

# Gun Series

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## Base Panel Construction - ZCTA-Week Level

### Hospital Data - ZCTA-Week level

```
hosp_zcta <- read_csv("Restricted Hospital Data/minnepop_1620_agg_zipfull_updated.csv") %>%  
  arrange(zipcode, year, weekofyr) %>%  
  select(-c(`_chk`, zippop_tag)) %>%  
  filter(!(year==2016 & weekofyr==53))
```

## ZCTAs and ACS 5-Year Estimates

```
#adding in 5-year ACS data  
census_api_key("ecda17575f4d914b502c70f2bae7a5f3d253792d")  
  
year <- lst(2016, 2017, 2018, 2019)  
  
acs <- map_dfr(  
  year,  
  ~ get_acs(geography = "zcta",  
            variables = c("B01001_001E", "B03003_003E",  
                          "B02001_003E", "B02001_002E",  
                          "B02001_004E", "B02001_008E",  
                          "B02001_005E", "B02001_006E",  
                          "B02001_007E", "B11001_003E",  
                          "B17001_002E", "B01002_001E",  
                          "B09010_002E", "B06009_005E",  
                          "B01001_002E", "B99233_005E"),  
            output = "wide",  
            survey = "acs5",  
            year = .x), .id = "year") %>%  
  rename(total_pop = B01001_001E,  
         white_pop = B02001_002E,  
         black_pop = B02001_003E,  
         na_pop = B02001_004E,  
         asian_pop = B02001_005E,  
         hpi_pop = B02001_006E,  
         other_pop = B02001_007E,
```

```

    biracial_pop = B02001_008E,
    hisp_pop = B03003_003E,
    ssi_snap = B09010_002E, #snap, ssi, public cash transfers
    med_age = B01002_001E,
    mar_fam = B11001_003E,
    povlevel = B17001_002E,
    bach_degree = B06009_005E,
    male = B01001_002E,
    nowork_12 = B99233_005E) %>%
select(-ends_with("M", ignore.case = F), -GEOID) %>%
mutate(zcta = str_sub(NAME, 6)) %>%
select(-NAME) %>%
select(zcta, everything()) %>%
mutate(year = as.numeric(year)) %>%
mutate_at(vars(-zcta, -year, -total_pop, -med_age), list(~(./total_pop)*100))

#LOCF imputation of 2020 until 2020 ACS release (12/9/2021)
acs_2020 <- acs %>%
  complete(zcta, year = 2016:2020) %>%
  group_by(zcta) %>%
  mutate_at(vars(-zcta, -year),
    funs(if(sum(!is.na(.))<1) {.} else{na_locf(., option = "locf")}))) %>%
  filter(year==2020)

acs_imp <- acs %>%
  rbind(acs_2020) %>%
  mutate(zcta = as.numeric(zcta))

#joining to hospital data
hosp_panel <- hosp_zcta %>%
  left_join(acs_imp, by = c("zipcode"="zcta", "year"))

#SF geometries - get all ZCTAs
zcta <- get_acs(geography = "zcta",
  variables = "B01001_001",
  output = "wide",
  year = 2019,
  geometry = T,
  survey = "acs5") %>%
  rename(zcta = GEOID,
    pop_2019 = B01001_001E) %>%
  select(-c(NAME, B01001_001M, pop_2019)) %>%
  mutate(zcta = as.numeric(zcta))

```

```
## |
```

```

#minneapolis shapefile (source: openminneapolis.gov)
mpls <- st_read("Data/mpls_city-shp/16cdbbfa-ad10-493c-afaf-52b61f2e76e42020329-1-180h9ap.whbo.shp") %>%
  st_set_crs(st_crs(zcta))

```

```

## Reading layer '16cdbbfa-ad10-493c-afaf-52b61f2e76e42020329-1-180h9ap.whbo' from data source 'C:\User
## using driver 'ESRI Shapefile'
## Simple feature collection with 1 feature and 4 fields

```

```
## Geometry type: POLYGON
## Dimension:      XY
## Bounding box:   xmin: -93.32911 ymin: 44.89059 xmax: -93.19433 ymax: 45.05125
## Geodetic CRS:   WGS 84
```

```
#zctas that intersect MPLS
zcta_intersect <- zcta %>%
  filter(ifelse(lengths(st_intersects(., mpls)) > 0, 1, 0)==1) %>%
  select(zcta)

#which zctas are not in hosp data but still intersect MPLS
setdiff(unique(zcta_intersect$zcta), unique(hosp_zcta$zipcode))
```

```
## [1] 55114 55105 55104 55113 55116 55111 55108
```

```
#joining to panel, filter to those ZCTAs intersecting MPLS
panel <- zcta %>%
  left_join(hosp_panel, by = c("zcta"="zipcode")) %>%
  filter(ifelse(lengths(st_intersects(., mpls)) > 0, 1, 0)==1 &
    zcta >= 55401) #queen contiguity

#creating date bookends
panel <- panel %>%
  group_by(zcta, year) %>%
  mutate(begin_date = ISOweek2date(paste(year, paste0("W", sprintf("%02d", weekofyr)), 1, sep = "-")),
    end_date = begin_date+weeks(1)-days(1))

#number of unique MPLS ZCTAs
n_zcta <- length(unique(panel$zcta))

#vector of intersecting ZCTAs for filtering downstream
zcta_universe <- unique(panel$zcta)
```

## ZCTA-Week Level Police Data

```
#Minneapolis Police Department - Use of Force Dashboard
uof_spatial <- read_csv("Data/Police_Use_Of_Force.csv") %>%
  mutate(date=ymd_hms(ResponseDate),
    year=isoyear(date),
    week=isoweek(date)) %>%
  select(OBJECTID, year, week, X, Y, Race) %>%
  st_as_sf(coords = c("X", "Y"), crs = "NAD83", remove=F) %>%
  mutate(intersection = as.integer(st_intersects(geometry, zcta)),
    zcta = ifelse(is.na(intersection), NA, zcta$zcta[intersection])) %>%
  st_drop_geometry() %>%
  filter(!is.na(zcta) & year >= 2016 & year <= 2020 & zcta %in% zcta_universe) %>%
  group_by(year, week, zcta, Race, .drop=F) %>%
  tally(name = "use_of_force") %>%
  filter(!is.na(Race) & Race!="not recorded") %>%
  ungroup() %>%
  complete(year, week, zcta=zcta_universe, Race, fill = list(use_of_force = 0)) %>%
```

```

arrange(year, week, zcta, Race) %>%
mutate(race = str_to_lower(Race)) %>%
select(-Race) %>%
pivot_wider(names_from = race,
             values_from = use_of_force,
             values_fill = 0,
             names_glue = "{race}_{.value}") %>%
mutate(total_use_of_force = asian_use_of_force+black_use_of_force+`native american_use_of_force`+
`other / mixed race_use_of_force`+`pacific islander_use_of_force`+unknown_use_of_force+
white_use_of_force)

#MPD Stop Dashboard
stop_spatial <- read_csv("Data/Police_Stop_Data.csv") %>%
  mutate(date=ymd_hms(responseDate),
         year=isoyear(date),
         week=isoweek(date)) %>%
  select(OBJECTID, year, week, lat, long, race) %>%
  st_as_sf(coords = c("long", "lat"), crs = "NAD83", remove=F) %>%
  mutate(intersection = as.integer(st_intersects(geometry, zcta)),
         zcta = ifelse(is.na(intersection), NA, zcta$zcta[intersection])) %>%
  st_drop_geometry() %>%
  filter(!is.na(zcta) & year >= 2016 & year <= 2020 & zcta %in% zcta_universe) %>%
  group_by(year, week, zcta, race, .drop=F) %>%
  tally(name = "police_stops") %>%
  filter(!is.na(race) & race!="not recorded") %>%
  ungroup() %>%
  complete(year, week, zcta=zcta_universe, race, fill = list(police_stops = 0)) %>%
  mutate(race = str_to_lower(race)) %>%
  arrange(year, week, zcta, race) %>%
  pivot_wider(names_from = race,
             values_from = police_stops,
             values_fill = 0,
             names_glue = "{race}_{.value}") %>%
  mutate(total_police_stops = asian_police_stops+black_police_stops+
`east african_police_stops`+latino_police_stops+`native american_police_stops`+
other_police_stops+unknown_police_stops+white_police_stops)

#Officer Involved Shootings - MPD
ois_spatial <- read_csv("Data/Police_Officer_Involved_Shootings.csv") %>%
  mutate(date=ymd_hms(IncidentDate),
         year=isoyear(date),
         week=isoweek(date)) %>%
  select(OBJECTID, year, week, CenterLatitude, CenterLongitude, SubjectOfForceRace) %>%
  rename(race = SubjectOfForceRace,
         lat = CenterLatitude,
         long = CenterLongitude) %>%
  st_as_sf(coords = c("long", "lat"), crs = "NAD83", remove=F) %>%
  mutate(intersection = as.integer(st_intersects(geometry, zcta)),
         zcta = ifelse(is.na(intersection), NA, zcta$zcta[intersection])) %>%
  st_drop_geometry() %>%
  filter(!is.na(zcta) & year >= 2016 & year <= 2020 & zcta %in% zcta_universe) %>%
  group_by(year, week, zcta, race, .drop=F) %>%
  tally(name = "police_shootings") %>%

```

```

filter(!is.na(race) & race!="not recorded") %>%
ungroup() %>%
complete(year=2016:2021, week=1:53, zcta=zcta_universe, race, fill = list(police_shootings = 0)) %>%
mutate(race = str_to_lower(race)) %>%
arrange(year, week, zcta, race) %>%
pivot_wider(names_from = race,
             values_from = police_shootings,
             values_fill = 0,
             names_glue = "{race}_{.value}") %>%
mutate(total_police_shootings = asian_police_shootings+black_police_shootings+
       hispanic_police_shootings+other_police_shootings+
       unknown_police_shootings+white_police_shootings)

panel <- panel %>%
  left_join(uof_spatial, by = c("year", "weekofyr"="week", "zcta"="zcta")) %>%
  left_join(stop_spatial, by = c("year", "weekofyr"="week", "zcta"="zcta")) %>%
  left_join(ois_spatial, by = c("year", "weekofyr"="week", "zcta"="zcta"))

#creating period indicators for panel
panel <- panel %>%
  mutate(post_floyd = as.numeric(begin_date >= as.Date("2020-05-25")),
         post_floyd_3 = as.numeric(begin_date >= as.Date("2020-05-25")+months(3)),
         stay_at_home = as.numeric(begin_date >= as.Date("2020-03-28")) &
         state_of_emerg = as.numeric(begin_date >= as.Date("2020-03-13")),
         period = factor(case_when(
           post_floyd==0 & post_floyd_3==0 ~ "Pre-Killing",
           post_floyd>=1 & post_floyd_3==0 ~ "0-3 Months Post-Killing",
           post_floyd>=1 & post_floyd_3>=1 ~ "3+ Months Post-Killing",
           levels = c("Pre-Killing", "0-3 Months Post-Killing", "3+ Months Post-Killing")))) %>%
  group_by(zcta) %>%
  arrange(year, weekofyr) %>%
  mutate(t = row_number(),
         uof_lag = dplyr::lag(total_use_of_force, 1),
         stops_lag = dplyr::lag(total_police_stops, 1),
         shoot_lag = dplyr::lag(total_police_shootings, 1))

```

## Time Series Construction - Week Level

### Aggregate Hospital Panel to Week-Level

```

#panel to week-level, aggregating over ZCTAs
hosp_series <- panel %>%
  group_by(year, weekofyr) %>%
  summarize(assault_tot = sum(assault_tot, na.rm = T),
            unintent_tot = sum(unintent_tot, na.rm = T),
            suicide_tot = sum(suicide_tot, na.rm = T),
            undeter_tot = sum(undeter_tot, na.rm = T),
            legal_tot = sum(legal_tot, na.rm = T),
            combined_tot = sum(combined_tot, na.rm = T),
            total_pop = sum(total_pop, na.rm = T)) %>%

```

```

mutate(assault_incid_c = (assault_tot/total_pop)*1000,
       unintent_incid_c = (unintent_tot/total_pop)*1000,
       suicide_incid_c = (suicide_tot/total_pop)*1000,
       undeter_incid_c = (undeter_tot/total_pop)*1000,
       legal_incid_c = (legal_tot/total_pop)*1000,
       combined_incid_c = (combined_tot/total_pop)*1000) %>%
ungroup() %>%
mutate(week_id = row_number()) %>%
st_drop_geometry()

```

## Police Data Week-Level

```

#Minneapolis Police Department - Use of Force Dashboard
uof <- read_csv("Data/Police_Use_Of_Force.csv") %>%
  mutate(date=ymd_hms(ResponseDate),
         year=isoyear(date),
         week=isoweek(date)) %>%
  group_by(year, week, .drop=F) %>%
  tally(name = "use_of_force") %>%
  arrange(year, week) %>%
  ungroup() %>%
  select(year, week, everything())

#merge onto series
series <- hosp_series %>%
  left_join(uof, by=c("year", "weekofyr"="week")) %>%
  mutate(use_of_force_rate = (use_of_force/total_pop)*1000)

#MPD Officer Involved Shootings
ois <- read_csv("Data/Police_Officer_Involved_Shootings.csv") %>%
  mutate(date=ymd_hms(IncidentDate),
         year=isoyear(date),
         week=isoweek(date)) %>%
  group_by(year, week, .drop=F) %>%
  tally(name = "off_inv_shooting") %>%
  arrange(year, week) %>%
  ungroup() %>%
  select(year, week, everything())

#merge onto series
series <- series %>%
  left_join(ois, by=c("year", "weekofyr"="week")) %>%
  mutate(off_inv_shooting = ifelse(is.na(off_inv_shooting), 0, off_inv_shooting),
         off_inv_shooting_rate = (off_inv_shooting/total_pop)*1000)

#Minneapolis Police Department - Police Stops Dashboard
stop <- read_csv("Data/Police_Stop_Data.csv") %>%
  mutate(date=ymd_hms(responseDate),
         year=isoyear(date),
         week=isoweek(date)) %>%

```

```

group_by(year, week, .drop=F) %>%
tally(name = "police_stops")

#merge onto series
series <- series %>%
  left_join(stop, by = c("year", "weekofyr"="week")) %>%
  mutate(police_stops = ifelse(is.na(police_stops), 0, police_stops),
         police_stop_rate = (police_stops/total_pop)*1000)

#creating date variable
series <- series %>%
  mutate(begin_date = ISOweek2date(paste(year, paste0("W", sprintf("%02d", weekofyr)), 1, sep = "-")),
         end_date = begin_date+weeks(1)-days(1))

```

## Weather Data

```

# Minnesota DNR Daily Date
# https://www.dnr.state.mn.us/climate/historical/daily-data.html?sid=mspthr&sname=Minneapolis/St%20Paul
# Station Name: Minneapolis/St Paul Threaded Record - Station ID: mspthr

weather <- read_csv("Data/dnr_weather.csv") %>%
  mutate(year=isoyear(Date),
         week=isoweek(Date),
         precip_in = as.numeric(ifelse(`Precipitation (inches)`=="T", .001, `Precipitation (inches)`)),
         snow_in = as.numeric(ifelse(`Snow (inches)`=="T", .001, `Snow (inches)`)),
         tmax_f = `Maximum Temperature degrees (F)`)%>%
  filter(year >= 2016 & year <= 2020) %>%
  select(year, week, precip_in, snow_in, tmax_f) %>%
  group_by(year, week) %>%
  summarize(precip_in = mean(precip_in, na.rm = T),
         snow_in = mean(snow_in, na.rm = T),
         tmax_f = mean(tmax_f, na.rm = T))

#join to series
series <- series %>% left_join(weather, by = c("year","weekofyr"="week"))

```

## Sunset Data

```

#setting lat-lon for MPLS
mpls_lonlat <- geocode("Minneapolis, MN", output = "latlon", source="google")

#scrape sunset times for each begin date
#mutate to UTC-6 CST
#calculate hours of darkness before midnight
sun_series <- getSunlightTimes(date = seq(min(series$begin_date),
                                         max(series$begin_date),
                                         "days"),

```

```

lat = 44.97775 ,
lon = -93.26501,
keep = "sunset",
tz = "UTC") %>%
mutate(sunset = sunset-hours(6),
midnight = as.POSIXlt(date+days(1), format = '%Y-%m-%d %H:%M:%S'),
dark = as.numeric(midnight-sunset),
year = year(date),
week = isoweek(date)) %>%
group_by(year, week) %>%
summarize(dark_before_12 = mean(dark, na.rm = T))

#joining to series
series <- series %>%
left_join(sun_series, by = c("year", "weekofyr"="week"))

```

## School Data

```

#created manually from online MPLS Public School Calendars: https://mpls.k12.mn.us/calendars
school <- series %>%
select(year, weekofyr, begin_date, end_date) %>%
mutate(days_in_week = as.numeric((end_date-begin_date))+1,
days_in_school = NA_integer_)

school[1,6] <- 5
school[2,6] <- 4
school[3,6] <- 3
school[4,6] <- 5
school[5,6] <- 5
school[6,6] <- 4
school[7,6] <- 4
school[8,6] <- 5
school[9,6] <- 5
school[10,6] <- 4
school[11,6] <- 4
school[12,6] <- 5
school[13,6] <- 0
school[14,6] <- 5
school[15,6] <- 5
school[16,6] <- 5
school[17,6] <- 5
school[18,6] <- 5
school[19,6] <- 5
school[20,6] <- 5
school[21,6] <- 5
school[22,6] <- 4
school[23,6] <- 2
school[24,6] <- 0
school[25,6] <- 0
school[26,6] <- 0

```



```
school[27,6] <- 0
school[28,6] <- 0
school[29,6] <- 0
school[30,6] <- 0
school[31,6] <- 0
school[32,6] <- 0
school[33,6] <- 0
school[34,6] <- 0
school[35,6] <- 5
school[36,6] <- 4
school[37,6] <- 5
school[38,6] <- 5
school[39,6] <- 5
school[40,6] <- 5
school[41,6] <- 5
school[42,6] <- 2
school[43,6] <- 5
school[44,6] <- 3
school[45,6] <- 5
school[46,6] <- 5
school[47,6] <- 2
school[48,6] <- 5
school[49,6] <- 5
school[50,6] <- 5
school[51,6] <- 0
school[52,6] <- 0
school[53,6] <- 4
school[54,6] <- 5
school[55,6] <- 4
school[56,6] <- 4
school[57,6] <- 4
school[58,6] <- 5
school[59,6] <- 4
school[60,6] <- 4
school[61,6] <- 5
school[62,6] <- 5
school[63,6] <- 5
school[64,6] <- 5
school[65,6] <- 3
school[66,6] <- 0
school[67,6] <- 5
school[68,6] <- 5
school[69,6] <- 5
school[70,6] <- 5
school[71,6] <- 5
school[72,6] <- 5
school[73,6] <- 5
school[74,6] <- 4
school[75,6] <- 5
school[76,6] <- 3
school[77,6] <- 0
school[78,6] <- 0
school[79,6] <- 0
```

```
school[80,6] <- 0
school[81,6] <- 0
school[82,6] <- 0
school[83,6] <- 0
school[84,6] <- 0
school[85,6] <- 0
school[86,6] <- 0
school[87,6] <- 5
school[88,6] <- 4
school[89,6] <- 5
school[90,6] <- 5
school[91,6] <- 5
school[92,6] <- 5
school[93,6] <- 5
school[94,6] <- 2
school[95,6] <- 5
school[96,6] <- 3
school[97,6] <- 5
school[98,6] <- 5
school[99,6] <- 2
school[100,6] <- 5
school[101,6] <- 5
school[102,6] <- 5
school[103,6] <- 5
school[104,6] <- 0
school[105,6] <- 0
school[106,6] <- 0
school[107,6] <- 5
school[108,6] <- 4
school[109,6] <- 3
school[110,6] <- 5
school[111,6] <- 5
school[112,6] <- 4
school[113,6] <- 4
school[114,6] <- 5
school[115,6] <- 5
school[116,6] <- 5
school[117,6] <- 5
school[118,6] <- 4
school[119,6] <- 0
school[120,6] <- 5
school[121,6] <- 5
school[122,6] <- 5
school[123,6] <- 5
school[124,6] <- 5
school[125,6] <- 5
school[126,6] <- 5
school[127,6] <- 4
school[128,6] <- 5
school[129,6] <- 0
school[130,6] <- 0
school[131,6] <- 0
school[132,6] <- 0
```

```
school[133,6] <- 0
school[134,6] <- 0
school[135,6] <- 0
school[136,6] <- 0
school[137,6] <- 0
school[138,6] <- 0
school[139,6] <- 0
school[140,6] <- 5
school[141,6] <- 4
school[142,6] <- 5
school[143,6] <- 5
school[144,6] <- 5
school[145,6] <- 5
school[146,6] <- 5
school[147,6] <- 2
school[148,6] <- 5
school[149,6] <- 3
school[150,6] <- 5
school[151,6] <- 5
school[152,6] <- 2
school[153,6] <- 5
school[154,6] <- 5
school[155,6] <- 5
school[156,6] <- 5
school[157,6] <- 0
school[158,6] <- 0
school[159,6] <- 5
school[160,6] <- 5
school[161,6] <- 2
school[162,6] <- 5
school[163,6] <- 5
school[164,6] <- 4
school[165,6] <- 4
school[166,6] <- 5
school[167,6] <- 5
school[168,6] <- 5
school[169,6] <- 5
school[170,6] <- 4
school[171,6] <- 0
school[172,6] <- 5
school[173,6] <- 5
school[174,6] <- 5
school[175,6] <- 5
school[176,6] <- 5
school[177,6] <- 5
school[178,6] <- 5
school[179,6] <- 4
school[180,6] <- 5
school[181,6] <- 0
school[182,6] <- 0
school[183,6] <- 0
school[184,6] <- 0
school[185,6] <- 0
```

```
school[186,6] <- 0
school[187,6] <- 0
school[188,6] <- 0
school[189,6] <- 0
school[190,6] <- 0
school[191,6] <- 0
school[192,6] <- 0
school[193,6] <- 4
school[194,6] <- 5
school[195,6] <- 5
school[196,6] <- 5
school[197,6] <- 5
school[198,6] <- 5
school[199,6] <- 2
school[200,6] <- 5
school[201,6] <- 4
school[202,6] <- 5
school[203,6] <- 5
school[204,6] <- 5
school[205,6] <- 2
school[206,6] <- 5
school[207,6] <- 5
school[208,6] <- 5
school[209,6] <- 0
school[210,6] <- 0
school[211,6] <- 5
school[212,6] <- 4
school[213,6] <- 4
school[214,6] <- 5
school[215,6] <- 5
school[216,6] <- 5
school[217,6] <- 3
school[218,6] <- 5
school[219,6] <- 5
school[220,6] <- 5
school[221,6] <- 5
school[222,6] <- 4
school[223,6] <- 0
school[224,6] <- 5
school[225,6] <- 5
school[226,6] <- 5
school[227,6] <- 5
school[228,6] <- 5
school[229,6] <- 5
school[230,6] <- 5
school[231,6] <- 4
school[232,6] <- 5
school[233,6] <- 0
school[234,6] <- 0
school[235,6] <- 0
school[236,6] <- 0
school[237,6] <- 0
school[238,6] <- 0
```

```

school[239,6] <- 0
school[240,6] <- 0
school[241,6] <- 0
school[242,6] <- 0
school[243,6] <- 0
school[244,6] <- 0
school[245,6] <- 4
school[246,6] <- 5
school[247,6] <- 5
school[248,6] <- 5
school[249,6] <- 5
school[250,6] <- 5
school[251,6] <- 3
school[252,6] <- 4
school[253,6] <- 5
school[254,6] <- 4
school[255,6] <- 5
school[256,6] <- 5
school[257,6] <- 2
school[258,6] <- 5
school[259,6] <- 5
school[260,6] <- 5
school[261,6] <- 0
school[262,6] <- 0

school <- school %>%
  mutate(school = days_in_school/days_in_week) %>%
  select(year, weekofyr, school)

series <- series %>% left_join(school, by = c("year", "weekofyr"))

```

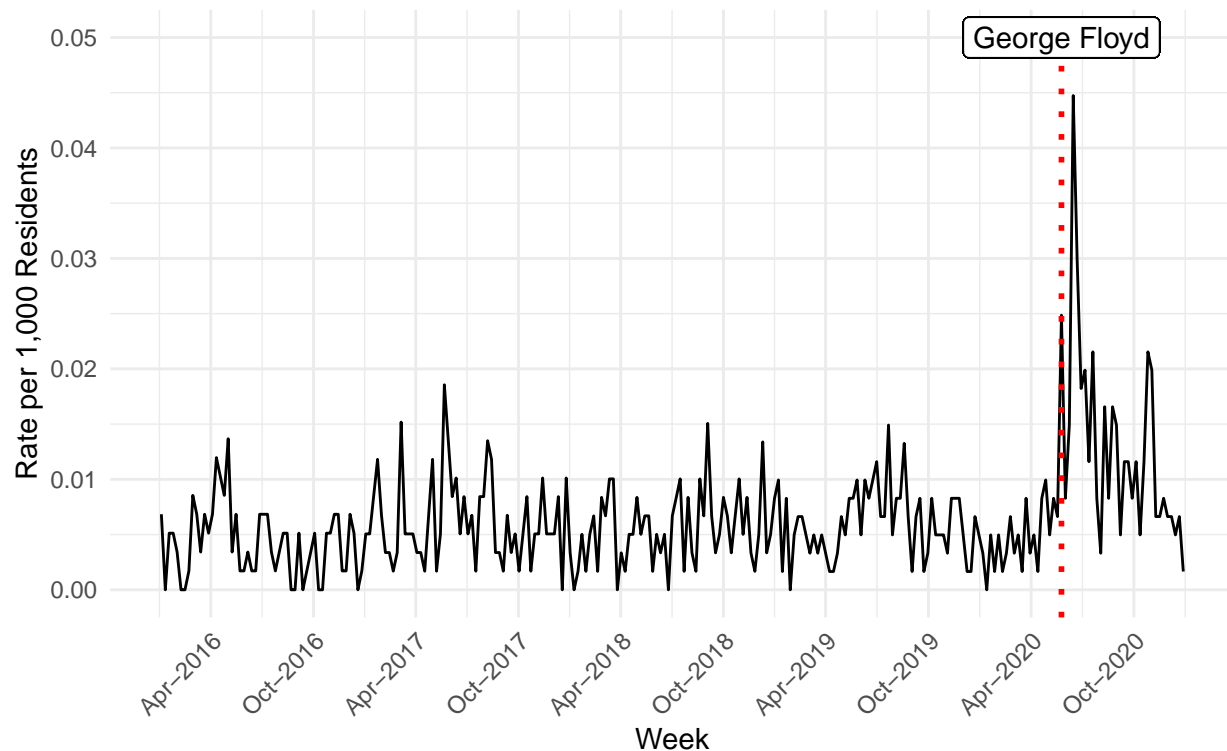
## Time Series Vizualization

```

ggplot(series)+
  geom_line(aes(x=begin_date, y=assault_incid_c))+
  scale_x_date(date_labels = "%b-%Y", date_breaks = "6 months")+
  geom_vline(xintercept=series$begin_date[series$year==2020 & series$weekofyr==isoweek(date("2020-05-25"))],
            linetype="dotted", color="red", size=1)+
  geom_label(aes(x=series$begin_date[series$year==2020 & series$weekofyr==isoweek(date("2020-05-25"))],
                y=0.050),
            label = "George Floyd", show.legend = FALSE)+
  labs(title = "Figure 1: Weekly Firearm Assault Injuries, 2016-2020",
        subtitle = "MHA Hospital Data",
        x = "Week",
        y = "Rate per 1,000 Residents")+
  theme_minimal()+
  theme(axis.text.x=element_text(angle=45, hjust=1))

```

Figure 1: Weekly Firearm Assault Injuries, 2016–2020  
MHA Hospital Data



## Time Series Analysis

```
series <- series %>%
  mutate(t = 1:length(assault_incid_c),
         post_floyd = as.numeric(begin_date >= as.Date("2020-05-25")),
         post_floyd_3 = as.numeric(begin_date >= as.Date("2020-05-25")+months(3)),
         stay_at_home = as.numeric(begin_date >= as.Date("2020-03-28") &
                                   state_of_emerg = as.numeric(begin_date >= as.Date("2020-03-13")),
         uof_lag=lag(use_of_force_rate,1),
         stops_lag = lag(police_stop_rate,1),
         shoot_lag = lag(off_inv_shooting_rate,1))

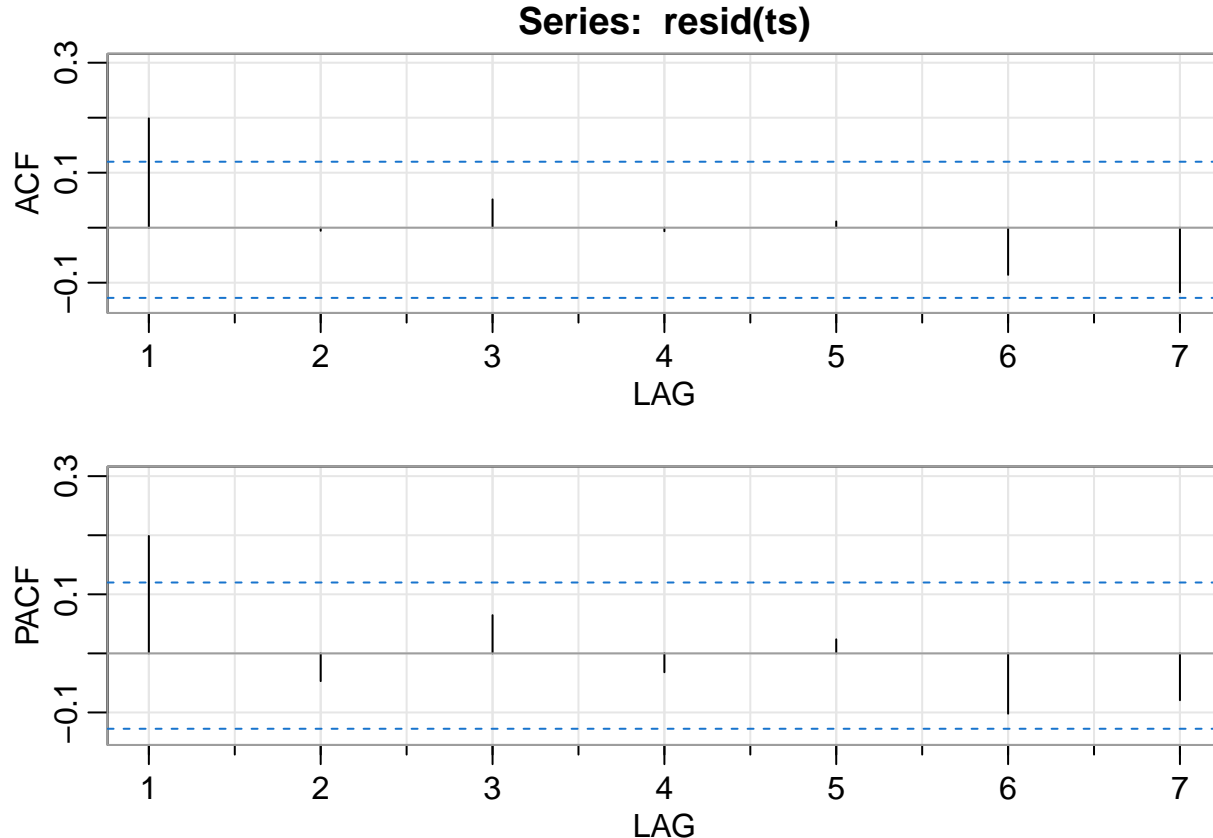
ts <- lm(assault_incid_c~t+state_of_emerg+stay_at_home+post_floyd+post_floyd_3+
        tmax_f+snow_in+precip_in+dark_before_12+school,
        data = series)

summary(ts)
```

```
##
## Call:
## lm(formula = assault_incid_c ~ t + state_of_emerg + stay_at_home +
##     post_floyd + post_floyd_3 + tmax_f + snow_in + precip_in +
##     dark_before_12 + school, data = series)
##
```

```
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.0139072 -0.0025135 -0.0002368  0.0018088  0.0273037
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   6.637e-03  3.557e-03   1.866   0.0633 .
## t              6.573e-06  4.524e-06   1.453   0.1475
## state_of_emerg -3.394e-03  2.573e-03  -1.319   0.1883
## stay_at_home   3.078e-03  2.646e-03   1.163   0.2459
## post_floyd     1.360e-02  2.631e-03   5.167 4.86e-07 ***
## post_floyd_3  -6.644e-03  1.633e-03  -4.068 6.36e-05 ***
## tmax_f         1.333e-05  2.415e-05   0.552   0.5814
## snow_in       -5.169e-04  7.725e-04  -0.669   0.5041
## precip_in     -2.400e-04  2.524e-03  -0.095   0.9243
## dark_before_12 -5.074e-04  4.136e-04  -1.227   0.2211
## school         7.116e-04  9.511e-04   0.748   0.4551
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.004188 on 250 degrees of freedom
## Multiple R-squared:  0.3509, Adjusted R-squared:  0.325
## F-statistic: 13.52 on 10 and 250 DF,  p-value: < 2.2e-16
```

```
acf2(resid(ts), max.lag = 7)
```



```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## ACF   0.2 -0.01 0.05 -0.01 0.01 -0.09 -0.12
## PACF  0.2 -0.05 0.06 -0.03 0.02 -0.10 -0.08
```

```
ts_ar1<- lm(assault_incid_c~t+state_of_emerg+stay_at_home+post_floyd+post_floyd_3+
            tmax_f+snow_in+precip_in+dark_before_12+school+
            dplyr::lag(assault_incid_c, 1), data = series)
summary(ts_ar1)
```

```
##
## Call:
## lm(formula = assault_incid_c ~ t + state_of_emerg + stay_at_home +
##     post_floyd + post_floyd_3 + tmax_f + snow_in + precip_in +
##     dark_before_12 + school + dplyr::lag(assault_incid_c, 1),
##     data = series)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
	-0.0122313	-0.0026183	-0.0002489	0.0020128	0.0277111

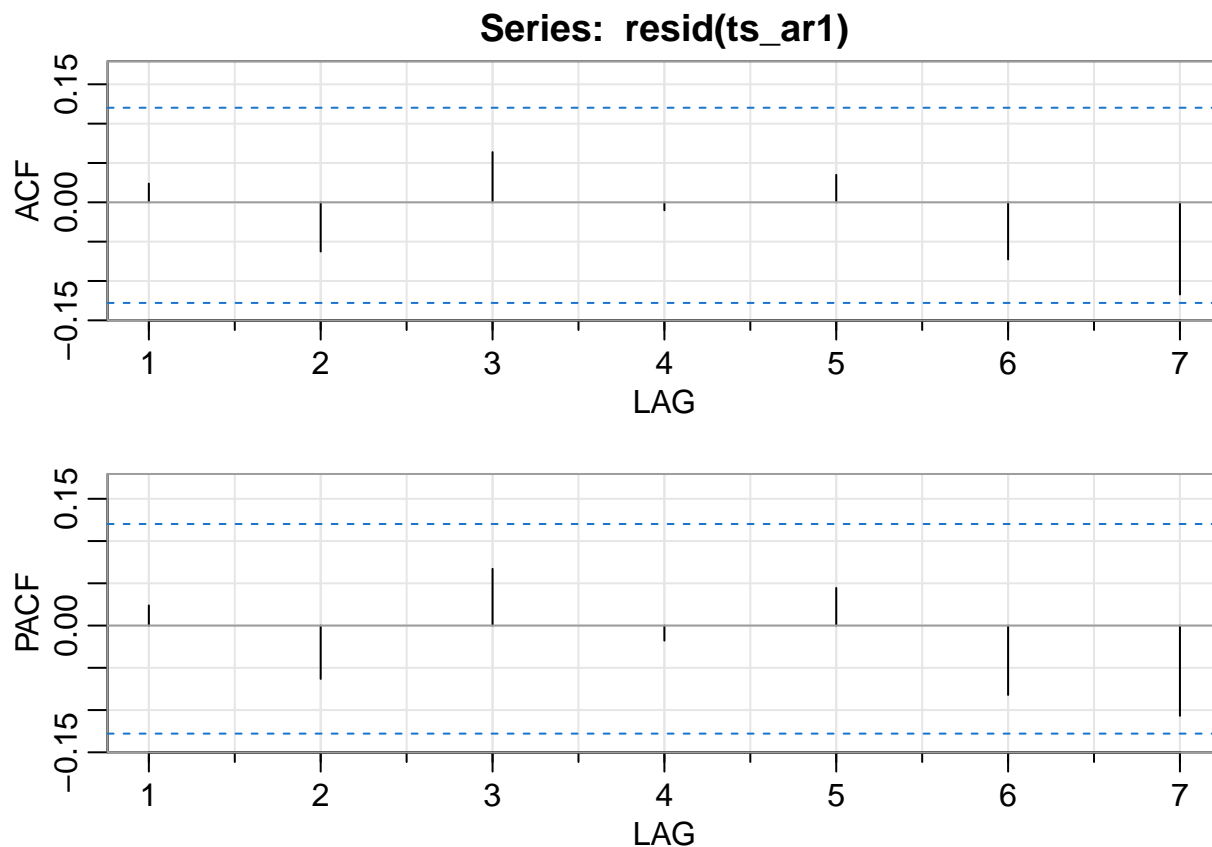
```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	5.373e-03	3.528e-03	1.523	0.12902
t	5.744e-06	4.512e-06	1.273	0.20425
state_of_emerg	-3.649e-03	2.535e-03	-1.439	0.15128
stay_at_home	3.529e-03	2.611e-03	1.352	0.17769
post_floyd	1.202e-02	2.642e-03	4.548	8.49e-06 ***
post_floyd_3	-5.645e-03	1.641e-03	-3.441	0.00068 ***
tmax_f	1.031e-05	2.383e-05	0.433	0.66551
snow_in	-4.415e-04	7.613e-04	-0.580	0.56252
precip_in	1.817e-04	2.490e-03	0.073	0.94190
dark_before_12	-4.423e-04	4.083e-04	-1.083	0.27983
school	8.569e-04	9.383e-04	0.913	0.36199
dplyr::lag(assault_incid_c, 1)	1.863e-01	6.179e-02	3.015	0.00284 **

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.004125 on 248 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.3753, Adjusted R-squared:  0.3476
## F-statistic: 13.54 on 11 and 248 DF, p-value: < 2.2e-16
```

```
acf2(resid(ts_ar1), max.lag = 7)
```





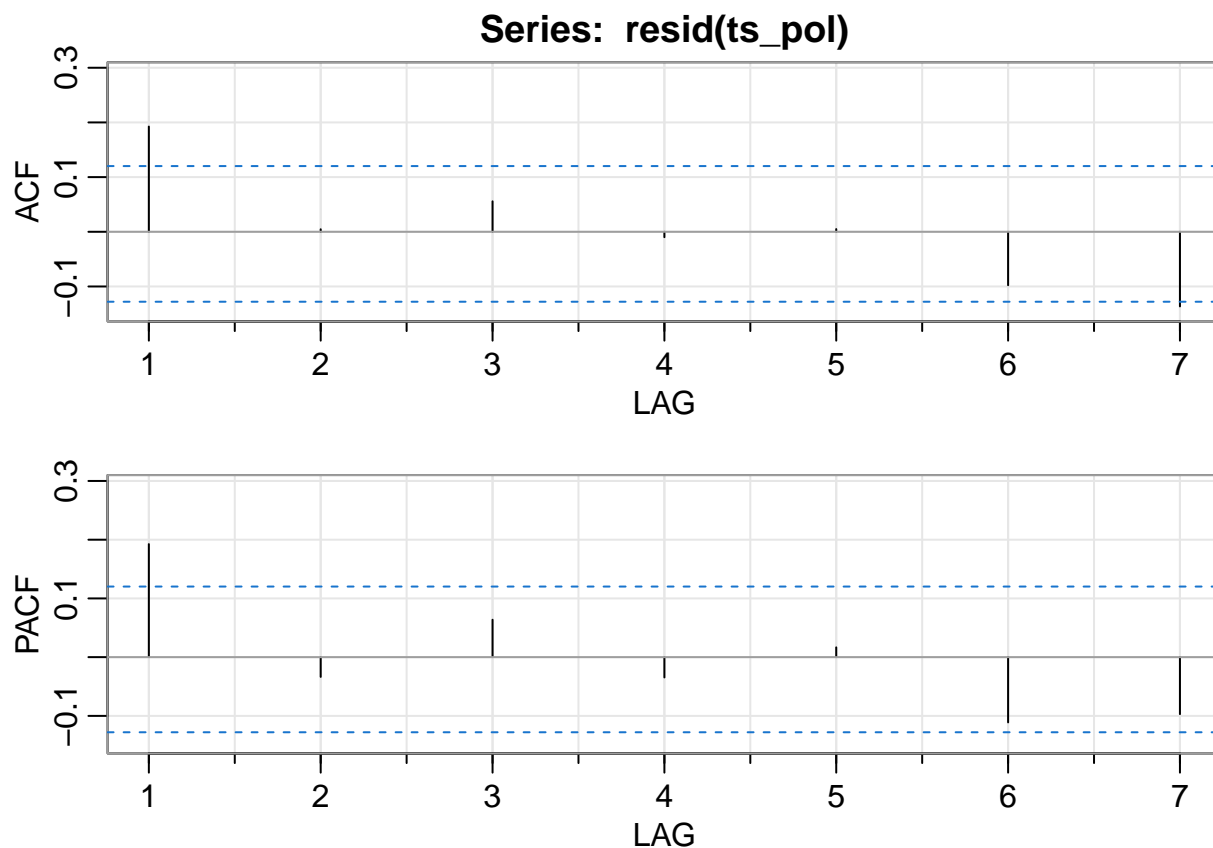
```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## ACF  0.02 -0.06  0.06 -0.01  0.03 -0.07 -0.12
## PACF 0.02 -0.06  0.07 -0.02  0.04 -0.08 -0.11
```

```
ts_pol <- lm(assault_incid_c~t+state_of_emerg+stay_at_home+post_floyd+post_floyd_3+
             tmax_f+snow_in+precip_in+dark_before_12+school+
             uof_lag+stops_lag+shoot_lag,
             data = series)
summary(ts_pol)
```

```
##
## Call:
## lm(formula = assault_incid_c ~ t + state_of_emerg + stay_at_home +
##     post_floyd + post_floyd_3 + tmax_f + snow_in + precip_in +
##     dark_before_12 + school + uof_lag + stops_lag + shoot_lag,
##     data = series)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.0140949 -0.0026548 -0.0002571  0.0018680  0.0271987
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   6.496e-03  3.622e-03   1.794  0.07410 .
## t             4.143e-06  4.997e-06   0.829  0.40788
```

```
## state_of_emerg -3.131e-03  2.595e-03 -1.207  0.22867
## stay_at_home   2.982e-03  2.647e-03  1.127  0.26095
## post_floyd     1.376e-02  2.641e-03  5.212  3.96e-07 ***
## post_floyd_3   -5.852e-03  1.803e-03 -3.246  0.00133 **
## tmax_f         1.403e-05  2.418e-05  0.580  0.56241
## snow_in        -5.236e-04  7.744e-04 -0.676  0.49959
## precip_in      7.839e-05  2.531e-03  0.031  0.97532
## dark_before_12 -4.575e-04  4.157e-04 -1.101  0.27217
## school         5.299e-04  9.556e-04  0.554  0.57974
## uof_lag        -8.392e-03  8.067e-03 -1.040  0.29920
## stops_lag      6.294e-04  4.709e-04  1.337  0.18256
## shoot_lag      -1.701e-01  2.259e-01 -0.753  0.45217
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.004187 on 246 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.3616, Adjusted R-squared:  0.3279
## F-statistic: 10.72 on 13 and 246 DF, p-value: < 2.2e-16
```

```
acf2(resid(ts_pol), max.lag = 7)
```



```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## ACF  0.19  0.00  0.06 -0.01  0.00 -0.10 -0.14
## PACF  0.19 -0.03  0.06 -0.03  0.02 -0.11 -0.10
```

```
ts_ar1_pol <- lm(assault_incid_c ~ t + state_of_emerg + stay_at_home + post_floyd + post_floyd_3 +
               tmax_f + snow_in + precip_in + dark_before_12 + school +
               uof_lag + stops_lag + shoot_lag +
               dplyr::lag(assault_incid_c, 1), data = series)
```

```
summary(ts_ar1_pol)
```

```
##
## Call:
## lm(formula = assault_incid_c ~ t + state_of_emerg + stay_at_home +
##     post_floyd + post_floyd_3 + tmax_f + snow_in + precip_in +
##     dark_before_12 + school + uof_lag + stops_lag + shoot_lag +
##     dplyr::lag(assault_incid_c, 1), data = series)
##
## Residuals:
```

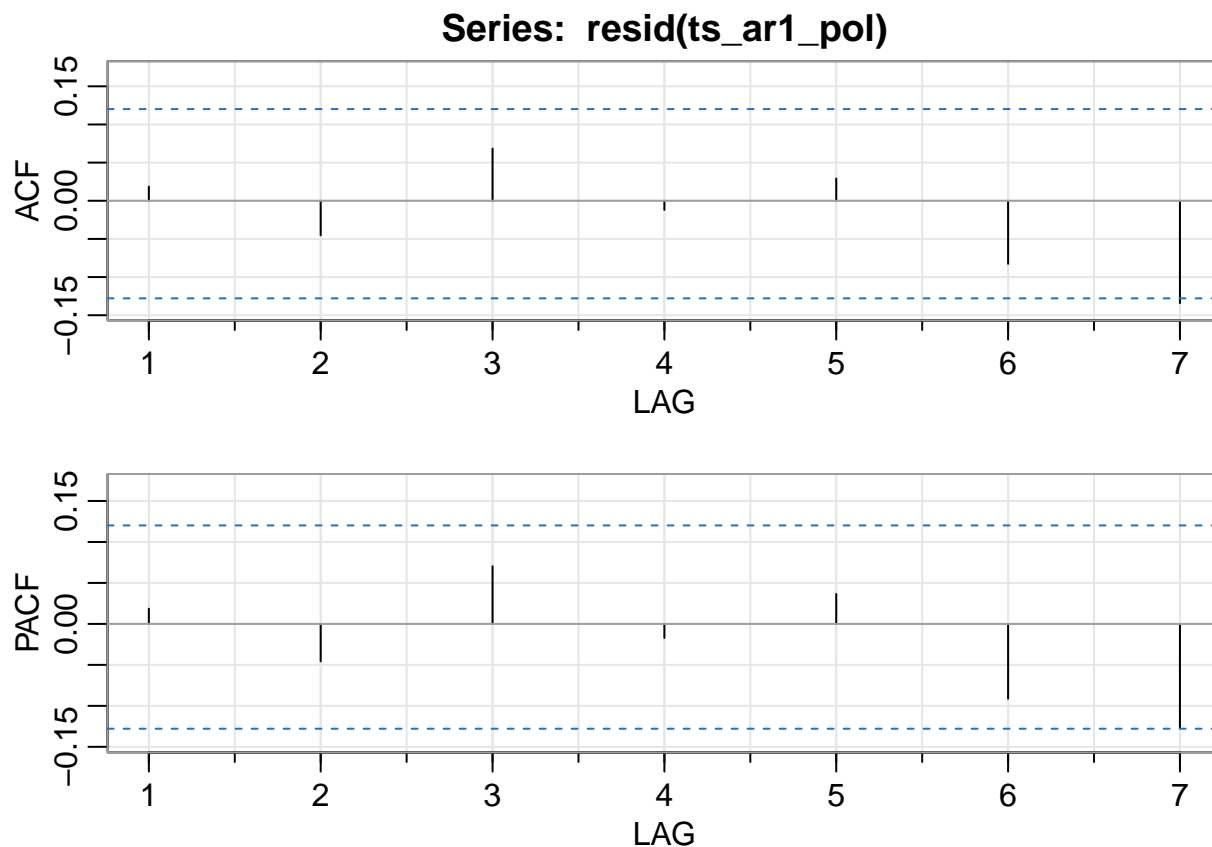
	Min	1Q	Median	3Q	Max
	-0.0124598	-0.0025174	-0.0003879	0.0021404	0.0275611

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	5.455e-03	3.584e-03	1.522	0.12924
t	3.311e-06	4.929e-06	0.672	0.50234
state_of_emerg	-3.428e-03	2.557e-03	-1.341	0.18129
stay_at_home	3.454e-03	2.611e-03	1.323	0.18712
post_floyd	1.218e-02	2.655e-03	4.590	7.09e-06 ***
post_floyd_3	-4.812e-03	1.810e-03	-2.659	0.00835 **
tmax_f	1.083e-05	2.384e-05	0.454	0.65014
snow_in	-4.529e-04	7.629e-04	-0.594	0.55331
precip_in	4.554e-04	2.495e-03	0.183	0.85534
dark_before_12	-3.921e-04	4.099e-04	-0.956	0.33979
school	6.973e-04	9.427e-04	0.740	0.46021
uof_lag	-9.394e-03	7.950e-03	-1.182	0.23852
stops_lag	5.243e-04	4.650e-04	1.128	0.26062
shoot_lag	-1.554e-01	2.225e-01	-0.698	0.48561
dplyr::lag(assault_incid_c, 1)	1.831e-01	6.200e-02	2.952	0.00346 **

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.004123 on 245 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.3836, Adjusted R-squared:  0.3483
## F-statistic: 10.89 on 14 and 245 DF,  p-value: < 2.2e-16
```

```
acf2(resid(ts_ar1_pol), max.lag = 7)
```



```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7]
## ACF  0.02 -0.05 0.07 -0.01 0.03 -0.08 -0.13
## PACF 0.02 -0.05 0.07 -0.02 0.04 -0.09 -0.13
```

## ZCTA-Week Level Analysis

```
#aggregate to zip-level over years
zip_level <- panel %>%
  group_by(zcta, period) %>%
  summarize(assault_tot = mean(assault_tot, na.rm = T),
            unintent_tot = mean(unintent_tot, na.rm = T),
            suicide_tot = mean(suicide_tot, na.rm = T),
            undeter_tot = mean(undeter_tot, na.rm = T),
            legal_tot = mean(legal_tot, na.rm = T),
            combined_tot = mean(combined_tot, na.rm = T),
            total_pop = mean(total_pop, na.rm = T)) %>%
  mutate(assault_incid_c = (assault_tot/total_pop)*1000,
         unintent_incid_c = (unintent_tot/total_pop)*1000,
         suicide_incid_c = (suicide_tot/total_pop)*1000,
         undeter_incid_c = (undeter_tot/total_pop)*1000,
         legal_incid_c = (legal_tot/total_pop)*1000,
         combined_incid_c = (combined_tot/total_pop)*1000) %>%
  ungroup() %>%
```

```

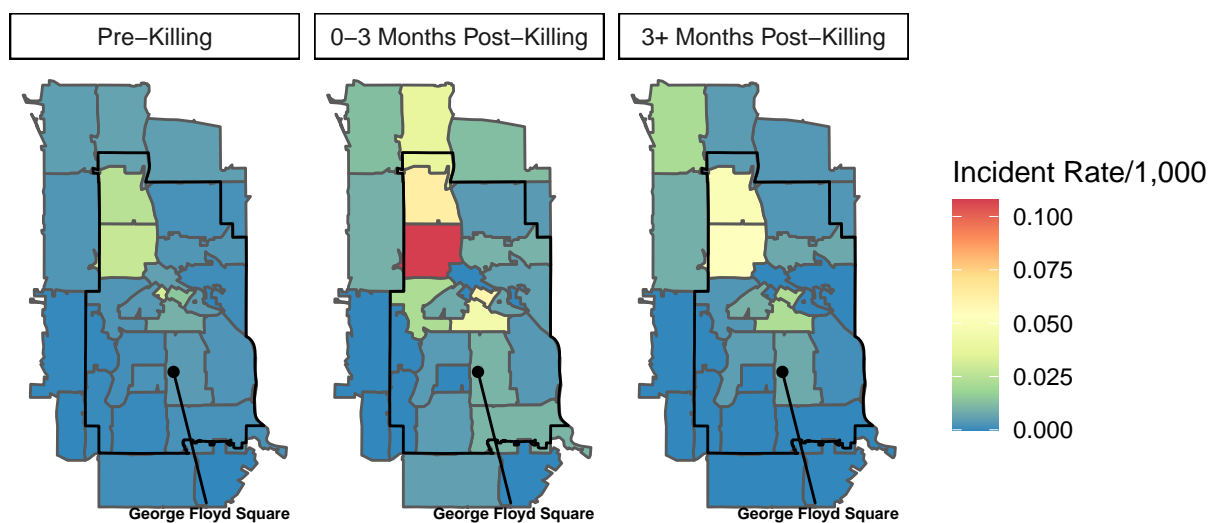
st_drop_geometry() %>%
left_join(zcta, by = "zcta")

#george floyd square
gfs <- geocode("George Floyd Square, Minneapolis", output = "latlon") %>%
  st_as_sf(coords = c("lon", "lat"), crs = "NAD83", remove=F) %>%
  mutate(name = "George Floyd Square")

ggplot() +
  geom_sf(data = zip_level, aes(geometry = geometry, fill = assault_incid_c)) +
  geom_sf(data = mpls, aes(geometry = geometry), color = "black", alpha = 0)+
  geom_sf(data = gfs, aes(geometry = geometry), color = "black")+
  geom_text_repel(data = gfs, aes(x=lon, y=lat, label = name),
    size = 2,
    fontface = "bold",
    nudge_x = 1, nudge_y = -1)+
  facet_wrap(~period)+
  scale_fill_distiller(palette = "Spectral")+
  labs(title = "Figure 2: Firearm Assault Injury Rates by ZCTA and Period",
    subtitle = "MHA Hospital Discharge Data",
    fill = "Incident Rate/1,000")+
  theme(axis.text = element_blank(),
    axis.line = element_blank(),
    axis.ticks = element_blank(),
    panel.border = element_blank(),
    panel.grid = element_blank(),
    axis.title = element_blank(),
    panel.background = element_blank(),
    panel.grid.major = element_line(colour="transparent"),
    plot.subtitle = element_text(face="italic"),
    strip.background = element_rect(fill = "white",
      colour = "black"))

```

Figure 2: Firearm Assault Injury Rates by ZCTA and Period  
MHA Hospital Discharge Data



## Panel Analysis

```
##
## Call:
## lm(formula = assault_incident_c ~ t + state_of_emerg + stay_at_home +
##      post_floyd + post_floyd_3 + tmax_f + snow_in + precip_in +
##      dark_before_12 + school + as.factor(zcta), data = panel)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.349  -0.625  -0.270   0.022  264.043
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.5572515   0.8381086   0.665  0.50614
## t               0.0022449   0.0009967   2.252  0.02433 *
## state_of_emerg -0.5300117   0.5660457  -0.936  0.34913
## stay_at_home    0.2300424   0.5822084   0.395  0.69277
## post_floyd      0.9772376   0.5789280   1.688  0.09145 .
## post_floyd_3   -0.4441449   0.3593825  -1.236  0.21655
## tmax_f          0.0028671   0.0053216   0.539  0.59007
## snow_in        -0.0850339   0.1700075  -0.500  0.61697
## precip_in      -0.1840307   0.5556500  -0.331  0.74050
## dark_before_12 -0.0804399   0.0911871  -0.882  0.37773
## school         -0.1524415   0.2093498  -0.728  0.46654
```

```

## as.factor(zcta)55402  2.0188457  0.4268702  4.729 2.29e-06 ***
## as.factor(zcta)55403  0.0171753  0.4268702  0.040 0.96791
## as.factor(zcta)55404  0.7732006  0.4268702  1.811 0.07013 .
## as.factor(zcta)55405 -0.0343970  0.4268702 -0.081 0.93578
## as.factor(zcta)55406 -0.1560144  0.4268702 -0.365 0.71476
## as.factor(zcta)55407  0.0582581  0.4268702  0.136 0.89145
## as.factor(zcta)55408 -0.2586522  0.4268702 -0.606 0.54458
## as.factor(zcta)55409 -0.2364360  0.4268702 -0.554 0.57968
## as.factor(zcta)55410 -0.4089703  0.4268702 -0.958 0.33806
## as.factor(zcta)55411  2.9258866  0.4268702  6.854 7.76e-12 ***
## as.factor(zcta)55412  2.4045153  0.4268702  5.633 1.84e-08 ***
## as.factor(zcta)55413 -0.0592224  0.4268702 -0.139 0.88966
## as.factor(zcta)55414 -0.3227576  0.4268702 -0.756 0.44961
## as.factor(zcta)55415  1.2416106  0.4268702  2.909 0.00364 **
## as.factor(zcta)55416 -0.4414446  0.4268702 -1.034 0.30110
## as.factor(zcta)55417 -0.2190085  0.4268702 -0.513 0.60793
## as.factor(zcta)55418 -0.2264396  0.4268702 -0.530 0.59581
## as.factor(zcta)55419 -0.3964493  0.4268702 -0.929 0.35306
## as.factor(zcta)55421  0.1187478  0.4268702  0.278 0.78088
## as.factor(zcta)55422 -0.0500345  0.4268702 -0.117 0.90670
## as.factor(zcta)55423 -0.3066886  0.4268702 -0.718 0.47250
## as.factor(zcta)55424 -0.4652388  0.4268702 -1.090 0.27580
## as.factor(zcta)55429  0.2386535  0.4268702  0.559 0.57613
## as.factor(zcta)55430  0.3444312  0.4268702  0.807 0.41976
## as.factor(zcta)55450 -0.4776894  0.4311256 -1.108 0.26790
## as.factor(zcta)55454  0.0086534  0.4268702  0.020 0.98383
## as.factor(zcta)55455 -0.4652388  0.4268702 -1.090 0.27580
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.876 on 7260 degrees of freedom
## (10 observations deleted due to missingness)
## Multiple R-squared:  0.03484,    Adjusted R-squared:  0.02992
## F-statistic: 7.083 on 37 and 7260 DF,  p-value: < 2.2e-16

##
## Call:
## lm(formula = assault_incid_c ~ t + state_of_emerg + stay_at_home +
##      post_floyd + post_floyd_3 + tmax_f + snow_in + precip_in +
##      dark_before_12 + school + uof_lag + stops_lag + shoot_lag +
##      as.factor(zcta), data = panel)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.610  -0.654  -0.273   0.036  263.804
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.670981   0.844991   0.794  0.42718
## t              0.002128   0.001053   2.021  0.04336 *
## state_of_emerg -0.524791   0.568540  -0.923  0.35601
## stay_at_home    0.220967   0.583276   0.379  0.70482
## post_floyd      0.973902   0.580683   1.677  0.09355 .
## post_floyd_3    -0.344853   0.362656  -0.951  0.34168

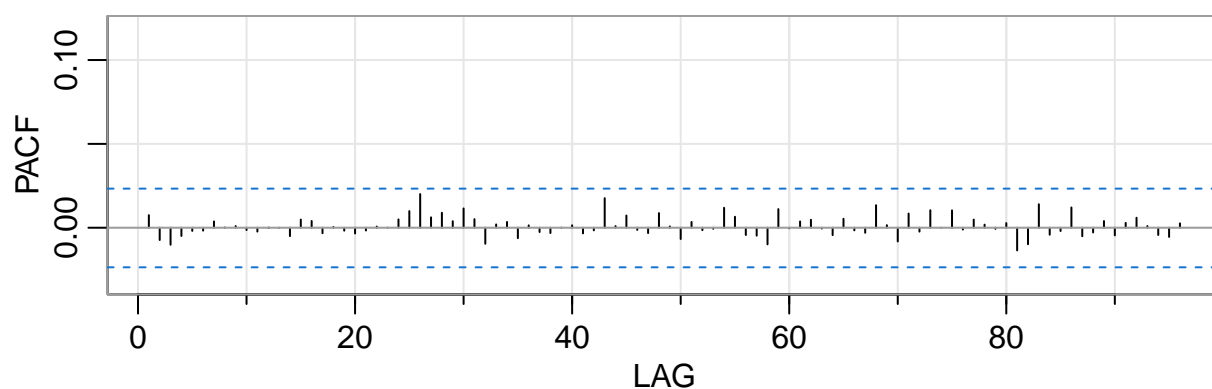
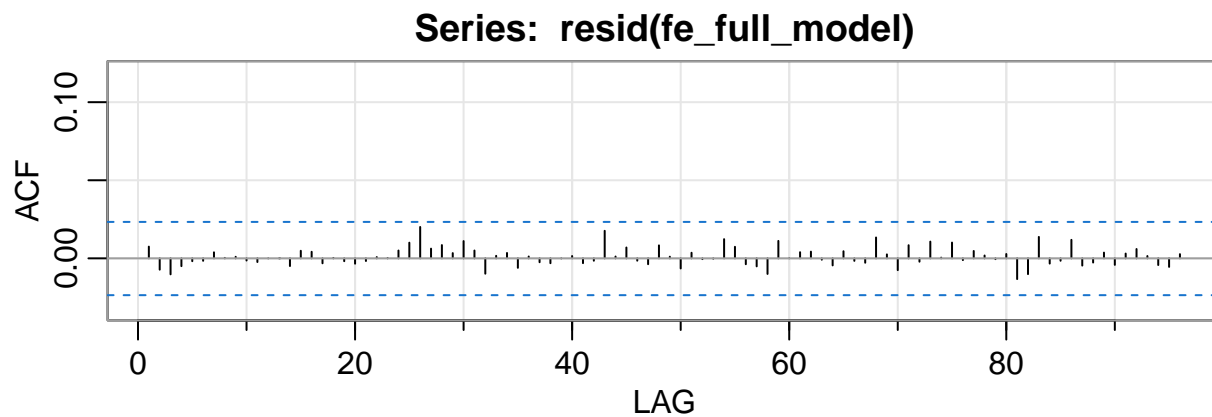
```

```

## tmax_f          0.003259    0.005334    0.611    0.54122
## snow_in         -0.079543    0.170338   -0.467    0.64053
## precip_in       -0.166529    0.556667   -0.299    0.76483
## dark_before_12  -0.079835    0.091488   -0.873    0.38290
## school          -0.149832    0.210028   -0.713    0.47563
## uof_lag         -0.051330    0.023112   -2.221    0.02639 *
## stops_lag        0.001729    0.003212    0.538    0.59031
## shoot_lag       -0.293985    0.431783   -0.681    0.49598
## as.factor(zcta)55402 2.088780    0.429819    4.860 1.20e-06 ***
## as.factor(zcta)55403 0.037945    0.429398    0.088    0.92959
## as.factor(zcta)55404 0.712232    0.433899    1.641    0.10074
## as.factor(zcta)55405 -0.130753    0.430381   -0.304    0.76128
## as.factor(zcta)55406 -0.267428    0.431376   -0.620    0.53532
## as.factor(zcta)55407 -0.027715    0.433533   -0.064    0.94903
## as.factor(zcta)55408 -0.324824    0.447391   -0.726    0.46784
## as.factor(zcta)55409 -0.371963    0.436166   -0.853    0.39380
## as.factor(zcta)55410 -0.542810    0.438870   -1.237    0.21619
## as.factor(zcta)55411 2.947904    0.488093    6.040 1.62e-09 ***
## as.factor(zcta)55412 2.354194    0.432437    5.444 5.38e-08 ***
## as.factor(zcta)55413 -0.205577    0.432992   -0.475    0.63496
## as.factor(zcta)55414 -0.421141    0.430667   -0.978    0.32817
## as.factor(zcta)55415 1.152529    0.434889    2.650    0.00806 **
## as.factor(zcta)55416 -0.580212    0.440058   -1.318    0.18738
## as.factor(zcta)55417 -0.347800    0.436608   -0.797    0.42571
## as.factor(zcta)55418 -0.377813    0.435679   -0.867    0.38587
## as.factor(zcta)55419 -0.535217    0.433294   -1.235    0.21679
## as.factor(zcta)55421 -0.015101    0.441480   -0.034    0.97271
## as.factor(zcta)55422 -0.210477    0.441619   -0.477    0.63366
## as.factor(zcta)55423 -0.442733    0.441771   -1.002    0.31629
## as.factor(zcta)55424 -0.601940    0.441756   -1.363    0.17305
## as.factor(zcta)55429 0.104753    0.441785    0.237    0.81258
## as.factor(zcta)55430 0.223424    0.437964    0.510    0.60997
## as.factor(zcta)55450 -0.614245    0.446083   -1.377    0.16856
## as.factor(zcta)55454 -0.109903    0.436008   -0.252    0.80100
## as.factor(zcta)55455 -0.597703    0.441157   -1.355    0.17551
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.884 on 7229 degrees of freedom
## (38 observations deleted due to missingness)
## Multiple R-squared:  0.03556,    Adjusted R-squared:  0.03022
## F-statistic: 6.663 on 40 and 7229 DF,  p-value: < 2.2e-16

```





```
##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8] [,9] [,10] [,11] [,12] [,13]
## ACF  0.01 -0.01 -0.01 -0.01  0    0    0    0    0    0    0    0    0
## PACF  0.01 -0.01 -0.01  0.00  0    0    0    0    0    0    0    0    0
##      [,14] [,15] [,16] [,17] [,18] [,19] [,20] [,21] [,22] [,23] [,24] [,25]
## ACF  0.00    0    0    0    0    0    0    0    0    0    0.01  0.01
## PACF -0.01    0    0    0    0    0    0    0    0    0    0.00  0.01
##      [,26] [,27] [,28] [,29] [,30] [,31] [,32] [,33] [,34] [,35] [,36] [,37]
## ACF  0.02  0.01  0.01    0  0.01  0.01 -0.01    0    0 -0.01    0    0
## PACF  0.02  0.01  0.01    0  0.01  0.01 -0.01    0    0 -0.01    0    0
##      [,38] [,39] [,40] [,41] [,42] [,43] [,44] [,45] [,46] [,47] [,48] [,49]
## ACF    0    0    0    0    0    0.02    0  0.01    0    0    0.01    0
## PACF    0    0    0    0    0    0.02    0  0.01    0    0    0.01    0
##      [,50] [,51] [,52] [,53] [,54] [,55] [,56] [,57] [,58] [,59] [,60] [,61]
## ACF -0.01    0    0    0  0.01  0.01    0 -0.01 -0.01  0.01    0    0
## PACF -0.01    0    0    0  0.01  0.01    0  0.00 -0.01  0.01    0    0
##      [,62] [,63] [,64] [,65] [,66] [,67] [,68] [,69] [,70] [,71] [,72] [,73]
## ACF    0    0    0  0.00    0    0  0.01    0 -0.01  0.01    0  0.01
## PACF    0    0    0  0.01    0    0  0.01    0 -0.01  0.01    0  0.01
##      [,74] [,75] [,76] [,77] [,78] [,79] [,80] [,81] [,82] [,83] [,84] [,85]
## ACF    0  0.01    0    0    0    0    0 -0.01 -0.01  0.01    0    0
## PACF    0  0.01    0    0    0    0    0 -0.01 -0.01  0.01    0    0
##      [,86] [,87] [,88] [,89] [,90] [,91] [,92] [,93] [,94] [,95] [,96]
## ACF  0.01  0.00    0    0    0    0  0.01    0    0 -0.01    0
## PACF  0.01 -0.01    0    0    0    0  0.01    0    0 -0.01    0
```

```
##
```

```
## Call:
## lm(formula = assault_incid_c ~ t + state_of_emerg + stay_at_home +
##     post_floyd + post_floyd_3 + tmax_f + snow_in + precip_in +
##     dark_before_12 + school + as.factor(zcta) + post_floyd:as.factor(zcta) +
##     post_floyd_3:as.factor(zcta), data = panel)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -10.790  -0.589  -0.269   0.004  263.627
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      6.896e-01  8.441e-01   0.817  0.41400
## t                2.243e-03  9.952e-04   2.254  0.02426 *
## state_of_emerg   -5.298e-01  5.652e-01  -0.937  0.34862
## stay_at_home     2.300e-01  5.813e-01   0.396  0.69232
## post_floyd       -5.560e-01  1.438e+00  -0.387  0.69894
## post_floyd_3     3.399e-01  1.741e+00   0.195  0.84523
## tmax_f           2.877e-03  5.313e-03   0.541  0.58819
## snow_in          -8.495e-02  1.697e-01  -0.500  0.61679
## precip_in        -1.840e-01  5.548e-01  -0.332  0.74012
## dark_before_12   -8.024e-02  9.105e-02  -0.881  0.37821
## school           -1.522e-01  2.090e-01  -0.728  0.46649
## as.factor(zcta)55402  2.301e+00  4.550e-01   5.057  4.37e-07 ***
## as.factor(zcta)55403  -1.105e-01  4.550e-01  -0.243  0.80811
## as.factor(zcta)55404   4.278e-01  4.550e-01   0.940  0.34719
## as.factor(zcta)55405  -2.036e-01  4.550e-01  -0.447  0.65464
## as.factor(zcta)55406  -2.168e-01  4.550e-01  -0.477  0.63369
## as.factor(zcta)55407  -6.147e-02  4.550e-01  -0.135  0.89254
## as.factor(zcta)55408  -3.636e-01  4.550e-01  -0.799  0.42426
## as.factor(zcta)55409  -2.695e-01  4.550e-01  -0.592  0.55372
## as.factor(zcta)55410  -4.661e-01  4.550e-01  -1.024  0.30569
## as.factor(zcta)55411   2.256e+00  4.550e-01   4.959  7.25e-07 ***
## as.factor(zcta)55412   1.964e+00  4.550e-01   4.317  1.60e-05 ***
## as.factor(zcta)55413  -1.900e-01  4.550e-01  -0.418  0.67631
## as.factor(zcta)55414  -4.043e-01  4.550e-01  -0.889  0.37426
## as.factor(zcta)55415   8.710e-01  4.550e-01   1.914  0.05565 .
## as.factor(zcta)55416  -5.031e-01  4.550e-01  -1.106  0.26888
## as.factor(zcta)55417  -3.153e-01  4.550e-01  -0.693  0.48838
## as.factor(zcta)55418  -3.157e-01  4.550e-01  -0.694  0.48780
## as.factor(zcta)55419  -4.831e-01  4.550e-01  -1.062  0.28843
## as.factor(zcta)55421   2.839e-02  4.550e-01   0.062  0.95025
## as.factor(zcta)55422  -1.887e-01  4.550e-01  -0.415  0.67830
## as.factor(zcta)55423  -3.856e-01  4.550e-01  -0.848  0.39674
## as.factor(zcta)55424  -5.303e-01  4.550e-01  -1.165  0.24393
## as.factor(zcta)55429   2.157e-02  4.550e-01   0.047  0.96220
## as.factor(zcta)55430   1.214e-01  4.550e-01   0.267  0.78963
## as.factor(zcta)55450  -5.414e-01  4.602e-01  -1.176  0.23951
## as.factor(zcta)55454  -3.370e-02  4.550e-01  -0.074  0.94096
## as.factor(zcta)55455  -5.303e-01  4.550e-01  -1.165  0.24393
## post_floyd:as.factor(zcta)55402 -2.301e+00  1.896e+00  -1.214  0.22488
## post_floyd:as.factor(zcta)55403  9.617e-01  1.896e+00   0.507  0.61197
## post_floyd:as.factor(zcta)55404  4.076e+00  1.896e+00   2.150  0.03159 *
## post_floyd:as.factor(zcta)55405  2.444e+00  1.896e+00   1.289  0.19739
```

```

## post_floyd:as.factor(zcta)55406 6.425e-01 1.896e+00 0.339 0.73468
## post_floyd:as.factor(zcta)55407 1.107e+00 1.896e+00 0.584 0.55917
## post_floyd:as.factor(zcta)55408 8.138e-01 1.896e+00 0.429 0.66772
## post_floyd:as.factor(zcta)55409 2.695e-01 1.896e+00 0.142 0.88697
## post_floyd:as.factor(zcta)55410 4.661e-01 1.896e+00 0.246 0.80578
## post_floyd:as.factor(zcta)55411 8.509e+00 1.896e+00 4.489 7.27e-06 ***
## post_floyd:as.factor(zcta)55412 4.383e+00 1.896e+00 2.312 0.02080 *
## post_floyd:as.factor(zcta)55413 1.192e+00 1.896e+00 0.629 0.52961
## post_floyd:as.factor(zcta)55414 1.001e+00 1.896e+00 0.528 0.59756
## post_floyd:as.factor(zcta)55415 5.063e+00 1.896e+00 2.671 0.00759 **
## post_floyd:as.factor(zcta)55416 5.031e-01 1.896e+00 0.265 0.79071
## post_floyd:as.factor(zcta)55417 1.390e+00 1.896e+00 0.733 0.46354
## post_floyd:as.factor(zcta)55418 7.871e-01 1.896e+00 0.415 0.67800
## post_floyd:as.factor(zcta)55419 9.939e-01 1.896e+00 0.524 0.60009
## post_floyd:as.factor(zcta)55421 1.221e+00 1.896e+00 0.644 0.51949
## post_floyd:as.factor(zcta)55422 1.146e+00 1.896e+00 0.605 0.54540
## post_floyd:as.factor(zcta)55423 9.760e-01 1.896e+00 0.515 0.60669
## post_floyd:as.factor(zcta)55424 5.303e-01 1.896e+00 0.280 0.77971
## post_floyd:as.factor(zcta)55429 1.259e+00 1.896e+00 0.664 0.50678
## post_floyd:as.factor(zcta)55430 3.680e+00 1.896e+00 1.941 0.05225 .
## post_floyd:as.factor(zcta)55450 5.414e-01 1.897e+00 0.285 0.77536
## post_floyd:as.factor(zcta)55454 7.463e-01 1.896e+00 0.394 0.69384
## post_floyd:as.factor(zcta)55455 5.303e-01 1.896e+00 0.280 0.77971
## post_floyd_3:as.factor(zcta)55402 6.514e-14 2.454e+00 0.000 1.00000
## post_floyd_3:as.factor(zcta)55403 1.419e-01 2.454e+00 0.058 0.95390
## post_floyd_3:as.factor(zcta)55404 -2.237e+00 2.454e+00 -0.912 0.36196
## post_floyd_3:as.factor(zcta)55405 -1.892e+00 2.454e+00 -0.771 0.44074
## post_floyd_3:as.factor(zcta)55406 -2.601e-01 2.454e+00 -0.106 0.91558
## post_floyd_3:as.factor(zcta)55407 -2.324e-01 2.454e+00 -0.095 0.92454
## post_floyd_3:as.factor(zcta)55408 7.504e-02 2.454e+00 0.031 0.97561
## post_floyd_3:as.factor(zcta)55409 5.140e-14 2.454e+00 0.000 1.00000
## post_floyd_3:as.factor(zcta)55410 5.021e-14 2.454e+00 0.000 1.00000
## post_floyd_3:as.factor(zcta)55411 -5.421e+00 2.454e+00 -2.209 0.02718 *
## post_floyd_3:as.factor(zcta)55412 -1.411e+00 2.454e+00 -0.575 0.56540
## post_floyd_3:as.factor(zcta)55413 -2.226e-01 2.454e+00 -0.091 0.92772
## post_floyd_3:as.factor(zcta)55414 -5.965e-01 2.454e+00 -0.243 0.80794
## post_floyd_3:as.factor(zcta)55415 -3.626e+00 2.454e+00 -1.478 0.13949
## post_floyd_3:as.factor(zcta)55416 5.801e-14 2.454e+00 0.000 1.00000
## post_floyd_3:as.factor(zcta)55417 -1.074e+00 2.454e+00 -0.438 0.66150
## post_floyd_3:as.factor(zcta)55418 -1.048e-01 2.454e+00 -0.043 0.96595
## post_floyd_3:as.factor(zcta)55419 -5.108e-01 2.454e+00 -0.208 0.83509
## post_floyd_3:as.factor(zcta)55421 -8.608e-01 2.454e+00 -0.351 0.72574
## post_floyd_3:as.factor(zcta)55422 -2.660e-02 2.454e+00 -0.011 0.99135
## post_floyd_3:as.factor(zcta)55423 -5.903e-01 2.454e+00 -0.241 0.80989
## post_floyd_3:as.factor(zcta)55424 5.047e-14 2.454e+00 0.000 1.00000
## post_floyd_3:as.factor(zcta)55429 9.103e-01 2.454e+00 0.371 0.71065
## post_floyd_3:as.factor(zcta)55430 -3.309e+00 2.454e+00 -1.349 0.17753
## post_floyd_3:as.factor(zcta)55450 5.857e-14 2.454e+00 0.000 1.00000
## post_floyd_3:as.factor(zcta)55454 -7.126e-01 2.454e+00 -0.290 0.77151
## post_floyd_3:as.factor(zcta)55455 6.736e-14 2.454e+00 0.000 1.00000
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.869 on 7206 degrees of freedom

```

```
## (10 observations deleted due to missingness)
## Multiple R-squared: 0.04495, Adjusted R-squared: 0.03289
## F-statistic: 3.727 on 91 and 7206 DF, p-value: < 2.2e-16
```

```
stargazer(ts_ar1_pol, fe_full_model,
  title = "Interrupted Time Series Models of Firearm Assault Injuries",
  covariate.labels = c("T", "COVID - State of Emergency", "COVID - Stay at Home",
    "Post-Killing", "Post-Killing 3 Months",
    "MPD Use of Force t-1", "MPD Stops t-1",
    "MPD Officer Involved Shootings t-1",
    "AR(1)"),
  dep.var.caption = "Firearm Assault Injuries",
  dep.var.labels = "Rate per 1,000",
  column.labels = c("Week-Level", "ZCTA-Week-Level"),
  model.numbers = TRUE,
  single.row = TRUE,
  align = T,
  omit = c("zcta", "tmax_f", "snow_in", "precip_in", "dark_before_12", "school"),
  omit.stat = "adj.rsq",
  star.cutoffs = c(.05, .01, .001), star.char = c(" ", "**", "***"),
  add.lines = list(c("ZCTA FE", "No", "Yes")),
  notes.label = "Models include controls for seasonality.")
```

% Table created by stargazer v.5.2.2 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu  
 % Date and time: Thu, Mar 03, 2022 - 4:22:18 PM % Requires LaTeX packages: dcolumn

Table 1: Interrupted Time Series Models of Firearm Assault Injuries

	Firearm Assault Injuries	
	Rate per 1,000	
	Week-Level	ZCTA-Week-Level
	(1)	(2)
T	0.00000 (0.00000)	0.002* (0.001)
COVID - State of Emergency	-0.003 (0.003)	-0.525 (0.569)
COVID - Stay at Home	0.003 (0.003)	0.221 (0.583)
Post-Killing	0.012*** (0.003)	0.974 (0.581)
Post-Killing 3 Months	-0.005** (0.002)	-0.345 (0.363)
MPD Use of Force t-1	-0.009 (0.008)	-0.051* (0.023)
MPD Stops t-1	0.001 (0.0005)	0.002 (0.003)
MPD Officer Involved Shootings t-1	-0.155 (0.223)	-0.294 (0.432)
AR(1)	0.183** (0.062)	
Constant	0.005 (0.004)	0.671 (0.845)
ZCTA FE	<i>No</i>	<i>Yes</i>
Observations	260	7,270
R <sup>2</sup>	0.384	0.036
Residual Std. Error	0.004 (df = 245)	4.884 (df = 7229)
F Statistic	10.889*** (df = 14; 245)	6.663*** (df = 40; 7229)

Models include controls for seasonality.

\*p<0.05; \*\*p<0.01; \*\*\*p<0.001

```

#maps of post_floyd and post_floyd_3 coefficients by zip - colored divergently
coef <- broom::tidy(fe_int_model$coefficients) %>%
  filter(str_detect(names, "post_floyd")) %>%
  mutate(period = ifelse(str_detect(names, "post_floyd_3"), "3+ Months Post-Killing", "0-3 Months Post-
    main_effect = ifelse(period=="3+ Months Post-Killing", round(-0.0301406,2), round(-0.5604477,2),
    zcta = as.numeric(str_sub(names, -5)),
    zcta = as.numeric(ifelse(is.na(zcta), "55401", zcta)),
    interaction_effect = ifelse(zcta=="55401", 0, round(x,2)),
    coef = main_effect+interaction_effect) %>%
  select(zcta, period, coef, main_effect, interaction_effect) %>%
  arrange(zcta, period)

```

```

## Warning: 'tidy.numeric' is deprecated.
## See help("Deprecated")

```

```

## Warning: 'data_frame()' was deprecated in tibble 1.1.0.
## Please use 'tibble()' instead.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_lifecycle_warnings()' to see where this warning was generated.

```

```

## Warning in mask$eval_all_mutate(quo): NAs introduced by coercion

```

```

#creating period rows in other spatial layers
coef_zip_level <- zip_level %>%
  filter(period!="Pre-Killing") %>%
  left_join(coef, by = c("zcta", "period"))
coef_gfs <- gfs
coef_gfs[2,] <- gfs[1,]
coef_gfs$period <- c("3+ Months Post-Killing", "0-3 Months Post-Killing")
coef_mpls <- mpls
coef_mpls[2,] <- mpls[1,]
coef_mpls$period <- c("3+ Months Post-Killing", "0-3 Months Post-Killing")

ggplot() +
  geom_sf(data = coef_zip_level, aes(geometry = geometry, fill = coef)) +
  geom_sf(data = mpls, aes(geometry = geometry), color = "black", alpha = 0)+
  geom_sf(data = coef_gfs, aes(geometry = geometry), color = "black")+
  geom_text_repel(data = gfs, aes(x=lon, y=lat, label = name),
    size = 2,
    fontface = "bold",
    nudge_x = 1, nudge_y = -1)+
  scale_fill_gradient2(trans="reverse")+
  facet_wrap(~period)+
  labs(title = "Figure 3: Treatment Effects by ZCTA",
    fill = "Coef.")+
  theme(axis.text = element_blank(),
    axis.line = element_blank(),
    axis.ticks = element_blank(),
    panel.border = element_blank(),
    panel.grid = element_blank(),
    axis.title = element_blank(),

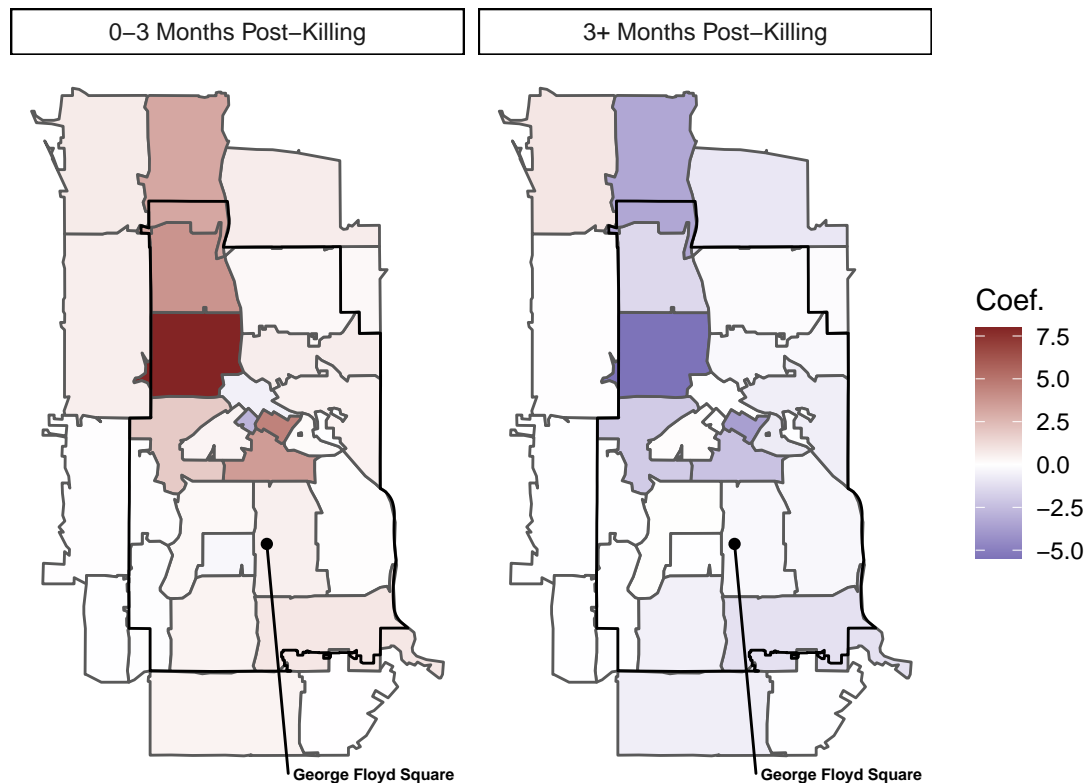
```

```

panel.background = element_blank(),
panel.grid.major = element_line(colour="transparent"),
plot.subtitle = element_text(face="italic"),
strip.background = element_rect(fill = "white",
                                colour = "black"))+
guides(fill = guide_colorbar(reverse = TRUE))

```

Figure 3: Treatment Effects by ZCTA



## MPD Murders: Figures 4 and 5

```

#pre-pims
mpd_2016 <- read_csv("Data/Police_Incidents_2016.csv")
mpd_2017 <- read_csv("Data/Police_Incidents_2017.csv")
mpd_2018a <- read_csv("Data/Police_Incidents_2018.csv")

#pims
mpd_2018b <- read_csv("Data/Police_Incidents_2018_PIMS.csv")
mpd_2019 <- read_csv("Data/Police_Incidents_2019.csv")
mpd_2020 <- read_csv("Data/Police_Incidents_2020.csv")
mpd_2021 <- read_csv("Data/Police_Incidents_2021.csv")

pre_pims_base <- mpd_2016 %>%
  rbind(mpd_2017) %>%

```

```

rbind(mpd_2018a) %>%
  rename(reportedDate = ReportedDate,
         centerLong = Long,
         centerLat = Lat) %>%
  select(FID, centerLong, centerLat, Offense, reportedDate) %>%
  rename(OBJECTID = FID,
         X = centerLong,
         Y = centerLat,
         offense = Offense)

post_pims_base <- mpd_2018b %>%
  rbind(mpd_2019) %>%
  rbind(mpd_2020) %>%
  rbind(mpd_2021) %>%
  select(OBJECTID, X, Y, offense, reportedDate)

mpd <- pre_pims_base %>%
  rbind(post_pims_base)

mpd_series <- mpd %>%
  mutate(date=ymd_hms(reportedDate),
         year=isoyear(date),
         week=isoweek(date)) %>%
  st_as_sf(coords = c("X", "Y"), crs = "NAD83", remove=F) %>%
  mutate(intersection = as.integer(st_intersects(geometry, zcta)),
         zcta = ifelse(is.na(intersection), NA, zcta$zcta[intersection])) %>%
  st_drop_geometry() %>%
  filter(offense=="MURDR" & zcta %in% zcta_universe) %>%
  group_by(year, week, .drop=F) %>%
  tally(name = "murder") %>%
  arrange(year, week) %>%
  filter(year <= 2021 & year >= 2016) %>%
  ungroup() %>%
  complete(year, week = 1:52, fill = list(murder = 0))

mpls_pops_year <- series %>%
  group_by(year) %>%
  summarize(total_pop = mean(total_pop, na.rm = T)) %>%
  add_row(year = 2021, total_pop = 603465)

mpd_series <- mpd_series %>%
  left_join(mpls_pops_year, by = "year") %>%
  mutate(murder_rate = (murder/total_pop)*1000,
         begin_date = ISOweek2date(paste(year, paste0("W", sprintf("%02d", week))), 1, sep = "-"),
         end_date = begin_date+weeks(1)-days(1))

ggplot(mpd_series)+
  geom_line(aes(x=begin_date, y=murder_rate))+
  scale_x_date(date_labels = "%b-%Y", date_breaks = "6 months")+
  geom_vline(xintercept=mpd_series$begin_date[mpd_series$year==2020 & mpd_series$week==isoweek(date("2020-05-01"))],
            linetype="dotted", color="red", size=1)+
  geom_label(aes(x=mpd_series$begin_date[mpd_series$year==2020 & mpd_series$week==isoweek(date("2020-05-01"))],
                y=murder_rate[mpd_series$year==2020 & mpd_series$week==isoweek(date("2020-05-01"))],
                text="2020-05-01"))

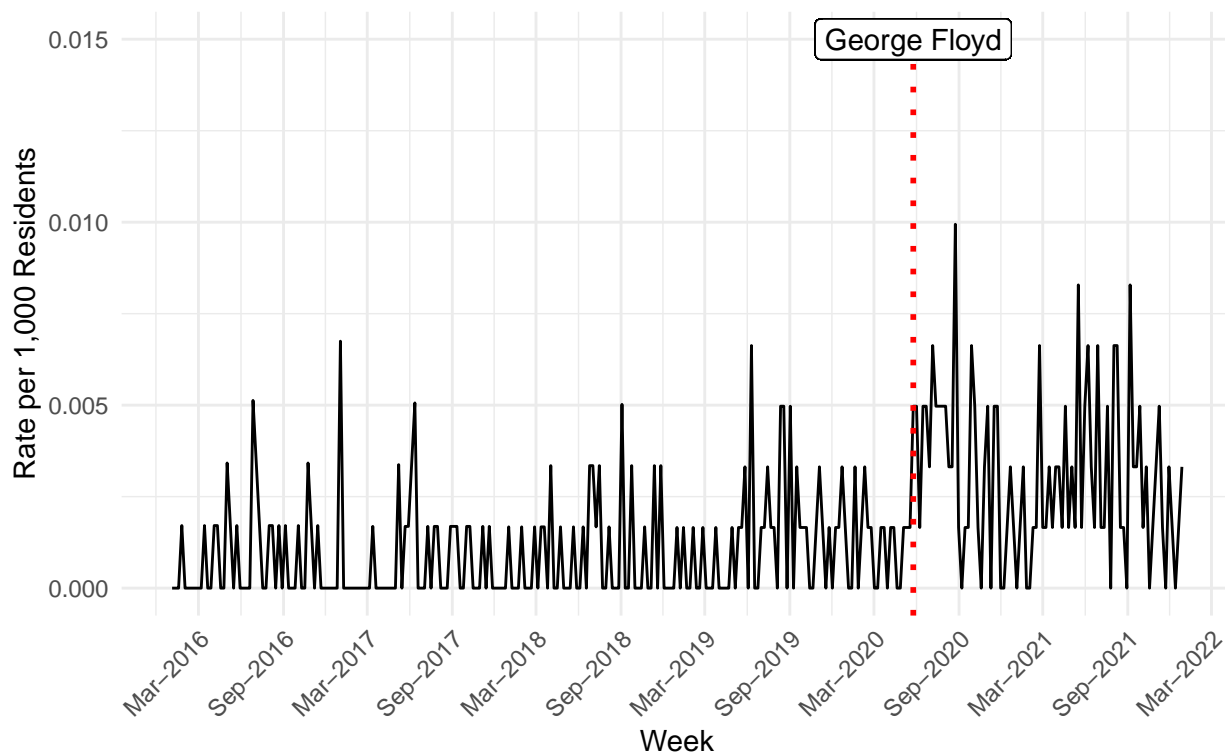
```

```

      y=0.0150),
      label = "George Floyd", show.legend = FALSE)+
scale_y_continuous(limits = c(0,.015))+
labs(title = "Figure 4: Weekly Murder Rate, 2016-2021",
      subtitle = "MPD Data",
      x = "Week",
      y = "Rate per 1,000 Residents")+
theme_minimal()+
  theme(axis.text.x=element_text(angle=45, hjust=1))

```

Figure 4: Weekly Murder Rate, 2016–2021  
MPD Data



```

mpd_series <- mpd_series %>%
  mutate(csma = forecast::ma(murder_rate, order=5,centre=TRUE),
         tsma = TTR::SMA(murder_rate, n=5))

ggplot(mpd_series)+
  geom_line(aes(x=begin_date, y=murder_rate))+
  scale_x_date(date_labels = "%b-%Y", date_breaks = "6 months")+
  geom_vline(xintercept=mpd_series$begin_date[mpd_series$year==2020 & mpd_series$week==isoweek(date("2020-05-06"))],
             linetype="dotted", color="red", size=1)+
  geom_label(aes(x=mpd_series$begin_date[mpd_series$year==2020 & mpd_series$week==isoweek(date("2020-05-06"))],
                y=0.0150),
             label = "George Floyd", show.legend = FALSE)+
  scale_y_continuous(limits = c(0,.015))+
  labs(title = "Figure 4: Weekly Murder Rate, 2016-2021",
       subtitle = "MPD Data",

```

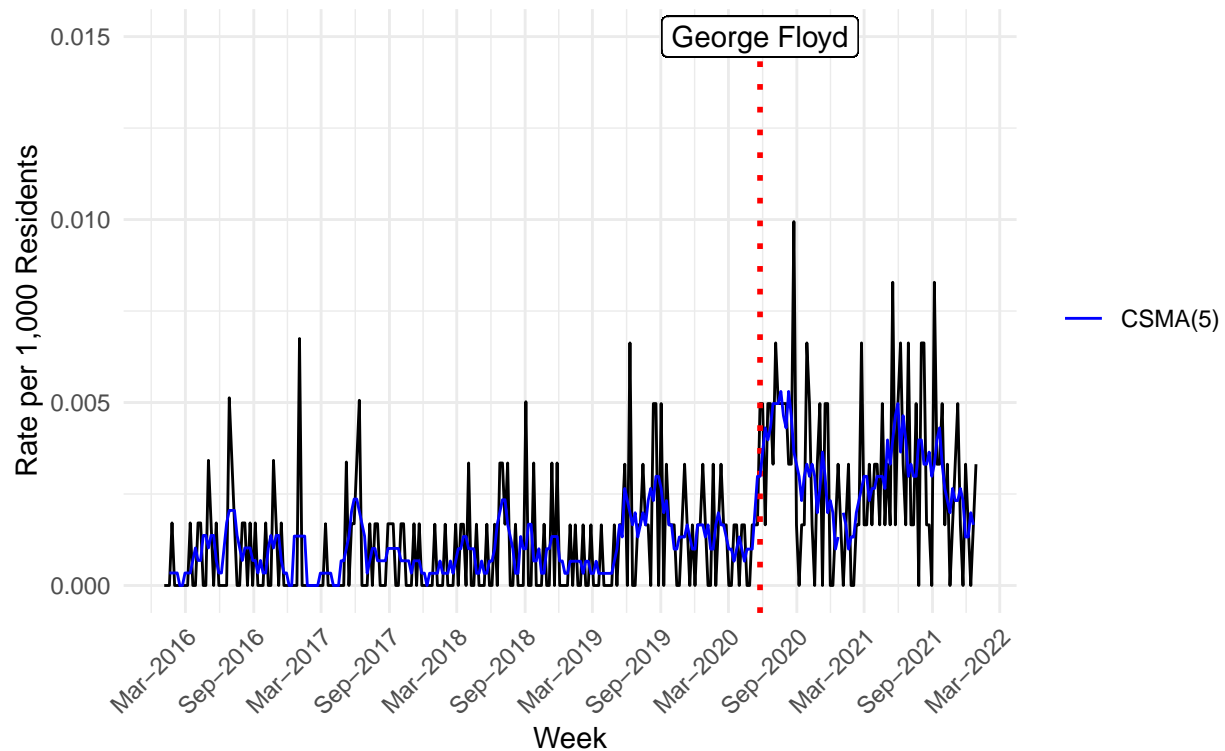


```

x = "Week",
y = "Rate per 1,000 Residents",
color = NULL)+
theme_minimal()+
theme(axis.text.x=element_text(angle=45, hjust=1)) +
geom_line(aes(x=begin_date, y=csma, color = "CSMA(5)"))+
#geom_line(aes(x=begin_date, y=tsma, color = "TSMA(5)"))+
#geom_ma(aes(x = begin_date, y = murder_rate, color = "MA4"), ma_fun = SMA, n = 4)
scale_color_manual(values = c("blue", "green"))

```

Figure 4: Weekly Murder Rate, 2016–2021  
MPD Data



```

mpls_pops_zcta <- panel %>%
  select(zcta, year, weekofyr, total_pop) %>%
  st_drop_geometry() %>%
  ungroup() %>%
  complete(zcta, year = 2016:2021, weekofyr = 1:52) %>%
  arrange(zcta, year, weekofyr) %>%
  mutate(total_pop = ifelse(is.na(total_pop), na_locf(total_pop), total_pop)) %>%
  group_by(zcta) %>%
  summarize(total_pop = mean(total_pop, na.rm = T))

mpd_zip <- mpd %>%
  mutate(date=ymd_hms(reportedDate),
         year=isoyear(date),
         week=isoweek(date)) %>%

```

```

st_as_sf(coords = c("X", "Y"), crs = "NAD83", remove=F) %>%
mutate(intersection = as.integer(st_intersects(geometry, zcta)),
      zcta = ifelse(is.na(intersection), NA, zcta$zcta[intersection])) %>%
st_drop_geometry() %>%
filter(offense=="MURDR" & zcta %in% zcta_universe) %>%
group_by(year, week, zcta, .drop=F) %>%
tally(name = "murder") %>%
arrange(zcta, year, week) %>%
ungroup() %>%
complete(year, week=1:52, zcta=zcta_universe, fill = list(murder = 0)) %>%
filter(year <= 2021 & year >= 2016) %>%
mutate(begin_date = ISOweek2date(paste(year, paste0("W", sprintf("%02d", week)), 1, sep = "-")),
      end_date = begin_date+weeks(1)-days(1),
      post_floyd = as.numeric(begin_date >= as.Date("2020-05-25")),
      post_floyd_3 = as.numeric(begin_date >= as.Date("2020-05-25")+months(3)),
      period = factor(case_when(
        post_floyd==0 & post_floyd_3==0 ~ "Pre-Killing",
        post_floyd>=1 & post_floyd_3==0 ~ "0-3 Months Post-Killing",
        post_floyd>=1 & post_floyd_3>=1 ~ "3+ Months Post-Killing",
        levels = c("Pre-Killing", "0-3 Months Post-Killing", "3+ Months Post-Killing")))) %>%
left_join(mpls_pops_zcta, by = "zcta") %>%
group_by(period, zcta, .drop=F) %>%
summarize(murder = mean(murder, na.rm = T),
          total_pop = mean(total_pop, na.rm = T)) %>%
left_join(zcta, by = "zcta") %>%
mutate(murder_rate = (murder/total_pop)*1000)

```

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

```

ggplot() +
  geom_sf(data = mpd_zip, aes(geometry = geometry, fill = murder_rate)) +
  geom_sf(data = mpls, aes(geometry = geometry), color = "black", alpha = 0)+
  geom_sf(data = gfs, aes(geometry = geometry), color = "black")+
  geom_text_repel(data = gfs, aes(x=lon, y=lat, label = name),
    size = 2,
    fontface = "bold",
    nudge_x = 1, nudge_y = -1)+
  facet_wrap(~period)+
  scale_fill_distiller(palette = "Spectral")+
  labs(title = "Figure 5: Murder Rates by ZCTA and Period",
    subtitle = "MPD Data",
    fill = "Murder Rate/1,000")+
  theme(axis.text = element_blank(),
    axis.line = element_blank(),
    axis.ticks = element_blank(),
    panel.border = element_blank(),
    panel.grid = element_blank(),
    axis.title = element_blank(),
    panel.background = element_blank(),
    panel.grid.major = element_line(colour="transparent"),
    plot.subtitle = element_text(face="italic"),
    strip.background = element_rect(fill = "white",

```

```
colour = "black"))
```

Figure 5: Murder Rates by ZCTA and Period

MPD Data

