

Original Article

Utility of Noncognitive Admission Variables in the Prediction of Academic Success

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Abstract

Background

Physician Assistant (PA) programs commonly rely on cognitive measures such as grade point average for admission to their programs. Noncognitive measures are also collected at the time of application, with less information known about their utility in prediction of academic outcomes. This analysis observed the following noncognitive admission domains and their relationship with PA student academic success: employment hours, shadowing experience hours, research hours, hours spent in extracurricular activities, healthcare related activity hours, leadership experience, patient care experience hours, teaching hours, and volunteerism hours.

Objective

This archival analysis aims to determine if noncognitive admission variables are predictive of Physician Assistant National Certifying Examination (PANCE) failure or academic attrition at a West Texas PA Program.

Methods: A series of multiple logistic regression models were constructed to predict PANCE failure and attrition using applicant self-reported cumulative experience hours in the noncognitive admission domains. Five cohorts (n=235) were evaluated using R statistical software (version 4.1.2).

Results

Patient care experience hours demonstrated a positive association with PANCE failure, while healthcare experience had positive trends with attrition. High GPA was protective against attrition when controlled for employment, research, and shadowing experiences, and also when controlled for leadership and patient care experience. High GPA was not a protective factor for PANCE failures.

Conclusion

Contrary to our hypothesis, increased self-reported experiences did not show a protective effect against PANCE failure or attrition but were instead detrimental.

Keywords: admission, attrition, physician assistant, PANCE, West Texas, healthcare experience

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Background

Attrition and Physician Assistant National Certifying Examination (PANCE) failure are significant challenges for physician assistant (PA) programs. In 2019, the Physician Assistant Education Association Program Report identified a 93% graduation rate and an attrition rate of 6.7%.¹ The National Commission on Certification of Physician (NCCPA) reported Assistants a 95% national first-time PANCE pass rate in 2020, and a 93% first-time PANCE pass rate in 2019.² Furthermore, success on the PANCE and graduation rates are vitally important to PA programs because it has implications for accreditation, overall student success, and for PA candidate recruitment.³⁻⁴ Attrition can also place financial strain on PA programs because the vacated seat often remains unreplaced.3

Attrition and first-time PANCE failure not only affect PA programs but can also have a lasting psychological impact on the PA student with one study describing it as an "emotional upheaval." 3 A student who is unable to complete the program is often left with unpaid student loans, loss of time spent actively employed, and an unlikely chance of ever realizing their goal of becoming a PA.³ To decrease PA program attrition and PANCE failure, admission committees are tasked with identifying any variables in their applicant pool that may be predictive of academic outcomes. There has been much research considering cognitive variables at admission predicting academic outcomes; however, less analysis has been performed variables.4-5 noncognitive considering

Cognitive variables that have been researched for their utility in predicting academic outcomes include undergraduate grade point average (GPA) and Graduate (GRE) Record Examination scores:³ however, increasingly the importance of holistic admission processes and placing a larger emphasis on noncognitive variables is gaining attention.6

Objective

Prior studies examining noncognitive variables before matriculation have focused on domains such as emotional intelligence8-⁹ or prior healthcare experience.^{6,10} PA programs often emphasize previous clinical experience before admission though there is little evidentary basis and at least one study specific to PA education found no association previous between clinical and academic outcomes.¹² experience There remains little research specific to PA students considering noncognitive variables and their utility in predicting academic outcomes. We hypothesize that noncognitive admissions variables in PA students could be predictive in reducing PA student attrition rates and first-time PANCE failure rates.

Methods

Institutional Review Board exemption was obtained to examine five cohorts of PA students from graduating class years 2017 to 2021 from one West Texas PA Program using total hours for nine noncognitive variables self-reported through the Application Centralized Service for Physician Assistants (CASPA). The admission demographics of the PA students for the cohorts examined are depicted in Table 1.

Table 1. GPA of attrition and PANCE failure of students by age					
Students <25yo					
Total Attrition	1				
Avg GPA Attrition	3.37				
Total PANCE failures	8				
Avg GOA PANCE failure	3.61				
Students 25-29yo					
Total Attrition	8				
Avg GPA Attrition	3.31				
Total PANCE failures	7				
Avg GOA PANCE failure	3.46				
Students 30-35yo					
Total Attrition	1				
Avg GPA Attrition	3.46				
Total PANCE failures	0				
Avg GOA PANCE failure	n/a				
Students >35yo					
Total Attrition	3				
Avg GPA Attrition	3.5				
Total PANCE failures	3				
Avg GOA PANCE failure	3.25				

Data was examined by total cohort (n=235) and by students with below the program acceptance average total GPA of 3.5 (n=108) defined as a low GPA. The following nine variables were analyzed for their potential utility in predicting either attrition from the PA program or PANCE failure: employment hours, shadowing experience hours, research hours, hours spent in extracurricular activities, healthcare-related activity hours, leadership experience, patient care experience hours, teaching hours, and volunteerism hours. Only attrition for academic reasons was included and did not include students who decelerated or had attrition for medical or personal reasons.

The data was cleaned and analyzed using R statistical software (version 4.1.2). A series of multiple logistic regression models were constructed to predict PANCE failures and attrition using self-reported total experience hours in each of the nine noncognitive variable categories. Cumulative GPA was examined in two categories: those with GPA

above and below the program average acceptance GPA of 3.5 on a 4.0 scale at the time of matriculation. Additional exploratory multiple logistic regression models were constructed to predict Attrition or PANCE failures using each of the self-reported noncognitive variables, GPA (i.e. High vs. Low), and their interaction. These multiple logistic regression models allowed us to examine the differential effects of noncognitive variables in predicting Attrition and PANCE failures among applicants with high vs. low GPAs.

Results

With consideration of the full data set, selfreported cumulative healthcare experience hours positively predicted attrition (β = 0.0001; OR = 1.0001, p = 0.048). This indicates that every additional hour of increase in healthcare experience increased the odds of attrition by approximately 0.01%. In addition, cumulative exposure to patient care and cumulative duration of employment had positive statistical trends in predicting PANCE failures and attrition respectively (β = 0.0001, OR = 1.0001, p = 0.054 and β = 0.0001, OR = 1.0001 = 0.098 respectively). All other univariate logistic regression models were not statistically significant.

The data was analyzed with consideration of students with low GPA only. The average total GPA at the time of application through CASPA for all five cohorts was 3.54 and the average total science GPA at the time of application was 3.46. Among students with low GPA (less than 3.5 on a 4.0 scale), selfreported patient care experiences showed a positively statistical trend of being associated with PANCE failures ($\beta = 0.0001$; OR = 1.0001, p = 0.052). This data represented in Table 2 indicates that every additional hour of increase in patient experiences increased the odds of PANCE failure by 0.01%.

Model	Dependent Variable	Independent Variable	β	SE	z-statistic	p-value	
1	PANCE Failures	Intercept	-3.1305	0.6687	-4.6811	< 0.0001	
		Employment	-0.0002	0.0003	-0.5834	0.5596	
		High GPA	-1.0389	1.0726	-0.9685	0.3328	
		Interaction	0.0002	0.0003	0.6211	0.5245	
				elihood (df=2) = -23			
2	PANCE Failures	Intercept	-2.2435	0.6254	-3.5873	0.0003	
		Extracurricular	-0.0039	0.0086	-0.4592	0.6461	
		High GPA	-1.1275	1.0098	-1.1166	0.2642	
		Interaction	0.0038	0.0086	0.4431	0.6577	
				elihood (df=2) = -18			
3	PANCE Failures	Intercept	-3.1511	0.6625	-4.7561	<0.0001	
-		Healthcare	-0.0003	0.0005	-0.5895	0.5555	
		Experience					
		High GPA	0.3593	1.0391	0.3458	0.7295	
		Interaction	-0.0156	0.0183	-0.8561	0.3919	
		N = 215, AIC = 50.496, log likelihood (df=2) = -21.248					
4	PANCE Failures	Intercept	-1.9521	0.6910	-2.8252	0.0047	
-		Leadership	-0.0041	0.0046	-0.8936	0.3715	
		High GPA	-1.5734	1.0758	-1.4626	0.1436	
		Interaction	0.0042	0.0046	0.9219	0.3566	
		N = 106, AIC = 45.424, log likelihood (df=2) = -18.712					
5	PANCE Failures	Intercept	-3.9911	0.7638	-5.2254	< 0.0001	
0		Patient Care	0.0001	0.0001	1.9415	0.0522	
		Experience				010022	
		High GPA	0.4321	1.1383	0.3796	0.7043	
		Interaction	-0.0009	0.0011	-0.8014	0.4229	
				ood (df=2) = -21.6		011220	
6	PANCE Failures	Intercept	-2.9087	0.5929	-4.9060	<0.0001	
•		Research	-0.8431	137.7127	-0.0061	0.9951	
		High GPA	-0.7676	0.9296	-0.8257	0.4090	
		Interaction	-0.0140	193.2997	-0.0001	0.9999	
		N = 215, AIC = 50.368, log likelihood (df=2) = -21.184					
7	PANCE Failures	Intercept	-3.4590	0.6597	-5.2428	< 0.0001	
'		Shadowing	0.0001	0.0008	0.1264	0.8994	
		High GPA	-0.0642	1.2198	-0.0527	0.9580	
		Interaction	-0.0042	0.0071	-0.5920	0.5538	
		N = 215, AIC= 54.458, log likelihood (df=2) = -23.229					
8	PANCE Failures	Intercept	26.5661	90577.6039	0003	0.9998	
-		Teaching	0.0000	255.5565	0.0000	1.0000	
		High GPA	0.0000	110250.0742	0.0000	1.0000	
		Interaction	0.0000	353.5806	0.0000	1.0000	
		N = 55, AIC = 8			0.0000	1.0000	
9	PANCE Failures	Intercept	-3.3964	0.6201	-5.4773	<0.0001	
3		Volunteerism	<0.0001	0.0003	-0.1186	0.9056	
			-0.7397				
		High GPA		1.0149	-0.7289	0.4661	
		Interaction	0.0001	0.0006 elihood (df=2) = -23	0.2124	0.8318	

Model 1 2 3 4	Dependent Variable Attrition Attrition Attrition Attrition Attrition Attrition	Intercept Extracurricular High GPA Interaction N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	-2.2745 -0.0003 -2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	SE 0.4213 0.0001 1.3992 0.0002 elihood (df=2) = -3 0.5557 0.0008 1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.4129 0.0001	-4.0933 -0.3919 -1.15975 0.3655	p-value <0.0001 0.3809 0.0132 0.0565 <0.0001 0.6951 0.1102 0.7148 <0.0001 0.0995			
2 3	Attrition	Employment High GPA Interaction N = 235, AIC = Intercept Extracurricular High GPA Interaction N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	-0.0001 -3.4674 0.0003 74.156, log like -2.2745 -0.0003 -2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	0.0001 1.3992 0.0002 elihood (df=2) = -3 0.5557 0.0008 1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	-0.8762 -2.4782 1.9068 3.078 -4.0933 -0.3919 -1.15975 0.3655 8.817 -6.6063 1.6472	0.3809 0.0132 0.0565 <0.00010.69510.11020.7148<0.0001			
3	Attrition	Employment High GPA Interaction N = 235, AIC = Intercept Extracurricular High GPA Interaction N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	-3.4674 0.0003 74.156, log like -2.2745 -0.0003 -2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	1.3992 0.0002 elihood (df=2) = -3 0.5557 0.0008 1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	-2.4782 1.9068 33.078 -4.0933 -0.3919 -1.15975 0.3655 8.817 -6.6063 1.6472	0.0132 0.0565 <0.0001			
3	Attrition	High GPAInteractionN = 235, AIC =InterceptExtracurricularHigh GPAInteractionN = 118, AIC =InterceptHealthcareExperienceHigh GPAInteractionN = 235, AIC =	-3.4674 0.0003 74.156, log like -2.2745 -0.0003 -2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	1.3992 0.0002 elihood (df=2) = -3 0.5557 0.0008 1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	-2.4782 1.9068 33.078 -4.0933 -0.3919 -1.15975 0.3655 8.817 -6.6063 1.6472	0.0132 0.0565 <0.0001			
3	Attrition	Interaction N = 235, AIC = Intercept Extracurricular High GPA Interaction N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	0.0003 74.156, log like -2.2745 -0.0003 -2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	0.0002 elihood (df=2) = -3 0.5557 0.0008 1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	1.9068 3.078 -4.0933 -0.3919 -1.15975 0.3655 8.817 -6.6063 1.6472	0.0565 <0.0001 0.6951 0.1102 0.7148 <0.0001			
3	Attrition	N = 235, AIC =InterceptExtracurricularHigh GPAInteractionN = 118, AIC =InterceptHealthcareExperienceHigh GPAInteractionN = 235, AIC =	74.156, log like -2.2745 -0.0003 -2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	elihood (df=2) = -3 0.5557 0.0008 1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	3.078 -4.0933 -0.3919 -1.15975 0.3655 8.817 -6.6063 1.6472	<0.0001 0.6951 0.1102 0.7148 <0.0001			
3	Attrition	Extracurricular High GPA Interaction N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	-0.0003 -2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	0.0008 1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	-0.3919 -1.15975 0.3655 8.817 -6.6063 1.6472	0.6951 0.1102 0.7148 <0.0001			
		High GPA Interaction N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	-2.0304 0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	1.2719 0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	-1.15975 0.3655 8.817 -6.6063 1.6472	0.1102 0.7148 <0.0001			
		Interaction N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	0.0004 45.635, log like -2.7276 0.0001 -1.4724 -0.0001	0.0011 elihood (df=2) = -1 0.4129 0.0001 0.9079	0.3655 8.817 -6.6063 1.6472	0.7148			
		N = 118, AIC = Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	45.635, log like -2.7276 0.0001 -1.4724 -0.0001	elihood (df=2) = -1 0.4129 0.0001 0.9079	8.817 -6.6063 1.6472	<0.0001			
		Intercept Healthcare Experience High GPA Interaction N = 235, AIC =	-2.7276 0.0001 -1.4724 -0.0001	0.4129 0.0001 0.9079	-6.6063 1.6472				
		Healthcare Experience High GPA Interaction N = 235, AIC =	0.0001 -1.4724 -0.0001	0.0001 0.9079	1.6472				
4	Attrition	Experience High GPA Interaction N = 235, AIC =	-1.4724 -0.0001	0.9079		0.0995			
4	Attrition	High GPA Interaction N = 235, AIC =	-0.0001		-1 6219				
4	Attrition	Interaction N = 235, AIC =	-0.0001		-1 6219	-			
4	Attrition	N = 235, AIC =			-1.0210	0.1048			
4	Attrition		00 FF0 · ···	0.0003	-0.2194	0.8263			
4	Attrition	Intercent		N = 235, AIC = 82.552 log likelihood (df=2) = -37.276					
		Intercept	-2.1352	0.6046	-3.5314	0.0004			
	1	Leadership	-0.0010	0.0016	-0.5910	0.5545			
		High GPA	-3.4178	1.9137	-1.7860	0.0741			
		Interaction	0.0020	0.0017	1.1383	0.2550			
		N = 118, AIC =	42.896, log like	elihood (df=2) = -1	7.448				
5	Attrition	Intercept	-2.5680	0.4318	-5.9469	<0.0001			
		Patient Care	0.0000	0.0001	0.3481	0.7278			
		Experience							
		High GPA	-1.5977	0.9396	-1.7004	0.0890			
		Interaction	0.0000	0.0003	-0.7047	0.9404			
				elihood (df=2) = -3					
6	Attrition	Intercept	-2.4961	0.3818	-6.5384	<0.0001			
		Research	0.0000	0.0003	0.0085	0.9932			
		High GPA	-1.6576	0.8245	-2.0104	0.0444			
		Interaction	<0.0001	0.0007	-0.0310	0.9753			
				elihood (df=2) = -3					
7	Attrition	Intercept	-2.3035	0.4473	-5.1502	<0.0001			
		Shadowing	-0.0008	0.0014	-0.5983	0.9932			
		High GPA	-1.6576	0.8245	-2.0104	0.0444			
		Interaction	0.0035	0.0021	1.7066	0.0879			
		N = 235, AIC= 8		elihood (df=2) = -3	7.167				
8	Attrition	Intercept	-3.2243	1.1479	-2.8088	0.0050			
		Teaching	0.0021	0.0013	1.6718	0.0946			
		High GPA	-18.3418	4977.4328	-0.0037	0.9971			
		Interaction	-0.0021	2.7407	-0.0008	0.9994			
		N = 58, AIC = 1	8.547, log likel	lihood (df=2) = -5.2	274				
9	Attrition	Intercept	-2.4387	0.04014	-6.0750	<0.0001			
		Volunteerism	-0.0001	0.0003	-0.2963	0.7670			
		High GPA	-1.2947	1.0084	-1.2840	0.1992			
		Interaction	-0.0011	0.0024	-0.4713	0.6374			

self-reported Similarly, cumulative healthcare and teaching experiences had positive trends of being associated with attrition among students with low gpas (β = 0.0001; OR = 1.0001, p = 0.099 and β = 0.0021; OR = 1.0021, p = 0.095indicating that per hour respectively), increase of health care and teaching experiences, the odds of failure among students with low gpas increased by 0.01% and 0.21% respectively. High GPA was a significant protective factor against attrition when controlled for employment, research, and shadowing experiences (P < 0.05) and also showed protective trends against attrition when controlled for leadership and patient care experiences (P < 0.1) but did not emerge as a protective factor for PANCE failures (P > 0.1). The attrition statistical analysis is listed in Table 3. None of the models revealed protective effects for selfreported experiences against PANCE failure or attrition.

Conclusion

Contrary to our hypotheses, increased selfreported experiences prior to matriculation into a PA program did not show a protective effect against PANCE failures or attrition but were detrimental. Taken together, prior employment hours, healthcare experience, leadership experience, research experience hours, or teaching experience hours at the time of admission are unlikely to be of value in predicting attrition and PANCE failures among PA students. In addition, a high GPA increased reported noncognitive and experience showed negative trends. These suggested that increased experience hours in the various domains may increase the risk of attrition among high GPA students to a greater extent than low GPA students. This suggests that if a student has a high GPA, but has stayed out of an academic setting for a long while, though their experience hours have increased, their risk of attrition is higher. Table 4 shows attrition events grouped by age range.

Table 4. Cumulative results of variablesacross multiple cohorts							
Year	2017	2018	2019	2020	2021		
Students per Cohort	55	58	58	58	60		
Age at Matriculation	35.6	25.9	26.8	25.1	26.1		
Gender	15(M) 40(F)	16(M) 42(F)	13(M) 45(F)	5(M) 53(F)	18(M) 42(F)		
Average GPA	3.56	3.5	3.53	3.55	3.56		
Average Science GPA	3.43	3.46	3.47	3.43	3.46		
Average Patient Care Hours	1805	1585	1530	1254	2892		

Our study did demonstrate that GPA is a significant predictor of protection from both attrition and PANCE failure. The finding that GPA is a strong predictor of academic success has been demonstrated in other studies.^{3,6,11} However, some of these same referenced studies' analyses revealed an opposite finding from this analysis that previous healthcare experience was positively associated with higher PANCE scores.^{3,6} There are previous analyses that consistent with our findings that are healthcare experience prior to matriculation is unlikely to be a significant predictor of academic outcomes.¹² Due to the discrepancies in the literature, there is a clear need for continued research evaluating noncognitive variables at the time of admission such as healthcare experience while controlling for further variables such as age at admission and specifics of the type of prior healthcare experience rather than a total quantifiable amount.

This study has several limitations. First, the noncognitive variables are self-reported

experience hours by applicants and are not verified by CASPA. Additionally, the only outcome measures were attrition for academic reasons and PANCE failure. Further analyses could be performed to include variables such as decelerations and attrition for personal reasons. Our secondary analysis had a few limitations. Even though detected significant associations we employment between prior hours. healthcare experience. leadership experience, research experience hours, teaching experience hours, and PANCE failures, the observed odds ratios were very close to 1. Furthermore, due to the small sample size, we did not attempt to adjust for the false discovery rate despite testing multiple closely related hypotheses using the same dataset. A final limitation is that the reported healthcare experience hours were not examined separately for differina professions but rather analyzed as total hours spent in any patient contact role. Moreover, this analysis uses only hours to reflect the complex life experiences of students and presupposes the question of whether the way that information is collected for noncognitive variables at the time of application is the most effective means.

This analysis has the strength of having a larger sample size than many of the previous similar analyses examining noncognitive variables at the time of admission. Also, to the best of our knowledge, this study is unique in that it examines previously unexamined domains such as volunteer leadership experience. hours. extracurricular activity hours, and teaching experience. This study was free of bias as quantified variables in the form of hours were used to perform the data analysis to assess the basic outcome measures of PANCE failure and attrition with all students from five cohorts included in the data set.

While healthcare experience hours specifically have been anecdotally weighted as important for success in allied health programs, this analysis did not support that Nonetheless. noncoanitive assertion. admission variables are an important part of the screening process for applicants to PA programs as they can be predictive of other traits such important as emotional intelligence⁸⁻⁹ which are required for success in physician assistant programs and this archival data analysis is not suggestive of the need for elimination of these variables from the selection process.

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