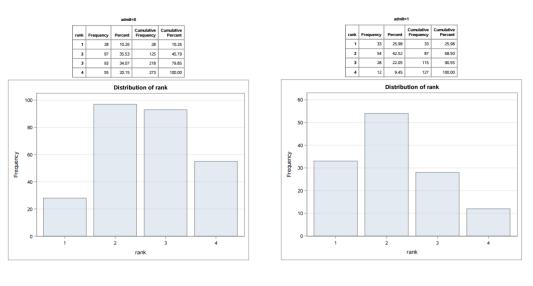
$\begin{array}{c} \textbf{Lab 4-Spring 2021} \\ 25^{th} \text{ March 2021} \\ \text{cavicchia@ese.eur.nl} \end{array}$

1 Introduction to Logit Model

A researcher is interested in how variables, such as GRE (Graduate Record Exam scores), GPA (grade point average) and prestige of the undergraduate institution, effect admission into graduate school. The response variable, admit/don't admit, is a binary variable. This data set has a binary response (outcome, dependent) variable called **admit**. There are three predictor variables: **gre**, **gpa** and **rank**. We will treat the variables gre and gpa as continuous. The variable rank takes on the values 1 through 4. Institutions with a rank of 1 have the highest prestige, while those with a rank of 4 have the lowest. Use the data set Admit.

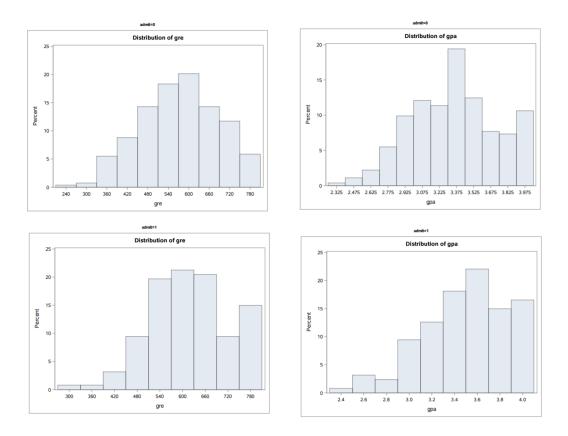
1.1 Descriptive Analysis

Let's start with the descriptive analysis.



uuniit-v											
Variable	N	N Miss	Minimum	Mean	Median	Maximum	Std Dev				
gre gpa	273 273	0	220.0000000 2.2600000	573.1868132 3.3436996	580.000000 3.3400000	800.0000000 4.0000000	115.8302427 0.3771330				

admit=1											
Variable	N	N Miss	Minimum	Mean	Median	Maximum	Std Dev				
gre gpa	127 127	0	300.0000000 2.4200000	618.8976378 3.4892126	620.000000 3.5400000	800.0000000 4.0000000	108.8848842 0.3701771				



The distributions Rank over the admit values are quite different. When admit takes value 0, most institutions have rank 2 or 4; on the other hand when admit takes value 1, institutions with rank 4 decrease and those with rank 1 increase. The distributions of gpa over the two groups (admit=0 and admit=1) is quite similar. The distribution of gre has higher mean and median when admit takes value 0, compared to the other group (admit=1). Looking at the histograms, the only distribution that result to be more symmetric is gpa when admit takes value 0.

1.2 Fitting a Logistic Model

In this section we fit a logistic model using all predictors. The output is given below.

• Overall significance test:

	Mo	del Fit	Statisti	CS			
Criterion	Intercep	t Only	Intercept and Covariates				
AIC	501.977		470.517				
SC	505.968		494.466				
-2 Log L	499.977		458.51				
5					1001011		
	ting Globa		lypoth	esis:			
	ting Globa			esis: DF			
Test	ting Globa od Ratio	al Null H Chi-So			BETA=0		
Test		al Null F Chi-So 41	quare	DF	BETA=0 Pr > ChiSq		

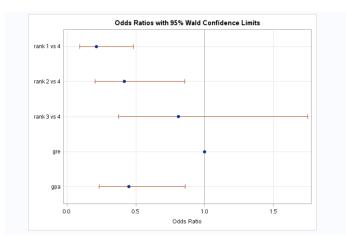
First, we note that the p-values are less than $\alpha = 0.05$, so we reject the null hypothesis. There is at least one predictor that is significant.

• Interpretation of coefficients

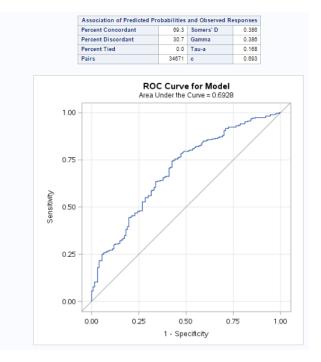
		Type 3 Analysis of Effects										
	Effect		t DF	Cł	Wal Chi-Squar							
	rank		3		20.894	9	0.0001					
	gre		1		4.284	4.2842		0.0385				
	gpa		1		5.871	4 0.0		0154				
Analysis of Maximum Likelihood Estimates												
Parameter		DF	Estim	Estimate		Standard Error		Wald Chi-Square		> ChiSq		
Intercept		1	5.5414		1.1381		2	3.7081		<.0001		
rank 1		1	-1.5514		0.4178		1	13.7870		0.0002		
rank	2	1	-0.8	760	0.3667			5.7056		0.0169		
rank	3	1	-0.2	112	0.3929			0.2891		0.5908		
rank	4 0			0								
gre	gre		-0.00226		0.00109		4.2842			0.0385		
gpa		1	-0.8040		0.3318		5.8714			0.0154		
	Odds Ratio Estimates											
	Effe	ect	Poi	Point Estimate			95% Wald Confidence Limits					
	rank 1 vs 4		1	0.212			0.093		0.481			
	rank	(2 vs 4	4	0.416		0.203		0.855				
	rank 3 vs 4		1	(0.375		1.748				
	gre			0.998			0.996 1.		00			
	gpa			0.448			0.234 0		58			

Both gre and gpa are statistically significant. The logistic regression coefficients give the change in the log odds of the outcome for a one unit increase in the predictor variable. For every one unit change in gre, the log odds of non-admission (versus admission) decreases by 0.002. For a one unit increase in gpa, the log odds of not being admitted to graduate school decreases by 0.804. The indicator variables for rank have a slightly different interpretation. For example, having attended an undergraduate institution with rank of 1, versus an institution with a rank of 4, changes the log odds of admission by -1.5514.

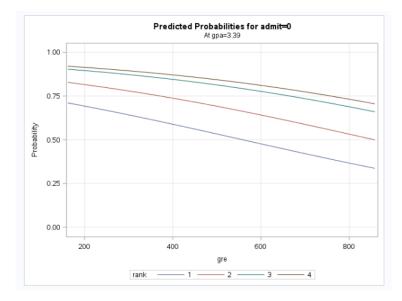
- note that the CI for gre has 1 as upper limit, indeed the p-value corresponding to its estimate is 0.0385, that is almost 0.05.
- Looking at the Odds Ratio Estimates, now we can say that for a one unit increase in gpa, the odds of not being admitted to graduate school (versus being admitted) increase by a factor of 0.45. For a one unit increase in gre, the odds of not being admitted to graduate school (versus being admitted) increase by a factor of 0.998. The odds of not being admitted to graduate school (versus being admitted) having attended an undergraduate institution with rank of 1 (2 or 3), versus an institution with a rank of 4, increase by a factor of 0.212 (0.416 or 0.810). This is depicted in the plot below.



• Model Performance: Looking at the ROC curve and the AUC, 0.69, we deduce that the classification performance is quite poor. Further variables (and thus information), are needed to boost the classification performance.



• Prediction



Fixing the gpa at 3.39 is is possible to note how the probability of being not admitted decreases with gre for every rank (although the gap between curves depends on the specific gre value).