Running head: TITLE 1

Final Project: Group 3

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Author Note

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Abstract 10

Sign Up: One or two sentences providing a basic introduction to the field, 11

comprehensible to a scientist in any discipline. 12

Two to three sentences of more detailed background, comprehensible to scientists 13

in related disciplines.

One sentence clearly stating the **general problem** being addressed by this particular 15

study. 16

One sentence summarizing the main result (with the words "here we show" or their 17

equivalent). 18

Two or three sentences explaining what the **main result** reveals in direct comparison

to what was thought to be the case previously, or how the main result adds to previous

knowledge.

One or two sentences to put the results into a more **general context**. 22

Two or three sentences to provide a **broader perspective**, readily comprehensible to 23

a scientist in any discipline.

Keywords: keywords 25

Word count: X 26

Final Project: Group 3

28 Introduction

Despite advances in access to health services, quality of care, and overall gains in life
expectancy, racial/ethnic disparities in health in the United States (U.S.) remains to
disproportionally affect the lives of racial/ethnic minority groups. Williams and
Mohammed (2009) refer to the finding of Levine et al. (2016) that approximately 100,000
African Americans who would not die if there were no racial disparities die prematurely
every year. Unfortunately, in the mental health arena, racial/ethnic disparities are no
exception.

Even though the burden and impact of physical diseases on different racial/ethnic subgroups have been far more studied than the impact and burden of mental health disorders, we know that globally, depression is the leading cause of disability and loss of productivity, and that its direct outcome, death via suicide, is on the rise (WHO, 2018 (accessed December 4, 2018)). It is well-known that mental health services are costly and thus a high proportion of the American population cannot afford them.

Given that depression is usually screened and treated first in primary care settings,
access to medical care is the first barrier to treatment that racial/ethnic minority groups
face (Williams & Mohammed, 2009). From there, racial/ethnic minority groups experience
barriers such as low detection rate of mental health disorders in comparison to Whites
(Borowsky et al., 2000); language barriers for non-English speakers (Fiscella, Franks,
Doescher, & Saver, 2002); use of screening measures not translated or validated for
racial/ethnic minority groups; issues of trust related but not limited to underrepresentation
of racial/ethnic minorities among mental health professionals, and cultural differences in
understanding and treating mental health disorders (Miranda & Cooper, 2004). Overall,
these and other barriers affect the access and quality of treatment racial/ethnic minority
groups receive in respect to their mental health.

—Maria will do drug use onset and health disparities (short 1 paragraph) — use of substances as an alternative treatment for depression that people may recur to—

In the present study, we aim to explore some of the health disparities among different racial/ethnic groups using a nationally representative sample, the National Health and Nutrition Examination Survey (NHANES) 2015 – 2016.

58 Methods

Sign up:

We report how we determined our sample size, all data exclusions (if any), all manipulations, and all measures in the study.

62 Participants

63 Sign up:

64 Measures

sign up: Alejandra will describe the measures for depression, use of mental health services, and insurance coverage.

Data analysis

We used R (Version 3.5.1; R Core Team, 2018) and the R-packages bindrcpp (Version 0.2.2; Müller, 2018), dplyr (Version 0.7.8; Wickham, François, Henry, & Müller, 2018), forcats (Version 0.3.0; Wickham, 2018a), ggplot2 (Version 3.0.0; Wickham, 2016), here (Version 0.1; Müller, 2017), kableExtra (Version 0.9.0; Zhu, n.d.), papaja (Version 0.1.0.9842; Aust & Barth, 2018), purrr (Version 0.2.5; Henry & Wickham, 2018), readr (Version 1.2.1; Wickham, Hester, & Francois, 2018), rio (Version 0.5.10; C.-h. Chan, Chan,

- Leeper, & Becker, 2018), stringr (Version 1.3.1; Wickham, 2018b), tibble (Version 1.4.2;
- Müller & Wickham, 2018), tidyr (Version 0.8.2; Wickham & Henry, 2018), and tidyverse

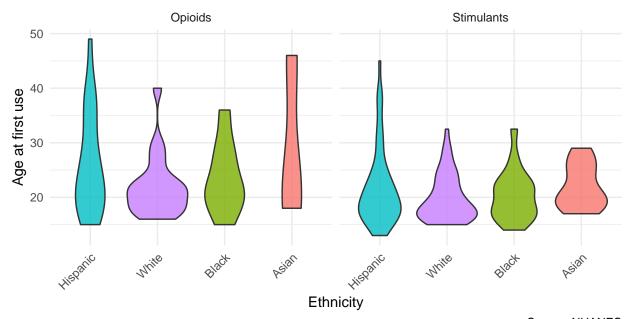
76 (Version 1.2.1; Wickham, 2017) for all our analyses. sign up:

77 Results

- We should use inline code here sign up:
- Alejandra will do the description of her plot (depression, health insurance coverage,
- visit to mental health, broken down by ethnicities).
- ## Warning: Removed 425 rows containing missing values (geom_col).
 - Shaina will describe her plot of drug use by ethnicities

Age of First Use by Ethnicity

Based on two drug types (opioids and stimulants)

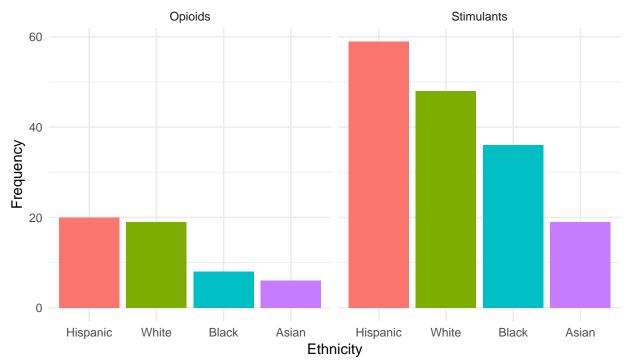


Source: NHANES

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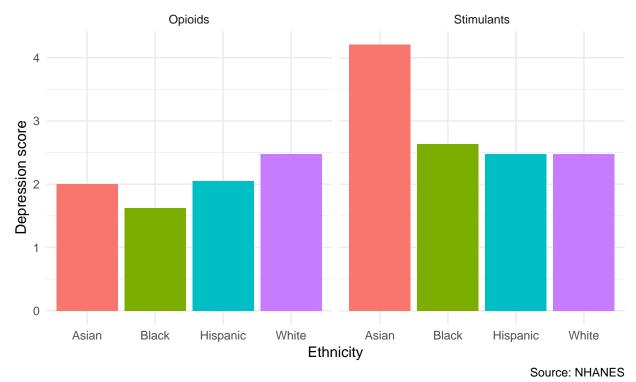
82

Number of Individuals Reporting Drug Use by Ethnicity Based on two drug types (opioids and stimulants)



Source: NHANES

Average Depression Score by Ethnicity Based on two drug types (opioids and stimulants)



ST table

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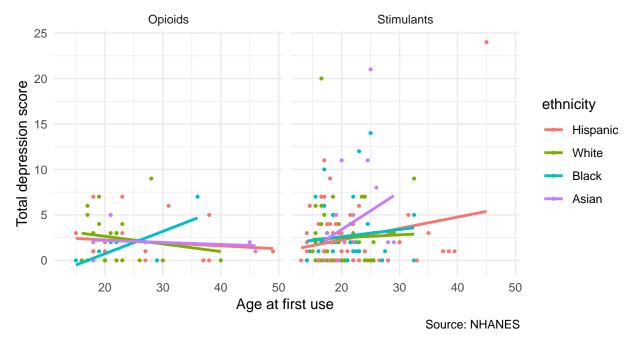
Warning: Expected 2 pieces. Missing pieces filled with NA in 10 rows [1, ## 2, 3, 4, 5, 6, 7, 8, 9, 10].

91 Discussion

sign up:

Insert one data visualization – we are using two, I believe Alejandra's viz. and one from Shaina Exploratory association plot?

Association between Age at First Use and Depression Colored to show differences among drug types



96 Insert Table – JP?

95

97 References

- sign up: Alejandra will include the references used in the intro.
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¹⁴⁴ Zhu, H. (n.d.). KableExtra: Construct complex table with 'kable' and pipe syntax.

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20-47.

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health: Evidence and needed research. Journal of Behavioral Medicine, 32(1),

Table 1

Average depression score and age of first opioid and stimulant use

Ethnicity	Depression Score	Opioid Use	Stimulant Use
Asian	3.01	27.86	21.73
Black	3.16	22.18	20.30
Hispanic	3.29	25.81	21.43
Other/Multiracial	3.47	22.33	18.93
White	3.53	21.60	20.23

Total Depression Score by Ethnicity and Usage of Mental Health Services Faceted by Insurance Coverage

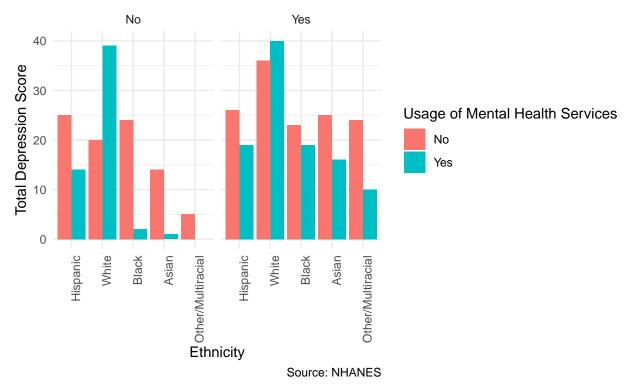


Figure 1