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# Predicting Student Attrition How Helpful are Surveys?

Predicting Student Attrition How Helpful are Surveys?

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### The Trouble With Surveys

o Partial coverage
o Low response rates
o Cost of administration
o Time to analyze data
o Non-response error



### Surveys – Are the Costs Justified

Does survey data improve the ability to predict attrition enough to justify the costs?



### Seven Models Tested

### Model

- 1 High school grade (HSG)
- 2 Records variables (8)
- 3 Records variables (8) & HSG
- 4 Survey variables (9)
- 5 Survey variables (9) & HSG
- 6 Records variables (8) & Survey variables (9)
- 7 Records variables (8) & Survey Variables (9) & HSG

### Variables – From Records

o High school grade o Country of birth o Language English placement test (level) Sector of enrolment (2 or 3 year) o Age o Sex o **Disability** Median income (Post code)

### Variables From Surveys – Demographic etc

 Level of motivation o First choice program Degree aspirations First generation college student o COB – Mother o COB - Father o Anticipated hrs paid work Hours in study – last yr of study Anticipated hours of study - college

### **Tools Used**

- Binary Logistic regression
- o Nagelkerke R2
- Probability of dropout
- Coefficient(s) to calculate probability for new sample
- Classification matrix (for different cutoffs)

		-/	
Observed	R	А	% C
Ret	553	273	66.9
Att	72	60	45.5
			64.0



 Plots Sensitivity
 vs false positive rate for each cutoff (probability)



### **Classification Matrix**

Cutoff = .4	Predicted				
			%		
Observed	Retention	Attrition	Correct		
					1- Specificity
Retention	553	273	66.9	Specificity	(False Positive)
				, , , , , , , , , , , , , , , , , , ,	
					1 – Sensitivity
Attrition	72	60	45.5	Sensitivity	(False Negative)
					· · · · · · · · ·
Overall					
Percentage			64.0		

### ROC Data

		Calculate				
Positive if Greater						
Than or Equal		1 -				
To(a)		Specificity	Predicted	False	Total	
(Cutoff or		(False	Attrition	Positive	Predicted	
Probability)	Sensitivity	Positive)	(Number)	(Number)	Attrition	% Correct
1 :			•			1
0.155	0.623	0.364	415	1274	1689	24.6%
0.156	0.620	0.357	413	1249	1662	24.8%
0.157	0.615	0.351	410	1229	1639	25.0%
0.158	0.608	0.346	405	1210	1615	25.1%
0.159	0.604	0.340	402	1190	1592	25.3%
0.160	0.594	0.333	396	1166	1562	25.3%
0.161	0.587	0.328	391	1147	1538	25.4%
0.162	0.582	0.322	388	1128	1516	25.6%
0.164	0.577	0.314	385	1099	1484	25.9%
0.165	0.571	0.310	380	1084	1464	26.0%
0.166	0.561	0.303	374	1060	1434	26.1%
0.167	0.548	0.297	365	1038	1403	26.0%
0.168	0.540	0.292	360	1020	1380	26.1%

# Receiver Operating Characteristic Curve (ROC)

Two-dimensional depiction of classifier performance. ROC Accuracy Ratio, a common technique for judging the accuracy of default probability models.



### Area Under the ROC Curve (AUC)

Null hypothesis Area = 0.5

.90-1 = excellent (A) .80-.90 = good (B) .70-.80 = fair (C) .60-.70 = poor (D) .50-.60 = fail (F)

St Error

0.023

0.025

0.00

0.00

Area

0.687

0.766



### 'Records' Model (8 Variables)



### Accuracy of 'Records' Model - AUC

Test Result Variable(s) Area under Curve (AUC) Error		Sig	Sig Asymptotic Confider Interv		
				Lower Bound	Upper Bound
High School Grade Alone (Poor)	.659	0.012	0.000	0.636	0.683
HS Grade + Records (8) (Poor)	.676	0.012	0.000	0.636	.686
8 Records Variables (Poor)	.608	0.012	0.000	0.585	0.631

### Classification Matrix 'Records Model'

Cut-off	= .16	N	Nagelkerke R <sup>2</sup>	% Drop Out Correctly Classified (Sensitivity)	% Retained Correctly Classified (Specificity)	%Total Correctly Classified
HS G Surve variat	rade & 8 ey ples	4153	.077	58.7%	69.5% (FP = 30.5%)	67.9%
8 Vaı (Exclu	riables ude SecV)	4427	.026	46.8%	70.3% (FP = 29.7%)	66.7%
High S Grade	School e Only	4164	.063	59.4%	66.7% (FP =33.3%)	65.6%

# Survey Model (9 Variables)



### 'Survey' Vs 'Records' Models

Test Results Variables	Area under Curve (AUC)	Std. Error	Sig	Asympto Confie Inte Lower Bound	otic 95% dence rval Upper Bound
1. High School Grade (HSG) (Poor)	.659	.012	.000	.636	.683
2. HSG & 8 Records Variables (Poor)	.676	.012	.000	.636	.686
3. Records Variables(8) (Poor)	.608	.012	.000	.585	.631
4. Survey Variables (9) (Poor)	.625	.017	.000	.592	.658
5. HSG & 9 Survey Variables (Fair)	.700	.025	.000	.652	.749
6. All Variables (17) (Poor)	.672	.024	.000	.626	.718
7. HSG & All Variables (Fair)	.715	.025	.000	.665	.764

# Variance Explained

	Medel	Nagelkerke
	Model	<b>K</b> <sup>2</sup>
1	HS Grade	0.063
2	Records Variables (8)	0.026
3	Records (8) + HS Grade	0.077
4	Survey Variables (9)	0.044
5	Survey Variables (9) + HS Grade	0.089
6	Survey & Records (17 variables)	0.070
7	Survey (9) & Records (8) & HS Grade	0.104

### **Classification Accuracy**

	Cutoff = .16				
	Model	Sensitivity	Specificity	1- Specificity	Overall
1	HS Grade	.594	.667	.333	.656
2	Records (8)	.468	.703	.297	.667
3	Records (8)+ HS Grade	.587	.695	.305	.679
4	Survey Variables (9)	.500	.687	.313	.659
5	Survey Variables (9) + HS Grade	.518	.723	.277	.695
6	Survey (9) & Records (8)	.514	.721	.279	.691
7	Survey (8) & Records (9) & SecV	.567	.742	.258	.718

### Application – The Best Model?

Known:	
Historical attrition Rate to 3 <sup>rd</sup> semester	16%
Historical retention rate to the 3 <sup>rd</sup> semester	84%
For each cutoff and model:	
The model coefficients – calculate probabilities for each student	
The accuracy of classifying attrition (the percent of students who do drop out who are classified correctly by the model)	eg 57%
The false positive rate (% of retained students who are classified as dropping out)	eg 33%

# Application – 1000 New Students Cutoff .16

M O d	Historical (16% Att)		Historical Cl (16% Att)		Classify	Model Predicted	Total Attrition Predicted	% Correct	
1	Att: Ret:	160 840	Sens: .594 FP .333	95 280	375	25.4%	1:2.9		
2	Att: Ret:	160 840	Sens: .468 FP .297	75 249	324	23.1%	1:3.3		
3	Att: Ret:	160 840	Sens: .587 FP .305	94 256	350	26.8%	1:2.7		
4	Att: Ret:	160 840	Sens: .500 FP .313	80 263	343	23.3%	1:3.3		
5	Att: Ret:	160 840	Sens: .518 FP .277	83 233	316	26.3%	1:2.8		
6	Att: Ret:	160 840	Sens: .514 FP .279	82 234	317	26.0%	1:2.8		
7	Att: Ret:	160 840	Sens: .567 FP .258	91 217	307	29.5%	1:2.4		

### Application – 1000 New Students 70 students for remediation program

Mod Cut- off	Histo (1 A	orical 6% tt)	Classify	Model Predicted	Total Attrition Predicted	% Correct	
1 (.25)	Att: Ret:	160 840	Sens: .148 FP: .059	24 50	74	32.3%	1:2.1
2 (.21)	Att: Ret:	160 840	Sens: .126 FP .059	20 50	70	28.9%	1:2.5
3 (.26)	Att: Ret:	160 840	Sens: .162 FP .052	26 44	70	37.2%	1:1.7
4 (.24)	Att: Ret:	160 840	Sens: .148 FP: .061	24 51	75	31.6%	1:2.2
5 (.28)	Att: Ret:	160 840	Sens: .164 FP: .052	26 44	70	37.5%	1:1.7
6 (.28)	Att: Ret:	160 840	Sens: .150 FP .055	24 46	70	34.2%	1:1.9
7 (.30)	Att: Ret:	160 840	Sens: .174 FP .050	28 42	70	40.0%	1:1.5

### **Optimizing Attrition Models**



### Compare Male and Female Models of Attrition

.16	Characteristics	F	Μ	F+M
а	Sensitivity	49.1%	63.1%	0.567
b	Specificity	77.5%	76.1%	0.742
С	1 - Specificity (False Positives)	22.5%	23.9%	0.258
d	Nagelkerke R <sup>2</sup>	0.105	0.195	0.104
е	Area Under ROC Curve	0.687	0.766	0.715
f	% New sample correct (Cutoff .16)	25.2%	29.8%	29.5%
g	Select 70 for remediation - % Correct	47.0%	49.1%	40.0%
h	Cutoff required for (g)	0.292	0.364	0.300

### Summary

- Variability explained by all the models tested was low (Nagelkerke R<sup>2</sup>)
- The accuracy of the models tested were judged to be poor to fair at best (Area under the ROC curve)
- Under certain conditions the HS grade and the more complete 'records' variables did as well or nearly as well as survey variables and high school grades
- Male and female models have different sensitivities at any given cutoff – and prediction can be improved by modeling the sexes separately

### Summary

- The models tended to more accurately predict attrition for males than for females (Area under ROC curve, classification matrices)
- All models tested gave better than chance prediction
- None of the models predicted drop out particularly well
- The survey data used did improve the ability to predict attrition to a greater extent than the records variables in some situations, but not to the extent that we believe warrants the costs and overcomes the limitations of data collected through survey administration

# Questions



	Differences in attrition rate between groups	Females	Males
	*Age – Was over 17 when starting college for the first time	17.9%	20.2%
Sig for both	*High school grade was < 75	16.0%	21.6%
	Expected hours of paid employment was > 15 hours/ week	9.3%	12.5%
	Study Time <12 hours in last yr of study	6.7%	5.2%
	Motivation – Low or Average	6.5%	8.3%
	*Language was French	6.2%	3.7%
males and females	*Median family income (post code) <\$60000	4.9%	5.7%
Temates	*English Placement Level - Low	2.9%	5.0%
	Place of birth father – in Canada	2.8%	4.5%
	*Diploma type - Technical	1.6%	3.9%
Sig for	Student was not in first choice program	10.4%	2.9%
	Anticipated study time at cegep	3.3%	1.8%
F only	*Country of birth – outside of Canada	2.7%	0.7%
Sig for	Degree aspirations were DEC or Bachelor	3.0%	10.8%
M only	Student was a first generation college student	1.4%	5.3%
Not Sig for either	Place of birth mother - Canada	1.8%	3.0%

### Psychosocial and Study Skills Variables (ACT Testing – Student Readiness Inventory)

o Academic discipline Academic self-confidence Commitment to college Communication skills Emotional control General determination o Goal striving Social activity o Social connection o Study skills