

94-827: SAS for Public Policy/Policy Analysis

Final 2022

Due 2022-03-04

121 points

The final exam has 4 parts. Part 1 requires you to read output and interpret the output. In Part 2, you must answer questions about existing code and answer questions about what the program does. Answers to Part 1 questions and Part 2 questions should be answered in comments.

Parts 3 and 4 are related. You are given a SAS program in Part 3 and you must run the code, and add descriptive comments. The output datasets of the program in Part 3 will be needed to complete your analysis in Part 4. The data and programs needed for the final can be found on Canvas and in the course folder under SASOnDemand. You may use lecture notes, SAS programs saved online/Canvas, your books, and the internet. Your work must be your own.

For Part 4, data sources will need to be prepared before getting to perform exploratory data analysis, run models and interpret results. All parts in RED should be answered and included in your program. Part 4 also includes graphical output. Export your final graphs to a PDF (with titles). Each student should turn in one PDF (from Part 4), one SAS program with the answers for Parts 1 and 2, and one SAS program for Parts 3 and 4 with the code and answers.

Part 1: Reading Output (30 points)

This SAS program called Parts_1and2.sas can be found on Canvas. Save the program to a personal location and answer the questions in comments. The datasets are in the FINAL library. You may run the program if you find it helpful. You may run other procedures to help you gain more information. If you do, include that code with your answers.

```
proc logistic data=Final.LoanApp;  
  class loanpurp mortapp/param=reference;  
  model loanapproved = loanpurp mortapp creditscore price;  
run;
```

| Model Information | | |
|---------------------------|------------------|-----------------------------|
| Data Set | FINAL.LOAN APP | |
| Response Variable | LoanApproved | Loan Approved (0=no, 1=yes) |
| Number of Response Levels | 2 | |
| Model | binary logit | |
| Optimization Technique | Fisher's scoring | |

| | |
|-----------------------------|------|
| Number of Observations Read | 4999 |
| Number of Observations Used | 4999 |

| Response Profile | | |
|------------------|--------------|-----------------|
| Ordered Value | LoanApproved | Total Frequency |
| 1 | 0 | 1880 |
| 2 | 1 | 3119 |

Probability modeled is LoanApproved=0.

| Class Level Information | | | | | |
|-------------------------|-------|------------------|---|---|---|
| Class | Value | Design Variables | | | |
| LoanPurp | 1 | | | | |
| MortApp | 1 | 1 | 0 | 0 | 0 |
| | 2 | 0 | 1 | 0 | 0 |
| | 3 | 0 | 0 | 1 | 0 |
| | 4 | 0 | 0 | 0 | 1 |
| | 5 | 0 | 0 | 0 | 0 |

| Model Convergence Status |
|---|
| Convergence criterion (GCONV=1E-8) satisfied. |

| Model Fit Statistics | | |
|----------------------|----------------|--------------------------|
| Criterion | Intercept Only | Intercept and Covariates |
| AIC | 6621.776 | 5942.847 |
| SC | 6628.293 | 5988.466 |
| -2 Log L | 6619.776 | 5928.847 |

| Testing Global Null Hypothesis: BETA=0 | | | |
|--|------------|----|------------|
| Test | Chi-Square | DF | Pr > ChiSq |
| Likelihood Ratio | 690.9290 | 6 | <.0001 |
| Score | 674.2830 | 6 | <.0001 |
| Wald | 578.6876 | 6 | <.0001 |

| Type 3 Analysis of Effects | | | |
|----------------------------|----|--------------------|------------|
| Effect | DF | Wald Chi-Square | Pr > ChiSq |
| LoanPurp | 0 | . | . |
| MortApp | 4 | 0.5195 | 0.9716 |
| CreditScore | 1 | 1.0123 | 0.3144 |
| Price | 1 | 577.7748 | <.0001 |

| Analysis of Maximum Likelihood Estimates | | | | | | |
|--|---|----|----------|-------------------|--------------------|------------|
| Parameter | | DF | Estimate | Standard Error | Wald Chi-Square | Pr > ChiSq |
| Intercept | | 1 | -2.1944 | 0.3683 | 35.5042 | <.0001 |
| MortApp | 1 | 1 | 0.0292 | 0.1450 | 0.0404 | 0.8406 |
| MortApp | 2 | 1 | 0.0234 | 0.1719 | 0.0186 | 0.8915 |
| MortApp | 3 | 1 | 0.0661 | 0.1961 | 0.1137 | 0.7359 |
| MortApp | 4 | 1 | 0.1230 | 0.2018 | 0.3714 | 0.5422 |
| CreditScore | | 1 | 0.000470 | 0.000467 | 1.0123 | 0.3144 |
| Price | | 1 | 0.00205 | 0.000085 | 577.7748 | <.0001 |

| Odds Ratio Estimates | | | |
|----------------------|--------|-------------------|-------------------------------|
| Effect | | Point Estimate | 95% Wald Confidence Limits |
| MortApp | 1 vs 5 | 1.030 | 0.775 1.368 |
| MortApp | 2 vs 5 | 1.024 | 0.731 1.434 |
| MortApp | 3 vs 5 | 1.068 | 0.727 1.569 |
| MortApp | 4 vs 5 | 1.131 | 0.761 1.680 |
| CreditScore | | 1.000 | 1.000 1.001 |
| Price | | 1.002 | 1.002 1.002 |

```
proc ttest data=Final.Study_GP;
  class section;
  var GPA;
run;
```

| Section | Method | N | Mean | Std Dev | Std Err | Minimum | Maximum |
|------------|---------------|----|--------|---------|---------|---------|---------|
| 01 | | 58 | 3.3014 | 0.3941 | 0.0517 | 2.4200 | 3.9400 |
| 02 | | 64 | 3.1013 | 0.4494 | 0.0562 | 1.9300 | 3.9100 |
| Diff (1-2) | Pooled | | 0.2001 | 0.4240 | 0.0769 | | |
| Diff (1-2) | Satterthwaite | | 0.2001 | | 0.0764 | | |

| Section | Method | Mean | 95% CL Mean | | Std Dev | 95% CL Std Dev | |
|------------|---------------|--------|-------------|--------|---------|----------------|--------|
| 01 | | 3.3014 | 3.1978 | 3.4050 | 0.3941 | 0.3332 | 0.4825 |
| 02 | | 3.1013 | 2.9890 | 3.2135 | 0.4494 | 0.3828 | 0.5443 |
| Diff (1-2) | Pooled | 0.2001 | 0.0479 | 0.3523 | 0.4240 | 0.3765 | 0.4854 |
| Diff (1-2) | Satterthwaite | 0.2001 | 0.0489 | 0.3514 | | | |

| Method | Variances | DF | t Value | Pr > t |
|---------------|-----------|--------|---------|---------|
| Pooled | Equal | 120 | 2.60 | 0.0104 |
| Satterthwaite | Unequal | 119.88 | 2.62 | 0.0099 |

| Equality of Variances | | | | |
|-----------------------|--------|--------|---------|--------|
| Method | Num DF | Den DF | F Value | Pr > F |
| Folded F | 63 | 57 | 1.30 | 0.3153 |

```
proc freq data=Final.birthwgt;
    table Drinking * lowbirthwgt;
    table AgeGroup * lowbirthwgt;
run;
```

| Table of Drinking by LowBirthWgt | | | |
|--|----------------------------------|-------------------------------------|-----------------|
| Drinking | LowBirthWgt | | |
| Frequency Percent Row Pct Col Pct | No | Yes | Total |
| No | 74173 78.58 91.67 85.57 | 6741 7.14 8.33 87.52 | 80914 85.72 |
| Yes | 12513 13.26 92.87 14.43 | 961 1.02 7.13 12.48 | 13474 14.28 |
| Total | 86686 91.84 | 7702 8.16 | 94388 100.00 |
| Frequency Missing = 5612 | | | |

| Table of AgeGroup by LowBirthWgt | | | |
|--|----------------------------------|---------------------------------------|------------------|
| AgeGroup | LowBirthWgt | | |
| Frequency Percent Row Pct Col Pct | No | Yes | Total |
| 1 | 9213 9.21 89.93 10.03 | 1032 1.03 10.07 12.68 | 10245 10.25 |
| 2 | 69836 69.84 92.34 76.03 | 5797 5.80 7.66 71.21 | 75633 75.63 |
| 3 | 12810 12.81 90.71 13.95 | 1312 1.31 9.29 16.12 | 14122 14.12 |
| Total | 91859 91.86 | 8141 8.14 | 100000 100.00 |

```
proc corr data=Final.Vite;
    var plaque sbp ldl hdl trig;
run;
```

| 5 Variables: | | | | Plaque SBP LDL HDL Trig | | | |
|-------------------|------|-----------|----------|---------------------------------|----------|-----------|---------------------------------|
| Simple Statistics | | | | | | | |
| Variable | N | Mean | Std Dev | Sum | Minimum | Maximum | Label |
| Plaque | 1500 | 0.63290 | 0.17252 | 949.35040 | 0.22090 | 1.08080 | Plaque measurement (mm) |
| SBP | 1500 | 141.87533 | 27.36373 | 212813 | 65.00000 | 234.00000 | Systolic blood pressure (mm/Mg) |
| LDL | 1500 | 135.52800 | 14.98582 | 203292 | 83.00000 | 185.00000 | LDL cholesterol (mg/DL) |
| HDL | 1500 | 45.86533 | 6.82776 | 68798 | 22.00000 | 71.00000 | HDL cholesterol (mg/DL) |
| Trig | 1500 | 173.56067 | 87.72554 | 260341 | 25.00000 | 503.00000 | triglycerides mg/dL |

| Pearson Correlation Coefficients, N = 1500 Prob > r under H0: Rho=0 | | | | | |
|--|--------------------|--------------------|--------------------|--------------------|--------------------|
| | Plaque | SBP | LDL | HDL | Trig |
| Plaque Plaque measurement (mm) | 1.00000 | -0.01398 0.5885 | 0.00029 0.9911 | -0.13821 <.0001 | 0.03225 0.2119 |
| SBP Systolic blood pressure (mm/Mg) | -0.01398 0.5885 | 1.00000 | 0.00766 0.7669 | -0.00086 0.9733 | -0.03266 0.2062 |
| LDL LDL cholesterol (mg/DL) | 0.00029 0.9911 | 0.00766 0.7669 | 1.00000 | -0.01074 0.6777 | 0.03352 0.1945 |
| HDL HDL cholesterol (mg/DL) | -0.13821 <.0001 | -0.00086 0.9733 | -0.01074 0.6777 | 1.00000 | 0.02471 0.3388 |
| Trig triglycerides mg/dL | 0.03225 0.2119 | -0.03266 0.2062 | 0.03352 0.1945 | 0.02471 0.3388 | 1.00000 |

```
proc tabulate data=Final.Population;
  class Continent;
  variable y1;
  table Continent, Y1*(ColPctN Mean Max);
run;
```

| | Population (in 100,000s) for 2013 | | |
|--|-----------------------------------|--------|----------|
| | ColPctN | Mean | Max |
| Continent name (AF - Africa, AS - Asia, EU - Europe, NA - North America, SA - South America, OC - Oceania, AN - Antarctica) | | | |
| AF | 24.55 | 205.39 | 1736.15 |
| AS | 23.18 | 811.26 | 13573.80 |
| EU | 23.64 | 163.41 | 1435.00 |
| NA | 15.00 | 169.77 | 3161.29 |
| OC | 8.18 | 21.12 | 231.31 |
| SA | 5.45 | 338.74 | 2003.62 |

Part 2: Analyzing Code (12 points)

In Part 2, answer the questions (1 to 8) regarding data steps and procedures using the dataset FINAL.CARS. Answer the questions using comments and submit them to Canvas.

Parts 3 and 4: Data Preparation, Original Program, and Policy Analysis (79 points)

Effects of marijuana dispensaries on crime and property transaction values

In 2010, Colorado legalized medical marijuana centers, after legalizing medical marijuana usage in 2000. Later in 2012, Colorado became one of the first states to end the ban on recreational marijuana. There have been several studies that look at the impacts of marijuana legalization, and specifically of the dispensaries. These studies are often geospatial in nature, analyzing incidents or conditions near the dispensaries. We will look at one study examining the location of dispensaries and crime incidents in Denver, Colorado for the final exam:

Marijuana Dispensaries and Neighborhood Crime and Disorder in Denver, Colorado.
Lorine A. Hughes, Lonnie M. Schaible, and Katherine Jimmerson. Justice Quarterly
(2019)

The paper is provided for your reference and may help with understanding why and how some steps should be completed. The final will not directly replicate the steps done in the paper due to time constraints and complexity. Some of the tables will be referenced, but is not necessary to read the entire paper.

Crime and Disorder (2019) concludes that “except for murder, the presence of at least one medical marijuana dispensary was associated with statistically significantly increased neighborhood crime and disorder.” Using the methods you’ve learned, you will analyze some of the same data sources and indicate whether you agree with the conclusions of the model paper.

Answers to questions and descriptions of the process can be typed into your program as comments. Six (6) points of your score reflects good coding practices, such as descriptive comments, usage of titles for procedures, and checks on new data or newly derived variables. (6 points)

During this portion of the final you should use, at least once (in no particular order):

- PROC SORT
- IF...THEN
- New variable creation
- IN=
- SAS functions
- ODS Output
- PROC CONTENTS
- Merging
- PROC MEANS or TABULATE
- PROC FREQ
- Any other procedure explicitly mentioned

Description of datasets

1. Unfortunately, the data for this analysis are in several datasets and in different formats. This is common when working with data, particularly administrative datasets that were not created for the sake of analysis like survey data. Below, each file needed to complete the analysis is listed.
 - a. *Tract_medical_disp_totals.sas7bdat*. This is a SAS dataset that has the number of medical marijuana dispensaries in each census tract of Denver for the years 2010 to 2019. For example, a tract may have had 2 dispensaries in 2016, and the same tract may have had 4 dispensaries in 2019.
 - b. *Tract_retail_disp_totals.sas7bdat*. This is a SAS dataset that has the number of retail marijuana dispensaries in each census tract of Denver for each year from 2010 to 2019.
 - c. *Years.sas7bdat*. This is a SAS dataset that lists each year from 2010 to 2019.
 - d. *Tract_nhood.sas7bdat*. This SAS dataset links each census tract to a neighborhood in Denver. Each census tract is fully contained in one neighborhood, but some neighborhoods contain multiple census tracts.
 - e. *Crime.csv*. The crime file contains incident level information on each crime in Denver from 2015 through part of 2020. The neighborhood in which the crime occurred is listed. This file will be imported for you by running Part3.sas.
 - f. *ACS2018_DenverTracts.sas7bdat*. This file contains selected demographic information for all of the tracts in Denver based on the ACS 5 year tract estimates from 2014 to 2018. The list of demographics was based on the *Crime and Disorder* paper.

Crime Data Preparation

2. Run Part3.sas and respond to the prompts (A – Q) in your comments. (23 points)
3. Merge the datasets containing retail dispensary totals, the medical dispensary totals, and TRACT_NHOOD_YEAR together by the fips code and year. If a tract does not have a record for a given year for either of the dispensary files, assume that the tract had 0 dispensaries of that type for that year. (3 points)
4. The analysis is *Crime and Disorder* is at the neighborhood level. Use appropriate procedures to calculate the total number of each dispensary type in each neighborhood in each year, using the file created in #3. Your output dataset (call it NEIGHBORHOOD_DISPENS) from this step should have one record per neighborhood – year combination. Next, calculate 3 binary variables, where 1 indicates there is at least 1 dispensary of that type in the neighborhood: (4 points)
 - a. Has_retail_dispensary
 - b. Has_medical_dispensary
 - c. Has_dispensary (at least one of any type)

5. Merge **CRIME03** and **CRIME_MONTH_NEIGHBORHOOD** together by neighborhood, year, month, and crime category. If a neighborhood does not have a record for a crime category in a given month, assume that the neighborhood had 0 crimes of that category for that month. Keep only the crimes listed in Table 1 of *Crime and Disorder* and keep only records from 2015 to 2019 in your dataset. (4 points)
 - a. Was this a 1 to 1, 1 to Many, or Many to Many merge?
6. Merge the dataset you created in #5 with **NEIGHBORHOOD_DISPENS** using neighborhood and year. Ensure the resulting files only has records between 2015 and 2019. We are attaching the number of dispensaries in a given year for a neighborhood, to that neighborhoods monthly crime totals. (3 points)
7. Merge the dataset created in #6 with **NHOOD_3** using neighborhood. The period over which the ACS data were collected, roughly aligns with the period in which crime data were collected. Using the crime totals and neighborhood population, compute crime rate (# of crimes per 100,000 people). This is your **CRIME_ANALYTIC_DATASET**. (3 points)

Analysis

8. Using appropriate procedures and your analytic dataset produce tables that show descriptive statistics similar to Table 1 of *Crime and Disorder*. (3 points)
9. Using appropriate procedures, create two charts of your choosing and output them to a PDF. Using footnotes, write 1 to 3 sentences describing what the charts show. Be sure to submit your PDF on Canvas! (5 points)
10. Use appropriate procedures to determine if the crime rate you calculated for given crime types are (approximately) normally distributed. EACH CRIME TYPE SHOULD BE CONSIDERED SEPARATELY. (4 points)
11. Use appropriate procedures to determine if any of the ACS neighborhood demographics are associated with crime rates or monthly totals for Public Disorder. (4 points)
 - a. List any significant correlations in your comments.
12. Macro creation. Create and invoke a macro that accomplishes the following. It does not have to have a %DO loop, but if that is your preference, that is fine. (6 points)
 - a. Use **PROC GLM** to run a linear regressions for EACH crime category, where the crime rate is the dependent variable in one set, and monthly totals is the dependent variable in the other set. The independent variables should include the number of each type of dispensary in the neighborhood. Crime has seasonal components (higher in some months while lower in others). Include in your regression dummy variables for months and years (HINT: These are categorical in

this case). Finally include the neighborhood demographic variables you believe make sense. (6 points)

- b. AT THE END OF YOUR MODEL STATEMENT, MAKE SURE YOU INCLUDE “/ solution” before the semi-colon. If you do not, SAS will not produce parameter estimates.

13. Discuss significant findings related to the demographics and your variables of interest. There will be a lot of regressions (16) and lots of regression coefficients. Some patterns should emerge. Don't try to discuss everything individually. If you want to focus on just one, choose Public Disorder. This discussion does not have to be long (2 to 4 paragraphs), and if you prefer to write it in Word, that is fine. Some guiding questions: (11 points)

- a. Based on the R-squared, generally speaking which set of models explains more of the variation: crime rates or monthly totals?
- b. Which crime types seem positively related to the number of retail dispensaries? Which are negatively related? Which are unrelated?
- c. Choose one of the demographic characteristics. Discuss whether it is significantly related to the any or most of the crime rates.
- d. In some cases, are there differences between the effects of medical versus retail dispensaries?
- e. Do you agree with the conclusion of the paper?