## **WEEK 5 ASSIGNMENT**

## Concepts of Statistics 2 – DATA-51200 | Spring 2 2020

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# 1. In your own words, discuss (in less than one page) the differences between Multiple Regression Analysis and Multiple Discriminant Analysis.

Multiple Regression Analysis (MRA) and Multiple Discriminant Analysis (MDA) are both multivariate statistical techniques examining dependence relationships among the implicated variables. The major distinction between the two approaches is the nature of the single dependent variable which in the case of MRA is metric and nonmetric for MDA. Thus, MDA allows to predict a categorical dependent variable based on a set of independent metric variables while MRA deals with a continuous dependent variable.

Another distinction between MRA and MDA is the number of variates in a single analysis. While MRA is limited to a single variate, MDA can deal with several variates through calculating a discriminant function for each one.

When assessing model fit, the classification matrix and hit ratio value replace the R<sup>2</sup> value in MDA. The error in MRA is calculated as the residual between actual value and predicted value. Errors in MDA are evident from the classification matrix as misclassified examples.

2. For the data set associated with this homework (HBAT and HBAT\_Test (you may use any software and programming language you feel comfortable dealing with. Make sure to include your codes, diagrams and results). Using X4 as the non-metric variable and (X6 up to X18) as the metric variables:

# a. What does each variable represent? (go back to chapter 1)

The dependent nonmetric variable X4 in the HBAT survey represents the region i.e. the location of the customer which can be either 0 for being a USA/North America resident or 1 for customers located outside North America.

The metric independent variables X6 to X18 encode different HBAT business functions which were rated by customers on a metric scale from 0 (poor) to 10 (excellent):

Variable X6 denotes the perceived level of quality of HBAT's paper products.

The overall image and user-friendliness of the HBAT's website is stored in variable X7.

Variable X8 assesses the extent to which technical support is offered to help solve product and service

The extent to which any complaints are resolved in a timely and complete manner is measured by variable X9 complaint resolution.

Variable X10 measures the perceptions of HBAT's advertising campaigns in all types of media.

The depth and breadth of HBAT's product line is stored in variable X11.

Variable X12 stands for the overall image of HBAT's salesforce.

Competitive pricing is represented by variable X13 and assesses the extent to which HBAT offers competitive prices.

Warranty and claims issues are stored in variable X14.

The extent to which HBAT develops and sells new products is perceived with variable X15.

Efficiency in ordering and billing is measured using variable X16.

Variable X17 denotes the perceived willingness of HBAT sales representatives to negotiate price on purchases of paper products.

Variable X18 measures the amount of time it takes to deliver the paper products once an order has been confirmed.

### b. How many groups does X4 has?

The variable X4 has two groups: 0 and 1, for customers located in USA/North America and outside North America, respectively.

#### c. Apply linear discriminant analysis to the data (HBAT) and find:

Using SAS studio, I applied Linear Discriminant Analysis using X4 as the dependent categorical variable and 13 independent metric variables X6 to X18.

The following SAS code was generated:

```
/*
 * Task code generated by SAS Studio 3.8
 * Generated on '18/04/2020 21:16'
 * Generated by 'sasdemo'
 * Generated on server 'LOCALHOST'
 * Generated on SAS platform 'Linux LIN X64 2.6.32-754.6.3.el6.x86_64'
 * Generated on SAS version '9.04.01M6P11072018'

* Generated on browser 'Mozilla/5.0 (Macintosh; Intel Mac OS X 10_13_6)
AppleWebKit/605.1.15 (KHTML, like Gecko) Version/13.1 Safari/605.1.15'
                  Generated
                                             on
                                                               web
                                                                                   client
http://localhost:10080/SASStudio/38/main?locale=en GB&zone=GMT%252B02%253A00&http%
3A%2F%2Flocalhost%3A10080%2FSASStudio%2F38%2F='
 */
ods noproctitle;
proc discrim data=WORK.IMPORT pool=yes crossvalidate crosslisterr distance
              posterr list listerr;
       class x4;
       var x6 x7 x8 x9 x10 x11 x12 x13 x14 x15 x16 x17 x18;
      priors prop;
run;
```

#### • The linear discriminant function for X4.

The discriminant function for the variable X4 and the two classes, 0 (outside inside the USA) and 1 (outside the USA) is shown in Table 1. For every independent variable, discriminant weights are calculated which represent the contribution of each variable. The strongest effect on the discriminant function exhibit variables X11, X14, X17 and X18. Thus, the product line, the salesforce image, price flexibility and delivery speed seem to strongly impact group distinction.

Linea	Discrim	inant Functio	n for x4
Variable	Label	0	1
Constant		-191.92135	-194.33409
х6	х6	8.32797	7.65248
x7	х7	4.20485	1.20101
x8	x8	-2.06370	-2.09852
х9	x9	-3.62428	-3.60295
x10	x10	-1.62571	-2.03642
x11	x11	58.68681	58.34999
x12	x12	1.69711	4.70052
x13	x13	3.64274	4.22872
x14	x14	13.54926	13.38761
x15	x15	0.00591	0.28730
x16	x16	-2.84573	-2.25264
x17	x17	62.42839	64.51335
x18	x18	-101.40047	-103.57123

Table 1. Linear Discriminant Function for X4.

• By applying the LDF to the training data (HBAT): How many observations were misclassified? What are they? Find the confusion matrix and the probability of (error) misclassification.

Table 2 shows the classification matrix listing the number of correct and wrong classifications. Out of 100 observations, 5 observations were misclassified. Although the true class of these observations (observations 3, 22, 38, 60, 94) is 1 (outside the USA/North America), the linear discriminant analysis (LDA) classified the observations as class 0 (outside USA/North America).

lassified into x4	e Posterior Probabilities C	Number of Observations and A
1	0	From x4
(	39 0.8790	0
56 0.9453	5 0.8463	1
56 0.9453	44 0.8753	Total
0.61	0.39	Priors

Table 2. Classification matrix for LDA of HBAT training data set.

The rate of making this mistake, a Type I error, is with 0.0820 (approx. 8%) relatively low. See Table 3 for the error count estimates of variable x4 and the two classes.

Error	Count E	stimates 1	for x4
	0	1	Total
Rate	0.0000	0.0820	0.0500
Priors	0.3900	0.6100	

Table 3. Probability of misclassification for LDA on HBAT training data set.

• By applying the LDF to the test data (HBAT\_Test): How many observations were misclassified? What are they? Find the confusion matrix and the probability of (error) misclassification.

Load the HBAT Test data into SAS (WORK.IMPORT1) and perform LDA using this test set.

Generated SAS code:

When using the previously trained LDA model on the unseen test data set, the number of misclassifications increased to 13 with 11 observations classified as class 0 although being class 1 (Type I error) and 2 observations classified as class 1 although being class 0 (Type II error). See Table 4 for the results of the classification and the error for misclassification. The rate of committing a Type I or Type II error is 0.1803 and 0.0513, respectively. Thus, the model does well on classifying customers in the USA/North America market, whereas it makes mistakes in classifying customers outside the USA/North America.

ed into x4	t Classifie	Percen	ations an	ber of Observ
Tota	1		0	From x4
39 100.00	2 5.13		37 94.87	0
61 100.00	50 1.97	8	11 18.03	1
100 100.00	52 2.00	5	48 48.00	Total
	0.61		0.39	Priors
	for x4	timates 1	Count Es	Error
	Total	1	0	
	0.1300	0.1803	0.0513	Rate
		0.6100	0.3900	Priors

Table 4. Classification matrix and probability of misclassification for LDA on HBAT test data set.

