

Factors Associated with Grade 3 Reading Outcomes of Students in the Commonwealth of the Northern Mariana Islands Public School System

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See <https://go.usa.gov/xMxVf> for the full report.

Appendix A. About the study

This appendix provides background on the current reading proficiency among students in the Commonwealth of the Northern Mariana Islands (CNMI) and the factors that might be associated with it and discusses Head Start and other CNMI early care and education programs.

Reading proficiency among students on the Commonwealth of the Northern Mariana Islands

In 2019, 24 percent of grade 3 students in the CNMI Public School System demonstrated proficiency on standardized measures of reading skills in English (Commonwealth of the Northern Mariana Islands Office of Accountability, Research, and Evaluation, 2019). Research indicates that meeting reading benchmarks in grade 3 positively correlates with later academic outcomes (Hein et al., 2013) and postsecondary enrollment. One study found that students who met grade 3 reading proficiency benchmarks graduated from high school and enrolled in college at higher rates than students who did not meet the benchmarks (Lesnick et al., 2010). The CNMI Public School System aims to support students not only in meeting academic achievement standards but also in acquiring “essential skills for lifelong learning, and ha[ving] the ability to transition into higher education and/or the workforce” (Commonwealth of the Northern Mariana Islands Public School System, 2021b). This means that gaining further insight into the relationship between reading performance and characteristics of grade 3 CNMI students is of critical interest.

Many CNMI students are members of historically under-resourced groups. For example, 44 percent of families in the CNMI have a household income below the federal poverty level compared with 12 percent of all U.S. families (Commonwealth of the Northern Mariana Islands Department of Commerce, 2018). Prior research indicates that students who live in lower income households or who reside in under-resourced communities are at higher risk for difficulties in school, including lower reading and math achievement (Burchinal et al., 2011; Pungello et al., 1996; Welsh et al., 2010). One study that used data from the U.S. Department of Education’s Early Childhood Longitudinal Study Kindergarten Cohort found that by the time children from the most disadvantaged socioeconomic group began kindergarten, they had cognitive scores that were 60 percent lower on average than those of children in the least disadvantaged socioeconomic group (Lee & Burkam, 2002).

Other student demographic characteristics have been shown to influence student performance. For example, a U.S. Government Accountability Office study conducted in the continental United States found that school mobility was associated with lower performance on measures of academic skills and a higher probability of dropping out of school (Ashby, 2010). The same study showed that by grade 8, 70 percent of students had attended two or fewer schools, 18 percent had attended three schools, and 13 percent had attended four schools (Ashby, 2010). A similar study conducted on early grades found that 36 percent of students changed schools once between kindergarten and grade 3, whereas 55 percent of students did not change schools at all (Burkam et al., 2009). Although the current study did not explore the reasons why CNMI students changed schools, prior research indicates that mobility is often associated with economic or social challenges (Mehana & Reynolds, 2004). Enrolling in a new school might compound these issues by creating gaps in attendance, introducing inconsistencies in exposure to curricula and pedagogy, and disrupting relationships with teachers and peers (Paik & Philips, 2002).

CNMI stakeholders have suggested that enrollment in an early childhood education program might be associated with grade 3 reading proficiency. Growing evidence suggests a positive relationship between enrollment in a high-quality early childhood education program and reading skills development (Yoshikawa et al., 2016). Although some evidence indicates that the impacts of programs like Head Start do not persist beyond kindergarten, other studies suggest that the impacts might correlate with later academic outcomes (Deming, 2009). For example, middle school students who had enrolled in Head Start were more likely than students who had not to have higher math scores on state tests and were less likely to be retained by the time they reached grade 8 (Philips et al., 2016).

The extent to which results from the continental United States hold true in the CNMI is unclear. CNMI Public School System stakeholders are interested in exploring how student demographic characteristics (such as mobility) and other education experiences (such as Head Start enrollment) are related to grade 3 reading proficiency. A better understanding of these relationships is expected to provide CNMI Public School System stakeholders with additional information to inform current and future initiatives, instruction, and student supports.

Head Start and other early care and education programs

Families in the CNMI have access to early childcare services through the CNMI Public School System, 25 licensed facilities, and nine license-exempt home care providers (Commonwealth of the Northern Mariana Islands Preschool Development Grant Birth through Five, 2020). The CNMI Head Start program has been in place for more than 30 years. As of 2019, nine Head Start centers were in operation across the islands of Saipan, Tinian, and Rota, with morning, afternoon, and full-day sessions available, depending on the number of students enrolled (Commonwealth of the Northern Mariana Islands Office of Accountability, Research, and Evaluation, 2019). The extent to which CNMI children not enrolled in Head Start attend other preschool programs is unknown because programs are spread out across the three islands and data collection is inconsistent and not centralized (Commonwealth of the Northern Mariana Islands Preschool Development Grant Birth through Five, 2020).

Head Start is the largest and longest running publicly funded early childhood education program for low-income families with preschool-age children in the United States. Founded in 1965 as part of President Lyndon Johnson's Great Society program, Head Start provides more than 900,000 children and their families access to preschool education and a range of health, nutrition, and education services each year. More than 1,600 local grantees administer Head Start programs across all 50 states, the District of Columbia, and U.S. territories.

Many state and local governments supplement federal Head Start programs and offer non-Head Start preschool programs to a substantial number of children. According to the National Institute for Early Education Research, total state funding for preschool in 2015 was \$6.2 billion (Barnett & Friedman-Krauss, 2016). These funds provide support for nearly 1.4 million children to attend state-funded preschool programs, with approximately 13 percent

of three-year-old children and 38 percent of four-year-old children enrolled in either Head Start or other publicly funded preschool programs (Barnett & Friedman-Krauss, 2016).

Despite the increased availability of publicly funded preschool programs, not all eligible children participate in them. In 2015 Head Start provided services to 18 percent of low-income three-year-old children and 21 percent of low-income four-year-old children (Barnett & Friedman-Krauss, 2016). Children from low-income families are less likely to attend preschool than their peers from higher income families (Crosnoe et al., 2016; López et al., 2017). Lower enrollment rates among children from low-income families are worth addressing, as growing evidence suggests a relationship between participating in a high-quality early childhood education program and starting kindergarten with the skills and abilities that support school success (Yoshikawa et al., 2016). Although the CNMI Head Start program has worked to recruit and enroll children—and many children in the CNMI are eligible for enrollment—enrollment rates have fallen in recent years (Commonwealth of the Northern Mariana Islands Office of Accountability, Research, and Evaluation, 2019). Even though Head Start eligibility requirements limit the number of participating families based on income levels (Sec. 645. [42 U.S.C. 9840]), a small percentage of CNMI families enroll in the program regardless of income (Commonwealth of the Northern Mariana Islands Public School System, 2021a; Early Childhood Learning and Knowledge Center, 2021). Per Head Start policy, this number cannot exceed 10 percent of enrolled children (Early Childhood Learning and Knowledge Center, 2021).

CNMI stakeholders are interested in examining the relationship between enrollment in the Head Start program and grade 3 reading performance, among other potential predictors. Because Head Start is under the auspices of CNMI Public School System, stakeholders are particularly interested in understanding outcomes associated with Head Start, but there is limited research on the program regarding its possible relationship to CNMI student K–12 outcomes.

References

- Ashby, C. M. (2010). *K–12 education: Many challenges arise in educating students who change schools frequently*. Report to congressional requesters (GAO-11-40). U.S Government Accountability Office. <https://eric.ed.gov/?id=ED514099>.
- Barnett, W. S., & Friedman-Krauss, A. H. (2016). *State(s) of Head Start*. National Institute for Early Education Research. https://nieer.org/wp-content/uploads/2016/12/HS_Full_Reduced.pdf.
- Burchinal, M., McCartney, K., Steinberg, L., Crosnoe, R., Friedman, S. L., McLoyd, V., Pianta, R., & NICHD Early Child Care Research Network (2011). Examining the Black–White achievement gap among low-income children using the NICHD study of early child care and youth development. *Child Development*, 82(5), 1404–1420. <https://doi.org/10.1111/j.1467-8624.2011.01620.x>.
- Burkam, D. T., Lee, V. E., & Dwyer, J. (2009). *School mobility in the early elementary grades: Frequency and impact from nationally-representative data*. Paper commissioned by the Committee on Impact of Mobility and Change on the Lives of Young Children, Schools, and Neighborhoods. Retrieved February 20, 2021, from <https://www.fcd-us.org/school-mobility-in-the-early-elementary-grades-frequency-and-impact-from-nationally-representative-data/>.
- Commonwealth of the Northern Mariana Islands Department of Commerce. (2018). *CNMI economic census 2017*. Retrieved September 18, 2020, from <http://ver1.cnmicommerce.com/sy-2017-table-5-12-15-poverty-level/>.
- Commonwealth of the Northern Mariana Islands Office of Accountability, Research, and Evaluation. (2019). *Annual report SY 2018–2019*. Commonwealth of the Northern Mariana Islands Public School System. https://www.cnmipssoare.org/application/files/8815/8031/8258/SY_2018-2019_CNMI_PSS_Annual_Report_1.24b_ForPrinters.pdf.
- Commonwealth of the Northern Mariana Islands Preschool Development Grant Birth through Five. (2020). *CNMI PDG B–5: Safe, healthy, and thriving children*. <http://www.cnmiclp.gov/wp-content/uploads/2020/06/PDG-B-5-APPLICATION.pdf>.

- Commonwealth of the Northern Mariana Islands Public School System. (2021a). *Head Start/Early Head Start executive summary*. Retrieved September 27, 2020, from <https://cnmipss.org/head-startearly-head-start>.
- Commonwealth of the Northern Mariana Islands Public School System. (2021b). *Office of Curriculum and Instruction executive summary*. Retrieved September 18, 2020, from <https://cnmipss.org/office-curriculum-and-instruction>.
- Crosnoe, R., Purtell, K. M., Davis-Kean, P., Ansari, A., & Benner, A. D. (2016). The selection of children from low-income families into preschool. *Developmental Psychology*, 52(4), 599–612. <https://doi.org/10.1037/dev0000101>.
- Deming, D. (2009). Early childhood intervention and life-cycle skill development: Evidence from Head Start. *American Economic Journal: Applied Economics*, 1(3), 111–134. <https://doi.org/10.1257/app.1.3.111>.
- Early Childhood Learning and Knowledge Center. (2021). *Head Start policy & regulations: Sec. 645 participation in Head Start programs*. Retrieved September 18, 2020, from <https://eclkc.ohs.acf.hhs.gov/policy/head-start-act/sec-645-participation-head-start-programs>.
- Hein, V., Smerdon, B., & Sambolt, M. (2013). *Predictors of postsecondary success*. American Institutes for Research College and Career Readiness and Success Center. <https://eric.ed.gov/?id=ED555671>.
- Lee, V. E., & Burkam, D. T. (2002). *Inequality at the starting gate: Social background differences in achievement as children begin kindergarten*. Economic Policy Institute.
- Lesnick, J., Goerge, R., Smithgall, C., & Gwynne J. (2010). *Reading on grade level in third grade: How is it related to high school performance and college enrollment?* Chapin Hall at the University of Chicago.
- López, M., Grindal, T., Zanoni, W., & Goerge, R. (2017). *Hispanic children's participation in early care and education: A look at utilization patterns of Chicago's publicly funded programs*. National Research Center on Hispanic Children and Families. Retrieved March 18, 2021, from <https://www.childtrends.org/publications/hispanic-children-chicago-ece>.
- Mehana, M., & Reynolds, A. J. (2004). School mobility and achievement: A meta-analysis. *Children and Youth Services Review*, 26(1), 93–119. <https://doi.org/10.1016/j.childyouth.2003.11.004>.
- Paik, S., & Phillips, R. (2002). *Student mobility in rural communities: What are the implications for student achievement?* North Central Regional Educational Laboratory. <https://eric.ed.gov/?id=ED466669>.
- Philips, D., Gormley, W., & Anderson, S. (2016). The effects of Tulsa's CAP Head Start program on middle school academic outcomes and progress. *Developmental Psychology*, 52(8), 1247–1261. <https://doi.org/10.1037/dev0000151>.
- Pungello, E. P., Kupersmidt, J. B., Burchinal, M. R., & Patterson, C. J. (1996). Environmental risk factors and children's achievement from middle childhood to early adolescence. *Developmental Psychology*, 32(4), 755–767. <https://doi.org/10.1037/0012-1649.32.4.755>.
- Welsh, J. A., Nix, R. L., Blair, C., Bierman, K. L., & Nelson, K. E. (2010). The development of cognitive skills and gains in academic school readiness for children from low-income families. *Journal of Educational Psychology*, 102(1), 43–53. <https://doi.org/10.1037/a0016738>.
- Yoshikawa, H., Weiland, C., & Brooks-Gunn, J. (2016). When does preschool matter? *Future of Children*, 26(2), 21–35. <https://eric.ed.gov/?id=EJ1118535>.

Appendix B. Methods

This appendix describes the study’s data sources, the process for matching and cleaning data, measures and the process for recoding variable, and analytic methods.

Data sources

District staff from the Commonwealth of the Northern Mariana Islands (CNMI) Public School System, including the Head Start program, provided the study team with the administrative data files used to conduct the analyses in this study. At the time of the study requests for CNMI data, furloughs at the CNMI Public School System and the spread of COVID-19 limited staff capacity to compile and share school- and district-level data with the study team. The data files lacked consistent information on teacher education, teacher core academic skills, student absenteeism, student English learner status, and student eligibility for special education services. As a result, the study was unable to analyze these data.

Student-level administrative data were collected by schools and shared with CNMI Public School System staff, who then shared the data files with the study team. The information was provided for all students enrolled in kindergarten, grade 1, grade 2, and grade 3 during the 2011/12–2018/19 school years. The study team received information on student enrollment in Head Start for the 2009/10–2015/16 school years. In addition to the administrative data, CNMI public schools collected student performance data on the ACT Aspire reading assessment. School leaders shared the files from 2011/12–2018/19 with the CNMI Public School System staff, who then provided the information to the study team.¹

Matching and cleaning data

The student identifier in the student-level files was not consistent across years. To link student records across years, the study team used students’ full name and date of birth and the distance function (COMPGED) in SAS. This function generates a distance measure that characterizes the difference between two strings. A distance of 0 represents an exact match, and any distance more than 5 was reviewed by the study team before a match was confirmed. This approach enabled 98 percent of student records to be matched.

Measures and variable recoding

The measures used in the study are discussed below. The description of each measure includes the methods used to recode the variables prior to analysis so that the coefficient for a given categorical variable could be compared with the grand mean outcome rather than with the outcomes for another categorical variable. Deviation coding was used for categorical variables. The specific values for the predictor variables used in the research question analyses are presented in table B1.

Outcome variable

Grade 3 reading proficiency. The reading proficiency of CNMI students was measured at the end of grade 3 using the reading domain of the ACT Aspire assessment. ACT Aspire is a computer-based standardized benchmark assessment that measures academic skills in five domains (English, math, reading, science, and writing) in grades 3–10. ACT Aspire reports an overall scale score and readiness and scale scores in four domains (writing is collapsed within the other content-area domains). The study team used the readiness and scaled scores from the reading domain of the spring 2015–19 grade 3 administrations as the outcome measure. The reading domain covers

¹ The CNMI Public School System also shared teacher-level data, such as teacher evaluation scores, but the teacher-level data were not used due to high rates of missing data and other quality concerns.

informational and literacy text skills relating to identifying key ideas and details in text, analyzing text structure and purpose, and integrating knowledge and ideas from text. Data were provided as a categorical variable indicating whether a student scored significantly below (Needs Support), below (Close), at (Ready), or above (Exceeding) grade-level benchmarks for proficiency. The study team then converted that categorical variable to a binary variable indicating whether a student demonstrated proficiency (received a benchmark score of Ready or Exceeding) or did not demonstrate proficiency (received a benchmark score of Close or Needs Support) in reading in grade 3.

Student demographic characteristics

Gender. CNMI Public School System administrative records reported two categories of gender—female and male. The analyses used the record for students’ kindergarten year. This variable was deviation coded as –1 (male) and 1 (female).

Free or reduced-price lunch receipt. Student status indicating receipt of meals provided during the school day for free or at a reduced cost. Students are eligible for free or reduced-price lunch if their household income is at or below 185 percent of the federal income poverty level and their families enroll them in the program. Once enrolled, individual student data records are marked as receiving the benefit. There might be families who were eligible for the benefit but did not apply for it. Only children of families who applied received the benefit. CNMI Public School System administrative records report whether a student received free or reduced-price lunch until the 2018/19 school year, when the CNMI Public School System started providing free meals for all students.² The analyses used the record for students’ kindergarten year. This variable was deviation coded as –1 (student did not receive free or reduced-price lunch) and 1 (student received free or reduced-price lunch).

Age at kindergarten entry. CNMI Public School System administrative records included student date of birth. The study team calculated age in both months and days as of August 5 of the year a student started kindergarten. This variable was mean centered. Five students in the sample were very young at kindergarten entry (between four years, two months and four years, five months). The study team conducted a sensitivity analysis that removed extreme age values, and the results did not change substantially (see appendix D).

Ethnicity. Data on student ethnicity were derived from CNMI Public School System administrative files. CNMI Public School System data files included records for 18 ethnicity categories, which families report on school entry forms. The study team collapsed these into three categories: Chamorro or Carolinian;³ Filipino; and other, which combined the remaining ethnicity categories (Bangladeshi, Chinese, Chuukese, Japanese, Korean, Kosraean, Marshallese, Other, Other Asian, Other Pacific Islander, Palauan, Pohnpeian, White, and Yapese). This variable was deviation coded into three separate ethnicity values. Category A was coded as –1 (Other), 0 (Chamorro or Carolinian), and 1 (Filipino). Category B was coded as –1 (Other), 0 (Filipino), and 1 (Chamorro or Carolinian).

Education experiences

Head Start enrollment. Information on enrollment in Head Start was collected by the CNMI Public School System when students enrolled in kindergarten. The study team used the information in these administrative records to create a binary variable equal to 1 if a student had enrolled in Head Start for one or more years and –1 if a student had not enrolled in Head Start.

Mobility. For each student the study team identified the school of record from CNMI Public School System administrative records for each year between kindergarten and grade 3. The study team created a dichotomous variable equal to 0 if the student changed schools one or more times during the period and 1 if the student was

² The study did not include kindergarten records from 2018/19.

³ This report uses *Chamorro* and *Carolinian* to align with the terms used by the CNMI Public School System. Some community members in CNMI, however, may prefer *CHamoru* and *Refaluwasch* respectively.

enrolled in the same school for the entire period. This variable was then deviation coded as -1 (student attended two or more schools between kindergarten and grade 3) and 1 (student attended only one school between kindergarten and grade 3).

Table B1. Predictor variable coding in analytic models

Measure	Centering method	Value label	Value
Gender	Deviation coding	Male	-1
		Female	1
Free or reduced-price lunch receipt in kindergarten	Deviation coding	Student did not receive free or reduced-price lunch in kindergarten	-1
		Student received free or reduced-price lunch in kindergarten	1
Age at kindergarten entry (months)	Mean centered	Continuous age centered on sample mean	-14.7 to 8.1
Ethnicity category A	Deviation coding	Other	-1
		Chamorro or Carolinian	0
		Filipino	1
Ethnicity category B	Deviation coding	Other	-1
		Filipino	0
		Chamorro or Carolinian	1
Head Start enrollment	Deviation coding	Student had not enrolled in Head Start	-1
		Student had enrolled in Head Start for one or more years	1
School mobility	Deviation coding	Student attended two or more schools between kindergarten and grade 3	-1
		Student attended only one school between kindergarten and grade 3	1

Source: Authors' compilation.

Analytic methods

This section describes the data analyses for the study.

Hierarchical modeling. The study team used hierarchical logistic regression modeling (HLM) to estimate the associations between student demographic characteristics and education experiences and the outcome of interest. Because students were nested in five cohorts and schools, it was necessary to account for the nesting structure of the dataset using HLM (Hedeker & Gibbons, 2006; Raudenbush & Bryk, 2002).

The two-level HLM model for the study's research question is as follows.⁴ For a dichotomous outcome variable Y_{ij} ~ Bernoulli distribution (π_{ij}), the logit link function is:

$$\eta_{ij} = \beta_1 + \beta_2(Age_{ij}) + \beta_3(FRPL_{ij}) + \beta_4(Gender_{ij}) + \beta_5(Ethnicity_{ij}) + \beta_6(Education_{ij}) + u_j + e_{ij}$$

where $\eta_{ij} = \ln\left(\frac{\pi_{ij}}{1-\pi_{ij}}\right)$, π_{ij} denotes the probability that student i in cohort j received a benchmark score of Ready or Exceeding on the grade 3 reading assessment; Age_{ij} , $FRPL_{ij}$, $Gender_{ij}$, and $Ethnicity_{ij}$ are child-level demographic characteristics variables; $Education_{ij}$ is a vector of education experience variables, including mobility and Head Start enrollment; β_2 to β_6 are regression coefficients for each corresponding demographic

⁴ A three-level HLM model that included school as a random intercept was tried first, but it was removed because the model would not converge and there was no estimate for school.

characteristic variable; u_j represents the random effects for cohort j ; and e_{ij} represents the random effects for student ij .

The study team conducted the analyses in R version 4.04 (R Core Team, 2021) and used the lme4 package version 1.1-26 (Bates et al., 2015) to conduct the logistic HLM analyses, the mice package version 3.13.0 (van Buuren & Groothuis-Oudshoorn, 2011) to impute and pool the HLM results, and the emmeans package version 1.5.5-1 (Lenth, 2021) to calculate predicted probabilities from the HLM models.

Predicted probabilities. The study team estimated the least square means of the group effect on the logit scale using the emmeans package in R (Lenth, 2021). This produced an estimated probability by mapping the logits onto the probability scale. Marginal probabilities were then predicted for each subgroup that was found to be statistically significant for a given outcome. For continuous predictor variables the probabilities were calculated at each quartile of the variable to show how the probabilities changed across the range of observed values. Other variables in the model were held at their mean.

Missing data. Data were missing for 17 percent of the analytic sample (table B2).

Table B2. Missing data for key variables in research question analyses

Variable	Number of students with a grade 3 ACT Aspire assessment record	Number of cases with data on this variable	Percentage of cases with data on this variable
Outcome: Grade 3 reading proficiency	2,095	2,095	100.0
Student demographics			
Gender	2,095	2,094	99.9
Free or reduced-price lunch receipt in kindergarten	2,095	2,095	100.0
Age at kindergarten entry	2,095	2,072	98.8
Ethnicity	2,095	2,093	99.9
Education experiences			
Head Start enrollment	2,095	2,095	100.0
School mobility	2,095	1,760	84.0
All variables used in the analyses	2,095	1,742	83.2

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

The sample of students used for the complete case analysis, in which students were dropped if they were missing data on any of the variables, was not statistically significantly different from the full sample with respect to any covariate. However, complete case analysis could have substantial weaknesses. First, listwise deletion limits the statistical power of the tests conducted because it uses only cases with complete information on all variables and thus limits the overall size of the sample (Allison, 2001; Olinsky et al., 2003). Second, if there is a systematic difference between the complete cases and the incomplete cases, the statistical inference from complete case analysis might not apply to the population of all cases. To eliminate this bias, the study team used multiple imputation for missing data as the main analysis. Missing data for predictors (student demographic characteristics and education experiences) were imputed using the predictive mean matching for continuous variables and polytomous regression imputation for unordered categorical variables via the R package mice (van Buuren & Groothuis-Oudshoorn, 2011). Data were not imputed for missing outcome measures.

Multiple imputation inference involved three distinct phases:

- The missing data were filled in 16 times based on the largest percentage of missingness for any of the variables, to generate 16 complete datasets.

- The 16 complete datasets were analyzed separately using HLM procedures.
- The results from the complete datasets were combined for inference using the mice package's pool() function, which combines the estimates and standard errors from analyses performed on each of the imputed datasets.

Appendix C presents results from the imputed data as the main analysis in the report, and appendix D presents unimputed data for supplementary sensitivity analyses.

Supplementary Head Start analyses

Information on student enrollment in Head Start was collected by the CNMI Public School System when students enrolled in kindergarten. Fewer than 3 percent of cases were missing information for this analysis. The study team used the information in these administrative records to create a binary variable, equal to 1 if a student had enrolled in Head Start for at least one year and 0 if a student had not. Using a three-level logistic HLM model (student-cohort-school), the study team used student demographic characteristics (gender, free or reduced-price lunch receipt, age at kindergarten entry, and ethnicity) to predict Head Start enrollment. Further discussion of findings from these analyses can be found in appendix D.

References

- Allison, P. D. (2001). *Missing data* (Vol. 136). Sage.
- Bates, D., Mächler, M., Bolker, B., & Walker, S. (2015). Fitting linear mixed-effects models using lme4. *Journal of Statistical Software*, 67(1), 1–48. <http://hdl.handle.net/10.18637/jss.v067.i01>.
- Hedeker, D., & Gibbons, R. (2006). *Longitudinal data analysis*. Wiley.
- Lenth, R. V. (2021). emmeans: *Estimated marginal means, aka least-squares means* (R package version 1.5.5-1). The Comprehensive R Archive Network. <https://CRAN.R-project.org/package=emmeans>.
- Olinsky, A., Chen, S., & Harlow, L. (2003). The comparative efficacy of imputation methods for missing data in structural equation modeling. *European Journal of Operational Research*, 151(1), 5–79. [https://doi.org/10.1016/S0377-2217\(02\)00578-7](https://doi.org/10.1016/S0377-2217(02)00578-7).
- R Core Team. (2021). *R: A language and environment for statistical computing*. R Foundation for Statistical Computing. Retrieved April 26, 2021, from <https://www.R-project.org/>.
- Raudenbush, S. W., & Bryk, A. S. (2002). *Hierarchical linear models: Applications and data analysis methods* (2nd ed.). Sage.
- van Buuren, S. V., & Groothuis-Oudshoorn, K. (2011). mice: Multivariate imputation by chained equations in R. *Journal of Statistical Software*, 45(3), 1–68. <https://doi.org/10.18637/jss.v045.i03>.

Appendix C. Supporting analyses

This appendix describes the study population’s sample descriptive statistics, as well as the results of the data analyses.

Sample descriptives

This section provides descriptive statistics of the study’s population.

Table C1. Student demographic characteristics and education experiences, by grade 3 reading proficiency status (percent)

Variable	All students (<i>n</i> = 2,095)	Scored Ready or Exceeding on the ACT Aspire Reading assessment in grade 3 (<i>n</i> = 430)	Did not score Ready or Exceeding on the ACT Aspire Reading assessment in grade 3 (<i>n</i> = 1,665)
All students	100	20.5	79.5
Gender			
Male	51.0	44.9	52.6
Female	49.0	55.1	47.4
Gender missing	0.1	0.0	0.1
Free or reduced-price lunch receipt at kindergarten entry			
Did not receive free or reduced-price lunch at kindergarten entry	89.0	90.7	88.5
Received free or reduced-price lunch at kindergarten entry	11.0	9.3	11.5
Age at kindergarten entry			
Mean age at kindergarten entry ^a (months)	64.5 (3.7)	64.9 (3.7)	64.4 (3.7)
Ethnicity			
Filipino	33.9	46.7	30.6
Chamorro or Carolinian	47.2	35.8	50.1
Other	18.8	17.2	19.2
Ethnicity missing	0.1	0.2	0.1
Head Start enrollment			
Had enrolled in Head Start	51.9	54.7	51.2
Had not enrolled in Head Start	48.1	45.4	48.8
School mobility			
Attended more than one school between kindergarten and grade 3	17.2	12.1	18.6
Attended one school between kindergarten and grade 3	66.8	70.0	66.0
Mobility missing	16.0	17.9	15.5

Note: Percentages might not sum to 100 due to rounding.

a. Information on age at kindergarten entry was missing for 32 students. About 15.6 percent of students who were missing information on age earned a score of Ready or Exceeding on the grade 3 reading assessment compared with 20.6 percent of students for whom information on age was not missing. Values in parentheses are standard deviations, which applied only to the continuous variable.

Source: Authors’ analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Results of the data analyses

The section provides detailed results of the regression analyses examining the relationship between students' demographic characteristics and education experiences and their grade 3 reading proficiency.

Table C2. Hierarchical logistic regression model results for earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment from analyses using multiple imputation of missing data

Variable	Estimate (log odds)	Standard error	p-value
Fixed effect			
Intercept	-1.67***	0.17	<.001
Female	0.18**	0.06	<.01
Received free or reduced-price lunch	-0.21*	0.10	<.05
Age at kindergarten entry	0.04*	0.02	<.05
Filipino	0.44***	0.08	<.001
Chamorro or Carolinian	-0.34***	0.08	<.001
Had enrolled in Head Start	-0.03	0.06	.61
Attended one school between kindergarten and grade 3	0.18*	0.08	<.05
Random effect			
Cohort	0.27	na	na

* Significant at $p < .05$; ** significant at $p < .01$; *** significant at $p < .001$.

na is not applicable

Note: $n = 2,095$.

Source: Authors' analysis based on Commonwealth of the Northern Mariana Islands Public School System 2011–19 administrative data files.

Table C3. Predicted probability for earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment from analyses using multiple imputation of missing data

Subgroup	Predicted probability	95 percent confidence interval
Gender—male	0.14	0.10–0.18
Gender—female	0.18	0.14–0.24
Received free or reduced-price lunch	0.13	0.09–0.19
Did not receive free or reduced-price lunch	0.19	0.15–0.24
Age at kindergarten entry—50 months	0.10	0.06–0.16
Age at kindergarten entry—61 months	0.14	0.11–0.19
Age at kindergarten entry—64 months	0.16	0.12–0.21
Age at kindergarten entry—68 months	0.18	0.13–0.23
Age at kindergarten entry—72 months	0.20	0.14–0.27
Filipino	0.23	0.17–0.30
Chamorro or Carolinian	0.12	0.09–0.16
Other ethnicities	0.14	0.10–0.20
Attended one school between kindergarten and grade 3	0.18	0.14–0.24
Attended two or more schools between kindergarten and grade 3	0.14	0.09–0.19

Note: $n = 2,095$. Only variables identified as statistically significant in table C2 are included in the table. The table includes predicted probabilities only for age values that are represented in figure 3 in the main report.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Appendix D. Supplementary analyses

This appendix provides the results of supplementary and sensitivity analyses. These analyses include the results of regression models using only complete case data, regression analyses examining the effects of extreme ages at kindergarten entry on the results where the values less than 2.5 and greater than 97.5 percentiles were removed, and regression moderator analyses.

Because this study compared the school performance of students who had enrolled in Head Start and students who had not, it is important to first understand which children in the study sample had enrolled in Head Start. This appendix also provides supplementary analyses on the demographic characteristics of students who had enrolled in Head Start and students who had not, analyses that were done to understand this population of students and to inform the study's main findings.

Sensitivity analyses

The results of the sensitivity analyses were all consistent with those of the analyses discussed in the main report. The results of the complete case analysis and the analysis using imputed data and the results of the analysis with all values for age at kindergarten entry and the analysis in which extreme age values were removed were similar. The differences in regression coefficients (also called regression estimates) on all predictors were less than or equal to 0.25 log odds. All analyses continued to indicate statistically significant higher probabilities of demonstrating reading proficiency in grade 3 for Filipino students, students who were older at kindergarten entry, female students, and students who did not receive free or reduced-priced lunch than for the average student (tables D1–D6).

The study team also examined whether the relationship between Head Start enrollment and grade 3 reading proficiency differed for students from different ethnicity groups and by age at kindergarten entry. An interaction term between Head Start enrollment and ethnicity was added to the analytic model that used multiple imputation for missing data (table D5), and an interaction term between age at kindergarten entry and Head Start enrollment was added to the analytic model that included the interaction term between Head Start enrollment and ethnicity (table D6). The interaction terms were not statistically significant, indicating that neither ethnicity nor age at kindergarten entry moderated the relationship between Head Start enrollment and grade 3 reading proficiency.

Table D1. Hierarchical logistic regression model results for earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment using multiple imputation for missing data and for the complete case analysis

Variable	Multiple imputation (<i>n</i> = 2,095)			Complete case (<i>n</i> = 1,742)		
	Estimate	Standard error	<i>p</i> -value	Estimate	Standard error	<i>p</i> -value
Fixed effect						
Intercept	-1.67***	0.17	<.001	-1.73***	0.19	<.001
Female	0.18**	0.06	<.01	0.13*	0.06	<.05
Received free or reduced-price lunch	-0.21*	0.10	<.05	-0.25*	0.10	<.05
Age at kindergarten entry (months)	0.04*	0.02	<.05	0.04*	0.02	<.05
Filipino	0.44***	0.08	<.001	0.49***	0.12	<.001
Chamorro or Carolinian	-0.34***	0.08	<.001	-0.31***	0.13	<.001
Had enrolled in Head Start	-0.03	0.06	.61	-0.05	0.06	.42
Attended one school between kindergarten and grade 3	0.18*	0.08	<.05	0.17*	0.16	<.05
Random effect						
Cohort	0.27	na	na	0.31	na	na

* Significant at $p < .05$; ** significant at $p < .01$; *** significant at $p < .001$.

na is not applicable.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D2. Predicted probability of earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment using multiple imputation for missing data and for the complete case analysis (based on the hierarchical logistic regression model)

Subgroup	Multiple imputation (<i>n</i> = 2,095)		Complete case (<i>n</i> = 1,742)	
	Predicted probability	95 percent confidence interval	Predicted probability	95 percent confidence interval
Student characteristics				
Gender—male	0.14	0.10–0.18	0.13	0.10–0.19
Gender—female	0.18	0.14–0.24	0.17	0.12–0.23
Received free or reduced-price lunch	0.13	0.09–0.19	0.12	0.08–0.19
Did not receive free or reduced-price lunch	0.19	0.15–0.24	0.18	0.14–0.24
Age at kindergarten entry—50 months	0.10	0.06–0.16	0.09	0.05–0.16
Age at kindergarten entry—61 months	0.14	0.11–0.19	0.14	0.10–0.19
Age at kindergarten entry—64 months	0.16	0.12–0.21	0.15	0.11–0.20
Age at kindergarten entry—68 months	0.18	0.13–0.23	0.17	0.12–0.23
Age at kindergarten entry—72 months	0.20	0.14–0.27	0.19	0.13–0.27
Filipino	0.23	0.17–0.30	0.23	0.16–0.30
Chamorro or Carolinian	0.12	0.09–0.16	0.11	0.08–0.16
Other ethnicities	0.14	0.10–0.20	0.13	0.09–0.19
Attended one school between kindergarten and grade 3	0.18	0.14–0.24	0.17	0.13–0.23
Attended two or more schools between kindergarten and grade 3	0.14	0.09–0.19	0.13	0.09–0.19

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D3. Hierarchical logistic regression model results for earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment for the full sample and excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles

Variable	All values for age at kindergarten entry included (<i>n</i> = 2,659)			Excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles (<i>n</i> = 2,594)		
	Estimate	Standard error	<i>p</i> -value	Estimate	Standard error	<i>p</i> -value
Fixed effect						
Intercept	-1.67***	0.17	<.001	-1.65***	0.17	<.001
Female	0.18**	0.06	<.01	0.18**	0.06	<.01
Received free or reduced-price lunch	-0.21*	0.10	<.05	-0.20*	0.10	<.05
Age at kindergarten entry (months)	0.04*	0.02	<.05	0.04*	0.02	<.05
Filipino	0.44***	0.08	<.001	0.45***	0.08	<.001
Chamorro or Carolinian	-0.34***	0.08	<.001	-0.34***	0.08	<.001
Had enrolled in Head Start	-0.03	0.06	.61	-0.04	0.06	.52
Attended one school between kindergarten and grade 3	0.18*	0.08	<.05	0.15*	0.08	.06
Random effect						
Cohort	0.27	na	na	0.28	na	na

* Significant at $p < .05$; ** significant at $p < .01$; *** significant at $p < .001$.

na is not applicable.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D4. Predicted probability of earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment for the full sample and excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles

Subgroup	All values for age at kindergarten entry included (<i>n</i> = 2,659)		Excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles (<i>n</i> = 2,594)	
	Predicted probability	95 percent confidence interval	Predicted probability	95 percent confidence interval
Student characteristics				
Gender—male	0.14	0.10–0.18	0.14	0.10–0.14
Gender—female	0.18	0.14–0.24	0.19	0.14–0.25
Received free or reduced-price lunch	0.13	0.09–0.19	0.14	0.09–0.20
Did not receive free or reduced-price lunch	0.19	0.15–0.24	0.19	0.15–0.24
Age at kindergarten entry—50 months	0.10	0.06–0.16	0.10	0.06–0.16
Age at kindergarten entry—61 months	0.14	0.11–0.19	0.14	0.11–0.19
Age at kindergarten entry—64 months	0.16	0.12–0.21	0.16	0.12–0.21
Age at kindergarten entry—68 months	0.18	0.13–0.23	0.18	0.13–0.24
Age at kindergarten entry—72 months	0.20	0.14–0.27	0.20	0.15–0.28
Filipino	0.23	0.17–0.30	0.23	0.17–0.30
Chamorro or Carolinian	0.12	0.09–0.16	0.12	0.09–0.16
Other ethnicities	0.14	0.10–0.20	0.15	0.10–0.21
Attended one school between kindergarten and grade 3	0.18	0.14–0.24	0.19	0.14–0.24
Attended two or more schools between kindergarten and grade 3	0.14	0.09–0.19	0.14	0.09–0.19

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D5. Hierarchical logistic regression model results for earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment from analyses using multiple imputation of missing data and including interactions between ethnicity and Head Start enrollment

Variable	Estimate (log odds)	Standard error	<i>p</i> -value
Fixed effect			
Intercept	-1.68***	0.17	<.001
Female	0.18**	0.06	<.01
Received free or reduced-price lunch	-0.21*	0.10	<.05
Age at kindergarten entry (months)	0.04*	0.02	<.05
Filipino	0.44***	0.08	<.001
Chamorro or Carolinian	-0.33***	0.08	<.001
Had enrolled in Head Start	-0.05	0.06	.44
Attended one school between kindergarten and grade 3	0.18*	0.08	<.05
Filipino: Had enrolled in Head Start	0.04	0.08	.59
Chamorro or Carolinian: Had enrolled in Head Start	0.01	0.08	.86
Random effect			
Cohort	0.27	na	na

* Significant at $p < .05$; ** significant at $p < .01$; *** significant at $p < .001$.

na is not applicable.

Note: $n = 2,095$. Variable names with : indicate an interaction term between two variables.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D6. Hierarchical logistic regression model results for earning a score of Ready or Exceeding on the grade 3 ACT Aspire reading assessment from analyses using multiple imputation of missing data and including interactions between ethnicity and Head Start enrollment and age at kindergarten entry and Head Start enrollment

Variable	Estimate (log odds)	Standard error	p-value
Fixed effect			
Intercept	-1.68***	0.17	<.001
Female	0.18**	0.06	.001
Received free or reduced-price lunch	0.21*	0.10	.03
Age at kindergarten entry (months)	0.04*	0.02	.01
Filipino	0.44***	0.08	<.001
Chamorro or Carolinian	-0.33***	0.08	<.001
Had enrolled in Head Start	-0.05	0.06	.45
Attended one school between kindergarten and grade 3	0.18*	0.08	<.05
Filipino: Had enrolled in Head Start	0.04	0.08	.60
Chamorro or Carolinian: Had enrolled in Head Start	0.01	0.08	.86
Enrolled in Head Start: Age at kindergarten entry	-0.004	0.02	.79
Random effect			
Cohort	0.29	na	na

* Significant at $p < .05$; ** significant at $p < .01$; *** significant at $p < .001$.

na is not applicable.

Note: $n = 2,095$. Variable names with: indicate an interaction term between two variables.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Supplementary analyses on Head Start enrollment

The study team first calculated descriptive statistics for students who had enrolled in Head Start and for students who had not (table D7). Hierarchical logistic regression models were then used to estimate the associations between student demographic characteristics and the outcome of interest. For this analysis the outcome of interest was whether the student had enrolled in the CNMI Head Start program for one or more years. Predicted probabilities were then calculated from the results of the hierarchical models for demographic characteristics that had statistically significant associations with the outcome variable. The study team refers to associations as statistically significant if the p -value associated with the coefficient was less than .05.

The main findings from these analyses show that:

- *Filipino students were more likely to have enrolled in Head Start than students of other ethnicities.* The predicted probability of having enrolled in Head Start was 67 percent for Filipino students compared with 45 percent for Chamorro or Carolinian students and 46 percent for students who reported another ethnicity, after other student demographic characteristics were controlled for (tables D8 and D9). The difference in probability between Filipino students and both Chamorro or Carolinian students and students of other ethnicities was minor.
- *Students who had enrolled in Head Start tended to enter kindergarten at an older age.* Students who were older at kindergarten entry were more likely to have enrolled in Head Start than students who were younger at kindergarten entry, after other student demographic characteristics were controlled for (see tables D8 and D9). The predicted probability of having enrolled in Head Start was 63 percent for students who were 6 years

old (72 months) at kindergarten entry compared with 42 percent for students who were 4 years, 8 months old (56 months; see tables D8 and D9). The difference between those probabilities was major.

- *Student gender and free or reduced-price lunch receipt were not associated with Head Start enrollment.* The odds of having enrolled in Head Start did not differ significantly between male and female students or between students who received free or reduced-price lunch and students who did not receive free or reduced-price lunch, after other student demographic characteristics and education experiences were controlled for (see table D8).

Sample descriptives

Table D7. Student demographic characteristics, by Head Start enrollment

Variable	All students (n = 2,659)	Had not enrolled in Head Start (n = 1,344)	Had enrolled in one or more years of Head Start (n = 1,315)
All students	100	50.6	49.5
Gender			
Male	50.7	51.2	50.3
Female	48.9	48.1	49.6
Gender missing	0.4	0.7	0.2
Free or reduced-price lunch receipt in kindergarten entry			
Did not receive free or reduced-price lunch at kindergarten entry	89.5	90.7	88.3
Received free or reduced-price lunch at kindergarten entry	10.5	9.3	11.7
Age at kindergarten entry			
Mean age at kindergarten entry ^a (months)	64.4 (3.9)	64.1 (4.0)	64.7 (3.7)
Ethnicity			
Filipino	32.0	22.0	42.4
Chamorro or Carolinian	47.4	54.6	40.1
Other	19.9	22.4	17.3
Ethnicity missing	0.6	1.04	0.2

Note: Percentages might not sum to 100 because of rounding.

a. Information on age at kindergarten entry was missing for 90 students. Of those students who were missing information on age at kindergarten entry, 16.7 percent had enrolled in one or more years of Head Start compared with 50.6 percent of those students for whom information on age was not missing. Values in parentheses are standard deviations, which apply only to continuous variables.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Head Start enrollment data analyses. This section provides detailed results of the regression analyses examining the relationship between Head Start enrollment and student demographic characteristics.

Table D8. Hierarchical logistic regression model results for predicting Head Start enrollment with multiple imputation for missing data

Variable	Estimate (log odds)	Standard error	p-value
Fixed effect			
Intercept	0.12	0.16	.43
Female	0.04	0.04	.36
Received free or reduced-price lunch	-0.13	0.07	.09
Age at kindergarten entry (months)	0.05***	0.01	<.001
Filipino	0.58***	0.06	<.001
Chamorro or Carolinian	-0.32***	0.06	<.001
Random effect			
School	0.41	na	na
Cohort	0.07	na	na

*** Significant at $p < .001$.

na is not applicable.

Note: $n = 2,659$.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D9. Predicted probabilities for Head Start enrollment with multiple imputation for missing data

Subgroup	Predicted probability	95 percent confidence interval
Age at kindergarten entry—56 months	0.42	0.34–0.50
Age at kindergarten entry—58 months	0.44	0.37–0.52
Age at kindergarten entry—60 months	0.47	0.40–0.55
Age at kindergarten entry—62 months	0.50	0.42–0.57
Age at kindergarten entry—64 months	0.53	0.45–0.60
Age at kindergarten entry—66 months	0.55	0.48–0.63
Age at kindergarten entry—68 months	0.58	0.50–0.65
Age at kindergarten entry—70 months	0.61	0.53–0.68
Age at kindergarten entry—72 months	0.63	0.55–0.71
Filipino	0.67	0.59–0.74
Chamorro or Carolinian	0.45	0.38–0.53
Other ethnicities	0.46	0.38–0.55

Note: $n = 2,659$. Only variables identified as statistically significant in table D8 are included in this table.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Head Start enrollment sensitivity analyses. The results of the analysis using complete case analysis and the analysis using imputed data and the results of the analysis with all values for age at kindergarten entry and the analysis in which extreme age values were removed were similar for supplementary analyses on Head Start enrollment, with differences on all predictors less than or equal to 0.25 log odds. These sensitivity analyses found that Filipino students and students who were older for their grade had a statistically significantly higher probability of having enrolled in Head Start than the average student. Both multiple imputation and complete case analyses also found that age at kindergarten entry is positively and statistically significantly associated with Head Start enrollment (tables D10–D13).

Table D10. Hierarchical logistic regression model results for Head Start enrollment using multiple imputation for missing data

Variable	Multiple imputation (<i>n</i> = 2,659)			Complete cases (<i>n</i> = 2,594)		
	Estimate	Standard error	<i>p</i> -value	Estimate	Standard error	<i>p</i> -value
Fixed effect						
Intercept	0.12	0.16	.43	0.14	0.15	.39
Female	0.04	0.04	.36	0.03	0.04	.38
Received free or reduced-price lunch	-0.13	0.07	.09	-0.12	0.08	.11
Age at kindergarten entry (months)	0.05***	0.01	<.001	0.06***	0.01	<.001
Filipino	0.58***	0.06	<.001	0.56***	0.06	<.001
Chamorro or Carolinian	-0.32***	0.06	<.001	-0.29***	0.06	<.001
Random effect						
School	0.41	na	na	0.43	na	na
Cohort	0.07	na	na	0.03	na	na

*** Significant at *p* < .001.

na is not applicable.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D11. Predicted probability of Head Start enrollment using multiple imputation for missing data and complete case analysis

Subgroup	Multiple imputation (<i>n</i> = 2,659)		Complete case (<i>n</i> = 2,594)	
	Predicted probability	95 percent confidence interval	Predicted probability	95 percent confidence interval
Student characteristics				
Age at kindergarten entry—56 months	0.42	0.34–0.50	0.42	0.34–0.51
Age at kindergarten entry—58 months	0.44	0.37–0.52	0.45	0.37–0.53
Age at kindergarten entry—60 months	0.47	0.40–0.55	0.48	0.40–0.55
Age at kindergarten entry—62 months	0.50	0.42–0.57	0.50	0.43–0.58
Age at kindergarten entry—64 months	0.53	0.45–0.60	0.53	0.46–0.60
Age at kindergarten entry—66 months	0.55	0.48–0.63	0.56	0.48–0.63
Age at kindergarten entry—68 months	0.58	0.50–0.65	0.59	0.51–0.66
Age at kindergarten entry—70 months	0.61	0.53–0.68	0.61	0.53–0.69
Age at kindergarten entry—72 months	0.63	0.55–0.71	0.64	0.56–0.71
Filipino	0.67	0.59–0.74	0.67	0.59–0.74
Chamorro or Carolinian	0.45	0.38–0.53	0.46	0.39–0.53
Other ethnicities	0.46	0.38–0.55	0.47	0.39–0.55

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D12. Hierarchical logistic regression model results for Head Start enrollment for the full sample and excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles

Variable	All values for age at kindergarten entry included (n = 2,659)			Excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles (n = 2,618)		
	Estimate	Standard error	p-value	Estimate	Standard error	p-value
Fixed effect						
Intercept	0.12	0.16	.43	0.13	0.15	.39
Female	0.04	0.04	.36	0.04	0.04	.34
Received free or reduced-price lunch	-0.13	0.07	.09	-0.12	0.07	.10
Age at kindergarten entry (months)	0.05***	0.01	<.001	0.05***	0.01	<.001
Filipino	0.58***	0.06	<.001	0.58***	0.06	<.001
Chamorro or Carolinian	-0.32***	0.06	<.001	-0.32***	0.06	<.001
Random effect						
School	0.41	na	na	0.41	na	na
Cohort	0.07	na	na	0.07	na	na

*** Significant at $p < .001$.

na is not applicable.

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.

Table D13. Predicted probability of Head Start enrollment for the full sample and excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles

Subgroup	All values for age at kindergarten entry included (n = 2,659)		Excluding students with values for age at kindergarten entry in the <2.5 and >97.5 percentiles (n = 2,618)	
	Predicted probability	95 percent confidence interval	Predicted probability	95 percent confidence interval
Student characteristics				
Age at kindergarten entry—56 months	0.42	0.34–0.50	0.42	0.34–0.50
Age at kindergarten entry—58 months	0.44	0.37–0.52	0.45	0.37–0.53
Age at kindergarten entry—60 months	0.47	0.40–0.55	0.47	0.40–0.55
Age at kindergarten entry—62 months	0.50	0.42–0.57	0.50	0.43–0.57
Age at kindergarten entry—64 months	0.53	0.45–0.60	0.53	0.46–0.60
Age at kindergarten entry—66 months	0.55	0.48–0.63	0.56	0.48–0.63
Age at kindergarten entry—68 months	0.58	0.50–0.65	0.58	0.51–0.65
Age at kindergarten entry—70 months	0.61	0.53–0.68	0.61	0.53–0.68
Age at kindergarten entry—72 months	0.63	0.55–0.71	0.63	0.55–0.71
Filipino	0.67	0.59–0.74	0.67	0.60–0.74
Chamorro or Carolinian	0.45	0.38–0.53	0.45	0.38–0.53
Other ethnicities	0.46	0.38–0.55	0.47	0.39–0.55

Source: Authors' analysis based on 2011–19 administrative data files provided by the Commonwealth of the Northern Mariana Islands Public School System.