30 degrees of freedom
Multiple R-squared:  0.6024, Adjusted R-squared:  0.5892 
F-statistic: 45.46 on 1 and 30 DF,  p-value: 1.788e-07

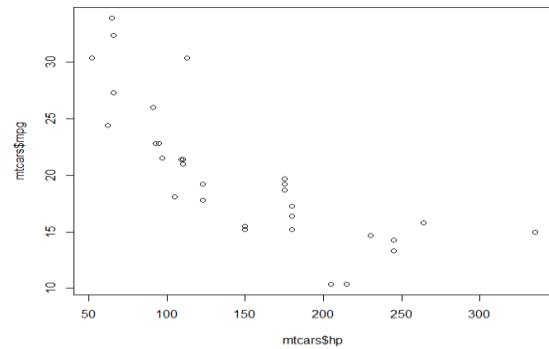
# preds <- predict(fit, newdata = TestData)
# df1 <-data.frame(preds,TestData$mpg)
# head(df1)
```

	preds	TestData.mpg
Fiat x1-9	25.59579	27.3
Lincoln Continental	15.42978	10.4
Hornet Sportabout	18.15891	18.7
Toyota Corona	23.48072	21.5
Chrysler Imperial	14.40636	14.7
Mazda RX4	22.59375	21.0

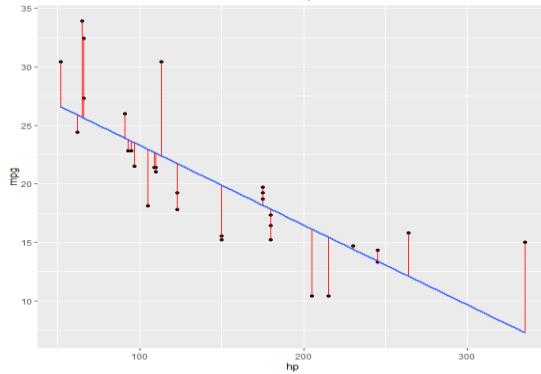
```
#Correlation :- cor(preds,TestData$mpg)
```

```
[1] 0.8784059
```

```
# plot(mtcars$hp, mtcars$mpg)
```



```
# ggplot(fit, aes(hp, mpg)) + geom_point() +
stat_smooth(method = lm, se = FALSE) +
geom_segment(aes(xend = hp, yend = .fitted),
color = "red", size = 0.3)
```



```
-----Better MODEL-----
```

```
# lmmodel1 <- lm(mpg ~ hp+cyl+gear+wt, data
=TrainData)
# summary(lmmodel1)
```

```

Call:
lm(formula = mpg ~ hp + cyl + gear + wt, data = TrainData)

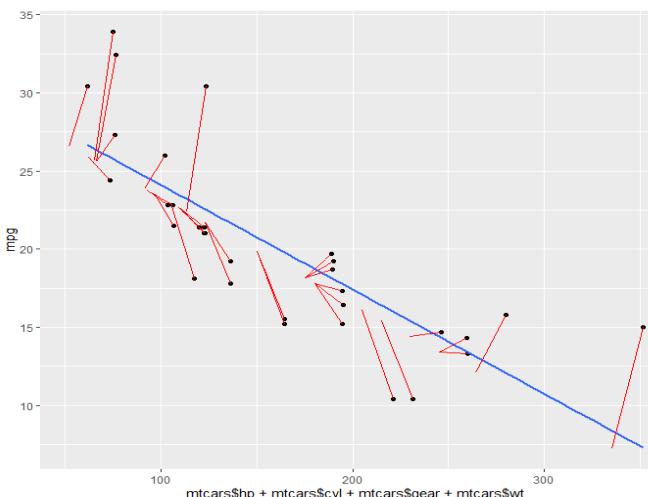
Residuals:
    Min      1Q  Median      3Q     Max 
-1.9431 -1.1795 -0.4437  0.4191  4.1244 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept) 50.719998   6.657175   7.619 1.56e-06 ***
hp          -0.001659   0.015130  -0.110 0.914155    
cyl         -1.684467   0.760197  -2.216 0.042588 ***  
gear        -1.310123   1.008465  -1.299 0.213515    
wt          -4.667897   1.082578  -4.312 0.000617 ***  
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1 

Residual standard error: 1.976 on 15 degrees of freedom
Multiple R-squared:  0.9138, Adjusted R-squared:  0.8908 
F-statistic: 39.73 on 4 and 15 DF,  p-value: 8.187e-08

# preds_new <- predict(lmmodel1, newdata =
# TestData)
# df2 <- data.frame(preds_new, TestData$mpg)
# head(df2)
      preds_new TestData$mpg
Fiat X1-9       29.599785    27.3
Lincoln Continental 7.638601    10.4
Hornet Sportabout  16.966057    18.7
Toyota Corona     28.384503    21.5
Chrysler Imperial  7.982484    14.7
Mazda RX4        22.960359    21.0
-----Corelation-----
# cor(preds_new, TestData$mpg)
[1] 0.8891933
# ggplot(fit,
aes(mtcars$hp+mtcars$cyl+mtcars$gear+mtcars
$wt, mpg)) + geom_point() +
# stat_smooth(method = lm, se = FALSE) +
# geom_segment(aes(xend = hp, yend = .fitted),
color = "red", size = 0.3)

```



## Practical 8

**Aim:** - Implementation and analysis of Classification algorithms like

1. Naive Bayesian,
2. K-Nearest Neighbor.

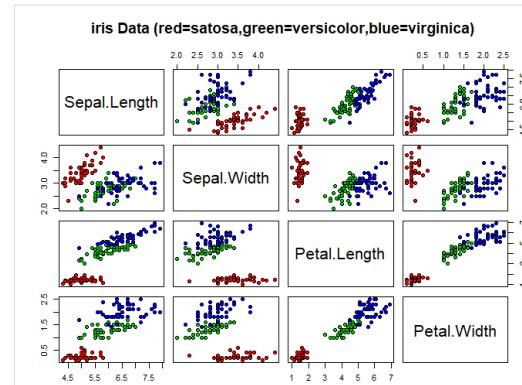
### Theory:-

#### Naive Bayes

- Based on the Bayes theorem
  - Predicts based on probabilities from training data
- $P(B|A) = P(A|B) P(B)/P(A)$  Gives posterior probability of 'B' given 'A' using prior probability of 'B' prior probability of 'A' and conditional probability of 'A' given 'B'
- Takes two step approach
    - Calculates the posterior probability of the Class given the input - for every class
    - Assigns the class with higher posterior probability
  - More suited when dimensionality of input is high the - widely used for document classification
  - Also good for the multiclass classifications
  - Works well with less datasets also, but the assumption that predictor variables are independent should hold ##NaiveBayes .

### R Commands:-

```
setwd("E:/R Orientation")
library("caret")
library(ggplot2)
data(iris)
head(iris)
> head(iris)
   Sepal.Length Sepal.Width Petal.Length Petal.Width Species
1          5.1         3.5          1.4         0.2  setosa
2          4.9         3.0          1.4         0.2  setosa
3          4.7         3.2          1.3         0.2  setosa
4          4.6         3.1          1.5         0.2  setosa
5          5.0         3.6          1.4         0.2  setosa
6          5.4         3.9          1.7         0.4  setosa
> |
unique(iris$Species)
pairs(iris[1:4], main="Iris Data
(red=setosa,green=versicolor,blue=virginica)",
pch=21,
bg=c("red","green3","blue")[unclass(iris$Species)])
```



```
index = sample(nrow(iris), floor(nrow(iris) *
0.7))
train = iris[index,] test = iris[-index,]
xTrain = train[,-5]
yTrain = train$Species
xTest = test[,-5]
yTest = test$Species
model =
train(xTrain,yTrain,'nb',trControl=trainControl(
method='cv',number=10))
model
> model
Naive Bayes
105 samples
 4 predictor
 3 classes: 'setosa', 'versicolor', 'virginica'
No pre-processing
Resampling: cross-validated (10 fold)
Summary of sample sizes: 96, 95, 94, 93, 94, 94, ...
Resampling results across tuning parameters:
usekernel Accuracy Kappa
FALSE     0.9336364 0.9003204
TRUE      0.9436364 0.9155895
Tuning parameter 'fL' was held constant at a value of 0
Tuning parameter 'adjust' was held
constant at a value of 1
Accuracy was used to select the optimal model using the largest value.
The final values used for the model were fL = 0, usekernel = TRUE and adjust = 1.
> |
prop.table(table(predict(model$finalModel,xTest)$class,yTest))

> prop.table(table(predict(model$finalModel,xTest)$class,yTest))
           yTest
setosa    versicolor  virginica
setosa     0.3111111 0.0000000 0.0000000
versicolor 0.0000000 0.2666667 0.0222222
virginica  0.0000000 0.0000000 0.4000000
> |
```

K-Nearest Neighbor  
df <- data(iris)  
head(iris)

```

> head(iris)
  Sepal.Length Sepal.Width Petal.Length Petal.Width Species
1          5.1         3.5          1.4         0.2   setosa
2          4.9         3.0          1.4         0.2   setosa
3          4.7         3.2          1.3         0.2   setosa
4          4.6         3.1          1.5         0.2   setosa
5          5.0         3.6          1.4         0.2   setosa
6          5.4         3.9          1.7         0.4   setosa
> |
ran <-sample(1:nrow(iris), 0.9 * nrow(iris))
nor <-function(x) { (x - min(x))/(max(x)-
min(x)) }
iris_norm <-
as.data.frame(lapply(iris[,c(1,2,3,4)], nor))
summary(iris_norm)
> summary(iris_norm)
  Sepal.Length     Sepal.Width     Petal.Length     Petal.Width    
Min.   :0.0000   Min.   :0.0000   Min.   :0.00000   Min.   :0.00000  
1st Qu.:0.2222  1st Qu.:0.3333  1st Qu.:0.1017   1st Qu.:0.08333  
Median :0.4167  Median :0.4167  Median :0.5678   Median :0.50000  
Mean   :0.4287  Mean   :0.4406  Mean   :0.4675   Mean   :0.45806  
3rd Qu.:0.5833  3rd Qu.:0.5417  3rd Qu.:0.6949   3rd Qu.:0.70833  
Max.   :1.0000  Max.   :1.0000  Max.   :1.00000  Max.   :1.00000 
> |
iris_train <-iris_norm[ran,]
iris_test <-iris_norm[-ran,]
iris_target_category <- iris[ran,5]
iris_test_category <- iris[-ran,5]
library(class)
pr <-
knn(iris_train,iris_test,cl=iris_target_category,k
=13)
tab <-table(pr,iris_test_category)
accuracy <-
function(x){ sum(diag(x)/(sum(rowSums(x)))) *
100}
accuracy(tab)
> accuracy(tab)
[1] 93.33333
> |

```

## Practical No: 9

**Aim:** - Implementation and analysis of Apriori Algorithm using Market Basket Analysis.

**Theory:** -

**Apriori algorithm** is used for finding frequent itemsets in a dataset for boolean association rule. Name of the algorithm is Apriori because it uses prior knowledge of frequent itemset properties. We apply an iterative approach or level-wise search where k-frequent itemsets are used to find k+1 itemsets.

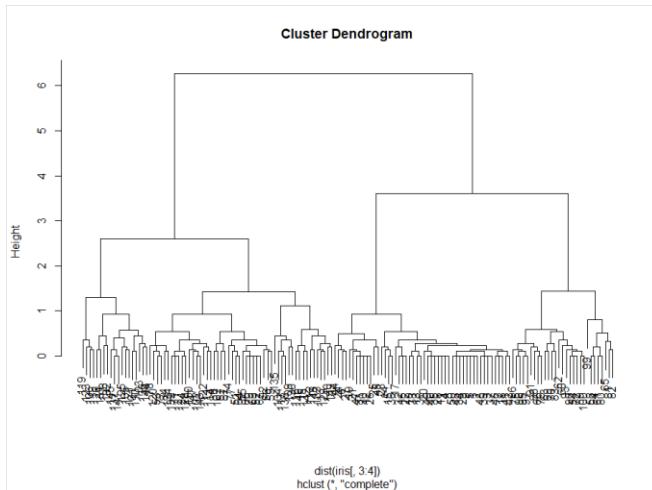
### R Commands:

```
setwd("D:/13_Fymca")
getwd()
mba_data<-read.csv("data_apriori.csv")
trans<-
split(mba_data$Products, mba_data$Customer_I
d, "transactions")
head(trans)
> head(trans)
$`1`
[1] "bread" "butter" "eggs"   "milk"
$`2`
[1] "beer"   "bread"  "cheese" "chips"  "mayo"   "soda"
$`3`
[1] "bread"  "butter" "eggs"   "milk"   "oranges"
$`4`
[1] "bread"  "butter" "eggs"   "milk"   "soda"
$`5`
[1] "buns"   "chips"  "beer"    "mustard" "pickels" "s
$`6`
[1] "bread"  "butter" "chocolate" "eggs"   "mil
```

```
install.packages("arules")
library(arules)
rules=apriori(trans, parameter =
list(support=0.5, confidence=0.9, maxlen=3, minl
en=2))
inspect(rules)
```

	lhs	rhs	support	confidence	coverage	lift	count
[1]	{eggs}	=> {milk}	0.6000000 1	0.6000000 1.6666667	0.6000000 1.6666667	9	
[2]	{milk}	=> {eggs}	0.6000000 1	0.6000000 1.6666667	0.6000000 1.6666667	9	
[3]	{butter}	=> {bread}	0.6000000 1	0.6000000 1.2500000	0.6000000 1.2500000	9	
[4]	{butter, eggs}	=> {milk}	0.5333333 1	0.5333333 1.6666667	0.5333333 1.6666667	8	
[5]	{butter, milk}	=> {eggs}	0.5333333 1	0.5333333 1.6666667	0.5333333 1.6666667	8	
[6]	{bread, eggs}	=> {milk}	0.5333333 1	0.5333333 1.6666667	0.5333333 1.6666667	8	
[7]	{bread, milk}	=> {eggs}	0.5333333 1	0.5333333 1.6666667	0.5333333 1.6666667	8	
[8]	{butter, eggs}	=> {bread}	0.5333333 1	0.5333333 1.2500000	0.5333333 1.2500000	8	
[9]	{bread, eggs}	=> {butter}	0.5333333 1	0.5333333 1.6666667	0.5333333 1.6666667	8	
[10]	{butter, milk}	=> {bread}	0.5333333 1	0.5333333 1.2500000	0.5333333 1.2500000	8	
[11]	{bread, milk}	=> {butter}	0.5333333 1	0.5333333 1.6666667	0.5333333 1.6666667	8	

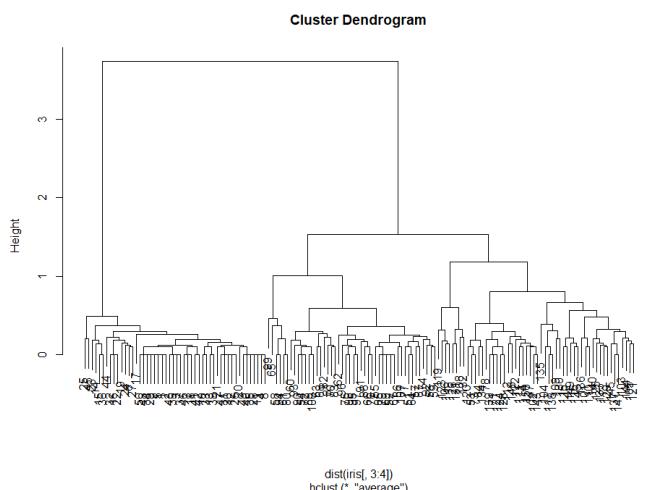




```
clusterCut<- cutree(clusters, 3)
table(clusterCut, iris$Species)
> table(clusterCut, iris$Species)

clusterCut setosa versicolor virginica
      1      50          0      0
      2       0         21     50
      3       0        29      0
```

```
> |
clusters <- hclust(dist(iris[, 3:4]), method =
'average')
plot(clusters)
```



```
clusterCut <-cutree(clusters, 3)
table(clusterCut, iris$Species)
ggplot(iris, aes(Petal.Length, Petal.Width, color
= iris$Species)) +geom_point(alpha = 0.4, size
= 3.5) + geom_point(col = clusterCut) +
scale_color_manual(values = c('black','red',
'green'))
```

