1) we have

$$p(x) = \frac{e\beta_0 + \beta_1 x}{1 + e\beta_0 + \beta_1 x}$$

Subtrout 1 tom both side,

$$1 - P(x) = 1 - \frac{e^{\beta_0 + \beta_1 x}}{1 + e^{\beta_0 + \beta_1 x}}$$

$$\Rightarrow 1 - p(x) = \frac{1 + e^{\beta_0 + \beta_1 x} - e^{\beta_0 + \beta_1 x}}{1 + e^{\beta_0} + \beta_1 x}$$

$$\Rightarrow 1 - P(x) = \frac{1}{1 + e^{B_0 + \beta_1 X}}$$

$$= \frac{1 - p(x)(e^{\beta_0 + \beta_1 x})}{1 - p(x)} = \frac{e^{\beta_0 + \beta_1 x}}{1 + e^{\beta_0 + \beta_1 x}} \Rightarrow p(x)$$

$$= \frac{1 - p(x)(e^{\beta_0 + \beta_1 x})}{1 - p(x)} = \frac{e^{\beta_0 + \beta_1 x}}{1 - p(x)} \Rightarrow p(x)$$

$$= \frac{p(x)}{1 - p(x)} = e^{\beta_0 + \beta_1 x}$$

$$= \frac{p(x)}{1 - p(x)} = e^{\beta_0 + \beta_1 x}$$

$$OX$$
,  $\frac{P(x)}{1-P(x)} = e^{B_0+B_1X}$ 

$$\beta(x) = \frac{e^{i\beta + \beta_{1}x}}{1 + e^{\beta_{0} + \beta_{1}x}}, \text{ biven } \beta_{0} = -6$$

$$\beta_{1} = 0.05$$

$$\beta_{2} = 4$$

$$x_{1} = 50$$

$$x_{2} = 0.8$$

$$\Rightarrow P(x) = \frac{e^{-6 + 0.05 \times 50} + 0.8}{1 + e^{-6 \times 0.05 \times 50} + 0.8}$$

$$= \frac{e^{-6 + 2.5 + 0.8}}{1 + e^{-6 + 2.5 + 0.8}} = \frac{e^{-2.7}}{1 + e^{-2.7}}$$

$$= \frac{0.06720}{1 + 0.06720} = 0.06296$$

(b) Griven 
$$p(x) = 50\% = 0.5, x_1 = ?$$

$$0.5 = \frac{(-6+0.05 \times x_1 + 0.8)}{(-6+0.05 \times x_1 + 0.8)}$$

$$1 + (e^{-6+0.05 \times x_1 + 0.8})$$

**Q3.** We now examine the differences between LDA and QDA.

a. If the Bayes decision boundary is linear, do we expect LDA or QDA to perform better on the training set? On the test set?

If the Bayes decision boundary is linear, we expect QDA to perform better on the training set because its higher flexibility may yield a closer fit. On the test set, we expect LDA to perform better than QDA, because QDA could overfit the linearity on the Bayes decision boundary.

b. If the Bayes decision boundary is non-linear, do we expect LDA or QDA to perform better on the training set? On the test set ?

If the Bayes decision bounary is non-linear, we expect QDA to perform better both on the training and test sets.

c. In general, as the sample size nn increases, do we expect the test prediction accuracy of QDA relative to LDA to improve, decline, or be unchanged? Why?

Roughly speaking, QDA (which is more flexible than LDA and so has higher variance) is recommended if the training set is very large, so that the variance of the classifier is not a major concern.

d. True or False: Even if the Bayes decision boundary for a given problem is linear, we will probably achieve a superior test error rate using QDA rather than LDA because QDA is flexible enough to model a linear decision boundary. Justify your answer.

False. With fewer sample points, the variance from using a more flexible method such as QDA, may lead to overfit, which in turns may lead to an inferior test error rate.

(4) Griven 
$$\bar{x}_1 = 8$$
,  $\bar{x}_2 = 0$ ,  $\delta^2 = 25$ ,  $75\%$ .

$$P_{1}(4) = \frac{0.35e^{-(1/72)(4-8)^{2}}}{0.75e^{-(1/72)(4-8)^{2}} + 0.25e^{-(1/72)(4-0)^{2}}}$$

$$= \frac{0.60055}{0.60055 + 0.20018}$$

= 0.7500

So 75% Is the Annuar.

(36) iven, Traing Data Error Pate = 201.

Test Data Error Pate = 261.

Ang Error Pate = 181.

K=1

In the case of KNN with K=1, error is o's.

But we have average faste of 18's, which implies a test rate of 36's, of KNN which is greater than test Data crow out 26's, So better to choose logistic fegression.

6 © we may write  $\frac{P(x)}{1-P(x)} = 0.32$  P(x) = 0.24So 247. Answer

(b) We ong write 
$$\frac{p(x)}{1-p(x)} = \frac{0.15}{1-0.15}$$

= 0.17-64 So odd 18 17.64 1/0 Anser

	A	В	С	D	E	F	G	Н	1	J
1	Q7.a									
2	LDA 2 Categor	ry 2 Dimentina	al Data							
3										
4	Trainiing Data set				Mean Corrected Data		Dimentional function		Results	
5										
6										
7	Class	x1	x2		x10	x20	f1	f2	Classification	
8	0	-6	9		-5.33333333	7			0	
9	0	-4	8		-3.33333333	6			0	
10	0	1	8		1.66666667	6			0	
11	0	-4	6		-3.33333333	4			0	
2	0	-4	2		-3.33333333	0			0	
3	0	-2	3		-1.33333333	1			0	
4	0	2	5		2.66666667	3			0	
15	0	-2	7		-1.33333333	5			0	
16	1	2	-2		2.66666667	-4			1	
17	1	3	-4		3.66666667	-6			1	
18	1	-1	-3		-0.33333333	-5			1	
19	1	2	2		2.66666667	0			1	
20	1	-2	-6		-1.33333333	-8			1	
21	1	3	1		3.66666667	-1			1	
22	1	2	-6		2.66666667	-8			1	
23										
24										
25	g	2								
26										
27										
28	m1	-2.375	6							
29	m2	1.28571429	-2.57142857							
30	m	-0.66666667	2							
31										
32										
33	corr1	7.4107	-0.8571							
34		-0.8751	6.2857							
35										
	corr2	3.9048	2.8571							
37		2.8571	9.9524							
38										
39	corr0	8.9524	-75714							
10		-75714	27							
11										
12	corr0-1	0	-0.1321							
13		-0.1321	0							
14										
15										
16	р		0.53333333							
17			0.46666667							
18										