



YARN

Resource
manages
Application
manages
containers

7/10/2023

* R programming

• Statistical Measures

⇒ Descriptive ; Inferential

1. Central tendency or moments (eg. Mean, mode, median)
2. Measures of positions → Quantiles (fractions)
3. " " Distribution → Skewness and kurtosis
4. " " Dispersion (Spread) → Variance, std. deviation, Co Variance, correlation Coeff

⇒ Inferential

• Probability

eg: population, sample

→ P-test

→ t-test

→ z-test

→ chi square test

* installing packages in R

install.packages("moments")

library(moments)

> x = c(4, 1, 8, 9, 10)

> min(x)

> max(x)

> sort(x)

> range(x)

> var(x)

> sd(x)

> cov(x, y)

> cor(x, y)

> skewness(x)

> kurtosis(x)

> mean(x)

> median(x)

⇒ mode has no predefined function in 'R'

> quantile(x, probs = 0.10)

Quantile = $\frac{\text{fraction per } 100}{(n+1)}$

Q₁ = quantile(x, probs = 0.25) → quartile

quantile(x, probs = 0.50) → median

Q₃ = quantile(x, probs = 0.75) → quartile

> IQR(Q₃ - Q₁)

> data() // display all predefined datasets

> data(iris) // iris dataset

> str(iris) // gives attributes

> summary(iris) // gives summary

> `iris[1:150,]` // entire dataset
 > `iris[1:10,]` // 10 rows
 > `iris[1:150, -5]` // displays only 4 columns
 > `iris[1:150, c(-4, -5)]` // removes last two col,
 // more than one value should
 give in vector

Programs

1. Linear Regression

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

> `y = c(2, 4, 6, 8)`

> `z = data.frame(x, y)`

> `lr = lm(y ~ x,`
`data = z)`

// linear model is predefined function

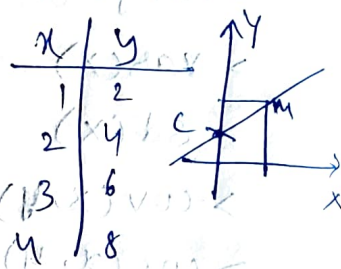
// multiple \Rightarrow `lm(y ~ (x1 + x2)`

> `print(lr)` // gives summary

> `new = data.frame(x = 5)`

> `p = predict(lr, new)`

> `print(p)`



2. Multiple Regression

3. Decision Tree Model (DTM) recursive partition regression trees

```

> install.packages("rpart")
> library(rpart)
> library(rpart.plot)

> x = iris[1:140,]
> y = iris[141:150,]

> dtm = rpart(Species ~ (sepal.length + sepal.width +
                        petal.length + petal.width),
              data = x, method = "class")
  
```

dtm = rpart(Species ~ , data = x, method = "class")

```

> plot(dtm)
> text(dtm) // gives summary based on branch
> P = predict(dtm, y[1:5], type = "class")
  
```

```
> print(P)
```

```
> table(y[1:5], P)
```

	S	vc	v
S	0	0	0
vc	0	0	0
v	0	0	10

• Random forest : iterative of multiple DTM's

14/07

Scala

⇒ Scalable Language

Hadoop 1.0	Hadoop 2.0	Hadoop 3.0
HDFS + MapReduce	HDFS + MapReduce + YARN	HDFS + MapReduce + Spark → load balancing
modules developed by R/Python		modules developed by: Scala

Scala Programs

Syntax

object main

{

def main (args: Array[String])

{

//

}

}

class classname

{

def f1()

{

//

def f2()

{

//

}

- Val: Value which is constant
- Var: Variable which is not fixed

Syntax: Val x: Int = 95; → immutable

Var x: Int = 100; → mutable

→ object declaration

Syntax: Var obj = new classname();

* object mca

{

def main (args: Array[String])

{

var x: Int = 50;

var y: Double = 49.99;

var z: String = "Rani";

Print(x + " " + y + "(" + z + ")");

}

* Reading input directly from keyboard

var x: Int = scala.io.StdIn.readLine.toInt;

var y: String = scala.io.StdIn.readLine;

Object peekedu

{

def main (args: Array[String])

{

var z: Double = scala.io.StdIn.readLine.toDouble;

Print(z);

}

}

* Static Single dimensional array

Object arraykoushik

d

```
def main (args : Array [String])
```

```
{
```

```
    var x = Array (3, 4, 5, 8, 6);
```

```
    var y = Array (Array (1, 2, 3, 4),  
                   Array (4, 3, 2, 1));
```

```
    // var x: Array [Int] = Array (1, 2, 3, 4);
```

```
    print
```

```
    for (i ← 0 to 4)
```

```
    {
```

```
        print (x(i));
```

```
    }
```

```
    for (i ← 0 to 1)
```

```
    {
```

```
        for (j ← 0 to 4)
```

```
        {
```

```
            print (y(i)(j));
```

```
        }
```

```
    }
```

```
}
```

```
}
```

* Dynamic Single D & Two D array

object nasawnensarta

{

```
def main (args: Array[String])
```

```
{
```

```
var x = Array[Int](5);
```

```
for (i ← 0 to 4)
```

```
{
```

```
  x(i) = scala.io.StdIn.readLine.toInt;
```

```
}
```

```
for (i ← 0 to 4)
```

```
{
```

```
  println(x(i));
```

```
}
```

```
var y = Array.ofDim[Int](2, 5);
```

```
for (i ← 0 to 1)
```

```
{
```

```
  for (j ← 0 to 4)
```

```
  {
```

```
    y(i)(j) = scala.io.StdIn.readLine.toInt;
```

```
  }
```

```
for (i ← 0 to 1)
```

```
{
```

```
  for (j ← 0 to 4)
```

```
  {
```

```
    println(y(i)(j));
```

```
  }
```

```
}
```

```
}
```

```
}
```

```
for (i ← 0 to 1;
```

```
      j ← 0 to 4)
```

```
{
```

```
  print(y(i)(j));
```

```
}
```


Exercise

- Create method overloading
- constructor overloading
- simple inheritance
- method overriding

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- Real time appl. for method overloading & constr. overloading

11/05

* Spark

→ Hadoop 3.0

• Spark Ecosystem :-

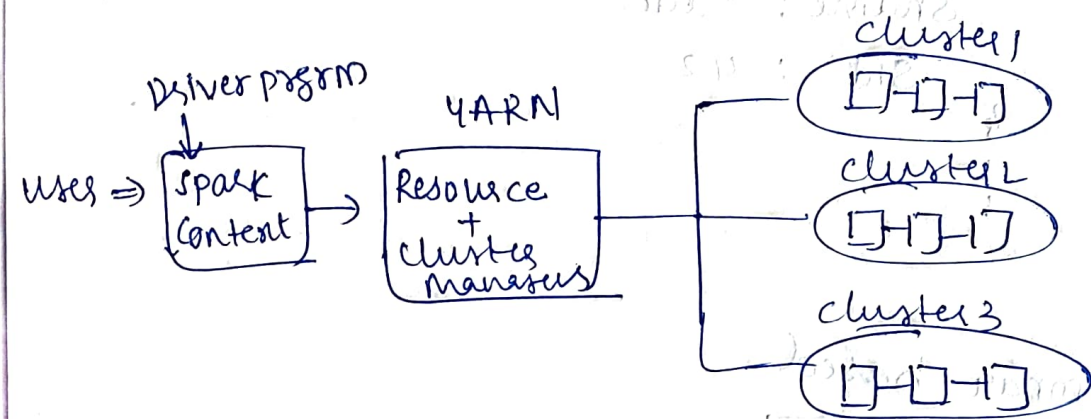
Data management	O33ie	zookeeper	R, Java Python, Scala
Data Accessing	Spark MLlib	Spark Streaming	Spark SQL
Data Processing	Map Reduce	Spark → RDD Case	
Data Storage	Rdbms	Spark HBASE	

• RDD : Re-silent Distributed databases

↳ Spark Context

• Re-silent \Rightarrow fault-tolerance

\rightarrow In-memory processing is done in Spark



* NOSQL

\rightarrow Not Only SQL Databases

1. Key - Value based

2. columnar - based

3. Graph - based

4. Document - based

11/03

\rightarrow BASE properties

BA \rightarrow Basically Available

S \rightarrow soft state

E \rightarrow Eventually consistency

1. Key - value based

Relation : attributekey = "value"

e.g Student : "Sname" = "xyz"

Student : "sid" = 45

2. Document - based

→ json / bson

student

{

{

sname: "adc"

sid : 42

}

}

Bson

either 1 or 0

3. columnar - based

→ Dynamically Create column for the row we needed instead of creating a column for entire table like in RDBMS

↙
row-based Segregation

abc	48	9888	
xyz	49	8568	9.6

column - based

abc	48	9888
xyz	49	8568

- row based takes more traversal time than column based
- row based → can increase as many columns as we need