## **COVIDSchoolClosures\***

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#### **Abstract**

COVID-19 was an outbreak of virus that forced many instituations to shut down for 2-3 years. Schools were no different and this paper aims to look at the effects of the said closures in school and how it affected the population. This paper finds that with more inperson schooling provided the less the enrollment rates drop(more in depth analysis in later sections).

### Introduction

### **Data**

### Data Source and Collection //Methodology

We use R Core Team (2022) to make this paper as well as the graphs and topic were taken inspiration from Jack and Oster (2023). Various helpful packages were used in order to clean, sort and graph this paper in a way

<sup>\*</sup>Code and some data from this paper are available at: https://github.com/NotSakura/COVIDSchoolClosures.git

such that the reader will not have difficulty undertsnading neither the topic nor the data sets of this paper. The packages are, Wickham (2016), Wickham et al. (2019), Wickham et al. (2023), Wickham, Hester, and Bryan (2023), Xie (2014), Firke (2023), Zhu (2021), Wickham, Vaughan, and Girlich (2024), Wickham and Miller (2021), Hyndman and O'Hara-Wild (2021).

#### **Data Cleaning**

What data set did we clean and why. Explain the variable here too ig

### Results

The graphs we made and describe the trends. Only talk about results not what they mean

#### First Graph

##Second Graph

#### **Third Graph**

```
Rows: 14967 Columns: 7
-- Column specification -------
Delimiter: ","
chr (3): StateAbbrev, DistrictName, StateAssignedDistrictID
dbl (4): NCESDistrictID, share_inperson, share_hybrid, share_virtual

i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.
```

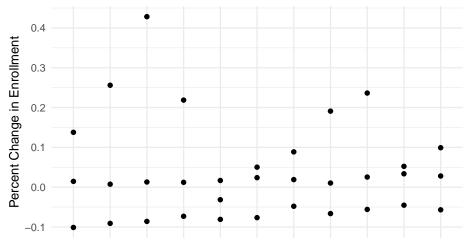
```
data <- raw_d |>
  filter(grade == 0 | grade == 99) |>
  filter(race == 99)
data <- subset(data, select = -race)</pre>
\# Drop observations where enrollment is less than 0
data <- data[data$enrollment >= 0, ]
# Drop observations where fips is greater than 56
data <- data[data$fips <= 56, ]</pre>
data <- data %>%
  filter(!is.na(sex))
# Reshape 'enrollment' wide by 'leaid', 'year', and 'grade'
data_wide <- pivot_wider(data,</pre>
                             names_from = sex,
                             values_from = enrollment) |>
 rename(
    enrollment1 = `1`,
    enrollment2 = `2`,
    enrollment9 = `9`,
    enrollment99 = `99`)
# Drop the 'enrollment9' variable
data_wide$enrollment9 <- NULL
\# Drop observations where 'enrollment99' is missing or less than or equal to 0
data_wide <- data_wide[!is.na(data_wide$enrollment99) & data_wide$enrollment99 > 0
```

```
# Reshape 'enrollment1', 'enrollment2', and 'enrollment99' wide by 'leaid' and 'ye
data_final <- pivot_wider(data_wide,</pre>
                             names_from = grade,
                             values_from = c(enrollment1, enrollment2, enrollment99
# Can get rid of fips
data_final <- subset(data_final, select = -fips)</pre>
# Rename columns
colnames(data_final) <- c("year", "leaid",</pre>
                               "enroll_male_kinder", "enroll_female_kinder",
                               "enroll_male_total", "enroll_female_total",
                               "enroll_all_kinder", "enroll_all_total")
# Convert 'leaid' to numeric (assuming it's a string)
data$leaid <- as.numeric(data$leaid)</pre>
other_raw$leaid <- as.numeric(other_raw$leaid)</pre>
write_csv(x = data_final,
          file = "../data/cleaned_raw_data.csv")
# Link data with another dataset (replace 'other_data' with your actual dataset)
# Assuming 'leaid' is the common identifier
merged_data <- merge(data_final, other_raw, by = "leaid")</pre>
write_csv(x = merged_data,
          file = "../data/merged_data.csv")
# Assuming 'year' is numeric (e.g., 2019, 2020)
merged_data$date <- as.Date(paste0(merged_data$year, "-01-01"))</pre>
```

```
my_tsibble <- as_tsibble(merged_data, key = leaid, index = date)</pre>
# Assuming you have a tsibble called my_tsibble with appropriate columns
# Calculate change_kinder
my_tsibble <- my_tsibble %>%
  mutate(change_kinder = enroll_all_kinder - lag(enroll_all_kinder),
         perc_change_kinder = change_kinder / lag(enroll_all_kinder))
# Calculate change_all
my_tsibble <- my_tsibble %>%
  mutate(change_all = enroll_all_total - lag(enroll_all_total),
         perc_change_all = change_all / lag(enroll_all_total))
# Calculate change_male
my_tsibble <- my_tsibble %>%
  mutate(change male = enroll male_kinder - lag(enroll male_kinder),
         perc_change_male = change_male / lag(enroll_male_kinder))
# Calculate change_female
my_tsibble <- my_tsibble %>%
  mutate(change_female = enroll_female_kinder - lag(enroll_female_kinder),
         perc_change_female = change_female / lag(enroll_female_kinder))
# Drop rows where absolute change_all is greater than enroll_all_total
my_tsibble <- my_tsibble %>%
  filter(abs(change_all) <= enroll_all_total)</pre>
# Drop rows where absolute change kinder is greater than enroll all total
```

```
my_tsibble <- my_tsibble %>%
  filter(abs(change_kinder) <= enroll_all_total)</pre>
my_tsibble <- my_tsibble %>%
  mutate(inperson_bin = cut(share_inperson, breaks = 11))
# Compute weighted mean change in enrollment for each bin
weighted_means_all <- my_tsibble %>%
  group_by(inperson_bin) %>%
  summarise(
    weighted_mean_change = weighted.mean(perc_change_all, enroll_all_total)
weighted_means_all <- subset(weighted_means_all, select = -date)</pre>
weighted_means_kinder <- my_tsibble %>%
  group_by(inperson_bin) %>%
  summarise(
    weighted_mean_change = weighted.mean(perc_change_kinder, enroll_all_kinder)
weighted_means_kinder <- subset(weighted_means_kinder, select = -date)</pre>
##Graph
ggplot(weighted_means_all, aes(x = inperson_bin , y = weighted_mean_change)) +
  geom_point() +
  labs(
    title = "Percent Change in Overall District Enrollment, Fall 2019-Fall 2020",
    x = "Share of School Year with In-Person Learning Offered, 2020-2021",
    y = "Percent Change in Enrollment"
  ) +
  theme_minimal()
```

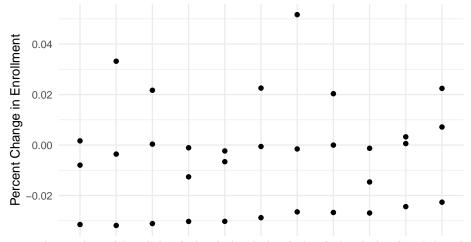
### Percent Change in Overall District Enrollment, Fall 2019-Fall 2



(-0.001,000999,001822,022373,02664,04955,05495,06866,07227,088818,0.909909,1] Share of School Year with In-Person Learning Offered, 2020–2021

```
ggplot(weighted_means_kinder, aes(x = inperson_bin, y = weighted_mean_change)) +
  geom_point() +
  labs(
    title = "Percent Change in Kindergarten Enrollment, Fall 2019-Fall 2020",
    x = "Share of School Year with In-Person Learning Offered, 2020-2021",
    y = "Percent Change in Enrollment"
  ) +
  theme_minimal()
```





 $(-0.001, 0.0039324, 0.01822, 0.02233, 0.03644, 0.04355, 0.05435, 0.05666, 0.07227, 0.08833, 0.902909, 1]\\ Share of School Year with In-Person Learning Offered, 2020–2021$ 

## Discussion

Interesting point 1

Intresting point 2

### interesting point 3

##Ethics and Bias could talk about mental health maybe but it might apply to other "interesting point"

### weakness and limitations

### how to solve the limitations

## Furthur questions?

# **Appendix**

```
weighted_means_all |> kable(format = "pipe", padding = 2, col.names = c("Share of
```

Share of inperson	average enrollment rate
(-0.001,0.0909]	0.1377654
(-0.001, 0.0909]	0.0145865
(-0.001, 0.0909]	-0.1012197
(0.0909, 0.182]	0.2560526
(0.0909, 0.182]	0.0075213
(0.0909, 0.182]	-0.0909043
(0.182, 0.273]	0.4282981
(0.182, 0.273]	0.0130138
(0.182, 0.273]	-0.0859602
(0.273, 0.364]	0.2186839
(0.273, 0.364]	0.0121407
(0.273, 0.364]	-0.0731328
(0.364, 0.455]	-0.0315128
(0.364, 0.455]	0.0168102
(0.364, 0.455]	-0.0808996
(0.455, 0.545]	0.0502807
(0.455, 0.545]	0.0238917
(0.455, 0.545]	-0.0765528
(0.545, 0.636]	0.0886407
(0.545, 0.636]	0.0190350

Share of inperson	average enrollment rate
(0.545, 0.636]	-0.0479435
(0.636, 0.727]	0.1909656
(0.636, 0.727]	0.0103950
(0.636, 0.727]	-0.0663055
(0.727, 0.818]	0.2364083
(0.727, 0.818]	0.0253973
(0.727, 0.818]	-0.0558533
(0.818, 0.909]	0.0523212
(0.818, 0.909]	0.0336325
(0.818, 0.909]	-0.0452392
(0.909,1]	0.0991443
(0.909,1]	0.0280538
(0.909,1]	-0.0568407

### Reference

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