For the data shown in Table 8.1, divide the chemical companies into two groups: group I consists of those companies with a P/E less than 9, and group II consists of those companies with a P/E greater than or equal to 9. Group I should be considered mature or troubled firms, and group II should be considered growth firms. Perform a discriminant function analysis, using ROR5, D/E, SALESGR5, EPS5, NPM1, and PAYOUTR1. Assume equal prior probabilities and costs of misclassification. Test the hypothesis that the population D2 = 0. Produce a graph of the posterior probability of belonging to group I versus the value of the discriminant function. Estimate the probabilities of misclassification by several methods.

#### CODE:

```
data firms;
set chem;
if pe<9 then pe= 0;
else if pe>=9 then pe= 1;
proc discrim distance data = firms;
class pe;
var ROR5 DE SALESGR5 EPS5 NPM1 PAYOUTR1;
run;
proc discrim data = firms crossvalidate;
class pe;
var ROR5 DE SALESGR5 EPS5 NPM1 PAYOUTR1;
run;
```

		The DIS	SCR	IM Pro	cedure			
	Total S	ample Size	30	DF To	otal		2	9
	Variable	es	6	DF W	ithin Cla	sses	2	8
	Classes		2	DF Be	etween (	lasse	s	1
		Number of C	)bse	ervatio	ns Read	42		
		Number of C	)bse	ervation	ns Used	30		
		Class L	eve	I Infor	mation			
	Variable		, v	loight	Description	.	DL	Prior
pe	Name	Frequency		vergiit	Proport	ion	Prob	ability
pe 0	Name _0	Frequency		2.0000	0.400			
•	200		2 1			000	0.4	500000 500000
0	_0	12 18 Pooled	2 1: 3 1: Cov	2.0000	0.400 0.600 e Matrix	000	0.4	500000
0	_0	12 18 Pooled	Covered I	2.0000 8.0000 arianco matior Natural	0.400 0.600 e Matrix	000 000 he	0.4	500000

The SAS System

#### The DISCRIM Procedure

Squared Distance to pe					
From pe	0	1			
0	0	1.82693			
1	1.82693	0			

F Statistics, NDF=6, DDF=23 for Squared Distance to pe					
From pe	0	1			
0	0	1.80083			
1	1.80083	0			

Prob > Mahalanobis Distance for Squared Distance to pe				
From pe	0	1		
0	1.0000	0.1434		
1	0.1434	1.0000		

Prob > Mahalanobis distance is within 0.15

 $\Rightarrow$  Fail To reject null hypothesis  $D^2 = 0$  (Bonferroni inequality)

Generalized Squared Distance to pe					
From pe	0	1			
0	0	1.82693			
1	1.82693	0			

Linear Discriminant Function for pe					
Variable	Label	0	1		
Constant		-31.90052	-38.28292		
ror5	ror5	1.24581	1.28715		
de	de	27.30944	25.74322		
salesgr5	salesgr5	1.37095	1.58963		
eps5	eps5	-0.19533	-0.25480		
npm1	npm1	0.64526	0.84870		
payoutr1	payoutr1	47.65038	55.07132		

## Discrimination Function Analysis Resubstitution Summary=>

#### The SAS System

The DISCRIM Procedure Classification Summary for Calibration Data: WORK.FIRMS Resubstitution Summary using Linear Discriminant Function

	ito p		Percent
		1	Tota
		2	12
3.3	16.6	57	100.00
	1	4	18
2.2	77.7	78	100.00
1	1	16	30
6.6	53.3	33	100.00
0	0	.5	

Error Count Estimates for pe					
	0	1	Total		
Rate	0.1667	0.2222	0.1944		
Priors	0.5000	0.5000			

Discrimination Function Analysis Cross-Validation Summary=>

### The SAS System

The DISCRIM Procedure Classification Summary for Calibration Data: WORK.FIRMS Cross-validation Summary using Linear Discriminant Function

Number of Observations and Percent Classified into pe						
From pe	0	1	Total			
0	9	3	12			
	75.00	25.00	100.00			
1	6	12	18			
	33.33	66.67	100.00			
Total	15	15	30			
	50.00	50.00	100.00			
Priors	0.5	0.5				

Error Count Estimates for pe					
	0	1	Total		
Rate	0.2500	0.3333	0.2917		
Priors	0.5000	0.5000			

Is it possible to distinguish between men and women in the depression data set on the basis of income and level of depression? What is the classification function? What are your prior probabilities? Test whether the following variables help discriminate: EDUCAT, EMPLOY, HEALTH.

```
CODE:
```

```
proc discrim data = dep1;
class SEX;
var INCOME CESD;
run;
proc discrim data = dep1;
class SEX;
var EDUCAT EMPLOY HEALTH;
run;
proc discrim data = dep1;
class SEX;
var INCOME CESD EDUCAT EMPLOY HEALTH;
run;
```

Taking variables INCOME and Level of Depression:

<b>Total Sample Size</b>	294	DF Total	293
Variables	2	DF Within Classes	292
Classes	2	DF Between Classes	1

Nun	nber of Observations Read	294
Nun	nber of Observations Used	294

	Class Level Information					
SEX	Variable Name	Frequency	Weight	Proportion	Prior Probability	
1	_1	111	111.0000	0.377551	0.500000	
2	_2	183	183.0000	0.622449	0.500000	

Pooled Covariance Matrix Information		
	Natural Log of the Determinant of the Covariance Matrix	
2	9.74816	

Generalized Squared Distance to SEX				
From SEX 1 2				
1	0	0.18425		
2	0.18425	0		

Linear Discriminant Function for SEX			
Variable 1 2			
Constant	-1.87083	-1.58347	
INCOME	0.11635	0.09326	
CESD	0.12511	0.14879	

Equations=>

Male = -1.87 +.116\* INCOME + .125 \* CESD Female = -1.58 + .09\*INCOME + .14 \* CESD

WORK.DEP1 Resubstitution Summary using using LDA

Number of Ol Clas	sified in		Percent
From SEX	1	2	Total
1	57	54	111
	51.35	48.65	100.00
2	60	123	183
	32.79	67.21	100.00
Total	117	177	294
	39.80	60.20	100.00
Priors	0.5	0.5	

<b>Error Count Estimates for SEX</b>				
	1 2 Total			
Rate	0.4865	0.3279	0.4072	
Priors	0.5000	0.5000		

# Taking variables EDUCAT, EMPLOY, HEALTH:

<b>Total Sample Size</b>	294	DF Total	293
Variables	3	DF Within Classes	292
Classes	2	DF Between Classes	1

Number of Observations Read	294
Number of Observations Used	294

	Class Level Information					
SEX	Variable Name	Frequency	Weight	Proportion	Prior Probability	
1	_1	111	111.0000	0.377551	0.500000	
2	_2	183	183.0000	0.622449	0.500000	

Pooled Covariance Matrix Information		
	Natural Log of the Determinant of the Covariance Matrix	
3	0.72043	

Generalized Squared Distance to SEX			
From SEX 1 2			
1	0	0.22465	
2	0.22465	0	

Linear Discriminant Function for SEX			
Variable	1	2	
Constant	-9.18416	-9.73107	
EDUCAT	3.13499	3.09751	
EMPLOY	1.16444	1.45247	
HEALTH	2.97652	3.03149	

## Equations=>

### Male = -9.18 + 3.13\*EDUCAT + 1.16 \* EMPLOY+2.97\*HEALTH Female = -9.73 + 3.09\*EDUCAT + 1.45 \* EMPLOY+3.03\*HEALTH

WORK.DEP1 Resubstitution Summary using using LDA:

Number of Observations and Percent Classified into SEX				
From SEX	1	2	Total	
1	86	25	111	
0.130	77.48	22.52	100.00	
2	101	82	183	
	55.19	44.81	100.00	
Total	187	107	294	
	63.61	36.39	100.00	
Priors	0.5	0.5		

<b>Error Count Estimates for SEX</b>					
	1 2 Total				
Rate	0.2252	0.5519	0.3886		
Priors	0.5000	0.5000			

Variables EDUCAT, EMPLOY, HEALTH for classification of sex gives error rate as 38.86 %.

# Taking variables INCOME, CESD, EDUCAT, EMPLOY and HEALTH:

<b>Total Sample Size</b>	294	DF Total	293
Variables	5	DF Within Classes	292
Classes	2	DF Between Classes	1

Number of Observations Read	294
Number of Observations Used	294

Class Level Information					
SEX	Variable Name	Frequency	Weight	Proportion	Prior Probability
1	_1	111	111.0000	0.377551	0.500000
2	_2	183	183.0000	0.622449	0.500000

	ovariance Matrix formation
	Natural Log of the Determinant of the Covariance Matrix
5	10.23090

Generalized Squared Distance to SEX				
From SEX	1	2		
1	0	0.33114		
2	0.33114	0		

Linear Discriminant Function for SEX				
Variable 1 2				
Constant	-9.65008	-10.13496		
INCOME	0.06030	0.04092		
CESD	0.06939	0.08893		
EDUCAT	2.87860	2.92996		
EMPLOY	1.17998	1.44978		
HEALTH	2.89879	2.90040		

Eqautions=>

MALE=-9.65+.06\*INCOME+.06\*CESD+2.87\*EDUCAT+1.17\*EMPLOY+2.89\*HEALTH FEMALE=-10.13+.04\*INCOME+.088\*CESD+2.92\*EDUCAT+1.44\*EMPLOY+2.90\*HEALTH

Number of Ol Clas	sified in		Percent
From SEX	1	2	Total
1	86	25	111
	77.48	22.52	100.00
2	84	99	183
****	45.90	54.10	100.00
Total	170	124	294
	57.82	42.18	100.00
Priors	0.5	0.5	

<b>Error Count Estimates for SEX</b>					
	1 2 Total				
Rate	0.2252	0.4590	0.3421		
Priors	0.5000	0.5000			

Variables INCOME, CESD, EDUCAT, EMPLOY and HEALTH for classification of sex gives error rate as 34.21%.