

**UNIVERSITY OF PENNSYLVANIA
SCHOOL OF SOCIAL POLICY & PRACTICE
Master of Science in Social Policy**

MSSP 897 Applied Linear Modeling

Fall 2019

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Course Description

This course deals with the underlying assumptions and applications of the general linear model with social science, education, and social policy related questions/data. The first half of the course begins by covering simple linear regression, multiple linear regression, and the assumptions of the general linear model, assumption diagnostics, consequences of violation, and how to correct for violated assumptions. Various aspects of regression analysis with multiple independent variables will be covered including categorical explanatory variables (e.g. to estimate group differences), non-linear effects, interaction effects, and mediating effects (e.g. to estimate the indirect effect of social processes). The course will then cover some of the applications of the general(ized) linear model including logistic regression, path analysis in the OLS regression framework, and the basics to the modeling of clustered data via multilevel analysis and growth modeling (of repeated measures). The course will be taught using R, but students are welcome to use any statistical package of comfort. Pre-requisite: MSSP 630, Introduction to Quantitative Social Policy Analysis, or another Introductory Graduate Statistics course.

Educational and Learning Objectives

By the end of Applied Linear Modeling students are expected to demonstrate:

1. a strong foundation in simple linear and multiple regression and how to arrive at the Best Linear Unbiased Estimates (B.L.U.E.);
2. their knowledge of the main assumptions of the general linear model and the implications they have on estimation when they are violated;
3. how to conduct diagnostics and correct for the violation of the assumptions of the general linear model;
4. how to interpret various coefficients;

5. how to estimate moderating (i.e. interaction) and mediating effects (and their relationship to path modeling);
6. how to analyze binary outcomes in logistic regression;
7. what clustered and nested data structures are and the basics of how to analyze them within multilevel (or hierarchical linear) modeling;
8. a strong competence in the selection of appropriate applications of the general(ized) linear model given particular research questions;
9. how to effectively communicate, in writing, the results for a professional publishable journal article (including tables, coefficients, etc.).

Course Requirements & Expectations

Expectations

Classroom learning is a fundamental component of your professional education. Students are therefore expected to attend each class, arrive to class on time, and be in attendance for the full class. In the event that you are unable to attend class for any reason, you must notify the instructor in advance and learn how you are to make up the content you missed. Excessive absenteeism (i.e., missing more than two classes) is considered a serious problem the instructor will handle by meeting with the student and determining whether the student's educational adviser should be notified. Excessive absenteeism could result in course failure. This applies to both the lecture and lab parts of the class.

Students are expected to: (A) participate substantively in class discussions; (B) read on a weekly basis and come to class prepared to apply and discuss the reading assignments; (C) submit assignments by the due date and in accordance with the specified format.

Grades will be based on lab exercises and four policy evaluation reports written as would be reported in the results section of a publishable manuscript. The lab exercises will be worth 15% of your course grade. The first three policy research reports will each be worth 15% of your course grade and the final policy research report will be worth 40% of your overall course grade.

Assignments

Students will be responsible for lab exercises and four, graded written policy research reports during the semester. These assignments will require a specific analytic task of applied linear modeling and students will be required to perform and show all work. This will include submitting print outs of the output to analyses and writing up the results as you would report in the results section of a publishable manuscript. Specific instructions for completing each assignment will be provided during the semester.

Format

All papers must be typewritten, in 12-point font, double-spaced, page-numbered, with 1" margins at the left, right, top and bottom. The cover page (not included in the page limit) should include the title of the paper, student's name, assignment number, professor's name and date submitted. Papers should be stapled, not paper-clipped. Papers **must be proofread** carefully for clarity, organization, spelling, punctuation, and other potential errors before submission.

In-text citations following APA style guidelines are required for all written assignments, with the specific source including authors' last names and year of publication, regardless of whether you are paraphrasing or using specific quotes. Direct quotes must have the specific source as above but with page number(s). **A list of references cited or consulted must be included at the end of each paper in proper APA bibliographic form.** Footnotes may be used where appropriate to further explicate a concept or issue. **American Sociological Association style may be used, but this must be consistent throughout the assignment.**

You should keep a copy of each paper submitted. The instructor will gladly answer any questions regarding format, citing or organization. Papers written for other classes may not be submitted for written assignments in this course. Direct substitution of papers between courses may result in a failing grade for that assignment.

Evaluation

Assignments will be evaluated based on the following criteria:

1. demonstrates understanding of the research question and how to appropriately analyze;
2. effective use of R or any other statistical software package;
3. shows all analyses;
4. utilizes tables and reports data/coefficients appropriately, as would be expected for the results section of a peer-review journal article;
5. demonstrates understanding of the particular task/application of the general(ized) linear model;
6. and, appropriately interprets data.

Grading Policies

The final course grade is based on the student's performance on all assignments. Students whose performance is minimal or failing at midterm will be notified in writing. Students will only be given completion credit for the first three policy evaluation reports and not a grade. This means that these assignments must be completed with satisfactory approval from the instructor or the instructor may ask for the assignment to be redone given the feedback provided by the instructor. The lab exercises will be graded based on the discretion of the lab instructor.

Academic Integrity

Students are expected to adhere to the University's Code of Academic Integrity, available at <https://catalog.upenn.edu/pennbook/code-of-academic-integrity/>. Care should be taken to avoid academic integrity violations, including plagiarism, fabrication of information, and multiple submissions (see descriptions below).** Students who engage in any of these actions will be referred to the Office of Student Conduct, which investigates and decides on sanctions in cases of academic dishonesty.

1. Plagiarism: using the ideas, data, or language of another person or source without specific or proper acknowledgment. Example: copying, in part or in its entirety, another person's paper, article, or web-based material and submitting it for an assignment; using someone else's ideas without attribution; not using quotation marks where appropriate; etc.

2. Fabrication: submitting contrived or altered information in any academic exercise. Example: making up data or statistics, citing nonexistent articles, contriving sources, etc.
3. Multiple submissions: submitting, without prior permission, any work submitted to fulfill another academic requirement.

**It is students' responsibility to consult the instructor if they are unsure about whether something constitutes a violation of the Code of Academic Integrity.

Readings

It is expected that students will read required class assignments along with the recommended texts and/or articles and come prepared to discuss in class. Students may also use relevant materials of their own choosing for the assignments.

Required Texts

Gordon, Rachel A. 2015. *Regression Analysis for the Social Sciences* (Second Edition). New York NY: Routledge.

Canvas: Additional readings will be posted online via Canvas at: <https://canvas.upenn.edu>

Recommended Texts

Cohen, Jacob, Patricia Cohen, Stephen G. West, and Leona S. Aiken, L.S. 2003. *Applied Multiple Regression/Correlation Analysis for the Behavioral Sciences*, Third Edition. Mahwah NJ: Lawrence Erlbaum Associates.

Maindonald, J., & Braun, J. (2006). *Data analysis and graphics using R: an example-based approach* (Vol. 10). Cambridge University Press.

[http://www.pindex.com/uploads/post_docs/Maindonald%20Data%20Analysis%20and%20Graphics\(PINDEX-DOC-6953\).pdf](http://www.pindex.com/uploads/post_docs/Maindonald%20Data%20Analysis%20and%20Graphics(PINDEX-DOC-6953).pdf)

Verzani, John. 2001-02. *simpleR – Using R for Introductory Statistics*.

<http://www.math.csi.cuny.edu/Statistics/R/simpleR/printable/simpleR.pdf>

R for regression tutorial: <http://tutorials.iq.harvard.edu/R/Rstatistics/Rstatistics.html#introduction>

Class Schedule and Required Readings

Section I: Introduction to Course and The General(ized) Linear Model

Week 1: August 27

Course Introduction. Review of the syllabus, and course requirements. What is the general(ized) linear model? What is correlation? What is regression? Introduction to the R program.

Week 2: September 3

Bivariate Correlation and Regression.

Readings:

Gordon, Ch. 5

Discussion Reading:

Ziliak, Stephen T., and Deirdre N. McCloskey. 2008. *The Cult of Statistical Significance: How the standard error costs us jobs, justice, and lives*. Ann Arbor MI: University of Michigan Press.

Introduction: "A Significant Problem" (Pp. 1-22)

Chapter 2: "The Sizeless State of Statistical Significance" (Pp. 33-41)

Recommended:

Gordon, Chs. 3 & 4

Cohen, et.al., text, Chs. 1 & 2

Section II: Multiple Regression and Assumptions of the General Linear Model

Week 3: September 10

Multiple Regression with Two or More Independent Variables.

Readings:

Gordon, Ch. 6

Discussion Reading:

Orr, Amy J. 2003. Black-White differences in achievement: The importance of wealth. *Sociology of Education*, 76:281-304.

Recommended:

Cohen, et.al., text, Chs. 3 & 5

Week 4: September 17

Multiple Regression with Categorical Independent Variables.

Readings:

Gordon, Ch. 7

Discussion Reading:

Bonilla-Silva, Eduardo, and Tukufu Zuberi. 2008. "Toward a Definition of White Logic and White Methods." in T. Zuberi and E. Bonilla-Silva (eds.) *White Logic, White Methods*. New York NY: Rowman & Littlefield Publishers. (Pp. 3-30)

Recommended:

Cohen, et.al., text, Ch. 8

Week 5: September 24

Policy Research Report 1 Due: Multiple Regression Assumption Diagnostics and Violation Correction.

Readings:

Gordon, page 115-17, 380-385, 351-354

Recommended:

Cohen, et.al., text, Ch. 4

Week 6: October 1

Assumption Diagnostics and Violation Correction Continued: Model Specification.

Readings:

Gordon, Ch. 9

Recommended:

Cohen, et.al., text, Ch. 6

Week 7: October 8

Outliers and Multicollinearity.

Readings:

Gordon, Ch. 11

Recommended:

Cohen, et.al., text, Ch. 10

Section III: Applications of the General(ized) Linear Model

Week 8: October 15

Interaction Effects.

Readings:

Gordon, Ch. 8

Recommended:

Cohen, et.al., text, Chs. 7 & 9

Week 9: October 22

Policy Research Report 2 Due: Assumption Diagnostics Indirect/Mediating Effects & Path Analysis.

Readings:

Gordon, Ch. 10

Recommended:

Cohen, et.al., text, Ch. 12

Loehlin, John C. 2004. *Latent Variable Models: An introduction to factor, path, and structural equation analysis* (Fourth Edition). Mahwah NJ: Lawrence Erlbaum Associates.
Chapter One: "Path models in factor, path, and structural equation analysis." (Pp. 1-34)

Poetz, Anneliese, John D. Eyles, Susan Elliott, Kathleen Wilson, and Susan Keller-Olaman. 2007. "Path analysis of income, coping and health at the local level in a Canadian context." *Health and Social Care in the Community*, 15:542-552.

Week 10: October 29

The Analysis of Binary Outcomes: Logistic Regression.

Readings:

Agresti, Alan. 1996. *An Introduction to Categorical Data Analysis*. New York NY: John Wiley & Sons.

Chapter Five: "Logistic Regression." (Pp. 103-144)

Recommended:

Cohen, et.al., text, Ch. 13

Shinn, Marybeth, Beth C. Weitzman, Daniela Stojanovic, James R. Knickman, Lucila Jimenez, Lisa Duchon, Susan James, and David H. Krantz. 1998. "Predictors of Homelessness Among Families in New York City: From Shelter Request to Housing Stability." *American Journal of Public Health*, 88:1651-1657.

Week 11: November 5

The Analysis of Nested Data Structures: Multilevel/Hierarchical Linear Modeling.

Readings:

Raudenbush, Stephen, and Anthony Bryk. 2002. *Hierarchical Linear Models: Applications and Data Analysis Methods* (Second Edition). Thousands Oaks CA: Sage Publications.

Chapter Two: "The Logic of Hierarchical Linear Models" (Pp. 16-37)

Recommended:

Singer, Judith. 1998. "Using SAS PROC MIXED to Fit Multilevel Models, Hierarchical Models, and Individual Growth Models." *Journal of Educational and Behavioral Statistics*, 24(4):323-355.

Raudenbush, Stephen, and Anthony Bryk. 1986. "A Hierarchical Model for Studying School Effects." *Sociology of Education*, 59:1-17.

Week 12: November 12 *Policy Research Report 3 Due: Estimating Interaction Effects*

The Analysis of Nested Data Structures: Multilevel/Hierarchical Linear Modeling.

Readings:

Continue same readings from Week 11.

Week 13: November 19**The Analysis of Nested Data Structures: Multilevel Growth Modeling.****Readings:**

Singer, Judith D., and John B. Willett. 2003. *Applied Longitudinal Data Analysis: Modeling Change and Event Occurrence*. New York NY: Oxford University Press.
Chapter One: "A Framework for Investigating Change over Time" (Pp. 3-15)
Chapter Three: "Introducing the Multilevel Model for Change" (Pp. 45-74)

Recommended:

Dixon-Román, E. 2013. The forms of capital and the developed achievement of Black males. *Urban Education* 48(6): 828-862.

Week 14: December 3**Material Make Up or Review.****Readings:**

No Readings

Policy Research Report 4 Due December 6th: Application of the General Linear Model

Lab Exercises (15%)
Due: One week after they are assigned

There are seven scheduled lab exercises. Lab exercises will provide you with an opportunity to apply material learned during lectures and gain hands-on experience with R.

Lab Exercises are scheduled for the following weeks:

- Week 1
- Week 2
- Week 3
- Week 4
- Week 5
- Week 7
- Week 10

How to Submit Lab Exercises

Submit your code (R script) and short written responses (in a Word document) to the lab instructor: Ama Nyame-Mensah (amanyame@upenn.edu).

Due Dates & Grading

Lab exercises are due one week after they are assigned, unless otherwise specified. Lab exercises will be graded on an attempted/not attempted basis.

Policy Research Report 1: Multiple Regression (15%)
Due: Tuesday September 24th

Increasing research evidence has indicated the meaningful effect nutrition and family income between the ages of 0 to 5 has on child development (Birch & Gussow 1970; Duncan & Brooks-Gunn 1997). You have been hired as a research consultant for an evaluation study examining the effect of the Women, Infant and Children (WIC) Nutrition Program and Aid to Families with Dependent Children (AFDC) program participation during pregnancy on child reading achievement. This is a large evaluation study using a national probability sample, the Child Development Supplement to the Panel Study of Income Dynamics, and thus has meaningful policy implications. Using the constructed data set (“good”), the evaluation team has asked you to examine the effect and unique contribution of (1) WIC program participation during pregnancy (WICpreg) on child reading achievement (readss97) over and above the child’s age in 1997 (age97), family income (faminc97), low birth weight status (bthwht), and parenting practices (HOME97) and (2) AFDC program participation during pregnancy (AFDCpreg) on child reading achievement (readss97) over and above the child’s age in 1997 (age97), family income (faminc97), low birth weight status (bthwht), and parenting practices (HOME97). The evaluation team is interested in (1) each variables individual effect, (2) the total variation of child reading achievement accounted for by the model, (3) the unique contribution of each program participations on the variance of child reading achievement, and (4) which program has a larger effect on child reading achievement.

Thus, you are asked to write a report similar to the results section of a journal article reporting the descriptive statistics (i.e. N, means/medians/proportions, sds, frequencies, and observed ranges) of the variables then multiple regression analyses and interpretation. For examples, please see the journal articles on Canvas on how to format tables and discuss results. Unlike traditional research reports and journal articles where you would focus your interpretation on the independent variables of interest, you will be expected to interpret the regression coefficients for each independent variable. You are also asked to provide in a one to two paragraph concluding statement your recommendations on the importance of each policy for child achievement. In addition, in an appendix, the evaluation team would like to see the syntax of your R program.

Definition of Variables:

WICpreg – Women, Infant and Children (WIC) Nutrition Program participant during pregnancy: 0 = No, 1 = Yes.

AFDCpreg – Aid to Families with Dependent Children (AFDC) program participant during pregnancy: 0 = No, 1 = Yes.

readss97 – Woodcock-Johnson Revised Reading Achievement Test Age Standardized Score. Minimum = 47.5, Maximum = 165.5.

AGE97 – The child’s age in 1997. Minimum = 3, Maximum = 13.

faminc97 – Total family income in 1997 (in 2002 constant dollars). Minimum = \$-72296.26, Maximum = \$784610.59.

bthwht – Low birth weight status of the child. 0 = non-low birth weight child, 1 = low birth weight child.

HOME97 – A composite total score of the emotional and cognitive stimulation at home. Minimum = 7, Maximum = 27.

Policy Research Report 2: Assumptions Diagnostics & Violation Corrections (15%)
Due: Tuesday October 22nd

You are still working as a research consultant for an evaluation study examining the effect of the Women, Infant and Children (WIC) Nutrition Program participation during pregnancy, but now on the raw scores of child mathematics achievement in 1997. Using the Child Development Supplement to the Panel Study of Income Dynamics (“good”), the evaluation team has asked you to (1) diagnose the assumptions of the general linear model for the following model:

$$\text{mathraw97} = B_0 + B_1 \text{Age97} + B_2 \text{faminc97} + B_3 \text{bthwht} + B_4 \text{WICpreg} + \varepsilon$$

that is, evaluate whether the assumptions of linearity, homoscedasticity, and normality of residuals are violated, (2) evaluate diagnostics for outliers, and (3) make any necessary and appropriate corrections. In addition, the evaluation team would like to assess whether parenting practices (HOME97) is a relevant variable for the model. The evaluation team would like for you to report the multiple regression model results before respecification/corrections and after respecification/corrections then discuss model differences.

The report should include: the descriptive statistics (i.e. N, means/medians/proportions, standard deviations, frequencies, observed ranges, and correlations) of the variables; a multiple regression analysis before assumption diagnostics and corrections; plots, diagnostic statistics, and full discussion of diagnostics and corrections; and, lastly, a multiple regression analysis after respecification and corrections. You are also asked to provide in a one to two paragraph concluding statement your thoughts on the changes in the results of the before and after analyses and what they suggests. In an appendix, the evaluation team would like to see the syntax of your R program.

Definition of Variables:

- WICpreg – Women, Infant and Children (WIC) Nutrition Program participant during pregnancy: 0 = No, 1 = Yes.
- mathraw97 – Woodcock-Johnson Revised Mathematics Achievement Test Raw Score. Minimum = 0, Maximum = 98.
- AGE97 – The child’s age in 1997. Minimum = 3, Maximum = 13.
- faminc97 – Total family income in 1997 (in 2002 constant dollars). Minimum = \$-72296.26, Maximum = \$784610.59.
- bthwht – Low birth weight status of the child. 0 = non-low birth weight child, 1 = low birth weight child.
- HOME97 – A composite total score of the emotional and cognitive stimulation at home. Minimum = 7, Maximum = 27.

Policy Research Report 3: Interaction Effects (15%)
Due: Tuesday November 12th

You are still working as a research consultant for an evaluation study examining the effect of the Women, Infant and Children (WIC) Nutrition Program participation during pregnancy on the raw scores of child mathematics achievement in 1997. Using the Child Development Supplement to the Panel Study of Income Dynamics (“good”), the evaluation team has asked you to examine if the effect of WIC program participation during pregnancy is moderated by (1) family income, (2) race, and (3) the current age of the child. Although assumptions still need to be evaluated it is not necessary to report the evaluation of the assumptions as was asked for in assignment 2. Simply state what changes were made to the data and model in order to account for assumption violations.

The report should include: the descriptive statistics (i.e. N, means/medians/proportions, standard deviations, frequencies, observed ranges, and correlations) of the variables and the multiple regression analyses with interaction effects. You are asked to provide in a one to two paragraph concluding statement your interpretation of the results and what they suggest. In an appendix, the evaluation team would like to see the syntax of your R program.

Use the following syntax to recode CHRACE to a centered binary variable called race:

```
data.frame$Race <- ifelse(data.frame$CHPACE== 9, NA,  
  ifelse(data.frame$CHPACE== 1, .5,  
  ifelse(data.frame$CHPACE== 2, -.5, NA)))
```

Definition of Variables:

- WICpreg – Women, Infant and Children (WIC) Nutrition Program participant during pregnancy: 0 = No, 1 = Yes.
- Race – Centered Binary Coding of Race: -0.5 = Black, 0.5 = White.
- mathraw97 – Woodcock-Johnson Revised Mathematics Achievement Test Raw Score. Minimum = 0, Maximum = 98.
- AGE97 – The child’s age in 1997. Minimum = 3, Maximum = 13.
- faminc97 – Total family income in 1997 (in 2002 constant dollars). Minimum = \$-72296.26, Maximum = \$784610.59.
- bthwht – Low birth weight status of the child. 0 = non-low birth weight child, 1 = low birth weight child.
- HOME97 – A composite total score of the emotional and cognitive stimulation at home. Minimum = 7, Maximum = 27.

Policy Research Report 4: Application of the General(ized) Linear Model in a Policy Research Report, Blog, Vlog, Op-Ed, or Podcast (40%)
Due: Friday December 6th

You are now a full-time researcher for a major social policy research institute. For this major social policy research institute you have received grant funding, have collected your data or are using pre-existing data, and are now ready to begin analyses. Using any data set (if you do not have any data to work with you are more than welcomed to use the PSID “good” data set, but you may also consider using the National Longitudinal Survey of Youth, the Health and Retirement Study, the National Longitudinal Study of Adolescent Health, Fragile Families, or any other pre-existing national probability sample), you are to examine any particular research question(s) of interests that will call for you to use one of the applications of the general(ized) linear model covered between Week 9 and Week 13 of the course. These applications include path modeling (using OLS regression), logistic regression, multilevel modeling, or growth modeling. You must use at least one application of the general(ized) linear model but are not limited to one.

You have the option to do a policy research report, blog, vlog, op-ed, or podcast. If you do a policy research report, you need to provide a brief introduction that frames and states your research question(s). Describe your data as you would in a methods section, report your analyses as you would in a results section, and provide a discussion of your interpretation and implications of the results. In the methods section you need to explain which modeling technique you are using, why it is appropriate for your research question(s), and present the equation for the model with all variables included in model. Assumptions need to be evaluated but it is not necessary to report everything done. Simply state if any assumptions were violated and what changes were made to the data and/or model in order to account for assumption violations. The results section should report: the descriptive statistics (i.e. N, means/medians/proportions, standard deviations, frequencies, observed ranges, and correlations) of the variables. If you do a policy blog or op-ed, the narrative of the blog or op-ed should refer to data from the research and you will also need submit a methods and results section of the analyses.

You may work in groups of two for this assignment but you must inform the instructor in advance. Given that this assignment is open to many research questions and the use of any data, please talk to the instructor first about what you plan to do. This assignment can be a good opportunity to work on the analyses for a journal article or an on-going research project including for the Data for Equitable Justice Lab. Please feel free to use this assignment as an opportunity to sharpen, strengthen, or simply complete the analyses for your existing on-going work.

GOOD LUCK & HAVE FUN! ☺