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Using R execute the basic commands, array list & frames.

* R-command prompt.

```
> text <- "Welcome R!"  
> print(text)  
[1] "Welcome R!"
```

* Datatypes

Datatype	Example	Verify
Logical	True, False	v <- TRUE print(class(v)) it produces the following result [1] "logical"
Numeric	12.3, 5, 999	v <- 23.5 print(class(v)) it produces the following o/p: [1] "numeric"
Integer	2L, 34L, 0L	v <- 2L print(class(v)) It produces the following o/p [1] "integer"

Complex	$3+2i$	$v \leftarrow 2+5i$ <code>print(class(v))</code> it produces the following o/p [1] "complex"
Character	"i", "good" "True"	$v \leftarrow "True"$ <code>print(class(v))</code> it produces the following o/p [1] character
Raw	"Hello" is stored as 48, 65, 6c 6c 6F	$v \leftarrow \text{charToRaw} ("Hello")$ <code>print(class(v))</code> it produces the following o/p [1] "raw"

* Vectors .

```
> # Create A vector
> car_model <- c("Skoda Rapid", "Nissan GTR", "Toyota", "Hyundai")
> print(car_model)
```

O/P :

```
[1] "Skoda Rapid", "Nissan GTR", "Toyota"
[4] "Hyundai"
```

* Lists

```
> # Create a list
> mylist <- list(x = c(1,5,7), c(4.4,2.5,6), max)
> print(mylist)
```

O/p:

[1] 1 5 7

[2]

[1] 4.4 25.0 6.0

[3]

function (..., na.rm = FALSE). primitive ("max")

Matrices

> # Create a Matrix

> M = matrix (c ('A', 'B', 'C', 'D', 'E', 'F', 'G', 'H', 'I'), nrow=3, ncol=3, byrow=TRUE)

O/P :-

	[, 1]	[, 2]	[, 3]
[1,]	"A"	"B"	"C"
[2,]	"D"	"E"	"F"
[3,]	"G"	"H"	"I"

array

> # Create an Array

> A <- array (c ('DN', 'DF'), dim = c(3,3,2))

> print(A)

O/P:

,	,1	[,1]	[,2]	[,3]
[,1]		"ON"	"OFF"	"ON"
[,2]		"OFF"	"ON"	"OFF"
[,3]		"ON"	"OFF"	"ON"
,	2			

,	,1	[,1]	[,2]	[,3]
[,1]		"OFF"	"ON"	"OFF"
[,2]		"ON"	"OFF"	"ON"
[,3]		"OFF"	"ON"	"OFF"

* Factors

```
> # Create a vector  
> permission <- c('Read', 'Write', 'Execute')  
>  
> # Create a Factor  
> F_permission <- factor(permission)  
>  
> # print the factor  
> print(F_permission)
```

O/P:

```
> print(F_permission)  
[1] Read Write Execute
```

Levels: Execute Read Write

```
>
> print(levels(f_permission))
[1] 3
```

* Data frames

```
> # Create database
> stud <- data.frame(
+ sno = c(101, 102, 103, 104, 105),
+ name = c("Madhuaya", "Reyansh", "Tarush",
+ "Saloni", "Idhant"),
+ age = c(22, 23, 26, 23, 24));
>> print(stud)
```

O/P :-

	Sno.	name	age
1	101	Madhuaya	22
2	102	Reyansh	23
3	103	Tarush	26
4	104	Saloni	23
5	105	Idhant	24

* R-Operators

i) Arithmetic operators

Operator	Description	Example
+	Add 2 Vectors	<pre>> a <- c(1, 4, 6) > b <- c(13, 44, 57)</pre> <p>It produces the following result:</p> <pre>> #print(a+b) [1] 14 48 63</pre>
*	Multiply 2 vectors	<pre>> a <- c(1, 4, 6) > b <- c(13, 44, 57)</pre> <p>It produces the following result</p> <pre>> print(a*b) [1] 13 176 342</pre>
/	Division of 2 vectors	<pre>> a <- c(1, 4, 6) > b <- c(13, 44, 57)</pre> <p>It produces the foll. result.</p> <pre>> print(a/b) [1] 0.07 0.09 0.105</pre>
-	Subtract one vector from another	<pre>> a <- c(1, 4, 6) > b <- c(13, 44, 57)</pre> <p>It produces the following result.</p> <pre>> print(b-a) [1] 12 40 5</pre>
% %	Give the remainder of the first vector with the 2nd.	<pre>> a <- c(1, 4, 6) > b <- c(13, 44, 57)</pre> <p>It produces the following result</p> <pre>> print(a %% b) [1] 1 4 6</pre>

$\% / \%$	The result of division of first vector with second(vector) intent)	<pre>> a <- c(1, 4, 6) > b <- c(13, 44, 57) It produces the following result. > print(a % / % b) [1] 0 0 0</pre>
$^$	The first vector raised to exponent of 2nd vector	<pre>> a <- c(1, 4, 6) > b <- c(13, 44, 57) It produces the following result. > print(a ^ b) [1] 1.0000e+00 3.04485e+26 2.26267e+44</pre>

Relational Operators

Operator	Description	Example
$<$	<p>checks if each element of the first vector is less than the corresponding element of the second vector</p>	<pre>> a <- c(12, 34, 44) > b <- c(65, 4, 16) It produces the following result. > print(a < b) [1] TRUE FALSE FALSE</pre>

~~corresponding element of the second vector~~

>	Checks if each element of the first vector is greater than corresponding element of second vector	> a <- c(12, 34, 44) > b <- c(65, 4, 16) It produces the following result > print(a > b) [1] FALSE TRUE TRUE
==	Checks if each element of the first vector is less than or equal to the corresponding element of the 2nd vector	> a <- c(12, 34, 44) > b <- c(65, 4, 16) It produces the following result > print(a == b) [1] FALSE FALSE FALSE
<=	Checks if each element of the first vector is less than or equal to the corresponding element of the 2nd vector	> a <- c(12, 34, 44) > b <- c(65, 4, 16) It produces the following result. > print(a <= b) [1] TRUE FALSE FALSE
>=	Checks if each element of the first vector is greater than or equal to the corresponding element of the 2nd vector	> a <- c(12, 34, 844) > b <- c(65, 4, 16) It produces the following result > print(a >= b) [1] FALSE TRUE TRUE
!=	Checks if each element of the first vector is unequal to the corresponding element of the 2nd vector	> a <- c(12, 34, 44) > b <- c(65, 4, 16) It produces the following result > print(a != b) [1] TRUE TRUE TRUE

Operator	Description	Example
&	<p>It is called Element-wise logical AND operator. It combines each element of the first vector with the corresponding element of 2nd vector & gives an output true if both the elements are TRUE</p>	<pre>>a <- c(12,34,TRUE,2+3i) >b <- c(65,4, FALSE,2+3i)</pre> <p>It produces the following result:-</p> <pre>>print(a&b) [1] TRUE TRUE FALSE TRUE</pre>
	<p>It is called Element wise logical OR operator. It combines each element of the 1st vector to the corresponding element of the 2nd vector. & gives an O/p 'true' if one element is 'true'. logical</p>	<pre>>a <- c(12,34,TRUE,2+3i) >b <- c(65,4, FALSE,2+3i)</pre> <pre>>print(a b) [1] TRUE TRUE TRUE TRUE</pre>
!	<p>It is called NOT operator. Takes each element of the vector & gives the opposite logical value</p>	<pre>>a <- e(12,34,TRUE,2+3i) >b <- c(65,4, FALSE,2+3i)</pre> <p>It produces the following result</p> <pre>>print(!b) [1] FALSE FALSE TRUE FALSE</pre>

Operators
:

%in%

Description

colon operator. It generates the series of numbers in sequences for a vector

This operator is used to identify if an element belongs to a vector

Example

```
> a <- 3:6  
> print(a)  
[1] 3 4 5 6
```

```
> a <- 5  
> b <- 10  
> t <- 1.5  
> print(a %in% t)  
[1] TRUE  
> print(b %in% t)  
[1] FALSE
```

Create a Matrix using R and Perform the operations addition, inverse, transpose & multiplication operations

Matrices

Syntax :

`matrix(data, nrow, ncol, byrow, dimnames)`

Example :-

> # Elements are arranged sequentially by rows
 > x <- matrix(c(1:10), nrow=2, byrow=TRUE)
 > print(x)

```
[,1] [,2] [,3] [,4] [,5]
[1,] 1     2     3     4     5
```

```
[2,] 6     7     8     9     10
```

> # Elements are arranged sequentially by column

> y <- matrix(c(1:10), nrow=2, byrow=FALSE)
 > print(y)

```
[,1] [,2] [,3] [,4] [,5]
[1,] 1     3     5     7     9
[2,] 2     4     6     8     10
```

> # Define the column & rows names

> rownames = c("row1", "row2")

> colnames = c("col1", "col2", "col3", "col4", "col5")

```

> P <- Matrix( c(1:10), nrow = 2, byrow = TRUE,
  dimnames = list(rownames, colnames))
> print(p)
      col1    col2    col3    col4    col5
row1     1       2       3       4       5
row2     6       7       8       9      10

```

Accessing Elements of a matrix

```

> # Create matrix P
> p <- matrix(c(10:21), nrow = 4, byrow = TRUE)

```

```

> print(p)
      [,1]    [,2]    [,3]
[1,] 10      11      12
[2,] 13      14      15
[3,] 16      17      18
[4,] 19      20      21
>

```

> # Access the element at 3rd row & 2nd column

```

> print(p[3,2])
[1] 17

```

> # Access the element at 4th row

```

> print(p[4,])
[1] 17 20 21

```

> # Access the element at 4th row & 2nd col

```

> print(p[4,2])
[1] 20

```

* Matrix Addition & Subtraction

> # Create Matrix P & Q

> P <- matrix(c(10:21), nrow = 4, byrow = TRUE)

> Q <- matrix(c(1:12), nrow = 4, byrow = TRUE)

>

> # Addition

> add = P + Q

>

> # Subtraction

> sub = P - Q

> > # Print Both Matrix

> print(P)

	[,1]	[,2]	[,3]
[1,]	10	11	12
[2,]	13	14	15
[3,]	16	17	18
[4,]	19	20	21

> print(Q)

	[,1]	[,2]	[,3]
[1,]	1	2	3
[2,]	4	5	6
[3,]	7	8	9
[4,]	10	11	12

> # Addition Matrix

> print(add)

	[, 1]	[, 2]	[, 3]
[1,]	11	13	15
[2,]	17	19	21
[3,]	23	25	27
[4,]	29	31	33

> # Subtraction Matrix
 > print(sub)

	[, 1]	[, 2]	[, 3]
[1,]	9	9	9
[2,]	9	9	9
[3,]	9	9	9
[4,]	9	9	9

* Multiplication & Division

> # print both Matrix

> print(p)

	[, 1]	[, 2]	[, 3]
[1,]	10	11	12
[2,]	13	14	15
[3,]	16	17	18
[4,]	19	20	21

> print(q)

	[, 1]	[, 2]	[, 3]
[1,]	1	2	3
[2,]	4	5	6
[3,]	7	8	9
[4,]	10	11	12

> #Addition Matrix

> print(multi)

[, 1] [, 2] [, 3]

[1,] 10 22 36

[2,] 52 70 90

[3,] 112 136 162

[4,] 190 220 252

>

> #Subtraction Matrix

> print(div)

[, 1] [, 2] [, 3]

[1,] 10.00000 5.50000 4.00

[2,] 3.25000 2.80000 2.50

[3,] 2.285714 2.12500 2.00

[4,] 1.90000 1.818182 1.75



Using R Execute the statistical functions: mean, median, mode, quartiles, range, inner quartile range histogram.

* Mean

Syntax :

```
mean(x, trim=0, na.rm=FALSE, ...)
```

Example

```
> # Create a Vector  
> v <- c(10, 3, 25, 6, 22, -33)  
> # find Mean  
> result <- mean(v)  
> print(result)  
[1] 5.5
```

Apply TRIM option

```
> # Create a vector  
> v <- c(10, 3, 25, 6, 22, -33)  
> # find mean  
> result <- mean(v) trim = 0.2  
> print(r)  
[1] 1.6
```

Applying NA option

```
> # Create a vector  
> v <- c(10, 3, 25, 6, 22, -33, -44, -44, NA)  
> # find Mean  
> r <- mean(v)
```

> print(g)

[1] NA

> g1 <- mean(v, na.rm=TRUE)

> print(g1)

[1] -1.571429

* Median

Syntax :-

median(x, na.rm = FALSE)

Example :-

> # Create a vector

> v <- c(10, 3, 25, 6, 22, -33, -44)

> # find Median

> g1 <- median(v)

> print(g1)

[1] 6

* Mode

> # Create the function

> getmode <- function(v)

{

+ unqv <- unique(v)

+ unqv[which.max(tabulate(match(v, unqv)))]

}

> # Create the no. vector

> v <- c(2, 3, 3, 4, 4, 1, 2, 3)

```
## calculate the mode using function  
> g <- getmode(v)  
> print(g)  
[1] 3  
> char v <- c("R", "(", "++", "R", "(++)")  
> ## calculate the mode using function  
> g <- getmode(char v)  
> print(g)  
[1] "R"
```

* Histogram

Syntax :-

```
hist(v, main, xlab, xlib, ylim, breaks, w1, border)  
> ## Create data for graph  
> v <- c(3, 4, 5, 23, 4, 5, 66)  
> png(file = "Histogram.jpg")  
> hist(v, xlab = "Weight", col = "red", border = "yellow")  
> dev.off()  
null device  
1  
> hist(v, xlab = "Weight", col = "red", border = "yellow")  
> 1
```

> Create Histogram.

```
> hist(v, xlab = "Weight", col = "red", border = "blue", xlim =  
  c(0, 60), ylim = (0, 7), breaks = 3)
```

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Date : | | | |

A hand-drawn histogram on lined paper. The vertical axis is labeled "Frequency" and ranges from 0 to 5. The horizontal axis is labeled "weight" and has tick marks at 20, 40, and 60. There are three bars: one bar between 20 and 40 with height 1, one bar between 40 and 60 with height 1, and one bar between 60 and 80 with height 1.

Weight Range	Frequency
20 - 40	1
40 - 60	1
60 - 80	1

A hand-drawn histogram on lined paper. The vertical axis is labeled "Frequency" and ranges from 0 to 6. The horizontal axis is labeled "weights" and has tick marks at 10, 20, 30, 40, 50, and 60. A red line starts at approximately (10, 2) and ends at approximately (60, 5). There is one bar between 20 and 30 with height 1.

Weights Range	Frequency
20 - 30	1

Scanned by CamScanner

Using R import the data from Excel / CSV file & perform =
the above func.

* Input as CSV file

input.csv

R.no	Age	name	dept
101	26	Saloni	IT
102	25	Ishant	HR
103	24	Rayansh	Finance
104	24	Tarush	IT
105	25	Madhurya	HR

* Reading a CSV file

```
> # first set dictionary of csv file location  
> setwd ("C:/R")  
> d <- read.csv ("input.csv")  
> print(d)
```

O/P:

	R.no.	Age	name	Dept
1	101	26	Saloni	IT
2	102	25	Ishant	HR
3	103	24	Rayansh	Finance
4	104	24	Tarush	IT
5	105	25	Madhurya	HR

* Analysing the CSV file

```
> d <- read.csv("input.csv")
> print(is.data.frame(d))
[1] TRUE
> print(ncol(d))
[1] 4
> print(nrow(d))
[1] 5
```

* Get the maximum age

```
> # Display max age from data frame
> data <- read.csv("input.csv")
> max.age <- max(data$age)
> print(max.age)
[1] 26
```

* Mean , Median , Mode

values.csv

A	B
12	45
3	76
18	22
9	11

```
> mean(d$A)
[1] 10.5
> median(d$A)
[1] 10.5
> getmode <- function(v)
+ {
+   unqv <- unique(v)
+   unique[which.max(tabulate(match(v, unqv)))]
+   y
> v <- c(2,3,3,4,4,1,2,3)
> g2 <- getmode(v)
> print(g2)
[1] 3
> charv <- c("R", "C.", "(++", "R", "(++)")
> g2 <- getmode(charv)
> print(g2)
[1]"R"
```

PRACTICAL 9

Using R perform binomial & normal distribution on the data.

* Normal distribution

i) `dnorm()`

> # Create a sequence of -5 to 5 increment by 0.2

> a <- seq(-5, 5, by = 0.2)

> # Choose mean as 1.5 & std deviation .

> b <- dnorm(a, mean = 1.5, sd = 0.5)

> png(file = "Dnorm.png")

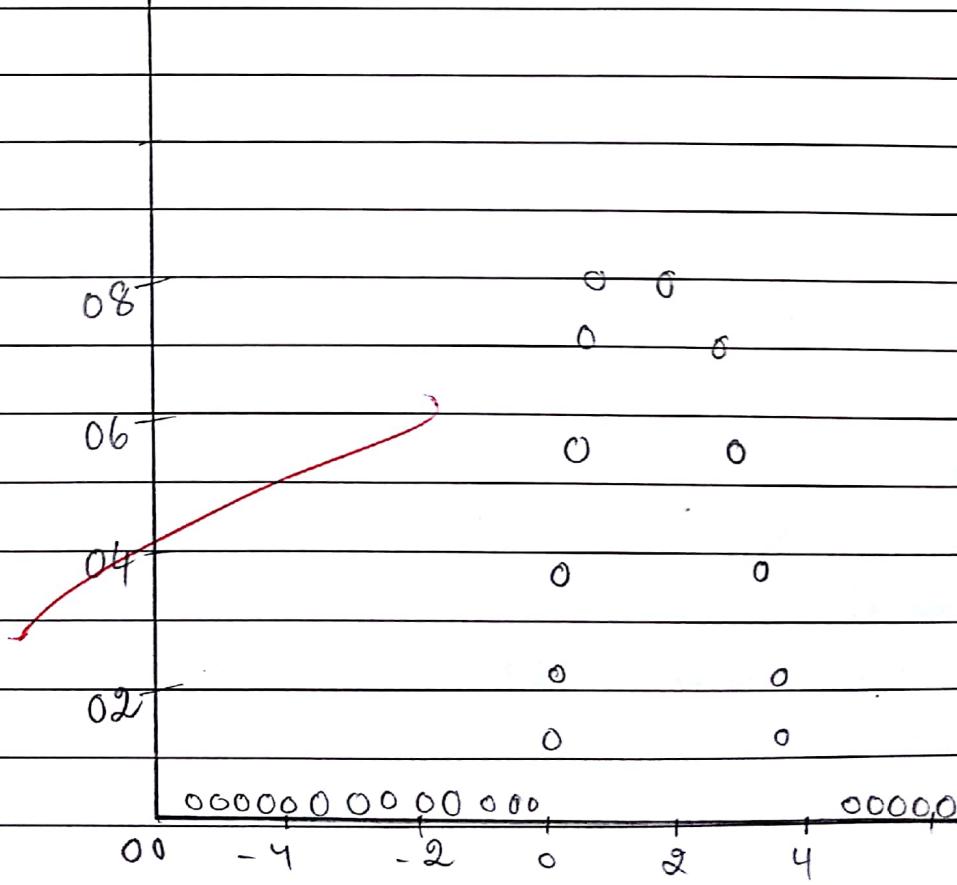
> plot(a, b)

> dev.off()

null device

> plot(a, b)

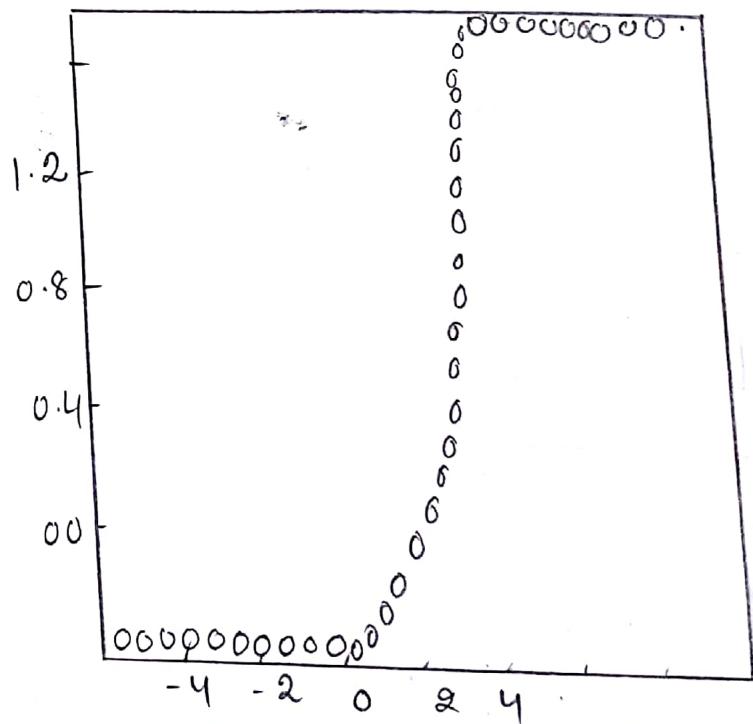
result :-



2) pnorm()

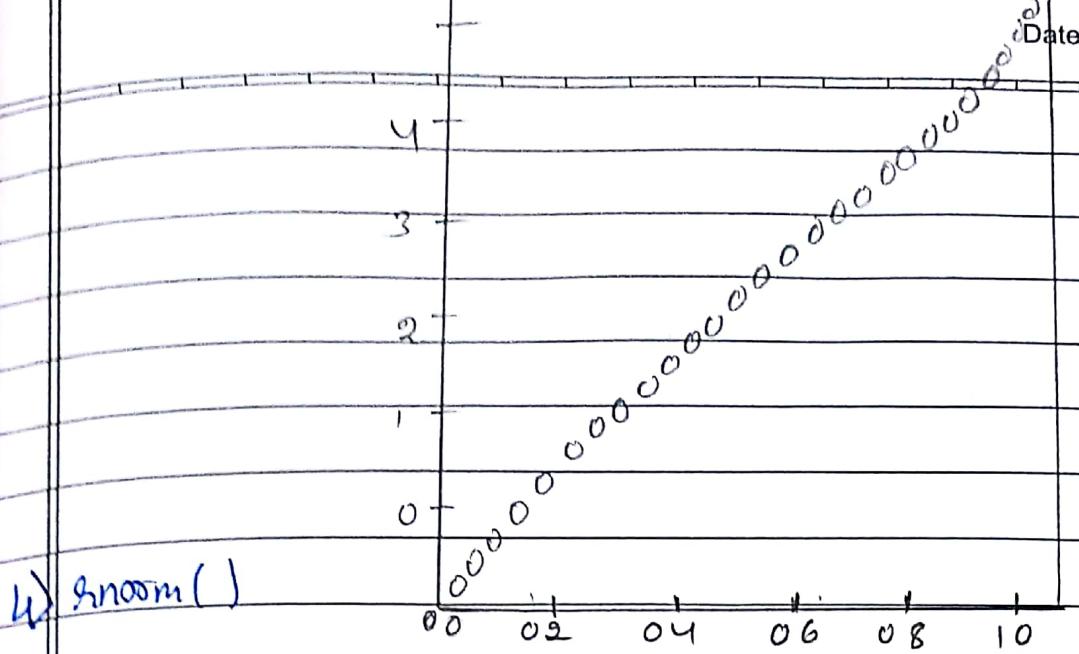
```
> # create a sequence of -5 to 5 increment by 0.2  
> a <- seq(-5, 5, by = 0.2)  
> png(file = "pnorm.png")  
> b <- pnorm(a, mean = 1.5, sd = 0.5)  
> plot(a, b)  
> dev.off()  
null device
```

> plot(a, b)



3) qnorm

```
> # Create a sequence  
> a <- seq(0, 1, by = 0.01)  
> b <- qnorm(a, mean = 2, sd = 1)  
> plot(a, b)
```



> ## Create a sample of 50 nos.

> a <- rnorm(50)

> png(file = "Rnorm.png")

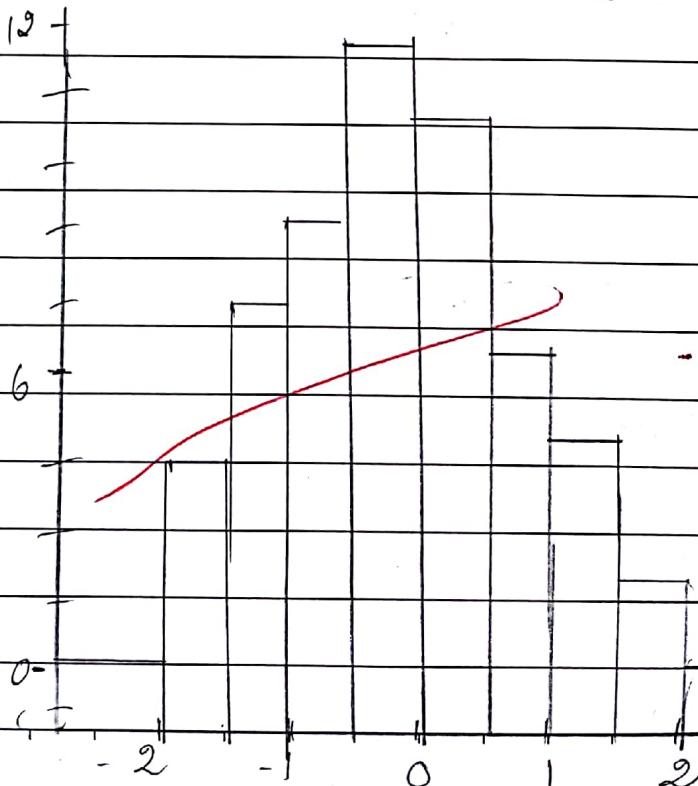
> hist(a, main = "Normal Distribution")

> dev.off()

null device

)

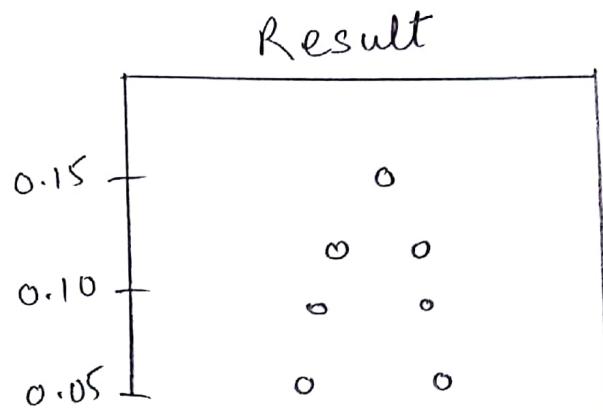
> hist(a, main = "Normal distribution")



* Binomial Distribution

1) dbinom()

```
> # Create a sample of 50 numbers  
> a <- seq(1, 30, by = 2)
```



2) pbinom

```
> # probability  
> a <- pbinom(12, 34, 0.4)  
> print(a)  
[c] 0.354171.
```

3) qbinom(c)

```
> a <- qbinom(0.12, 34, 1/4)  
> print(a)  
[c] 6
```

4) `abinom()`

`> # find 5 random values from`

`> a <- abinom(5, 30, 0.4)`

`> print(a)`

`[1] 10 13 8 9 18.`

Practical 10 perform the linear Regression using R.

• Linear Regression

$$y = ax + b$$

• lm() function

lm(formula, data)
create relation model and get the coffice.

> # Create data of diff length

> v <- c(23, 45, 66, 50)

> u <- c(2, 3, 56)

> r <- lm(v ~ u)

> print(r)

Call:

lm(formula = v ~ u)

Coefficients:

(intercept)

12.7

u

8.7

get the summary of relationship

> # Create data of diff length

> v <- c(23, 45, 66, 50)

> u <- c(2, 3, 5, 6)

> r <- lm(v ~ u)

> print(summary(r))

call:

lm (formula = $\sqrt{v} \sim u$)

Residuals

	1	2	3	4
	7.1	6.2	9.8	-8.9

Coefficients :

	estimate	std.error	t value	p > t
(Intercept)	12.700	15.630	0.813	0.502
u	8.700	3.634	2.394	0.139

Residual standard error 11.49 on 2 DF

Multiple R-squared: 0.7413

Adjusted R-squared: 0.612

F-statistic: 5.732 on 1 & 2 DF

P-value : 0.139

Predict @:

DynLab - (predict object newdata)

predict the value from series

\rightarrow # create data of diff. length.

$\rightarrow v \leftarrow c(23, 45, 66, 58)$

$\rightarrow u \leftarrow c(2, 3, 5, 6)$

$\rightarrow r \leftarrow lm(\sqrt{v} \sim u)$

\rightarrow # find relational value for u from

$\rightarrow s \leftarrow date$ from ($u = 0.4$)

$\rightarrow r \leftarrow lm(\sqrt{v} \sim u)$

$\rightarrow result \leftarrow predict(r, s)$

$\rightarrow print(result)$

47.5

Visualize the Regression Graphically:

```
> # create predictor and response  
values.  
> m <- c(1, 2, 3, 4, 5, 6)  
> t <- c(25, 22, 30, 34, 45, 26)  
> # Label to chart  
> png(file = "linear.png")  
> plot(t, m, col = "red", main = "maths,  
temperature  
+ abline(lm(m ~ t))  
+ lmx = 1.6,  
+ pch = 19  
+ xlab = "month"  
+ ylab = "temperature")
```

PRACTICAL 11

Computer the least square means using R

. Least square mean

- > instal packages ("lsmeans")
- > setwd ("c:/r folder")
- > library ("lsmeans")

Loading required packages ("lsmean")

- > Instal packages ("lsmean")
- > Setwd ("c:/r folder")
- > library ("lsmeans")

Loading required package estimability.

```
# Add option -quiet example add
> lsmeans <- crosstabs(motsif)
# This creates my own local copy of this function
> lsmeans <- update(lsmeans, . ~ . + nitrof:Variety)
> lsmeans <- lsmeans(lsmeans, list(repairing = Variety ~ nitrof, poly = nitrof ~ Variety))
> lsmeans <- lmmeans(lsmeans, list(poly = nitrof, pairwise = Variety))
> lmmeans(motsif, list(poly = nitrof))

fmeans of nitrof
  Variety   lsmeans    SE df lower.CL upper.CL
  Golden Rain 80.5000000 9.106954224 5 56.58961889 103.91017111
  Marvellous  86.6666667 9.106954224 5 83.25649555 110.07983777
  Victory     71.5000000 9.106954224 5 49.06982059 94.91017111

Results are averaged over the levels of: Variety
Confidence level used: 0.95

f polynomial contrasts of contrast
  contrast   estimate    SE df t.ratio P.value
  linear     147.3333333 13.439539716 5 10.963 <.0001
  quadratic  -10.3333333 6.010344878 5  -1.719  0.0916
  cubic      -2.0000000 13.439539716 5   -0.149  0.8623

Results are averaged over the levels of: Variety
Confidence level used: 0.95
```

```
# # Custom contrasts
> lmmeans(motsif, my.cov ~ Variety,
+         contrast = list(my.cov = list(G.vz.M = c(1, -1, 0), CH.vz.V = c(0, 1, -1))))
> lsmeans
  Variety   lsmeans    SE df lower.CL upper.CL
  Golden Rain 104.5000000 7.79746565 5 84.45591753 124.5540025
  Marvellous 102.7916667 7.70746565 5 80.747150420 129.9357491
  Victory     97.6250000 7.70746565 5 77.50021753 117.6690025

Results are averaged over the levels of: nitrof
Confidence level used: 0.95

contrasts
```

```
Model with interaction, follow-ups at levels of other factors
> lmmeans(motsif.lme, list(repairing = Variety ~ nitrof, poly = nitrof ~ Variety))
> lmmeans(Variety ~ nitrof)
  Variety   lsmeans    SE df lower.CL upper.CL
  Golden Rain 80.5000000 9.106954224 5 56.58961889 103.91017111
  Marvellous  86.6666667 9.106954224 5 83.25649555 110.07983777
  Victory     71.5000000 9.106954224 5 49.06982059 94.91017111

nitrof = 0:
  Variety   lsmeans    SE df lower.CL upper.CL
  Golden Rain 80.5000000 9.106954224 5 56.58961889 103.91017111
  Marvellous  86.6666667 9.106954224 5 83.25649555 110.07983777
  Victory     71.5000000 9.106954224 5 49.06982059 94.91017111

nitrof = 0.2:
  Variety   lsmeans    SE df lower.CL upper(CL)
  Golden Rain 88.5000000 9.106954224 5 75.0000000 122.91017111
  Marvellous 108.5000000 9.106954224 5 85.06981009 131.91017111
  Victory    89.8888889 9.106954224 5 66.23645556 112.71637777

nitrof = 0.4:
  Variety   lsmeans    SE df lower.CL upper(CL)
  Golden Rain 114.6666667 9.106954224 5 91.25649555 139.79583777
  Marvellous 117.1666667 9.106954224 5 93.75649555 140.57637777
  Victory    110.8333333 9.106954224 5 87.42316223 134.24350444

nitrof = 0.6:
  Variety   lsmeans    SE df lower(CL) upper(CL)
  Golden Rain 110.8333333 9.106954224 5 87.42316223 134.24350444
  Marvellous 117.1666667 9.106954224 5 93.75649555 140.57637777
  Victory    110.8333333 9.106954224 5 87.42316223 134.24350444

# The following will produce the above results for Victory only
> lmmeans(motsif.lme, poly = nitrof, fam.redund = functions(covf, lev), covf(3, 1))
NOTE: Results may be misleading due to involvement in interactions.

fmeans
  nitrof   lsmeans    SE df lower(CL) upper(CL)
  0       71.5000000 9.106954224 5 49.06982059 94.91017111
  0.2     89.6666667 9.106954224 5 68.19448146 111.07663777
  0.4     110.8333333 9.106954224 5 87.42316223 134.24350444
  0.6     118.5000000 9.106954224 5 93.75649555 140.57637777

Results are averaged over the levels of: Variety
Confidence level used: 0.95

contrasts
  contrast   estimate    SE df t.ratio P.value
  linear     162.1666667 24.23581697 45  5.575 <.0001
  quadratic -10.5000000 10.8533457 45  -0.988  0.330
  cubic     -16.5000000 24.23581697 45  -0.679  0.5035

Results are averaged over the levels of: Variety
```

I a fake covariance that has some predictive ability
 seed(271028, "default", "default") # for reproducibility of result
 fakeX = sqrt(covsfyield) + rnorm(nrow(motsif), 0, .25)
 cov.lme = update(motsif.lme, . ~ . + poly(fakeX, 1))

PRACTICAL - 12

Computer the linear square Regression

> # Five pairs consists of a year & the mean interest rate :

> Year \leftarrow c(2000, 2001, 2002, 2003, 2004)

> Rate \leftarrow c(9.34, 8.50, 7.62, 6.93, 6.60)

> # find the correlation between the year & mean interest rates.

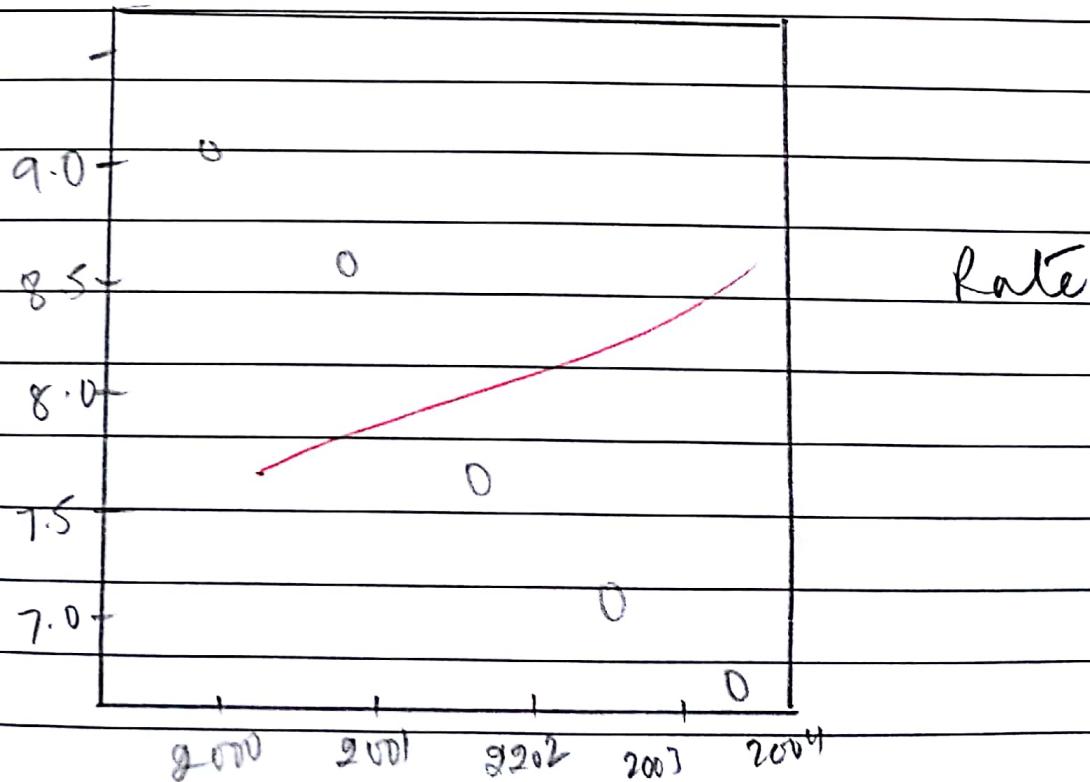
> plot(year, rate).

+ main(" commercial Banks, Interest Rates for year car loan")

url = " http://www.federalreserve.gov/release1of19(2085805)")

> car(cyear, rate)

[1] 0.9880813155.



```

> fit <- lm(rate ~ year)
> fit

Call:
lm(formula = rate ~ year)

Coefficients:
(Intercept)      year
1419.208       0.705

```

you would like to know what else is stored in the variable you can use the attributes command:

```

> attributes(fit)
$names
[1] "coefficients"   "residuals"      "effects"        "rank"
[5] "assign"          "contrasts"     "df.residual"    "xlevels"
[9] "terms"           "model"
$class
[1] "lm"

```

One of the things you should notice is the coefficients variable within fit. You can print out the y-intercept and slope by accessing this part of the variable:

```

> fit$coefficients[1]
(Intercept)
1419.208
> fit$coefficients[[1]]
[1] 1419.208
> fit$coefficients[2]
year
-0.705
> fit$coefficients[[2]]
[1] -0.705
>

```

Note that if you just want to get the number you should use two square braces. So if you want to get an estimate of the interest rate in the year 2015 you can use the formula for a line:

```

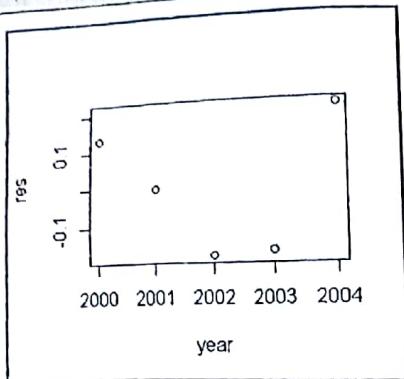
> fit$coefficients[[2]] * 2015 + fit$coefficients[[1]]
[1] -1.367
>

```

1.8 A Practical Approach to R

A better use for this formula would be to calculate the residuals and plot them:

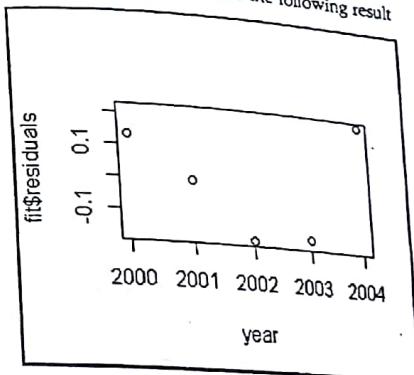
```
> res <- rate - (fit$coefficient[1] * year + fit$coefficient[2])
> res
[1] 0.13200000001 -0.003999999 -0.17800000000 -0.16300000000 0.21200000000
> plot(year,res)
>
```



That is a bit messy, but fortunately there are easier ways to get the residuals. Two other ways are shown below:

```
> residuals(fit)
   1    2    3    4    5
0.132 -0.003 -0.178 -0.163 0.212
> fit$residuals
   1    2    3    4    5
0.132 -0.003 -0.178 -0.163 0.212
> plot(year,fit$residuals)
>
```

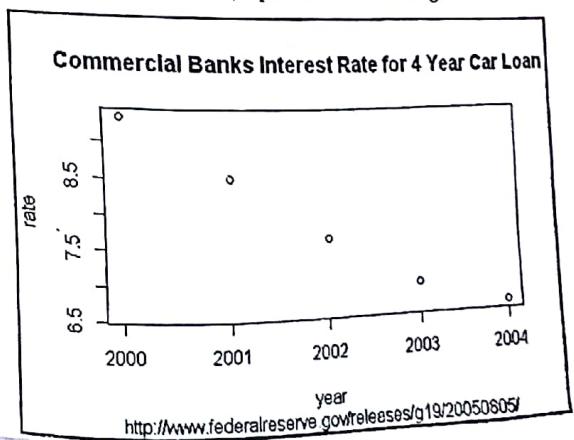
When we execute the above code, it produces the following result



If you want to plot the regression line on the same plot as your scatter plot you can use the abline function along with your variable fit:

```
> plot(year,rate)
+ main="Commercial Banks Interest Rate for 4 Year Car Loan",
+ sub="http://www.federalreserve.gov/releases/g19/20050805/"
+
```

When we execute the above code, it produces the following result



170A Practical Approach to R

Finally, as a teaser for the kinds of analyses you might see later, you can get the results of an F-test by asking R for a summary of the fit variable:

```
> summary(fit)

Call:
lm(formula = rate ~ year)

Residuals:
    1     2     3     4     5 
 0.132 -0.003 -0.178 -0.163  0.212 

Coefficients:
            Estimate Std. Error t value Pr(>|t|)    
(Intercept) 1419.2080000 126.94956528 11.17931 0.0015341 ***
year        -0.7050000  0.06341136 -11.11788 0.0015592 ***
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.2005243 on 3 degrees of freedom
Multiple R-squared:  0.9763047, Adjusted R-squared:  0.9684062 
F-statistic: 123.6073 on 1 and 3 DF,  p-value: 0.001559188
```

PRACTICAL - 8

Page No. _____

Date: / /

- > Import the data from Excel / csv file and the chp - squared Test syntax: chisq.test(data)
- > Install - packages("maiss")
- > library("MASS")
- > print cs @ (cars93)
- > # Create a data frame from the main data
- > car - data: data frame (cars 93 & Air bags Cars 93 & Type)
- > car - data: Table (cars 93 & Airbags, cars 93 & Type)
- > print (car - data)

compact large Midsize small sporty van

driver & Passanger	2	4	7	0	3	0
driver only	9	7	11	5	8	3
None	5	0	4	16	3	6

> print (str) (cars93)

• data frame": 93 obs. of 27 variables \Rightarrow

• Manufacturer: Factor w/ 32 levels 'Audi' "BMW" ...

• Model: Factor w/ 93 levels '100' "190 E" "112" "314" "444" ...

240" ... "49" "56" "91"

62 45 41 47 35

¶ Type: factor w/ levels compact' large, y_3, y_7, y_{12}

min - Parce: num 12.9. 29.2 25.9. 308 21,
14.21, 9.9

22.6 26.3000

\$ Price-: num 15.9 33.9 29.1 37.7 3015,
20.8 23.7 26.3

§ Max, Price num 18.8, 38.7, 32.3,
44.6, 36.9, 17.3, 12.7
24.9 26.3

\$ MPG. City : int 25 18 20 19 22 19 16 16

\$ MPG highway : int 31 25 26 26 30 31 28 25

§ Air bag and factors w/3 levels, gives
.. passengers... 31 21 222, 222,

\$ Drivetrain : factors w/6, levels "4 '40" front -

§ Cylinders : Factors w/6, levels '3 "4" 5" 6" -
2, 2.4 4 4 2 2

\$ Engine size: int. 8 12.28. 28 15.22
3.8, 5.73. 8

\$ Horsepower : int 140 200 172 172 208 16
170 180 170

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Date : 1 1 1

RPM : Int Mo 200 172 172 208 110 170
180 170

Fuel tanks : capacity num 13.2.18 16.9 21.7
21.1 21.116 4 18

Passengers : Pst 5 5 5 6 4 6 6 6 56000

PRACTICAL - 5

Using R repeat the data from Excel
car file and calculate the
standard deviation, variance,
co-variance

Standard Deviation:

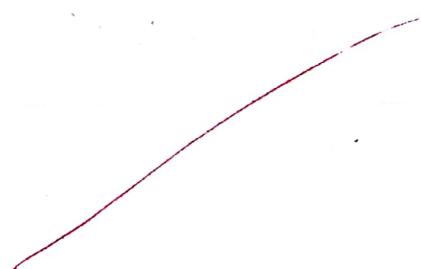
The standard deviation of a observation
variable is the square root of its
variance problem

Find the standard deviation of the
eruption duration in the data
set faithful.

Solution:

We apply the set function to
compute the standard deviation
of eruptions

- > duration = faithful \$ eruption
- # the durations
- > Var (duration)
- [1] 1.3027



co variance .

The co variance of two variables x & y in a data measures how the two are linearly related.

positive covariance would indicate a positive linear relationship between the variables and a negative covariance would indicate the opposite .

The sample covariance is defined in terms of the sample mean as

$$S_{xy} = \frac{1}{n-1} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

Similarly, the population covariance is defined in terms of population mean

μ_x μ_y as :

$$\sigma_{xy} \cdot \text{cov}_{xy} = \frac{1}{N} \sum_{i=1}^N (x_i - \mu_x)(y_i - \mu_y)$$

problem . -

Find the covariance of eruption duration & waiting time in the data set set faithful observe is there is any linear relation ship by the 2 variables .

Solution

We apply the cor function to compute the covariance of eruptions and waiting.

- > duration -> faithful & eruptions
- > Waiting period: faithful & waiting duration
- > cov (duration, waiting) # apply the cor function (t) 12.978.

PRACTICAL 6.

→ using import the data from excel and csv file draw the skewness.

The Skewness of a data population & is self defined by following formula, where U_2 & U_3 are the second & third central moments.

Intuitively the skewness is a measure of symmetric rule the negative skewness indicates that the mean data values is less than the median & the data distribution is left skewed positively skewness would indicate that the mean of the data value is larger than the median, & the data distributed is right:-

Problem:-

Find the skewness of eruption duration is the data faithful.

Solutions :-

We apply the function skewness from elv211 to computer the stenren wofficient of eruption package is not in one lib.
It has to see installed & loaded in R workspace

> Install packages ("Elo71")

> Library (el071)

> duration = faithful & eruptions #
eruptions duration.

skewness (duration)

EIJ - 0.41355.

