## RESEARCH BRIEF

## Early Identification of "At-Risk" Students at NOVA: Volume 1

This Research Brief is a part of the ongoing institutional effectiveness program at Northern Virginia Community College (NOVA).

- The Persistence Rate (retention, graduation, and transfer-out) is a measure of NOVA's institutional effectiveness, student success, and commitment to its students.
- The Attrition Rate (stop-out) is the percentage of students who did not return to NOVA or graduate from NOVA or transfer-out. The most significant loss of students, as a result of attrition, occurs during the first term and the first year.

On average, 19 percent of the NOVA first-time in college (FTIC), program-placed students stop-out (do not return, graduate, or transfer) after the first term, and an additional 16 percent of the students stop-out by the end of the first year ( 35 percent). Therefore, intervening early in the first year to retain students past the first term and first year is an efficient way to increase the Institution's persistence rates.

Figure 1. First-Term and First-Year Attrition Rates of First-Time in College, Program-Placed Students: Fall 2014, Fall 2015, and Fall 2016 Cohorts ${ }^{1}$


[^0]To enable early identification of students who are at risk for attrition, three logit models were developed by analyzing the following:

1. The predictive relationship of entering student characteristics with persistence after first term.
2. The predictive relationship of entering student characteristics and first-term momentum metrics with persistence after first year.
3. The predictive relationship of entering student characteristics and first year momentum metrics with persistence after third year.

This volume (Volume 1) presents the results of the first logit model (persistence after first term). Volumes 2 and 3 will present the results of second and third logit models. The study population for the first logit model consisted of 22,805 first-time in college (FTIC), program-placed students who entered NOVA in the fall terms of 2014, 2015, and 2016. To increase the generalizability of the model beyond the data on which the model is fit, three years of data were used in this analysis.

## Key Findings

$>$ An odds ratio of greater than one indicates that the likelihood of a target event occurring is more likely in the focus group than in the non-focus group. An odds ratio of one implies that the target event is equally likely for the two groups. An odds ratio of less than one suggests that the target event is less likely to occur in the focus group than in non-focus group (Meyers et al., 2017).

In this study population, the odds ratios indicate the following (controlling for all other predictors).

- Gender: The odds of a male student persisting after the first term were less than ( 0.70 times) the odds of a female student persisting.
- Race/Ethnicity:
- The odds of an Asian student persisting after the first term were 1.62 times higher than the odds of a White student persisting.
- Conversely, the odds of a Hispanic student ( 0.96 times) or a student of other/unknown ethnicity ( 0.98 ) persisting after the first term were slightly lower than the odds of a White student persisting.
- The odds of a Black student ( 0.99 times) persisting after the first term were nearly equal to the odds of a White student persisting.
- Degree Program:
- The odds of persisting after the first term for a student pursuing an A.A.A. or A.A.S. degree were 0.76 times the odds of persisting for a student pursuing an A.A. or A.S. degree.
- The odds of persisting after the first term for a student pursuing a certificate were 0.69 times the odds of persisting for a student pursuing an A.A. or A.S. degree.
- Major: The odds of persisting after the first term for a student enrolled in Science (A.S.) (1.37 times) or Business Administration (A.S.) (1.11 times) or Social Sciences (A.S.) (1.06 times) were higher than the odds of persisting for a student enrolled in General Studies (A.S.).
- SDV Course Enrollment: The odds of persisting after the first term for a student who enrolled in an SDV course in the first term were 1.38 times higher than the odds of persisting for a student who did not enroll in an SDV course.
- Developmental Course Enrollment: The odds of persisting after the first term for a student who enrolled in a developmental course in the first term were 0.86 times the odds of persisting for a student who did not enroll in a developmental course.
- Day Course Enrollment:
- The odds of persisting after the first term for a student who enrolled in only day courses were 1.24 times higher than the odds of persisting for a student who enrolled in a combination of both day and night courses.
- The odds of persisting after the first term for a student who enrolled in only night courses were about the same ( 0.99 times) as the odds of a student enrolled in a combination of both day and night courses.
- College-Level Course Enrollment:
- The odds of persisting after the first term for a student who enrolled in a college-level English course in the first term were 1.22 times higher than the odds of persisting for a student who did not enroll in college-level English.
- Likewise, the odds of persisting after the first term for a student who enrolled in a college-level math course in the first term were 1.17 times higher than the odds of persisting for a student who did not enroll in college-level math.
- Credits Enrolled: For every one credit increase in the number of credits enrolled during the first term, there was a 17 percent increase in the odds of persisting after the first term.
- Instruction Mode:
- The odds of persisting after the first term for a student enrolling only in online courses were 0.67 times the odds of persisting for students enrolling in online and on-campus courses.
- The odds of persisting after the first term for students enrolling in only on-campus courses were 0.94 times the odds of persisting for students enrolling in online and oncampus courses.

To identify students who are at risk for attrition, a binary logit model was developed by analyzing the predictive relationship between student persistence after first term and each of the following entering student characteristics (Table 1):

Table 1. Entering Student Characteristics

| Gender | Distance from home to campus |
| :--- | :--- |
| Race/ethnicity | Residence (in-state or out-of-state) |
| Age at initial enrollment | Enrollment in developmental courses |
| Entering major | Enrollment in an SDV course |
| Degree (A.A./A.S., A.A.A./A.A.S., or certificate) | Enrollment in college-level math |
| Recent high school graduate | Enrollment in college-level English |
| Day/Night Status | Enrollment in NOVA Online courses |
| First-Generation status | Credits enrolled in the first term |

Data availability and previous research on student success in community colleges provided the rationale for the selection of variables. Due to large sample size, the significance level cutoff was set at $p<0.05$ for a variable to stay in the model. This allowed for the development of a more parsimonious model without losing explanatory power. Listwise deletion was used wherever there were missing data. In essence, any student who had a missing value on any of the predictors was
excluded from the study. Thirty-six students were excluded from regression analyses because of missing data.

Table 2 presents the statistics assessing the model fit. The results indicate that the model provided a statistically significant prediction of persistence, $\mathrm{x} 2(20)=2,286.37, \mathrm{p}<0.001$. The Max-rescaled RSquare indicated that the model accounted for approximately 15 percent of the total variance. Classification accuracy for the cases based on a classification cutoff value of 0.65 for predicting membership in the persisted group was high. The overall rate of correct classification was 80 percent, with 92 percent of persisted (sensitivity), and 31 percent of not persisted (specificity) being correctly classified. In the opinion of Hosmer and Lemeshow (2000), "the classification table is most appropriate when classification is a stated goal of the analysis; otherwise it should only supplement more rigorous methods of assessment of fit" (p. 160). The Hosmer and Lemeshow goodness of fit test which yields a test statistic significance level above 0.05 leads to a failure to reject the null hypothesis for the model that there is no difference between the observed and model predicted values of the dependent variable; therefore, the model's estimates are acceptable (Menard, 1995). The Hosmer and Lemeshow test in this model resulted in a goodness-of-fit statistic of 15.21, distributed as a chi square value, and is associated with a $p$ value of 0.055 , indicating an acceptable match between predicted and observed probabilities. The area under the curve (AUC) was 0.723 suggesting that the fit of the logistic regression model is acceptable (Meyers et al., 2017).

Table 2. Model Fit Statistics ( $\mathrm{N}=22,769$, Persisted $=18,510$ )

| Model Fit Statistics |  |  |
| :--- | :---: | :---: |
| Likelihood Ratio X2 | $2,286.37$ | $\mathrm{p}<0.001$ |
| Max-rescaled R-Square | 0.154 |  |
| -2 Log likelihood | 21945.711 |  |
| AIC | 21947.711 |  |
| SC | 21955.744 |  |
| Hosmer and Lemeshow Goodness-of-Fit Test | 15.208 | $\mathrm{p}=0.055$ |
| Correctly Classified \% | 80.3 |  |
| Sensitivity | 91.8 |  |
| Specificity | 30.5 |  |
| Area Under Curve (AUC) | 0.723 |  |

Table 3 (next page) presents the standard results produced by logistic regression using the study population (coefficient estimates and their respective standard errors, Wald statistics, p-values, and odds ratios).
> The Wald test indicated that gender, race/ethnicity, degree type, major, credits enrolled in first term, enrollment in developmental, SDV, college-level math and English courses, enrollment in only day courses, and instruction mode were statistically significant predictors of persistence after the first term in the study population.

Table 3. Logistic Regression Predicting First Term Persistence

| Predictors | B | S.E. | Wald | p | OR | $\mathbf{9 5 \%} \mathbf{C l}$ |  |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Intercept | -0.537 | 0.094 | 32.466 | $<0.001$ |  |  |  |
| Male vs Female | -0.363 | 0.038 | 90.946 | $<0.001$ | 0.696 | 0.646 | 0.750 |
| Black vs White | -0.008 | 0.052 | 0.025 | 0.875 | 0.992 | 0.896 | 1.098 |
| Hispanic vs White | -0.045 | 0.046 | 0.968 | 0.325 | 0.956 | 0.873 | 1.046 |
| Asian vs White | 0.481 | 0.065 | 54.889 | $<0.001$ | 1.617 | 1.424 | 1.837 |
| Other Unknown vs White | -0.021 | 0.073 | 0.083 | 0.773 | 0.979 | 0.848 | 1.130 |
| Cert vs A.A./A.S. | -0.371 | 0.107 | 11.988 | 0.001 | 0.690 | 0.559 | 0.851 |
| A.A.A./A.A.S. vs A.A./A.S. | -0.273 | 0.062 | 19.587 | $<0.001$ | 0.761 | 0.674 | 0.859 |
| Online Only vs Mixed | -0.403 | 0.100 | 16.173 | $<0.001$ | 0.668 | 0.549 | 0.813 |
| On-campus Only vs Mixed | -0.064 | 0.074 | 0.748 | 0.387 | 0.938 | 0.812 | 1.084 |
| Day Only vs Day and Night | 0.218 | 0.044 | 24.426 | $<0.001$ | 1.243 | 1.140 | 1.355 |
| Night Only vs Day and Night | -0.008 | 0.102 | 0.007 | 0.935 | 0.992 | 0.812 | 1.211 |
| Business Admin vs General Studies | 0.100 | 0.063 | 2.530 | 0.112 | 1.105 | 0.977 | 1.249 |
| Social Sciences vs General Studies | 0.059 | 0.064 | 0.852 | 0.356 | 1.061 | 0.935 | 1.204 |
| Science vs General Studies | 0.316 | 0.076 | 17.454 | $<0.001$ | 1.372 | 1.183 | 1.592 |
| All Other Majors vs General Studies | 0.231 | 0.052 | 20.121 | $<0.001$ | 1.260 | 1.139 | 1.394 |
| Credits Enrolled in Fall | 0.156 | 0.006 | 776.669 | $<0.001$ | 1.169 | 1.156 | 1.181 |
| Enrolled in Developmental Course | -0.148 | 0.043 | 11.647 | 0.001 | 0.862 | 0.792 | 0.939 |
| Enrolled in SDV Course | 0.322 | 0.039 | 67.275 | $<0.001$ | 1.380 | 1.278 | 1.491 |
| Enrolled in College-level Math | 0.158 | 0.048 | 10.705 | 0.001 | 1.171 | 1.066 | 1.288 |
| Enrolled in College-level English | 0.202 | 0.042 | 23.144 | $<0.001$ | 1.224 | 1.127 | 1.329 |

Notes: B=coefficient estimates; SE=standard error; Wald=Wald statistic; p=p-value; OR=odds ratio; Cl=confidence interval for odds ratio. All Other Majors: Includes all majors other than General Studies (A.S.), Business Administration (A.S.), Social Sciences (A.S.), and Science (A.S.).

The odds ratio is known as odds change, which describes the proportionate change in the odds for one-unit difference in the explanatory variable (Hosmer and Lemeshow, 2000; and Menard, 1995). Logistic regression models the log odds of a positive response (probability modeled is persisted=1) as a linear combination of the predictor variables. In logistic regression, the dependent variable (persisted) is a logit, exponentiating logit gives us odds, dividing the odds by odds+1 gives probability.

In this model, the estimated logit is given by the following expression:

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Log [p / (1-p)] = - 0.5366-(0.3629*male) + (0*White) - (0.0082*Black) - (0.0454*Hispanic) +
(0.4808*Asian) - (0.0211*Other Race/Ethnicity) + (0*AA/AS) - (0.2730*AAA/AAS) -
(0.3708*Certificate) + (0.2310*Other Major) + (0.0997*Business Admin) + (0.0595*Social Sciences)
+ (0.3163*Science) + (0*General Studies) - (0.4034*Online Only) - (0.0636*On-campus Only) +
(0*On-Campus and Online) + (0.2176*Day Only) - (0.0083*Night Only) + (0*Day and Night) +
(0.1558*Credits Enrolled in fall) - (0.1480*Enrolled in Developmental Course) + (0.3224*Enrolled in
SDV Course) + (0.1583*Enrolled in College-level Math) + (0.2022*Enrolled in College-level English)
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Estimates produced by the model were validated by computing the predicted probability of persisting past the first term for each of the Fall 2017 FTIC program-placed students (Table 4, next page). Error rate and accuracy are the most common and intuitive measures derived from the confusion matrix.

- Error rate is calculated as the number of all incorrect predictions (cells in yellow) divided by the total number of the dataset. The best error rate is 0.0 (or 0 percent), whereas the worst is 1.0 (or 100 percent).
- Accuracy is calculated as the number of all correct predictions (cells in green) divided by the total number of the dataset. The best accuracy is 1.0 (or 100 percent), whereas the worst is 0.0 (or 0 percent). It can also be calculated by 1 - Error Rate.

Table 4 present the results of model performance on the Fall 2017 cohort.
Table 4. Confusion Matrix: Fall 2017 Cohort

| Predicted by the Model | Observed (Actual) |  | Total |
| :--- | ---: | ---: | ---: |
|  | Persisted past $\mathbf{1}^{\text {st }}$ Term | Did not persist |  |
| Persisted past $1^{\text {st }}$ Term | 5,124 | 975 | 6,099 |
| Did not persist | 442 | 394 | 836 |
| Total | $\mathbf{5 , 5 6 6}$ | $\mathbf{1 , 3 6 9}$ | $\mathbf{6 , 9 3 5}$ |

Note: First-term persistence is calculated as the number of students who graduated from NOVA or transferred to a four-year institution by the end of the first term, or who returned to NOVA for the second term of study.

The error rate based on Fall 2017 test data was 20 percent (cells in yellow) and accuracy was 80 percent (cells in green). When compared to the training data (i.e., the data with which the model was developed), the overall rate of correct classification ( 80 percent) and sensitivity (i.e., percent correctly classified as persisted) ( 92 percent) were same with test data. The specificity (i.e., percent correctly classified as not persisted) decreased marginally by 2 percentage points from 31 percent to 29 percent. Therefore, we can conclude that model performed reasonably well with the new data.

## Use of Regression Results

Using the estimates produced by the model, the Office of Institutional Research will compute the predicted probability of persisting after first term for the Fall 2019 FTIC, program-placed cohort of students. Students with low predicted probability (i.e., lower than the classification cutoff value) will be categorized as "at-risk." The list of at-risk students will be provided to the College for targeting intervention strategies.

## Limitations

Many variables affect a student's decision to remain at or leave an institution of higher education. Variables that affect persistence are generally categorized as cognitive (intellectual), non-cognitive, and environmental. The challenge of student retention is complex. Not all students are alike, nor are programs and institutions. This analysis did not examine the relationship between student support services and student persistence. The type and quality of counseling, advising, orientation, and learning assistance services can have positive effect on student persistence. Additionally, this analysis examined only the data that was available to the Office of Institutional Research. For example, a preliminary review of the data revealed that 35 percent of the study population did not submit the Free Application for Federal Student Aid (FAFSA). Therefore, after reviewing the various techniques for handling missing data (listwise and pairwise deletions, mean substitution, last observation carried forward, maximum likelihood, expectation-maximization, and multiple imputation), it was decided not to include financial aid variables. High school GPA and admission test scores were not available; therefore, these variables were not included.
$\square$
Retaining students is fundamental to an institution's ability to carry out its mission. Institutions of higher education have a distinct responsibility to help students complete their goals. That responsibility can be fulfilled effectively when institutions use available information to properly counsel students and to implement programs and policies that work for student persistence. Literature on student persistence suggests that institutions use research specific to their institution and to their student populations as a basis for designing and implementing student persistence programs. Early identification of at-risk students is crucial. If not identified, institutions will notice students who are struggling academically only after they start to fail and interventions may only be attempted after several failures, which reduces the likelihood of persistence. Therefore, in order to improve academic performance and, thus, improve student persistence, an early warning system for at-risk students is recommended.

The Office of Institutional Research welcomes comments and suggestions regarding this report.

## References

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## Appendix

Table A1. First-Term Persistence by Student Characteristics: Fall 2014 through Fall 2016 Cohorts

|  |  | Fall 201 |  |  | Fall 201 |  |  | Fall 201 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | N | Persisted past $1^{\text {st }}$ Term |  | N | Persisted past $1^{\text {st }}$ Term |  | N | Persisted past $1^{\text {st }}$ Term |  |
|  |  | \# | \% |  | \# | \% |  | \# | \% |
| Gender |  |  |  |  |  |  |  |  |  |
| Male | 4,059 | 3,187 | 78.5\% | 4,224 | 3,390 | 80.3\% | 3,883 | 3,100 | 79.8\% |
| Female | 3,670 | 3,051 | 83.1\% | 3,523 | 2,952 | 83.8\% | 3,446 | 2,859 | 83.0\% |
| Total | 7,729 | 6,238 | 80.7\% | 7,747 | 6,342 | 81.9\% | 7,329 | 5,959 | 81.3\% |
| Race/Ethnicity |  |  |  |  |  |  |  |  |  |
| White | 2,482 | 2,003 | 80.7\% | 2,470 | 1,975 | 80.0\% | 2,334 | 1,882 | 80.6\% |
| Black/African American | 1,382 | 1,059 | 76.6\% | 1,364 | 1,071 | 78.5\% | 1,255 | 1,015 | 80.9\% |
| Hispanic/Latino | 2,151 | 1,721 | 80.0\% | 2,151 | 1,745 | 81.1\% | 2,107 | 1,669 | 79.2\% |
| Asian | 1,175 | 1,026 | 87.3\% | 1,160 | 1,055 | 90.9\% | 1,152 | 1,018 | 88.4\% |
| American Indian/Alaska Native | 14 | 13 | 92.9\% | 14 | 10 | 71.4\% | 18 | 11 | 61.1\% |
| Hawaiian/Pacific Islander | 42 | 34 | 81.0\% | 32 | 27 | 84.4\% | 39 | 27 | 69.2\% |
| Not Specified | 2 | 2 | 100.0\% | 13 | 13 | 100.0\% | 6 | 6 | 100.0\% |
| Two or More Races | 375 | 291 | 77.6\% | 430 | 346 | 80.5\% | 339 | 262 | 77.3\% |
| Unknown | 106 | 89 | 84.0\% | 113 | 100 | 88.5\% | 79 | 69 | 87.3\% |
| Total | 7,729 | 6,238 | 80.7\% | 7,747 | 6,342 | 81.9\% | 7,329 | 5,959 | 81.3\% |
| Program Placement |  |  |  |  |  |  |  |  |  |
| AAA/AAS | 942 | 689 | 73.1\% | 874 | 651 | 74.5\% | 727 | 525 | 72.2\% |
| AA/AS | 6,550 | 5,395 | 82.4\% | 6,720 | 5,582 | 83.1\% | 6,468 | 5,348 | 82.7\% |
| Certificate | 237 | 154 | 65.0\% | 153 | 109 | 71.2\% | 134 | 86 | 64.2\% |
| Total | 7,729 | 6,238 | 80.7\% | 7,747 | 6,342 | 81.9\% | 7,329 | 5,959 | 81.3\% |
| Developmental Course Enrollment |  |  |  |  |  |  |  |  |  |
| Enrolled | 2,773 | 2,247 | 81.0\% | 2,692 | 2,200 | 81.7\% | 2,531 | 2,045 | 80.8\% |
| Did not Enroll | 4,956 | 3,991 | 80.5\% | 5,055 | 4,142 | 81.9\% | 4,798 | 3,914 | 81.6\% |
| Total | 7,729 | 6,238 | 80.7\% | 7,747 | 6,342 | 81.9\% | 7,329 | 5,959 | 81.3\% |
| SDV Course Enrollment |  |  |  |  |  |  |  |  |  |
| Enrolled | 5,370 | 4,515 | 84.1\% | 5,193 | 4,402 | 84.8\% | 5,334 | 4,558 | 85.5\% |
| Did not Enroll | 2,359 | 1,723 | 73.0\% | 2,554 | 1,940 | 76.0\% | 1,995 | 1,401 | 70.2\% |
| Total | 7,729 | 6,238 | 80.7\% | 7,747 | 6,342 | 81.9\% | 7,329 | 5,959 | 81.3\% |
| College-level English Course Enrollment |  |  |  |  |  |  |  |  |  |
| Enrolled | 5,464 | 4,631 | 84.8\% | 5,610 | 4,798 | 85.5\% | 5,356 | 4,513 | 84.3\% |
| Did not Enroll | 2,265 | 1,607 | 70.9\% | 2,137 | 1,544 | 72.3\% | 1,973 | 1,446 | 73.3\% |
| Total | 7,729 | 6,238 | 80.7\% | 7,747 | 6,342 | 81.9\% | 7,329 | 5,959 | 81.3\% |
| College-level Math Course Enrollment |  |  |  |  |  |  |  |  |  |
| Enrolled | 3,100 | 2,727 | 88.0\% | 3,191 | 2,836 | 88.9\% | 3,056 | 2,683 | 87.8\% |
| Did not Enroll | 4,629 | 3,511 | 75.8\% | 4,556 | 3,506 | 77.0\% | 4,273 | 3,276 | 76.7\% |
| Total | 7,729 | 6,238 | 80.7\% | 7,747 | 6,342 | 81.9\% | 7,329 | 5,959 | 81.3\% |

Note: First-term persistence is calculated as the number of students who graduated from NOVA or transferred to a four-year institution by the end of the first term, or who returned to NOVA for the second term of study.


[^0]:    ${ }^{1}$ The first-term attrition rate is calculated as the number of students who did not return to NOVA for the second term of study nor did they graduate from NOVA or transfer to a four-year institution by the end of the first term, divided by the number of students who started at NOVA. The first-year attrition rate is calculated as the number of students who did not return to NOVA in the following Fall term of study nor did they graduate from NOVA or transfer to a four-year institution by the end of the first year, divided by the number of students who started at NOVA.

