## Preliminary Analyses

```
# loading census data
az_census <- read.csv("data/az_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group by(ctvname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac male) + sum(nac female))/tot pop,
           hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  mutate(state = "AZ")
ak_census <- read.csv("data/ak_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group_by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "AK")
ct_census <- read.csv("data/ct_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group_by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h male) + sum(h female))/tot pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
```

```
select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "CT")
fl_census <- read.csv("data/fl_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
           hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "FL")
mo census <- read.csv("data/mo census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "MO")
nc_census <- read.csv("data/nc_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
           na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
```

```
ungroup() %>%
  mutate(state = "NC")
nd_census <- read.csv("data/nd_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group_by(ctyname) %>%
  mutate(tot pop = sum(tot pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
           na = (sum(nac_male) + sum(nac_female))/tot_pop,
           hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "ND")
ny_census <- read.csv("data/ny_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group_by(ctyname) %>%
  mutate(tot pop = sum(tot pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac male) + sum(bac female))/tot pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
           na = (sum(nac_male) + sum(nac_female))/tot_pop,
           hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "NY")
or_census <- read.csv("data/or_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group_by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "OR")
```

```
pa_census <- read.csv("data/pa_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac male) + sum(iac female))/tot pop,
            asian = (sum(aac male) + sum(aac female))/tot pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "PA")
sd_census <- read.csv("data/sd_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "SD")
ut_census <- read.csv("data/ut_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "UT")
wa_census <- read.csv("data/wa_census.csv") %>%
  clean_names() %>%
```

```
filter(year == 6) %>%
  group_by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac male) + sum(nac female))/tot pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "WA")
wi_census <- read.csv("data/wi_census.csv") %>%
  clean_names() %>%
  filter(year == 6) %>%
  group_by(ctyname) %>%
  mutate(tot_pop = sum(tot_pop)) %>%
  mutate(white = (sum(wa_male) + sum(wa_female))/tot_pop,
            black = (sum(bac_male) + sum(bac_female))/tot_pop,
            ia = (sum(iac_male) + sum(iac_female))/tot_pop,
            asian = (sum(aac_male) + sum(aac_female))/tot_pop,
            na = (sum(nac_male) + sum(nac_female))/tot_pop,
            hisp = (sum(h_male) + sum(h_female))/tot_pop,
         hba = (sum(hbac_male) + sum(hbac_female))/tot_pop) %>%
  select(stname, ctyname, tot_pop, white, black, ia, asian, na, hisp, hba) %>%
  unique() %>%
  ungroup() %>%
  mutate(state = "WI")
prosecutors <- readxl::read_xlsx("data/Prosecutor-Elected-Officials-2019.xlsx") %>%
  clean names() %>%
  filter(state %in% c("AZ", "AK", "CT", "FL", "MO", "NY", "NC", "ND", "OR", "PA", "SD", "UT", "WA", "WI
  mutate(ctyname = ifelse(state == "PA", paste(electoral_district, "County"), paste(electoral_district)
  select(state, ctyname, office_name, official_name, party_roll_up, white_non_white, race, gender)
## measures for justice
az_loc <- read.csv("data/Arizona_State_Data/locations-az.csv")</pre>
ak_loc <- read.csv("data/Arkansas_State_Data/locations-ar.csv")</pre>
ct_loc <- read.csv("data/Connecticut_State_Data/locations-ct.csv")</pre>
fl_loc <- read.csv("data/Florida_State_Data/locations-fl.csv")</pre>
mo_loc <- read.csv("data/Missouri_State_Data/locations-mo.csv")</pre>
ny_loc <- read.csv("data/New_York_State_Data/locations-ny.csv")</pre>
nc_loc <- read.csv("data/North_Carolina_State_Data/locations-nc.csv")</pre>
nd_loc <- read.csv("data/North_Dakota_State_Data/locations-nd.csv")</pre>
or_loc <- read.csv("data/Oregon_State_Data/locations-or.csv")</pre>
pa_loc <- read.csv("data/Pennsylvania_State_Data/locations-pa.csv")</pre>
sd_loc <- read.csv("data/South_Dakota_State_Data/locations-sd.csv")</pre>
ut_loc <- read.csv("data/Utah_State_Data/locations-ut.csv")</pre>
wa_loc <- read.csv("data/Washington_State_Data/locations-wa.csv")</pre>
wi loc <- read.csv("data/Wisconsin State Data/locations-wi.csv")</pre>
```

```
az_mfj <- read.csv("data/Arizona_State_Data/data-2011-2015-az.csv") %>%
  rename(id = location_id) %>%
 left join(az loc) %>%
 filter(type == "county") %>%
 mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "AZ") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left join(az census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
ak_mfj <- read.csv("data/Arkansas_State_Data/data-2011-2015-ar.csv") %>%
 rename(id = location_id) %>%
 left_join(ak_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
 mutate(state = "AK") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(ak_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
ct_mfj <- read.csv("data/Connecticut_State_Data/data-2011-2015-ct.csv") %>%
 rename(id = location_id) %>%
 left_join(ct_loc) %>%
 filter(type == "county") %>%
 mutate(ctyname = paste(name, "County")) %>%
 mutate(state = "CT") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left join(ct census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
fl_mfj <- read.csv("data/Florida_State_Data/data-2009-2013-fl.csv") %>%
 rename(id = location_id) %>%
 left_join(fl_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "FL") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(fl_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
```

```
mo_mfj <- read.csv("data/Missouri_State_Data/data-2011-2015-mo.csv") %>%
  rename(id = location_id) %>%
 left_join(mo_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "MO") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left join(mo census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
ny_mfj <- read.csv("data/New_York_State_Data/data-2011-2015-ny.csv") %>%
 rename(id = location_id) %>%
 left_join(ny_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "NY") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(ny_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
nc_mfj <- read.csv("data/North_Carolina_State_Data/data-2009-2013-nc.csv") %>%
 rename(id = location_id) %>%
 left_join(nc_loc) %>%
 filter(type == "county") %>%
 mutate(ctyname = paste(name, "County")) %>%
 mutate(state = "NC") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(nc_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
nd_mfj <- read.csv("data/North_Dakota_State_Data/data-2011-2015-nd.csv") %%
 rename(id = location_id) %>%
 left_join(nd_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "ND") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(nd_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
```

```
or_mfj <- read.csv("data/Oregon_State_Data/data-2011-2015-or.csv") %>%
  rename(id = location_id) %>%
 left_join(or_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "OR") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left join(or census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
pa_mfj <- read.csv("data/Pennsylvania_State_Data/data-2009-2013-pa.csv") %>%
 rename(id = location_id) %>%
 left_join(pa_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "PA") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(pa_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
sd_mfj <- read.csv("data/South_Dakota_State_Data/data-2013-2017-sd.csv") %>%
 rename(id = location_id) %>%
 left_join(sd_loc) %>%
 filter(type == "county") %>%
 mutate(ctyname = paste(name, "County")) %>%
 mutate(state = "SD") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(sd_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
ut_mfj <- read.csv("data/Utah_State_Data/data-2013-2017-ut.csv") %>%
 rename(id = location_id) %>%
 left_join(ut_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "UT") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(ut_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
```

```
wa_mfj <- read.csv("data/Washington_State_Data/data-2009-2013-wa.csv") %>%
  rename(id = location_id) %>%
 left_join(wa_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "WA") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left join(wa census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
wi_mfj <- read.csv("data/Wisconsin_State_Data/data-2011-2015-wi.csv") %>%
  rename(id = location_id) %>%
 left_join(wi_loc) %>%
 filter(type == "county") %>%
  mutate(ctyname = paste(name, "County")) %>%
  mutate(state = "WI") %>%
  select(state, ctyname, measure_id, filter_id, numerator, denominator, value) %>%
 left_join(wi_census)
## Joining, by = "id"
## Joining, by = c("state", "ctyname")
# binding all
full <- rbind(az_mfj, ak_mfj, ct_mfj, fl_mfj, mo_mfj, ny_mfj, nc_mfj, nd_mfj,</pre>
              or_mfj, pa_mfj, sd_mfj, ut_mfj, wa_mfj, wi_mfj) %>%
 mutate(value = as.numeric(value)) %>%
 left_join(prosecutors)
## Warning: NAs introduced by coercion
## Joining, by = c("state", "ctyname")
# filtering for measures of all defendants
all_defendants <- full %>%
 filter(filter_id == 1)
```

## Sentencing Length: Notes

The relationship between Black population share and length of imposed prison sentence for nonviolent felonies and misdemeanors is very statistically significant and very large.

```
## Length of Imposed Prison Sentence: Nonviolent Felonies ##

measure_26 <- all_defendants %>%
  filter(measure_id == 28) %>%
  filter(value > 0)
```

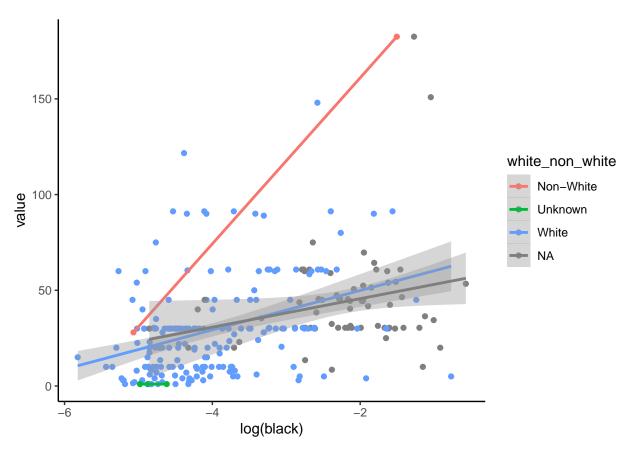
```
black_26 <- lm(value ~ black, data = measure_26)</pre>
summary(black_26)
##
## Call:
## lm(formula = value ~ black, data = measure_26)
## Residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
## -72.800 -18.299 -3.747
                             8.107 131.401
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) 26.502
                             2.026 13.079 < 2e-16 ***
                           19.468 5.687 3.51e-08 ***
## black
               110.710
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## Residual standard error: 26.53 on 257 degrees of freedom
     (1 observation deleted due to missingness)
## Multiple R-squared: 0.1118, Adjusted R-squared: 0.1083
## F-statistic: 32.34 on 1 and 257 DF, p-value: 3.514e-08
measure_26 %>%
  ggplot(aes(y = value, x = black, color = white_non_white)) +
  geom_point() +
  geom_smooth(method = "lm") +
 theme_classic()
## 'geom_smooth()' using formula 'y ~ x'
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
## Warning in qt((1 - level)/2, df): NaNs produced
## Warning: Removed 1 rows containing missing values (geom_point).
## Warning in max(ids, na.rm = TRUE): no non-missing arguments to max; returning -
## Inf
```

```
white_non_white
Non-White
Unknown
White
NA
```

```
measure_26 %>%
   ggplot(aes(y = value, x = log(black), color = white_non_white)) +
   geom_point() +
   theme_classic() +
   geom_smooth(method = "lm")
## 'geom_smooth()' using formula 'y ~ x'
```

```
## 'geom_smooth()' using formula 'y ~ x'
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
## Warning in qt((1 - level)/2, df): NaNs produced
## Warning: Removed 1 rows containing missing values (geom_point).
```

## Warning in max(ids, na.rm = TRUE): no non-missing arguments to max; returning ## Inf



```
## Length of Imposed Jail Sentence: Nonviolent Misdemeanors ##

measure_28 <- all_defendants %>%
    filter(measure_id == 28) %>%
    filter(value > 0)

black_28 <- lm(value ~ black, data = measure_28)
black_wpros_28_log <- lm(value ~ log(black)*white_non_white, data = measure_28) # huge distortions bc b
black_wpros_28 <- lm(value ~ black*white_non_white, data = measure_28) # huge distortions bc basically
summary(black_28)</pre>
```

```
##
## Call:
## lm(formula = value ~ black, data = measure_28)
##
## Residuals:
##
      Min
                                3Q
                1Q Median
                                       Max
##
   -72.800 -18.299 -3.747
                             8.107 131.401
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
## (Intercept)
                26.502
                             2.026 13.079 < 2e-16 ***
## black
                110.710
                            19.468
                                     5.687 3.51e-08 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
```

```
##
## Residual standard error: 26.53 on 257 degrees of freedom
     (1 observation deleted due to missingness)
## Multiple R-squared: 0.1118, Adjusted R-squared: 0.1083
## F-statistic: 32.34 on 1 and 257 DF, p-value: 3.514e-08
measure_28 %>%
  ggplot(aes(y = value, x = black, color = white_non_white)) +
  geom_point() +
  geom_smooth(method = "lm") +
  theme_classic()
## 'geom_smooth()' using formula 'y ~ x'
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
## Warning in qt((1 - level)/2, df): NaNs produced
## Warning: Removed 1 rows containing missing values (geom_point).
## Warning in max(ids, na.rm = TRUE): no non-missing arguments to max; returning -
## Inf
   150
                                                                         white_non_white
value
value
                                                                             Non-White
                                                                             Unknown
                                                                             White
                                                                             NA
    50
```

0.4

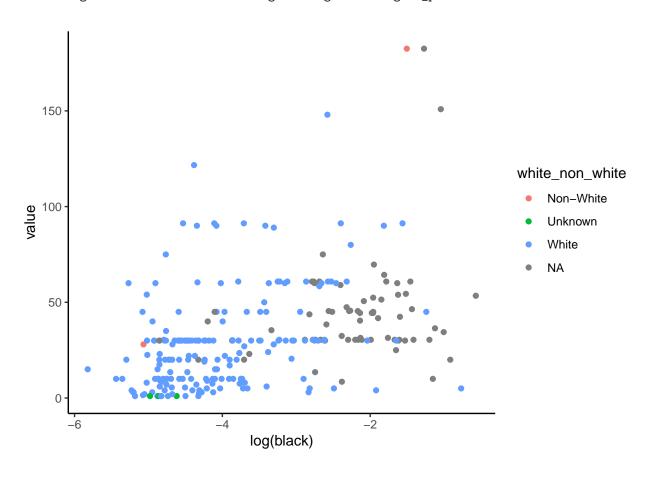
0.2

black

0.0

```
measure_28 %>%
   ggplot(aes(y = value, x = log(black), color = white_non_white)) +
   geom_point() +
   theme_classic()
```

## Warning: Removed 1 rows containing missing values (geom\_point).



## Gender

```
gender_26 <- lm(value ~ gender, data = measure_26)
summary(gender_26)</pre>
```

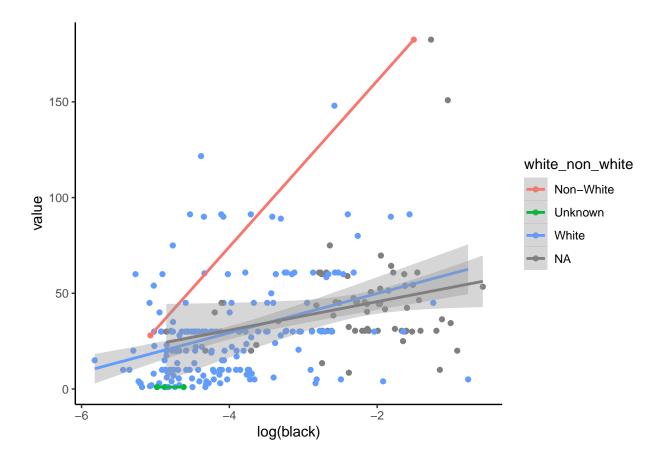
```
##
## Call:
## lm(formula = value ~ gender, data = measure_26)
##
## Residuals:
## Min    1Q Median   3Q Max
## -30.737 -20.161 -1.911   1.506 150.763
##
## Coefficients:
```

```
Estimate Std. Error t value Pr(>|t|)
##
                           3.902 8.133 5.54e-14 ***
## (Intercept)
                  31.737
                  -2.826
                              4.567 -0.619
## genderMale
                                               0.537
## genderUnknown -30.737
                             20.088 -1.530
                                               0.128
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 27.87 on 188 degrees of freedom
     (69 observations deleted due to missingness)
## Multiple R-squared: 0.01301,
                                   Adjusted R-squared: 0.00251
## F-statistic: 1.239 on 2 and 188 DF, p-value: 0.292
measure 26 %>%
 ggplot(aes(y = value, x = black, color = white_non_white)) +
 geom_point() +
 geom_smooth(method = "lm") +
theme_classic()
## 'geom_smooth()' using formula 'y ~ x'
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
## Warning in qt((1 - level)/2, df): NaNs produced
## Warning: Removed 1 rows containing missing values (geom_point).
## Warning in max(ids, na.rm = TRUE): no non-missing arguments to max; returning -
## Inf
```

```
white_non_white
Non-White
Unknown
White
NA
```

```
measure_26 %>%
  ggplot(aes(y = value, x = log(black), color = white_non_white)) +
  geom_point() +
  theme_classic() +
  geom_smooth(method = "lm")
```

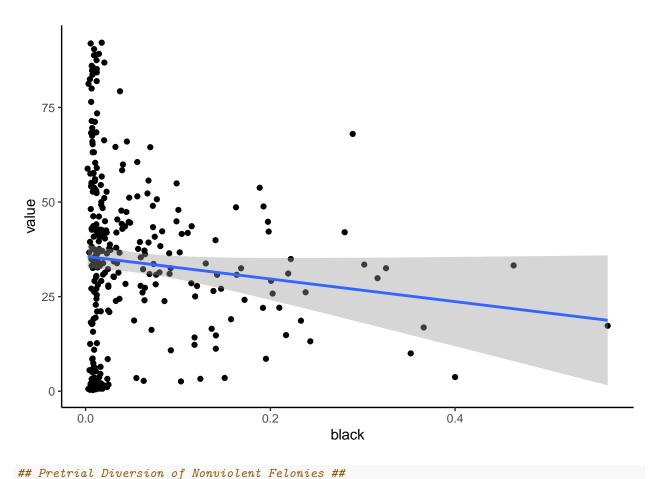
```
## 'geom_smooth()' using formula 'y ~ x'
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
## Warning in qt((1 - level)/2, df): NaNs produced
## Warning: Removed 1 rows containing missing values (geom_point).
## Warning in max(ids, na.rm = TRUE): no non-missing arguments to max; returning -
## Inf
```



## Other

```
# Nonviolent Misdemeanors Sentenced to Jail
measure_27 <- full %>%
  filter(filter_id == 1 & measure_id == 27) %>%
  filter(value > 0)
black_27 <- lm(value ~ black, data = measure_27)</pre>
black_wpros_27 <- lm(value ~ black*white_non_white, data = measure_27)</pre>
# not significant
summary(black_27)
##
## Call:
## lm(formula = value ~ black, data = measure_27)
##
## Residuals:
##
       Min
                1Q Median
                                 ЗQ
                                        Max
## -35.116 -17.140 -0.399 13.291 56.993
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
```

```
## (Intercept)
              35.647
                           1.602 22.256
                                            <2e-16 ***
## black
               -29.899
                           16.808 -1.779
                                            0.0762 .
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 23.57 on 308 degrees of freedom
    (1 observation deleted due to missingness)
## Multiple R-squared: 0.01017,
                                  Adjusted R-squared: 0.006956
## F-statistic: 3.164 on 1 and 308 DF, p-value: 0.07625
summary(black_wpros_27)
##
## Call:
## lm(formula = value ~ black * white_non_white, data = measure_27)
## Residuals:
     Min
             1Q Median
                           3Q
                                 Max
                 0.11 15.23 56.60
## -35.09 -20.83
## Coefficients:
##
                               Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                                 33.684
                                            18.557
                                                     1.815
                                                             0.0708 .
## black
                                  6.265
                                           144.577
                                                   0.043
                                                            0.9655
## white_non_whiteUnknown
                                -34.285
                                           25.317 -1.354
                                                            0.1770
                                            18.658
                                                             0.9375
## white_non_whiteWhite
                                  1.464
                                                     0.078
## black:white_non_whiteUnknown 1487.632
                                           809.307
                                                     1.838
                                                             0.0673 .
## black:white_non_whiteWhite
                                 24.775
                                           149.389
                                                     0.166
                                                            0.8684
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 25.56 on 236 degrees of freedom
    (69 observations deleted due to missingness)
## Multiple R-squared: 0.02222,
                                  Adjusted R-squared: 0.001508
## F-statistic: 1.073 on 5 and 236 DF, p-value: 0.3761
measure_27 %>%
 ggplot(aes(y = value, x = black)) +
 geom point() +
 geom_smooth(method = "lm") +
 theme_classic()
## 'geom_smooth()' using formula 'y ~ x'
## Warning: Removed 1 rows containing non-finite values (stat_smooth).
## Warning: Removed 1 rows containing missing values (geom_point).
```



```
measure_125 <- all_defendants %>%
  filter(measure_id == 125) %>%
  filter(value > 0)
black_125 <- lm(value ~ black, data = measure_125)</pre>
summary(black_125) # p = 0.069
##
## Call:
## lm(formula = value ~ black, data = measure_125)
##
## Residuals:
##
      Min
              1Q Median
                            ЗQ
                                  Max
## -7.861 -4.774 -1.181 3.239 24.439
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 8.1131
                            0.4945
                                      16.41
                                              <2e-16 ***
                -6.8711
                            3.7543
                                      -1.83
                                              0.0687 .
## black
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
```

## Residual standard error: 6.083 on 196 degrees of freedom

```
(1 observation deleted due to missingness)
## Multiple R-squared: 0.0168, Adjusted R-squared: 0.01179
## F-statistic: 3.35 on 1 and 196 DF, p-value: 0.06874
## Non-Custodial Promise to Appear Instead of Custodial Arrest ##
measure_1 <- all_defendants %>%
  filter(measure_id == 1) %>%
  filter(value > 0)
black_1 <- lm(value ~ black, data = measure_1)</pre>
summary(black_1) # not significant
##
## Call:
## lm(formula = value ~ black, data = measure_1)
## Residuals:
       Min
                 1Q Median
                                    3Q
## -25.1680 -11.0532 -0.0737 12.1084 29.0652
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                26.260
                            3.986
                                   6.588 1.92e-08 ***
                            21.541 -0.241
## black
                -5.184
                                              0.811
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## Residual standard error: 15.91 on 54 degrees of freedom
## Multiple R-squared: 0.001071, Adjusted R-squared: -0.01743
## F-statistic: 0.05792 on 1 and 54 DF, p-value: 0.8107
## Nonviolent Misdemeanor Cases with Nonmonetary Release ##
measure_7 <- all_defendants %>%
  filter(measure_id == 7) %>%
  filter(value > 0)
black_7 <- lm(value ~ black, data = measure_7)</pre>
summary(black_7) # not statistically significant
##
## Call:
## lm(formula = value ~ black, data = measure_7)
## Residuals:
##
       Min
               1Q Median
                                ЗQ
                                       Max
## -47.354 -6.806
                   3.272
                             9.713 20.595
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
```

```
## (Intercept) 72.556    1.377 52.709 <2e-16 ***
## black    5.920    19.646    0.301    0.764
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 13.15 on 131 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared: 0.0006927, Adjusted R-squared: -0.006936
## F-statistic: 0.0908 on 1 and 131 DF, p-value: 0.7636</pre>
```