DARE 1

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A. Data Management Tasks (1 point)

For these tasks, no write up is required. The code you submit will be sufficient.

A1. Convert the raw counts of enrollment by race/ethnicity into percentages (i.e., divide the enrollment count for each ethno-racial category by total enrollment). For programming efficiency, can you use a function to do this task?

```
dare1 <- dare1 %>%
  mutate(across(c(10:15), ~ . / !! dare1$enroll * 100))
```

A2. Generate dichotomous policy predictor variables that take the value of 1 in state-year observations in which the policy is in place. Call them eval, class remove and suspension. They should take the value of 0 in years during which these policies were not in place.

Also, generate a running time variable (run time) that reflects how far or close the state-year observation is from the implementation of higher stakes teacher evaluation and a variable that permits the effects of the evaluation policy to vary (linearly) over time (evalXyear). How will you deal with states that never implement evaluation? Do that too.

```
dare1 <- dare1 %>%
  mutate(run_time = ifelse(is.na(eval_year), -99, school_year-eval_year)) %>%
# -99 for states that never implement evaluation
mutate(evalXyear = eval*run_time)
```

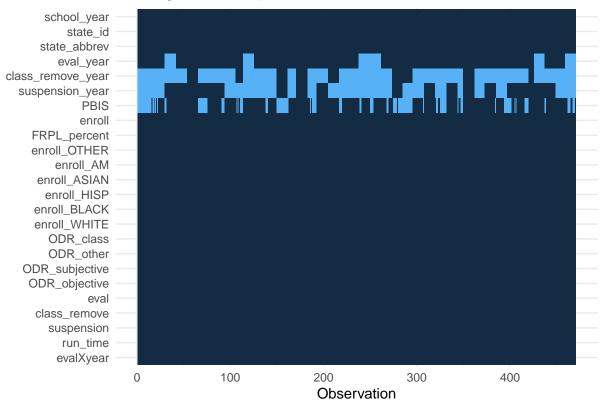
B. Understanding the Data and Descriptive Statistics (3 points)

For the following tasks, give your best attempt at completing the analysis and write-up. If you are unable to conduct the programming or analysis, describe what you are attempting to do and what your results would mean.

B1. Inspect your data. What sorts of missingness exist within the data file? What sorts of missingness should concern you? Which do not? In this assignment, please restrict your sample to state-years with non-missing outcomes.

dare1 %>%
 drop_na(ODR_class, ODR_objective, ODR_other, ODR_objective)%>%
 missing_plot()

Missing values map



dare1 %>% summary()

```
##
     school_year
                       state_id
                                     state_abbrev
                                                            eval_year
##
    Min.
            :2006
                    Min.
                            : 2.00
                                     Length:516
                                                         Min.
                                                                 :2011
    1st Qu.:2009
##
                    1st Qu.:18.00
                                     Class : character
                                                          1st Qu.:2013
    Median:2012
                    Median :29.00
                                                         Median:2014
                                     Mode : character
            :2012
                    Mean
                            :29.16
                                                         Mean
                                                                 :2014
##
    Mean
                                                          3rd Qu.:2014
##
    3rd Qu.:2014
                    3rd Qu.:41.00
##
    Max.
            :2017
                    Max.
                            :56.00
                                                         Max.
                                                                 :2016
##
                                                          NA's
                                                                 :72
##
                                              PBIS
                                                               enroll
    class_remove_year suspension_year
##
    Min.
            :2009
                       Min.
                               :2007
                                        Min.
                                                :0.0000
                                                                      216
                                                           Min.
    1st Qu.:2009
##
                       1st Qu.:2011
                                        1st Qu.:0.0000
                                                           1st Qu.:
                                                                     2891
##
    Median :2012
                       Median:2014
                                        Median :1.0000
                                                           Median :
                                                                     9764
##
    Mean
            :2012
                       Mean
                               :2013
                                        Mean
                                                :0.7214
                                                           Mean
                                                                  : 21897
                                                           3rd Qu.: 26510
##
    3rd Qu.:2015
                       3rd Qu.:2016
                                        3rd Qu.:1.0000
##
    Max.
            :2018
                       Max.
                               :2018
                                        Max.
                                                :1.0000
                                                           Max.
                                                                  :207879
##
    NA's
            :408
                       NA's
                                        NA's
                                                           NA's
                               :288
                                                :175
                                                                  :46
##
     FRPL_percent
                        enroll_OTHER
                                              enroll_AM
                                                                enroll_ASIAN
##
    Min.
            :0.07763
                       Min.
                              : 0.00000
                                           Min.
                                                   : 0.0000
                                                               Min.
                                                                      : 0.000
    1st Qu.:0.44201
                       1st Qu.: 0.00000
                                            1st Qu.: 0.3189
                                                               1st Qu.: 1.076
    Median : 0.53159
                       Median : 0.00000
                                           Median : 0.5504
                                                               Median: 1.965
##
```

```
:20.81448
                                                    :86.8996
                                                                       :17.611
##
    Max.
            :1.00000
                       Max.
                                            Max.
                                                                Max.
##
    NA's
            :46
                       NA's
                               :46
                                            NA's
                                                    :46
                                                                NA's
                                                                       :46
##
     enroll HISP
                       enroll BLACK
                                          enroll WHITE
                                                               ODR class
##
    Min.
            : 0.000
                              : 0.000
                                         Min.
                                                : 9.607
                                                                    :0.1612
                      Min.
                                                            Min.
    1st Qu.: 3.760
                      1st Qu.: 2.860
                                         1st Qu.: 47.575
##
                                                            1st Qu.:0.9673
    Median: 8.697
                      Median : 6.094
##
                                         Median: 67.423
                                                            Median :1.4329
##
    Mean
            :13.744
                      Mean
                              :11.663
                                         Mean
                                                : 62.068
                                                            Mean
                                                                    :1.6872
##
    3rd Qu.:18.147
                      3rd Qu.:18.487
                                         3rd Qu.: 78.330
                                                            3rd Qu.:1.9747
##
    Max.
            :76.691
                      Max.
                              :88.201
                                         Max.
                                                 :137.468
                                                            Max.
                                                                    :9.8629
    NA's
                                         NA's
##
            :46
                      NA's
                              :46
                                                 :46
                                                            NA's
                                                                    :46
                                                                   eval
##
      ODR_other
                      ODR_subjective
                                          ODR_objective
            :0.1533
##
    Min.
                      Min.
                              :0.09597
                                          Min.
                                                  :0.04506
                                                             Min.
                                                                     :0.0000
##
                      1st Qu.:0.59837
    1st Qu.:0.9565
                                          1st Qu.:0.37276
                                                             1st Qu.:0.0000
##
    Median :1.4003
                      Median :0.89286
                                          Median :0.52533
                                                             Median :1.0000
                              :1.09670
##
    Mean
            :1.5334
                                          Mean
                                                  :0.60468
                                                             Mean
                                                                     :0.6124
                      Mean
##
    3rd Qu.:1.8548
                      3rd Qu.:1.29252
                                          3rd Qu.:0.76524
                                                             3rd Qu.:1.0000
                                                  :3.06346
##
    Max.
            :7.9305
                              :6.84706
                                          Max.
                                                             Max.
                                                                     :1.0000
                      Max.
##
    NA's
            :46
                      NA's
                              :46
                                          NA's
                                                  :46
##
     class_remove
                       suspension
                                           run_time
                                                             evalXyear
##
            :0.000
                             :0.0000
                                               :-99.00
                                                                  :-10.000
    Min.
                     Min.
                                        Min.
                                                          Min.
                     1st Qu.:0.0000
                                        1st Qu.: -7.00
                                                          1st Qu.: -5.000
##
    1st Qu.:0.000
                     Median :0.0000
                                        Median : -3.00
##
    Median : 0.000
                                                          Median : -1.000
##
    Mean
            :0.126
                     Mean
                             :0.2888
                                        Mean
                                               :-15.57
                                                          Mean
                                                                  : -2.384
    3rd Qu.:0.000
                     3rd Qu.:1.0000
                                        3rd Qu.:
                                                  0.00
                                                          3rd Qu.:
                                                                     0.000
##
            :1.000
                             :1.0000
                                                  6.00
                                                                     0.000
    Max.
                     Max.
                                        Max.
                                               :
                                                          Max.
##
```

: 0.32800

3rd Qu.: 0.00492

: 3.1194

3rd Qu.: 1.2069

Mean

: 3.091

3rd Qu.: 3.826

Mean

:0.54094

3rd Qu.:0.62681

Mean

##

##

Mean

```
dare1_clean <- dare1 %>%
  drop_na(ODR_class, ODR_objective, ODR_other, ODR_objective)
```

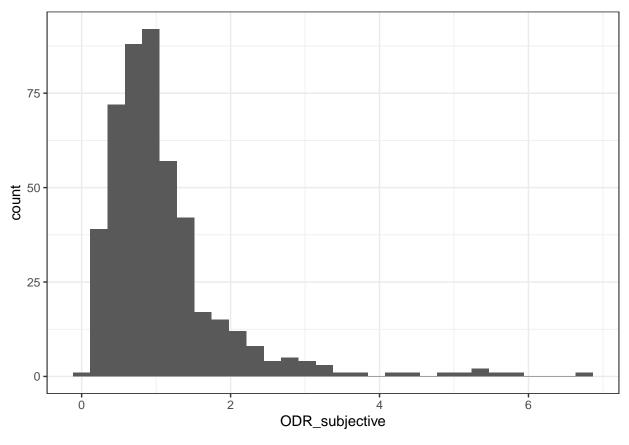
After excluding row with missing outcomes, there are only 470 observations left. Missing values found in these following variable: Var eval_year = 71, class_remove_year = 374, suspension_year = 259, PBIS = 129, based on the missingness pattern reflected in the plot above we see that missing data values do not relate to any other data in the dataset and there is no pattern to the actual values of the missing data themselves. Therefore we can conclude that this is Missing Completely at Random (MCAR). We should be concerned if there is specific pattern of the missingness.

B2. Graphically display the distribution of the outcome data. What do you notice about the distribution of outcomes? Are there any actions, transformations or sensitivity tests you would like to conduct based on this evidence?

```
# pivot_longer --> values to "ODR" could be another approach

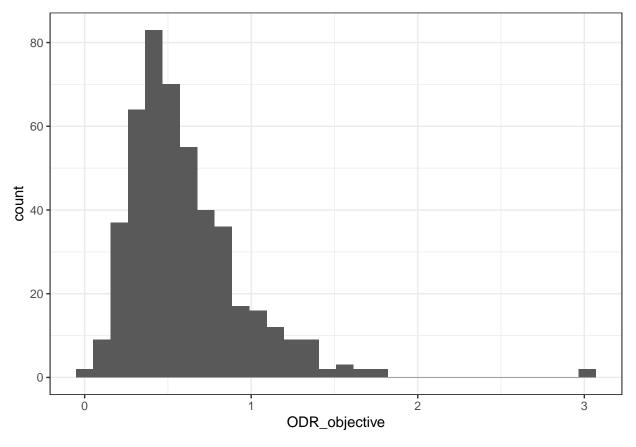
dare1_clean %>%
    ggplot(aes(ODR_subjective)) +
    geom_histogram() +
    theme_bw()
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



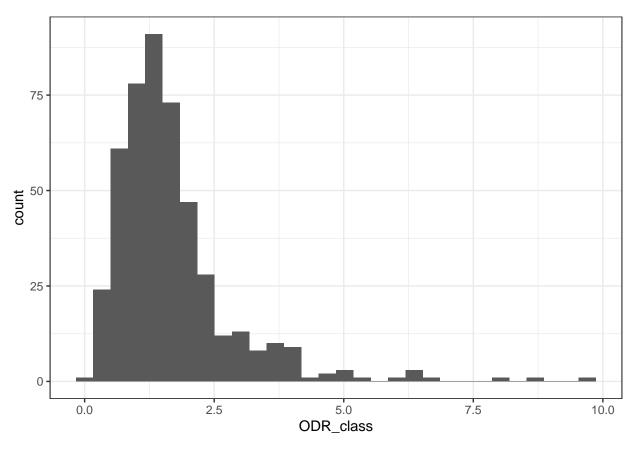
```
dare1_clean%>%
  ggplot(aes(ODR_objective)) +
  geom_histogram()+
  theme_bw()
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



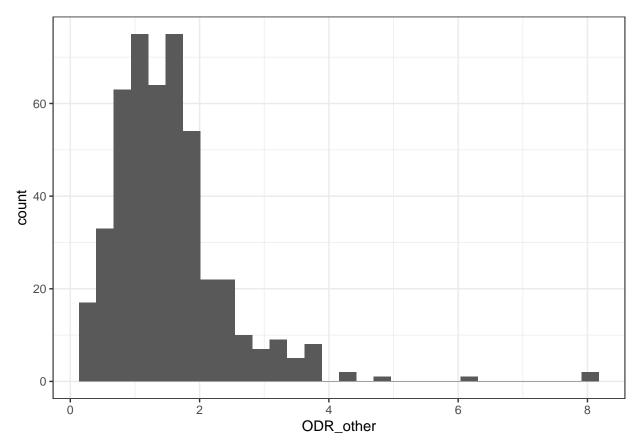
```
dare1_clean %>%
  ggplot(aes(ODR_class)) +
  geom_histogram()+
  theme_bw()
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



```
dare1_clean %>%
  ggplot(aes(ODR_other)) +
  geom_histogram()+
  theme_bw()
```

'stat_bin()' using 'bins = 30'. Pick better value with 'binwidth'.



Each ODR outcome data is right-skewed. Based on these histograms, we would suggest 1) testing for normality (applying transformations as needed) and 2) removing outliers, particularly for ODR_objective and ODR_class.

B3. What is the analytic sample from which you will draw your inferences? To what population are you drawing these inferences? For this analytic sample, reproduce Column 1 of Table 1 from Liebowitz, Porter & Bragg (2022) to create a summary of descriptive statistics for the following data elements. All of these statistics (except for state-year and year enrollment) should be weighted by the state-year population:

- Mean state-year enrollment
- Mean year enrollment
- % low-income (FRPL)
- % Am. Indian/Alask. Native
- % Asian/PI
- % Black
- % Hispanic
- % White
- % state-year observations in which PBIS was successfully implemented
- Classroom ODR rate
- Other location ODR rate
- Subjective-Classroom ODR rate
- Objective-Classroom ODR rate

stargazer(attitude)

##

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harv

```
## % Date and time: Mon, Jan 17, 2022 - 23:16:17
## \begin{table}[!htbp] \centering
     \caption{}
     \label{}
##
## \begin{tabular}{@{\extracolsep{5pt}}lccccccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
## Statistic & \multicolumn{1}{c}{N} & \multicolumn{1}{c}{Mean} & \multicolumn{1}{c}{St. Dev.} & \multi
## \hline \\[-1.8ex]
## rating & 30 & 64.633 & 12.173 & 40 & 58.8 & 71.8 & 85 \\
## complaints & 30 & 66.600 & 13.315 & 37 & 58.5 & 77 & 90 \
## privileges & 30 & 53.133 & 12.235 & 30 & 45 & 62.5 & 83 \\
## learning & 30 & 56.367 & 11.737 & 34 & 47 & 66.8 & 75 \\
## raises & 30 & 64.633 & 10.397 & 43 & 58.2 & 71 & 88 \\
## critical & 30 & 74.767 & 9.895 & 49 & 69.2 & 80 & 92 \\
## advance & 30 & 42.933 & 10.289 & 25 & 35 & 47.8 & 72 \\
## \hline \\[-1.8ex]
## \end{tabular}
## \end{table}
stargazer(dare1, header= FALSE, type = 'latex')
##
## \begin{table}[!htbp] \centering
    \caption{}
##
    \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lccccccc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
## Statistic & \multicolumn{1}{c}{N} & \multicolumn{1}{c}{Mean} & \multicolumn{1}{c}{St. Dev.} & \multi
## \hline \\[-1.8ex]
## school\_year & 516 & 2,011.500 & 3.455 & 2,006 & 2,008.8 & 2,014.2 & 2,017 \\
## state\_id & 516 & 29.163 & 14.760 & 2 & 18 & 41 & 56 \\
## eval\_year & 444 & 2,013.541 & 1.446 & 2,011.000 & 2,013.000 & 2,014.000 & 2,016.000 \\
## class\_remove\_year & 108 & 2,012.333 & 3.349 & 2,009.000 & 2,009.000 & 2,015.000 & 2,018.000 \\
## suspension\_year & 228 & 2,012.895 & 3.455 & 2,007.000 & 2,011.000 & 2,016.000 & 2,018.000 \\
## PBIS & 341 & 0.721 & 0.449 & 0.000 & 0.000 & 1.000 & 1.000 \\
## enroll & 470 & 21,897.410 & 32,425.870 & 216.000 & 2,890.750 & 26,509.500 & 207,879.000 \\
## FRPL\_percent & 470 & 0.541 & 0.150 & 0.078 & 0.442 & 0.627 & 1.000 \\
## enroll\_OTHER & 470 & 0.328 & 1.222 & 0.000 & 0.000 & 0.005 & 20.814 \\
## enroll\_AM & 470 & 3.119 & 8.959 & 0.000 & 0.319 & 1.207 & 86.900 \\
## enroll\_ASIAN & 470 & 3.091 & 3.195 & 0.000 & 1.076 & 3.826 & 17.611 \\
## enroll\_HISP & 470 & 13.744 & 14.047 & 0.000 & 3.760 & 18.147 & 76.691 \\
## enroll\_BLACK & 470 & 11.663 & 12.370 & 0.000 & 2.860 & 18.487 & 88.201 \\
## enroll\_WHITE & 470 & 62.068 & 21.141 & 9.607 & 47.575 & 78.330 & 137.468 \\
## ODR\_class & 470 & 1.687 & 1.182 & 0.161 & 0.967 & 1.975 & 9.863 \\
## ODR\_other & 470 & 1.533 & 0.905 & 0.153 & 0.956 & 1.855 & 7.931 \\
## ODR\_subjective & 470 & 1.097 & 0.857 & 0.096 & 0.598 & 1.293 & 6.847 \\
## ODR\_objective & 470 & 0.605 & 0.351 & 0.045 & 0.373 & 0.765 & 3.063 \\
## eval & 516 & 0.612 & 0.488 & 0 & 0 & 1 & 1 \\
## class\_remove & 516 & 0.126 & 0.332 & 0 & 0 & 0 & 1 \\
## suspension & 516 & 0.289 & 0.454 & 0 & 0 & 1 & 1 \\
## run\_time & 516 & $-$15.570 & 33.808 & $-$99 & $-$7 & 0 & 6 \\
## evalXyear & 516 & $-$2.384 & 2.813 & $-$10 & $-$5 & 0 & 0 \\
```

```
## \hline \\[-1.8ex]
## \end{tabular}
## \end{table}

#Mean State-Year Enrollment
dare1_clean %>%
    group_by(state_abbrev) %>% #use abbrev for readability
    summarise(mean_state = mean(enroll))
```

A tibble: 43×2

state_abbrev mean_state 1 AK 230. 2 AR 3859. 3 AZ 2983. 4 CA 111710. 5 CO 26989. 6 CT 26782. 7 FL 2165. 8 GA 21169. 9 IA 19158. 10 ID 2768 # . . . with 33 more rows

```
#Mean Year Enrollment
dare1_clean %>%
  group_by(school_year) %>%
  summarise(mean_year = mean(enroll))
```

A tibble: 12×2

```
#Mean State-Year Enrollment
dare1_clean %>%
  group_by(state_abbrev, school_year) %>%
  summarise(mean_state = mean(enroll))
```

'summarise()' has grouped output by 'state_abbrev'. You can override using the '.groups' argument.

A tibble: 470×3

Groups: state abbrev [43]

state_abbrev school_year mean_state 1 AK 2010 229 2 AK 2011 224 3 AK 2012 221 4 AK 2013 255 5 AK 2014 216 6 AK 2016 229 7 AK 2017 234 8 AR 2006 782 9 AR 2007 3131 10 AR 2008 3491 # . . . with 460 more rows

```
## Table printed with 'knitr::kable()', not {gt}. Learn why at
## http://www.danieldsjoberg.com/gtsummary/articles/rmarkdown.html
## To suppress this message, include 'message = FALSE' in code chunk header.
```

Characteristic	N = 341
% low-income (FRPL)	0.52 (0.13)
% Am. Indian/Alask. Native	1.44 (3.70)
% Asian/PI	3.4(3.2)
% Black	11 (11)
% Hispanic	13 (12)
% White	65 (18)
% Schools by Year Implementing PBIS	246 (72%)
Classroom ODR Rate	1.65 (0.91)
Other location ODR Rate	1.55(0.74)
Subjective-Classroom ODR rate	$1.06 \ (0.63)$
Objective-Classroom ODR rate	$0.61 \ (0.30)$

Table 1: Table 1. Summary Statistics

Describe the characteristics of your sample as you would report these statistics in an academic paper. How are the characteristics of the sample you will be using for this replication exercise different from the sample in Liebowitz, Porter & Bragg (2022)? How, if at all, do you anticipate this will affect your results?

This sample primarily differs at level of detail. This sample is at the year and state level while the Liebowitz, Porter & Bragg include school and grade level, as well. As a result, the characteristics of the above variables will look different: for example, one school may have much higher FRPL than another, but that variation may look different year to year.

B4. Optional Extension Plot the average classroom (ODR class) and classroom-subjective ODRs (ODR subjective) by how close the stateyear observation is to the implementation of the teacher evaluation policy for the states that implemented evaluation reform. (Note: this is similar to Figure 2 in the original paper). What do you notice about the raw outcome data plotted against the secular trend? Are there any actions, transformations or sensitivity tests you would like to conduct based on this evidence? Why do we stress plotting these raw averages only for states that implemented evaluation reform? How would including these states alter the interpretation of this figure?

C. Replication and Extension (6 points)

For the following tasks, give your best attempt at completing the analysis and write-up. If you are unable to conduct the programming or analysis, describe what you are attempting to do and what your results would

mean.

C1. Estimate the effects of the introduction of higher-stakes teacher evaluation reforms on Office Disciplinary Referrals. In one of your models, assume that the effects are constant and in another relax this assumption to allow the effects to differ (linearly) over time. Present these difference-in-differences estimates in a table and the associated writeup as you would report these results in an academic paper. Do you notice any important differences in these results and those reported in the original paper? If so, how would you consider addressing them (it is not necessary at this point for you to actually conduct the analysis, just describe approaches you might take)?

For classroom ODRs: Assume effects are constant

NOTE: 46 observations removed because of NA values (LHS: 46, Weights: 46).

```
summary(mod_class_constant)
```

```
## OLS estimation, Dep. Var.: ODR_class
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (state_id)
## Estimate Std. Error t value Pr(>|t|)
## eval 0.035771 0.060818 0.588161 0.55957
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 41.9 Adj. R2: 0.801764
## Within R2: 0.001163
```

Allow effects to differ over time

NOTE: 46 observations removed because of NA values (LHS: 46, Weights: 46).

```
summary(mod_class_time)
```

```
## OLS estimation, Dep. Var.: ODR_class
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (school_year^state_id)
## Estimate Std. Error t value Pr(>|t|)
## eval 0.035771 0.045199 0.791413 0.4291
## ---
```

```
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## RMSE: 41.9
                  Adj. R2: 0.801764
                Within R2: 0.001163
##
For subjective ODRs:
Assume effects are constant
mod_subj_constant <- feols(ODR_subjective ~ eval |</pre>
              state_id + school_year,
              data = dare1,
              weights = dare1$enroll)
## NOTE: 46 observations removed because of NA values (LHS: 46, Weights: 46).
summary(mod_subj_constant)
## OLS estimation, Dep. Var.: ODR_subjective
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (state_id)
        Estimate Std. Error t value Pr(>|t|)
## eval 0.03298
                 0.043522 0.757784 0.45281
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## RMSE: 29.3
                 Adj. R2: 0.791287
##
                Within R2: 0.002023
Allow effects to differ over time
mod_subj_time <- feols(ODR_subjective ~ eval |</pre>
              state_id + school_year,
              data = dare1,
              vcov = ~school_year^state_id,
              weights = dare1$enroll)
## NOTE: 46 observations removed because of NA values (LHS: 46, Weights: 46).
summary(mod_subj_time)
## OLS estimation, Dep. Var.: ODR_subjective
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (school_year^state_id)
       Estimate Std. Error t value Pr(>|t|)
## eval 0.03298 0.029894 1.10321 0.2705
```

Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1

Within R2: 0.002023

RMSE: 29.3 Adj. R2: 0.791287

##

Table 2: Table X. Effects of Teacher Eval Reforms on Discipline Referrals

	1	2	3	4
eval	0.036	0.036	0.033	0.033
	(0.061)	(0.045)	(0.044)	(0.030)
Num.Obs.	470	470	470	470
R2	0.825	0.825	0.815	0.815

Notes: 1 - Class, Constant Effects; 2 - Class, Time Effects, 3 - Subj. Constant Effects, 4 - Subj, Time Effects

Warning: In version 0.8.0 of the 'modelsummary' package, the default significance markers produced be ## This warning is displayed once per session.

Differences between the values in the table above and the paper can be come from controls not being accounted for in this model. Since the authors of the paper mention preferring a simpler model, a simpler model was used here. Secondly, these results are at the state-year level, which is a different sample than at the school-year level. As a result, differences are expected and can be remedied with more data (or a more advanced analysis).

C2. Liebowitz et al. (2022) conduct a broad set of robustness checks. For this DARE assignment, you will conduct two (2). First test whether the main results you present in Question C1 are robust to the introduction of potentially simultaneous discipline policy reforms. Present the table and associated write-up as you would report these results in an academic paper. Then select an additional robustness check (either from the paper or not) and present evidence on whether your findings are sensitive to this test.

The first set of robustness check is test the effect of evaluation policy implementation on rates on suspension from the Civil Right Data Collection (B11 - B13 of Figure 4)

For ODR Class

```
weights = dare1_clean$enroll)
summary(rc_b11)
## OLS estimation, Dep. Var.: ODR_class
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (state_id)
              Estimate Std. Error t value Pr(>|t|)
## suspension -0.069408
                       0.083888 -0.827386 0.41269
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## RMSE: 41.9
                 Adj. R2: 0.802307
               Within R2: 0.0039
##
#Robustness check with CRDC and Controls (B12)
rc_b12 <- feols(ODR_class ~ suspension + FRPL_percent + enroll_OTHER + enroll_AM + enroll_AM + enroll_H
                       state_id + school_year,
                     data = dare1_clean,
                     weights = dare1_clean$enroll)
summary(rc_b12)
## OLS estimation, Dep. Var.: ODR_class
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (state_id)
##
               Estimate Std. Error t value Pr(>|t|)
## suspension -0.053887 0.084717 -0.636092 0.5281680
## FRPL_percent -0.027840 0.395350 -0.070418 0.9441949
## enroll_OTHER -0.005647 0.021916 -0.257644 0.7979399
## enroll_AM
               0.016950 0.024893 0.680926 0.4996557
## enroll_HISP -0.000645 0.007778 -0.082870 0.9343485
## enroll_BLACK 0.030102 0.010863 2.770956 0.0082916 **
## enroll_WHITE 0.005572 0.004948 1.125922 0.2665912
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## RMSE: 40.6
                 Adj. R2: 0.8105
               Within R2: 0.061285
##
#Robustness check with CRDC, controls, and Time (B13)
rc_b13 <- feols(ODR_class ~ suspension + FRPL_percent + enroll_OTHER + enroll_AM + enroll_AM + enroll_H
                       state_id + school_year,
                     data = dare1_clean,
                     weights = dare1_clean$enroll)
summary(rc_b13)
## OLS estimation, Dep. Var.: ODR_class
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
```

Standard-errors: Clustered (state_id)

```
##
              Estimate Std. Error t value Pr(>|t|)
## suspension -0.049651 0.087273 -0.568921 0.5724397
## FRPL percent 0.017980 0.388742 0.046251 0.9633297
0.012607 0.028456 0.443042 0.6600089
## enroll_AM
## enroll_HISP -0.003277 0.010160 -0.322551 0.7486359
## enroll_BLACK 0.029763 0.010876 2.736620 0.0090596 **
## enroll_WHITE 0.005682 0.004939 1.150313 0.2565225
## run_time
            -0.015746 0.030840 -0.510578 0.6123197
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## RMSE: 40.6
               Adj. R2: 0.810682
              Within R2: 0.064489
##
For ODRs Subjective
#Robustness check with CRDC (B11)
rc_b11s <- feols(ODR_subjective ~ suspension |</pre>
            state_id + school_year, #default clustering on state id
            data = dare1_clean,
            weights = dare1_clean$enroll)
summary(rc_b11s)
## OLS estimation, Dep. Var.: ODR_subjective
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (state id)
            Estimate Std. Error t value Pr(>|t|)
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## RMSE: 29.3
               Adj. R2: 0.790883
##
              Within R2: 9.199e-5
#Robustness check with CRDC and Controls (B12)
rc_b12s <- feols(ODR_subjective ~ suspension + FRPL_percent + enroll_OTHER + enroll_AM + enroll_AM + en
                     state_id + school_year,
                   data = dare1_clean,
                   weights = dare1_clean$enroll)
summary(rc_b12s)
## OLS estimation, Dep. Var.: ODR_subjective
## Observations: 470
```

Estimate Std. Error t value Pr(>|t|) 0.014078 0.063985 0.220019 0.8269224

0.005387 0.611401 0.5442294

Fixed-effects: state_id: 43, school_year: 12

Standard-errors: Clustered (state id)

0.003294

##

suspension

enroll_AM

enroll_HISP

```
## enroll_BLACK 0.023673 0.008287 2.856477 0.0066334 **
## enroll WHITE 0.005224 0.003964 1.317919 0.1946740
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## RMSE: 28.2
              Adj. R2: 0.803454
            Within R2: 0.076053
#Robustness check with CRDC, controls, and Time (B13)
rc_b13s <- feols(ODR_subjective ~ suspension + FRPL_percent + enroll_OTHER + enroll_AM + enroll_AM + en
                   state_id + school_year,
                 data = dare1_clean,
                 weights = dare1_clean$enroll)
summary(rc_b13s)
## OLS estimation, Dep. Var.: ODR_subjective
## Observations: 470
## Fixed-effects: state_id: 43, school_year: 12
## Standard-errors: Clustered (state_id)
             Estimate Std. Error t value Pr(>|t|)
             ## suspension
## FRPL percent 0.034728 0.245331 0.141557 0.8881069
## enroll_AM
             0.023820 0.015699 1.517306 0.1366811
             ## enroll_HISP
## enroll_BLACK 0.023356 0.008332 2.803175 0.0076263 **
## enroll_WHITE 0.005326 0.003931 1.354890 0.1826960
## run_time
                     0.021005 -0.697938 0.4890622
            -0.014660
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## RMSE: 28.1
              Adj. R2: 0.804183
##
            Within R2: 0.081735
```

Another Robustness check mentioned in the paper is to use ODRs from locations other than the classroom and ODRs for behavioral infraction that involved objective reasons to send students to the office (B1 - B6)

OLS estimation, Dep. Var.: ODR_other Observations: 470 Fixed-effects: state_id: 43, school_year: 12 Standard-errors: Clustered (state_id) Estimate Std. Error t value $\Pr(>|t|)$ eval -0.04855 0.037355 -1.29969 0.2008 — Signif. codes: 0 '' 0.001 '' 0.01 " 0.05 '' 0.1 '' 1 RMSE: 26.8 Adj. R2: 0.896358 Within R2: 0.005222

```
#Other and controls (B2)
rc_b2 <- feols(ODR_other ~ eval + FRPL_percent + enroll_OTHER + enroll_AM + enroll_AM + enroll_HISP + enroll_AM + enroll
```

```
state_id + school_year,
                        data = dare1_clean,
                        weights = dare1 clean$enroll)
summary(rc b2)
OLS estimation, Dep. Var.: ODR_other Observations: 470 Fixed-effects: state_id: 43, school_year: 12
Standard-errors: Clustered (state_id) Estimate Std. Error t value Pr(>|t|)
eval -0.038656 0.046233 -0.836121 0.407819
FRPL percent 0.061635 0.297257 0.207347 0.836741
enroll_OTHER -0.017391 0.021072 -0.825316 0.413856
enroll AM 0.002127\ 0.012003\ 0.177202\ 0.860202
enroll_HISP 0.003719 0.005812 0.639876 0.525728
enroll ASIAN -0.008396 0.021224 -0.395569 0.694425
enroll BLACK 0.016501 0.006219 2.653176 0.011208 ^{*}
enroll WHITE 0.004592 0.003668 1.251791 0.217573
— Signif. codes: 0 ' ' 0.001 '' 0.01 '' 0.05 ': 0.1 ' ' 1 RMSE: 26.4 Adj. R2: 0.89794 Within R2: 0.036929
#Other, controls, and time (B3)
rc_b3 <- rc_b2 <- feols(ODR_other ~ eval + FRPL_percent + enroll_OTHER + enroll_AM + enroll_AM + enroll_AM
                          state_id + school_year,
                        data = dare1 clean,
                        weights = dare1_clean$enroll)
summary(rc_b3)
OLS estimation, Dep. Var.: ODR other Observations: 470 Fixed-effects: state id: 43, school year: 12
Standard-errors: Clustered (state_id) Estimate Std. Error t value Pr(>|t|)
eval -0.046307 0.052382 -0.884024 0.381716
FRPL_percent 0.071396\ 0.303894\ 0.234936\ 0.815400
enroll OTHER -0.018051 0.021328 -0.846318 0.402172
enroll AM 0.001342\ 0.012364\ 0.108553\ 0.914074
enroll_HISP 0.003213\ 0.006392\ 0.502673\ 0.617820
enroll ASIAN -0.007552 0.022370 -0.337595 0.737350
enroll BLACK 0.016308 0.006515 2.503110 0.016286 *
enroll WHITE 0.004693 0.003711 1.264728 0.212942
run time -0.003883 0.016017 -0.242402 0.809649
— Signif. codes: 0 '' 0.001 '' 0.001 '' 0.05 '' 0.1 '' 1 RMSE: 26.4 Adj. R2: 0.897727 Within R2: 0.037286
#Objective (B4)
rc_b4 <- feols(ODR_objective ~ eval|
               state_id + school_year, #default clustering on state id
               data = dare1_clean,
               weights = dare1_clean$enroll)
summary(rc_b4)
OLS estimation, Dep. Var.: ODR objective Observations: 470 Fixed-effects: state id: 43, school year:
12 Standard-errors: Clustered (state_id) Estimate Std. Error t value Pr(>|t|) eval -0.010117 0.016461 -
0.614613 0.54213 — Signif. codes: 0 '' 0.001 "' 0.01 " 0.05 '. 0.1 ' ' 1 RMSE: 10.7 Adj. R2: 0.888442
Within R2: 0.00143
```

state_id + school_year,

rc b5 <- feols(ODR objective ~ eval + FRPL percent + enroll OTHER + enroll AM + enroll AM + enroll HISP

#Objective and controls (B5)

```
data = dare1_clean,
                       weights = dare1_clean$enroll)
summary(rc b5)
OLS estimation, Dep. Var.: ODR objective Observations: 470 Fixed-effects: state id: 43, school year: 12
Standard-errors: Clustered (state_id) Estimate Std. Error t value Pr(>|t|)
FRPL percent 0.036497 0.119482 0.305459 0.761526
enroll_OTHER -0.001820 0.006672 -0.272762 0.786373
enroll AM -0.002449 0.006751 -0.362800 0.718574
enroll HISP 0.000679 0.002207 0.307692 0.759837
enroll_ASIAN -0.001799 0.007370 -0.244080 0.808357
enroll BLACK 0.005165 0.001944 2.656704 0.011108 *
enroll WHITE 0.001255 0.001312 0.956997 0.344045
— Signif. codes: 0 '' 0.001 '' 0.01 '' 0.05 ': 0.1 '' 1 RMSE: 10.6 Adj. R2: 0.888813 Within R2: 0.02154
#Objective, controls, and time (B6)
rc_b6 <- rc_b2 <- feols(ODR_objective ~ eval + FRPL_percent + enroll_OTHER + enroll_AM + enroll_AM + en
                         state_id + school_year,
                       data = dare1_clean,
                       weights = dare1_clean$enroll)
summary(rc_b6)
OLS estimation, Dep. Var.: ODR_objective Observations: 470 Fixed-effects: state_id: 43, school_year: 12
Standard-errors: Clustered (state id) Estimate Std. Error t value Pr(>|t|)
FRPL\_percent\ 0.034462\ 0.123933\ 0.278070\ 0.782323
enroll OTHER -0.001682 0.006557 -0.256581 0.798755
enroll AM -0.002286 0.006769 -0.337687 0.737281
enroll HISP 0.000784 0.002453 0.319777 0.750723
enroll ASIAN -0.001975 0.007935 -0.248855 0.804686
enroll BLACK 0.005205 0.002015 2.583252 0.013359 *
enroll_WHITE 0.001234\ 0.001304\ 0.946253\ 0.349432
run time 0.000809 \ 0.005512 \ 0.146841 \ 0.883960
— Signif. codes: 0 '' 0.001 '' 0.01 '' 0.05 '' 0.1 '' 1 RMSE: 10.6 Adj. R2: 0.888551 Within R2: 0.021638
compare_performance(rc_b11, rc_b12,rc_b13, rc_b11s, rc_b12s, rc_b13s, rc_b1, rc_b2, rc_b3, rc_b4, rc_b5
  print md()
```

Warning: When comparing models, please note that probably not all models were fit from ## same data.

Table 3: Comparison of Model Performance Indices

Name	Model	R2	R2 (adj.)	RMSE	Sigma	AIC weights	BIC weights	Performance-Score
rc_b4	fixest	0.90	0.89	0.19	11.38	0.996	1.000	96.81%
rc_b5	fixest	0.90	0.89	0.19	11.36	0.002	< 0.001	63.94%
rc_b2	fixest	0.90	0.89	0.19	11.38	< 0.001	< 0.001	63.86%
rc_b6	fixest	0.90	0.89	0.19	11.38	< 0.001	< 0.001	63.86%
rc_b3	fixest	0.91	0.90	0.48	28.34	< 0.001	< 0.001	49.66%
rc b1	fixest	0.91	0.90	0.48	28.53	< 0.001	< 0.001	48.69%

Name	Model	R2	R2 (adj.)	RMSE	Sigma	AIC weights	BIC weights	Performance-Score
rc_b13s	fixest	0.83	0.80	0.54	30.21	< 0.001	< 0.001	18.18%
rc_b12s	fixest	0.83	0.80	0.54	30.26	< 0.001	< 0.001	17.82%
rc_b11s	fixest	0.81	0.79	0.54	31.21	< 0.001	< 0.001	12.87%
rc_b13	fixest	0.84	0.81	0.75	43.61	< 0.001	< 0.001	7.21%
rc_b12	fixest	0.84	0.81	0.75	43.63	< 0.001	< 0.001	7.06%
rc_b11	fixest	0.83	0.80	0.75	44.56	< 0.001	< 0.001	3.53%

C3. Write a discussion paragraph in which you present the substantive conclusions of your results about the effects of the introduction of higher-stakes teacher evaluation on ODRs.

According to this analyses, there is insufficient evidence to suggest that relationship between the introduction of higher-stakes teacher evaluation reforms on Office Disciplinary Referrals is significant.

C4. Optional Extension Use an event-study approach to this difference-in-differences research design to estimate the effects of the introduction of higher-stakes teacher evaluation reforms on Office Disciplinary Referrals (ODRs). Present these findings in an event-study graph. Present the figure and associated write-up as you would report these results in an academic paper. Do you notice any important differences in these results and those reported in the original paper? If so, how would you consider addressing them (At this point, it is not necessary for you to actually conduct the analysis. Just describe approaches you might take.)?

C5. Optional Extension Use one (or more) approaches to present the extent to which the successful implementation of Positive Behavioral Intervention and Supports (PBIS) framework moderating the effects of the introduction of higher-stakes teacher evaluation policies. Present these difference-in-differences estimates and associated write-up as you would report these results in an academic paper. Do you notice any important differences in these results and those reported in the original paper? If so, how would you consider addressing them?