

# 1. Run a linear and quadratic discriminant analysis for the Remote Sensing Data.

```
> rm(list=ls())
> crop <- read.csv("~/Documents/Rutgers/Spring 2020/Stat 467/crop.csv")
> View(crop)
> library(MASS)
> set.seed(1601)
> n=nrow(crop)
> n
[1] 36
> ni=sample(n,.5*n,rep=F)
> ni
[1] 1 8 4 20 5 28 12 31 3 17 19 9 29 27 18 23 6 30
> croptr=crop[-ni,]
> croptst=crop[ni,]
> croplda=lda(crop~,data=croptr)
> croplda
Call:
lda(crop ~ ., data = croptr)
```

Prior probabilities of groups:

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
	0.3333333	0.1111111	0.1666667	0.1666667	0.2222222

Group means:

	x1	x2	x3	x4
CLOVER	36.83333	22.83333	32.33333	43.16667
CORN	13.50000	19.00000	23.00000	51.50000
COTTON	31.33333	29.33333	29.00000	35.66667
SOYBEANS	23.33333	34.00000	26.00000	26.33333
SUGARBEET	27.50000	36.75000	18.50000	36.75000

Coefficients of linear discriminants:

	LD1	LD2	LD3	LD4
x1	-0.10881361	-0.06540671	-0.01258370	-0.01702223
x2	-0.01897551	-0.08345353	-0.03360946	0.04872427
x3	-0.07044174	0.04842217	0.01557119	0.04687245
x4	-0.02958988	0.04467764	-0.04251665	0.02852162

Proportion of trace:

	LD1	LD2	LD3	LD4
	0.5334	0.4429	0.0235	0.0001

```
> table(predict(croplda)$class,crop$crop[-ni])
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
CLOVER	5	0	2	1	1
CORN	0	2	0	0	0
COTTON	0	0	0	0	0
SOYBEANS	0	0	1	1	1
SUGARBEET	1	0	0	1	2

```
> table(predict(croplda,newdata=croptst)$class,crop$crop[ni])
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
CLOVER	3	0	2	0	1
CORN	0	1	0	1	0
COTTON	0	0	0	0	0
SOYBEANS	0	4	1	2	1
SUGARBEET	2	0	0	0	0

```

> ni2=sample(n,.2*n,rep=F)
> croptr=crop[-ni2,]
> croptst=crop[ni2,]
> cropqda=qda(crop~,data=crop)
> cropqda
Call:
qda(crop ~ ., data = crop)

```

Prior probabilities of groups:

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
	0.3055556	0.1944444	0.1666667	0.1666667	0.1666667

Group means:

	x1	x2	x3	x4
CLOVER	46.36364	32.63636	34.18182	36.63636
CORN	15.28571	22.71429	27.42857	33.14286
COTTON	34.50000	32.66667	35.00000	39.16667
SOYBEANS	21.00000	27.00000	23.50000	29.66667
SUGARBEET	31.00000	32.16667	20.00000	40.50000

```
> table(predict(cropqda)$class,crop$crop)
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
CLOVER	9	0	0	0	0
CORN	0	7	0	0	0
COTTON	0	0	6	0	1
SOYBEANS	0	0	0	6	1
SUGARBEET	2	0	0	0	4

```
> table(predict(cropqda,newdata=croptst)$class,crop$crop[ni])
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
CLOVER	5	0	0	0	0
CORN	0	5	0	0	0
COTTON	0	0	3	0	0
SOYBEANS	0	0	0	3	1
SUGARBEET	0	0	0	0	1

2. Fit a tree model to the above data.

```
> library(tree)
```

```
> tree=tree(formula=crop~x1+x2+x3+x4,data=crop,na.action=na.exclude,mincut=5,minsize=10,mindev=0.01)
```

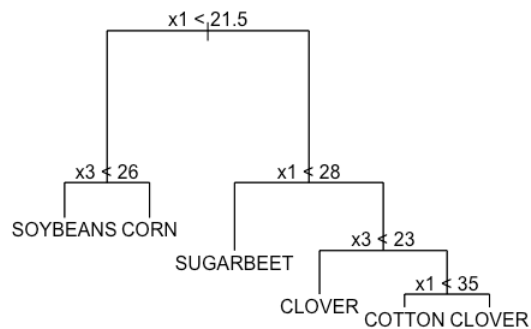
```
> plot(tree)
```

```
> text(tree)
```

node), split, n, deviance, yval, (yprob)

\* denotes terminal node

```
1) root 36 113.500 CLOVER ( 0.30556 0.19444 0.16667 0.16667 0.16667 )
2) x1 < 21.5 11 18.920 CORN ( 0.09091 0.63636 0.00000 0.27273 0.00000 )
4) x3 < 26 5 6.730 SOYBEANS ( 0.00000 0.40000 0.00000 0.60000 0.00000 ) *
5) x3 > 26 6 5.407 CORN ( 0.16667 0.83333 0.00000 0.00000 0.00000 ) *
3) x1 > 21.5 25 65.300 CLOVER ( 0.40000 0.00000 0.24000 0.12000 0.24000 )
6) x1 < 28 9 21.870 SUGARBEET ( 0.11111 0.00000 0.11111 0.33333 0.44444 ) *
7) x1 > 28 16 30.310 CLOVER ( 0.56250 0.00000 0.31250 0.00000 0.12500 )
14) x3 < 23 5 6.730 CLOVER ( 0.60000 0.00000 0.00000 0.00000 0.40000 ) *
15) x3 > 23 11 15.160 CLOVER ( 0.54545 0.00000 0.45455 0.00000 0.00000 )
30) x1 < 35 6 7.638 COTTON ( 0.33333 0.00000 0.66667 0.00000 0.00000 ) *
31) x1 > 35 5 5.004 CLOVER ( 0.80000 0.00000 0.20000 0.00000 0.00000 ) *
```



3. Fit a neural Net using the commands below.

```
> library(nnet)
> n=nrow(crop)
> ni=sample(n,.7*n,rep=F)
> ni
[1] 34 4 17 27 15 14 35 23 18 9 31 1 16 3 12 30 22 20 21 13 8 11 32 33 6
> croptr=crop[-ni,]
> croptst=crop[ni,]
> cropnnet=nnet(crop~,data=croptr,size=10)
# weights: 105
initial value 24.612731
iter 10 value 9.819991
iter 20 value 4.855405
iter 30 value 4.274512
iter 40 value 3.297687
iter 50 value 3.295839
final value 3.295837
converged
> pretr=apply(predict(cropnnet,class=T),1,which.max)
> pretr
2 5 7 10 19 24 25 26 28 29 36
2 2 2 5 3 5 5 5 1 1 1
> table(pretr,crop$crop[-ni])
pretr CLOVER CORN COTTON SOYBEANS SUGARBEET
1      3      0      0      0      0
2      0      3      0      0      0
3      0      0      1      0      0
5      1      0      0      1      2
> pretst=apply(predict(cropnnet,newdata=croptst),1,which.max)
> pretst
34 4 17 27 15 14 35 23 18 9 31 1 16 3 12 30 22 20 21 13 8 11 32 33 6
1 2 1 5 1 5 2 1 1 2 1 2 2 5 2 1 3 2 5 2 5 5 2 2 5
> table(pretst,crop$crop[ni])
pretst CLOVER CORN COTTON SOYBEANS SUGARBEET
1      3      0      3      0      1
2      3      2      1      3      1
3      0      0      0      0      1
5      1      2      1      2      1
```

#### 4. Compare the four methods.

```
> table(predict(croplda,newdata=crop)$class,crop[,1])
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
CLOVER	8	0	4	1	2
CORN	0	3	0	1	0
COTTON	0	0	0	0	0
SOYBEANS	0	4	2	3	2
SUGARBEET	3	0	0	1	2

```
> table(predict(cropqda,newdata=crop)$class,crop[,1])
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
CLOVER	9	0	0	0	0
CORN	0	7	0	0	0
COTTON	0	0	6	0	1
SOYBEANS	0	0	0	6	1
SUGARBEET	2	0	0	0	4

```
> table(tree$y,crop[,1])
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
CLOVER	11	0	0	0	0
CORN	0	7	0	0	0
COTTON	0	0	6	0	0
SOYBEANS	0	0	0	6	0
SUGARBEET	0	0	0	0	6

```
> table(apply(predict(cropnnet,newdata=crop),1,which.max),crop[,1])
```

	CLOVER	CORN	COTTON	SOYBEANS	SUGARBEET
1	6	0	3	0	1
2	3	5	1	3	1
3	0	0	1	0	1
5	2	2	1	3	3