SCM 651 Homework 4

Loan Analysis

Using the Universal Bank data, determine the factors which influence whether a customer takes out a loan

Resources

Use the dataset SCM 651 Homework 4 Universal Bank.csv.

Outline and grading criteria:

- 1. Perform a logit and probit analysis of the variables that affect whether a customer takes out a loan. Consider only main effects.
 - a. Which variables are significant?

Using logit, when considering age, income, credit card average, family size, education, mortgage, and securities account, all of the variables are statistically significant (when alpha = 0.05) **except** age (just barely, at p-value = 0.055) and mortgage (p-value = 0.2).

Using probit, we see the same variables are statistically significant (except age at p-value = 0.15, and mortgage at p-value = 0.14), but with slightly different p-values.

b. How do the significant variables influence the likelihood of taking out a loan?

Using logit, all of the significant variables have a positive effect. When income, family size, credit card average, education, and securities increases, the likelihood of taking out a loan also increases. No statistically significant variables had a negative effect that would decrease the likelihood of taking out a loan.

Using probit, we see the same positive effect for all of the statistically significant variables. No statistically significant variables had a negative effect that would decrease the likelihood of taking out a loan.

c. Copy screen snapshots of your analysis in R to your report.

Logit

```
Coefficients:
                  Estimate Std. Error z value Pr(>|z|)
              -14.1116316  0.6216320 -22.701 < 2e-16 ***
(Intercept)
                 0.0117206 0.0060994 1.922 0.054656
Age
                 0.0544416   0.0024091   22.599   < 2e-16 ***
Income
                                     3.463 0.000534 ***
CCAvg
                 0.1263517 0.0364847
                 Family
                 1.6597624 0.1044880 15.885 < 2e-16 ***
Education
                 0.0006615 0.0005213 1.269 0.204465
Mortgage
SecuritiesAccount
                0.5173873 0.2111582
                                     2.450 0.014276 *
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Probit

```
Coefficients:
                   Estimate Std. Error z value Pr(>|z|)
(Intercept)
                 -7.1134039 0.2915124 -24.402
                                               < 2e-16
                  0.0045947 0.0031577
                                       1.455
                                               0.14565
Age
                  0.0277167 0.0011750 23.588
Income
                                               < 2e-16 ***
                  0.0792593 0.0194629
                                       4.072 0.0000465 ***
CCAvg
                                               < 2e-16 ***
                  0.3398349 0.0351090
                                       9.679
Family
                                                < 2e-16 ***
Education
                  0.8074032 0.0512303 15.760
                  0.0004028 0.0002762
                                        1.458
                                                0.14474
Mortgage
SecuritiesAccount 0.2941015 0.1094193
                                                0.00719 **
                                        2.688
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

2. Add moderating effects (interactions of variables).

a. Which interactions make sense conceptually?

Income and CCAvg make sense to compare, because the data roughly shows that as income increases, the average monthly spending on credit cards also increases. People with higher amounts of credit card debt often look to consolidate their credit card balances using a person loan to lower the monthly interest charges and pay one monthly amount compared to all their credit cards separately. The monthly loan payment is also potentially less money than their separate credit card payments combined, which frees up more cash for other expenses or savings each month. Those with a lower income and a higher credit card debt would be more inclined to try and consolidate debt with a personal loan in order to decrease their monthly payments and increase their monthly cash flow. However, people with a higher income and little to no credit card debt would not be inclined to take out a personal loan.

b. Which interactions are statistically significant?

All three variables (*income*, CCAvg, *income**CCAvg) were statistically significant using alpha = 0.05.

c. How do you interpret the coefficients on these variables?

When *income* and *CCAvg* increase, the likelihood of taking out a loan also increases, since both of these coefficients are positive. However, *income*CCAvg* has a negative effect. This is because *income* and *CCAvg* are including all the other variables, include age, mortgage, experience, and family size.

d. Copy screen snapshots of your analysis in R to your report.

```
Coefficients:
               Estimate Std. Error z value Pr(>|z|)
             -9.5556476 0.4309632
                                    -22.17
                                             <2e-16
(Intercept)
              0.0600142
                         0.0031175
                                     19.25
Income
                                             <2e-16 ***
              1.3398815 0.1215377
                                     11.02
                                             <2e-16
CCAvg
Income:CCAvg -0.0086107
                                    -10.76
                                             <2e-16 ***
                         0.0008001
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Signif. codes:
```

- 3. Create a final regression model with the variables that you feel are important (both main effects and interaction terms). Create a spreadsheet prediction of the model.
 - a. Which variables have the greatest influence on the customers' loan behavior (combined main effects and interaction effects)?

We chose *income*, *CCAvg*, *family*, *education*, and a moderating effect of *Income*CCAvg* for a final regression model. All of the coefficients were statistically significant. In the model we created, the variables that seemed to effect the biggest changes in the probability were *CCAvg*, *family*, and *Education*. The effect of *income* was not as large of a probably change as the other variables.

	Input	ts		Outp	ut	
Variable	Value		Variable	Coefficient	Value	Coeff*Value
			Intercept	-20.14606	1	-20.14606
Income	90	Annual income in \$	Income	0.098543	90	8.86887
CCAvg	4	Avg spending on CC/month	CCAvg	2.2201	4	8.8804
Family	4	Family size	Family	0.770788	4	3.083152
Education	2	Education level	Education	1.818407	2	3.636814
Income*CCAvg	360	Interaction	Income*CCAvg	-0.014008	360	-5.04288
					Sum	-0.719704
					Exp(sum)	0.48689636
					Probability	33%

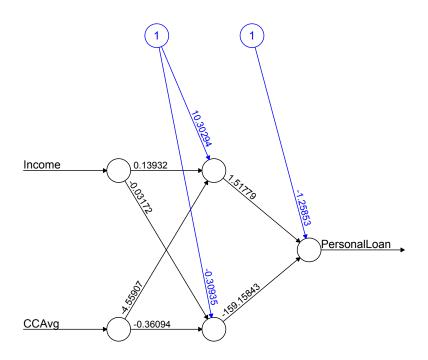
b. Perform a sensitivity analysis as seen earlier in the semester.

Sensitivity Analysis										
		CCAvg								
	33%	0	2	4	6	8	10			
Income	8	0%	0%	1%	50%	99%	100%			
	25	0%	0%	3%	56%	98%	100%			
	42	0%	0%	6%	62%	98%	100%			
	59	0%	1%	12%	68%	97%	100%			
	76	0%	3%	21%	73%	96%	100%			
	93	1%	8%	36%	78%	96%	99%			
	110	7%	23%	53%	82%	95%	99%			
	127	29%	49%	70%	85%	93%	97%			
	144	68%	76%	83%	88%	92%	94%			
	161	92%	91%	91%	90%	90%	89%			
	178	98%	97%	95%	92%	87%	80%			
	195	100%	99%	98%	94%	85%	66%			
	212	100%	100%	99%	95%	81%	49%			
	224	100%	100%	99%	96%	79%	37%			

c. Copy screen snapshots of your analysis in R to your report.

```
Coefficients:
               Estimate Std. Error z value Pr(>|z|)
(Intercept)
             -20.146060
                          0.883313
                                   -22.81
                                             <2e-16 ***
                                     20.13
Income
               0.098543
                          0.004896
                                             <2e-16 ***
CCAvg
               2.220100
                          0.167626
                                     13.24
                                             <2e-16 ***
Family
               0.770788
                          0.074187
                                    10.39 <2e-16 ***
Education
                                     16.65
               1.818407
                          0.109227
                                             <2e-16 ***
Income:CCAvg
                          0.001091
                                    -12.84
                                             <2e-16 ***
             -0.014008
Signif. codes:
                0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

4. Perform a neural network analysis of the variables found to be significant in the logit and probit analysis above. Copy screen snapshots of your final neural network model in R to your report.



Error: 144.281108 Steps: 21291

5. Create a prediction model of the neural network. Using the prediction model, perform a sensitivity analysis for the neural network model similar to the logit and probit sensitivity analysis.

Inputs			Hidden Nodes					Output		
Variable	Value		Hidden Node 1				Varia	able Coefficient	Value	Coeff*Value
			Variable	Coefficient	Value	Coeff*Value	Inter	cept -1.25852732	1 1	-1.25852732
Income	80	Annual income in \$	Intercept	10.30293778	1	10.30293778	Hidd	en 1 1.51779477	1 0.00271567	0.00412183
CCAvg	6	Avg spending on CC/month	Income	0.139318425	80	11.145474	Hidd	en 2 -159.158431	2 0.00661004	-1.05204387
			CCAvg	-4.559067935	6	-27.35440761				
									Sum	-2.30644936
					Sum	-5.905995828			Exp(sum)	0.09961431
					Exp(sum)	0.002723069			Probability	99
					Probability	0.00272				
			Hidden Node							
			Variable	Coefficient	Value	Coeff*Value				
			Intercept	-0.30934774	1					
			Income	-0.031719444	80					
			CCAvg	-0.360938344	6	-2.165630064				
					Sum	-5.012533324				
					Exp(sum)	0.006654025				
					Probability	0.00661				

Sensitivity Analysis										
		CCAvg								
	9%	0	2	4	6	8	10			
Income	8	0%	0%	0%	0%	0%	2%			
	25	0%	0%	0%	0%	2%	6%			
	42	0%	0%	0%	1%	5%	11%			
	59	0%	0%	1%	4%	10%	15%			
	76	0%	1%	9%	8%	14%	18%			
	93	0%	7%	23%	13%	17%	19%			
	110	4%	19%	36%	19%	19%	21%			
	127	14%	32%	44%	38%	20%	21%			
	144	28%	42%	49%	51%	21%	22%			
	161	39%	48%	52%	54%	22%	22%			
	178	46%	51%	54%	55%	27%	22%			
	195	50%	54%	55%	56%	46%	22%			
	212		55%	56%	56%	55%	22%			
	224		55%	56%	56%	56%	23%			