

# Final Project

*Justin Hsie, Bingyu Sun, Eleanor Zhang, Annie Yu*

*12/15/2018*

## Data Import

```
cancer_raw =  
  read_csv("./data/Cancer_Registry.csv") %>%  
  janitor::clean_names() %>%  
  dplyr::select(target_death_rate, geography, everything()) %>%  
  separate(geography, into = c("county", "state"), sep = ",")
```

## Data variable dictionary:

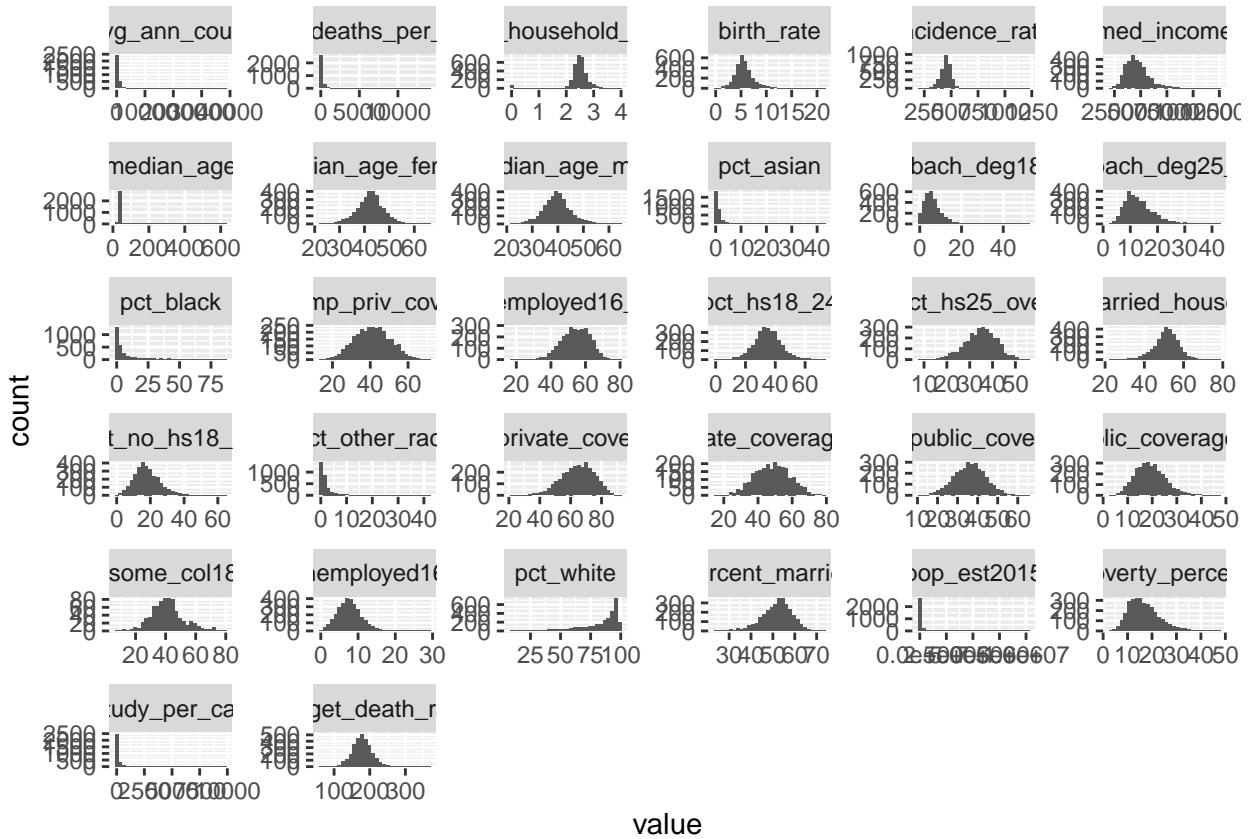
- **target\_death\_rate:** mean per capita (100,000) cancer mortalities (a)
- **avg\_ann\_count:** mean number of reported cases of cancer diagnosed annually (a)
- **avg\_deaths\_per\_year:** mean number of reported mortalities due to cancer (a)
- **incidence\_rate:** mean per capita (100,000) cancer diagnoses (a)
- **med\_income:** median income per county (b)
- **pop\_est2015:** population of county (b)
- **poverty\_percent:** percent of population in poverty (b)
- **study\_per\_cap** per capita number of cancer-related clinical trials per county (a)
- **binned\_inc:** median income per capita binned by decile (b)
- **median\_age:** median age of county residents (b)
- **median\_age\_male:** median age of male county residents (b)
- **median\_age\_female:** median age of female county residents (b)
- **geography:** county name (b)
- **avg\_household\_size:** mean household size of county (b)
- **percent\_married:** percent of county residents who are married (b)
- **pct\_no\_hs18\_24:** percent of county residents ages 18-24 highest education attained: less than high school (b)
- **pct\_hs18\_24:** percent of county residents ages 18-24 highest education attained: high school diploma (b)
- **pct\_some\_col18\_24:** percent of county residents ages 18-24 highest education attained: some college (b)
- **pct\_bach\_deg18\_24:** percent of county residents ages 18-24 highest education attained: bachelor's degree (b)
- **pct\_hs25\_over:** percent of county residents ages 25 and over highest education attained: high school diploma (b)
  
- **pct\_bach\_deg25\_over:** percent of county residents ages 25 and over highest education attained: bachelor's degree (b)
- **pct\_employed16\_over:** percent of county residents ages 16 and over employed (b)
  
- **pct\_unemployed16\_over:** percent of county residents ages 16 and over unemployed (b)
  
- **pct\_private\_coverage:** percent of county residents with private health coverage (b)
- **pct\_private\_coverage\_alone:** percent of county residents with private health coverage alone (no public assistance) (b)

- **pct\_emp\_priv\_coverage:** percent of county residents with employee-provided private health coverage (b)
- **pct\_public\_coverage:** percent of county residents with government-provided health coverage (b)
- **pct\_public\_coverage\_alone:** percent of county residents with government-provided health coverage alone (b)
- **pct\_white:** percent of county residents who identify as White (b)
- **pct\_black:** percent of county residents who identify as Black (b)
- **pct\_asian:** percent of county residents who identify as Asian (b)
- **pct\_other\_race:** percent of county residents who identify in a category which is not White, Black, or Asian (b)
- **pct\_married\_households:** percent of married households (b)
- **birth\_rate:** number of live births relative to number of women in county (b)

**Look at the distribution of all variables:**

```
cancer_raw %>%
  keep(is.numeric) %>%
  gather() %>%
  ggplot(aes(value)) +
    facet_wrap(~ key, scales = "free") +
    geom_histogram(bins = 30)

## Warning: Removed 3046 rows containing non-finite values (stat_bin).
```



Choose variables:

```
cancer_county =
  cancer_raw %>%
  janitor::clean_names() %>%
  dplyr::select(target_death_rate, incidence_rate, med_income, poverty_percent, median_age:median_age_fem)
  dplyr::select(-pct_hs25_over, -pct_bach_deg25_over, -pct_employed16_over, -percent_married) %>%
  mutate(pct_up_to_hs18_24 = pct_no_hs18_24 + pct_hs18_24,
    pct_above_hs18_24 = 100 - pct_up_to_hs18_24,
    pct_with_coverage = pct_private_coverage + pct_public_coverage_alone,
    income_cat = ifelse(med_income < 35000, 0, 1)) %>%
  dplyr::select(-(pct_no_hs18_24:pct_bach_deg18_24), -pct_above_hs18_24, -(pct_private_coverage:pct_pub))

```

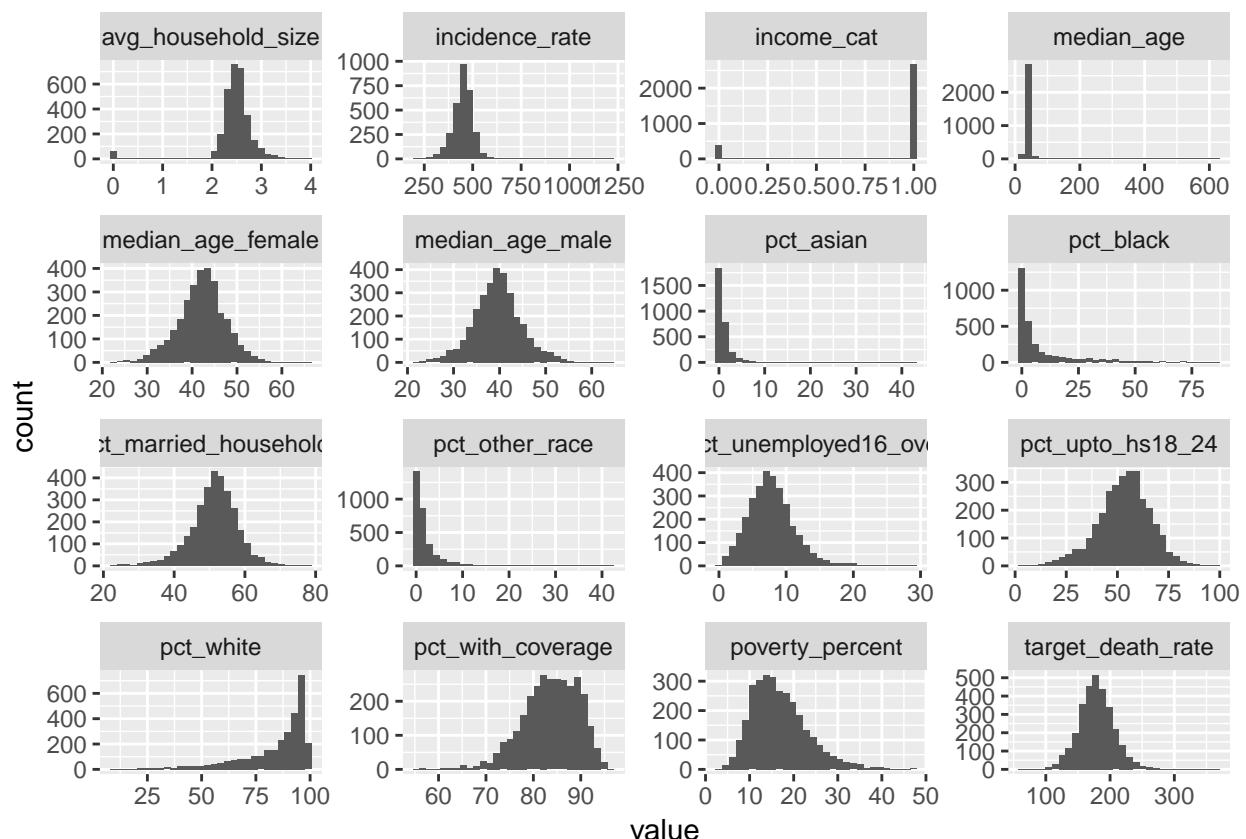
Check correlation and distribution:

```
cor(cancer_county) %>%
  knitr::kable()
```

	target_death_rate	incidence_rate	poverty_percent	median_age	median_age_male	med_income
target_death_rate	1.0000000	0.4494317	0.4293890	0.0043751	-0.0219294	
incidence_rate	0.4494317	1.0000000	0.0090463	0.0180892	-0.0147332	
poverty_percent	0.4293890	0.0090463	1.0000000	-0.0292800	-0.2140010	
median_age	0.0043751	0.0180892	-0.0292800	1.0000000	0.1291195	

	target_death_rate	incidence_rate	poverty_percent	median_age	median_age_male	median_age_female
median_age_male	-0.0219294	-0.0147332	-0.2140010	0.1291195	1.0000000	
median_age_female	0.0120484	-0.0091056	-0.1481635	0.1246784	0.9336961	
avg_household_size	-0.0369053	-0.1184000	0.0743076	-0.0319441	-0.3431887	
pct_unemployed16_over	0.3784124	0.0999795	0.6551481	0.0185904	-0.1427375	
pct_white	-0.1774000	-0.0145098	-0.5094328	0.0350094	0.3980444	
pct_black	0.2570236	0.1134890	0.5115297	-0.0171732	-0.2427481	
pct_asian	-0.1863311	-0.0081234	-0.1572887	-0.0384239	-0.2383224	
pct_other_race	-0.1898936	-0.2087483	0.0470959	-0.0302765	-0.2666554	
pct_married_households	-0.2933253	-0.1521763	-0.6049528	0.0145036	0.2222777	
pct_upto_hs18_24	0.2443042	-0.0929669	0.2517431	0.0401926	0.2371826	
pct_with_coverage	-0.2292798	0.2302489	-0.6516658	0.0049621	0.1497702	
income_cat	-0.3030288	0.0110839	-0.6344122	0.0103377	0.0266055	

```
cancer_county %>%
  gather() %>%
  ggplot(aes(value)) +
  facet_wrap(~ key, scales = "free") +
  geom_histogram(bins = 30)
```



## The descriptive statistics:

```
state_summary = function(x){
  mean = mean(x)
```

```

max = max(x)
min = min(x)
median = median(x)
var = var(x)
sd = sd(x)
sample_size = length(x) - sum(is.na(x))
tibble(mean, max, min, median, var, sd, sample_size)
}

#cancer_county %>% dplyr::select(-income_cat) %>% gather() %>% group_by(key) %>% nest() %>% mutate(summ

df_target_death_rate = state_summary(cancer_county$target_death_rate)
df_incidence_rate = state_summary(cancer_county$incidence_rate)
df_poverty_percent = state_summary(cancer_county$poverty_percent)
df_median_age = state_summary(cancer_county$median_age)
df_median_agemale = state_summary(cancer_county$median_age_male)
df_median_agedfemale = state_summary(cancer_county$median_age_female)
df_avg_household_size = state_summary(cancer_county$avg_household_size)
df_pct_unemployed16_over = state_summary(cancer_county$pct_unemployed16_over)
df_pct_white = state_summary(cancer_county$pct_white)
df_pct_black = state_summary(cancer_county$pct_black)
df_pct_asian = state_summary(cancer_county$pct_asian)
df_pct_other_race = state_summary(cancer_county$pct_other_race)
df_pct_married_households = state_summary(cancer_county$pct_married_households)
df_pct_upto_hs18_24 = state_summary(cancer_county$pct_upto_hs18_24)
df_pct_with_coverage = state_summary(cancer_county$pct_with_coverage)

state_des <- bind_rows(df_target_death_rate,
                       df_incidence_rate,
                       df_poverty_percent,
                       df_median_age,
                       df_median_agemale,
                       df_median_agedfemale,
                       df_avg_household_size,
                       df_pct_unemployed16_over,
                       df_pct_white,
                       df_pct_black,
                       df_pct_asian,
                       df_pct_other_race,
                       df_pct_married_households,
                       df_pct_upto_hs18_24,
                       df_pct_with_coverage)

variable = c("target_death_rate", "incidence_rate", "poverty_percent", "median_age", "median_agemale", "median_agedfemale", "avg_household_size", "pct_unemployed16_over", "pct_white", "pct_black", "pct_asian", "pct_other_race", "pct_married_households", "pct_upto_hs18_24", "pct_with_coverage")

state_wholedes = cbind(variable, state_des)

knitr::kable(state_wholedes)

```

variable	mean	max	min	median	var	sd	sample_size
target_death_rate	178.664063	362.80000	59.70000	178.100000	770.1463805	27.7515113	3047
incidence_rate	448.268586	1206.90000	201.30000	453.5494221	2976.8735549	54.5607327	3047
poverty_percent	16.878175	47.40000	3.20000	15.900000	41.0763948	6.4090869	3047
median_age	45.272333	624.00000	22.30000	41.000000	2052.4959218	45.3044802	3047

variable	mean	max	min	median	var	sd	sample_size
median_agemale	39.570725	64.70000	22.40000	39.600000	27.3112504	5.2260167	3047
median_agefemale	42.145323	65.70000	22.30000	42.400000	28.0142518	5.2928491	3047
avg_household_size	2.479662	3.97000	0.02210	2.5000000	0.1841906	0.4291744	3047
pct_unemployed16_over	7.852412	29.40000	0.40000	7.6000000	11.9188640	3.4523708	3047
pct_white	83.645286	100.00000	10.19916	90.0597742	268.3052265	16.3800252	3047
pct_black	9.107978	85.94780	0.00000	2.2475763	211.2527926	14.5345379	3047
pct_asian	1.253965	42.61942	0.00000	0.5498117	6.8135428	2.6102764	3047
pct_other_race	1.983523	41.93025	0.00000	0.8261852	12.3742846	3.5177101	3047
pct_married_households	51.243872	78.07540	22.99249	51.6699411	43.2018812	6.5728138	3047
pct_upto_hs18_24	53.226518	100.00000	4.80000	53.9000000	160.1814371	12.6562805	3047
pct_with_coverage	83.595011	95.70000	54.60000	84.0000000	35.3664623	5.9469709	3047

## Model building:

Stepwise:

```
# building full model
full_model <- lm(target_death_rate ~ ., data = cancer_county)
summary(full_model)

##
## Call:
## lm(formula = target_death_rate ~ ., data = cancer_county)
##
## Residuals:
##      Min       1Q   Median       3Q      Max 
## -129.883  -11.469    0.163   12.142  126.523 
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 72.058821 12.583888  5.726 1.13e-08 ***
## incidence_rate 0.219657  0.007547 29.104 < 2e-16 ***
## poverty_percent 1.036531  0.120008  8.637 < 2e-16 ***
## median_age     -0.004080  0.008394 -0.486 0.626969  
## median_age_male -0.241798  0.212259 -1.139 0.254726  
## median_age_female 0.046110  0.209093  0.221 0.825476  
## avg_household_size -0.117704  1.007617 -0.117 0.907015  
## pct_unemployed16_over 0.781391  0.153410  5.093 3.73e-07 ***
## pct_white        0.067342  0.058298  1.155 0.248127  
## pct_black         -0.016016  0.057020 -0.281 0.778818  
## pct_asian         -0.628303  0.178918 -3.512 0.000452 *** 
## pct_other_race   -0.940645  0.124524 -7.554 5.56e-14 *** 
## pct_married_households -0.220555  0.088361 -2.496 0.012611 *  
## pct_upto_hs18_24  0.392918  0.038241 10.275 < 2e-16 ***
## pct_with_coverage -0.194259  0.101672 -1.911 0.056145 .  
## income_cat        -4.104213  1.507502 -2.723 0.006516 ** 
## ---                
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 20.77 on 3031 degrees of freedom
## Multiple R-squared:  0.4427, Adjusted R-squared:  0.4399 
## F-statistic: 160.5 on 15 and 3031 DF,  p-value: < 2.2e-16
```

```

# Using the stepwise
stepwise_model = stepAIC(full_model, direction = "both", trace = FALSE)
summary(stepwise_model)

##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##     median_age_male + pct_unemployed16_over + pct_white + pct_asian +
##     pct_other_race + pct_married_households + pct_upto_hs18_24 +
##     pct_with_coverage + income_cat, data = cancer_county)
##
## Residuals:
##      Min        1Q    Median        3Q       Max
## -129.978   -11.453    0.127   12.095   126.386
##
## Coefficients:
##                               Estimate Std. Error t value Pr(>|t|)
## (Intercept)                70.812109  11.698781  6.053 1.60e-09 ***
## incidence_rate              0.219430   0.007509 29.223 < 2e-16 ***
## poverty_percent              1.038613   0.119140  8.718 < 2e-16 ***
## median_age_male             -0.202308   0.086539 -2.338 0.019465 *
## pct_unemployed16_over       0.774783   0.152081  5.095 3.71e-07 ***
## pct_white                   0.080440   0.034847  2.308 0.021043 *
## pct_asian                   -0.613378   0.167491 -3.662 0.000254 ***
## pct_other_race              -0.933301   0.119239 -7.827 6.84e-15 ***
## pct_married_households     -0.222728   0.086039 -2.589 0.009680 **
## pct_upto_hs18_24            0.394452   0.037753 10.448 < 2e-16 ***
## pct_with_coverage           -0.194337   0.099299 -1.957 0.050430 .
## income_cat                  -4.069227   1.500008 -2.713 0.006709 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 20.76 on 3035 degrees of freedom
## Multiple R-squared:  0.4426, Adjusted R-squared:  0.4406
## F-statistic: 219.1 on 11 and 3035 DF,  p-value: < 2.2e-16

vif(stepwise_model)

##
##          incidence_rate      poverty_percent      median_age_male
##                 1.186642                  4.122196                  1.446080
##          pct_unemployed16_over      pct_white      pct_asian
##                 1.948976                  2.303407                  1.351383
##          pct_other_race pct_married_households      pct_upto_hs18_24
##                 1.243892                  2.261059                  1.614089
##          pct_with_coverage      income_cat
##                 2.465515                  1.729209

# Cp and AIC and Adjusted R2
model_dig = glance(stepwise_model) %>%
  as.data.frame() %>%
  dplyr::select(adj.r.squared, sigma, p.value, AIC, BIC) %>%
  rename(RES = sigma) %>%
  mutate(cp = ols_mallows_cp(stepwise_model, full_model))

model_dig

```

```

##   adj.r.squared      RES p.value      AIC      BIC      cp
## 1      0.4405848 20.75648      0 27143.23 27221.51 8.35992

```

## Cross validation and Criterion method summary

```

cross_df = crossv_mc(cancer_county, n = 100, test = 0.2)

cross_result =
  cross_df %>%
  mutate(
    step_mod = map(train, ~lm(target_death_rate ~ incidence_rate + poverty_percent +
    median_age_male + pct_unemployed16_over + pct_white + pct_asian +
    pct_other_race + pct_married_households + pct_upto_hs18_24 +
    pct_with_coverage + income_cat, data = .x)),
    rmse_train = map2_dbl(step_mod, train, ~rmse(model = .x, data = .y)),
    rmse_test = map2_dbl(step_mod, test, ~rmse(model = .x, data = .y))
  )

mse_results = cross_result %>%
  dplyr::select(rmse_train, rmse_test) %>%
  summarize(mse_train = (mean(rmse_train))^2,
            mse_test = (mean(rmse_test))^2) #mse results

#LOOCV
glm.fit = glm(target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + pct_unemployed16_over + pct_white + pct_asian +
  pct_other_race + pct_married_households + pct_upto_hs18_24 +
  pct_with_coverage + income_cat, data = cancer_county)

cv.err = cv.glm(cancer_county, glm.fit)

# The two delta values should be similar: we use the first one
# The second value is bias corrected
cv.err$delta #434.8355

## [1] 434.8355 434.8346
anova(stepwise_model)[12, 3] #MSE: 431

## [1] 430.8316

tibble(
  mse_model = anova(stepwise_model)[12, 3],
  mse_LOOCV = cv.err$delta[1],
  mse_CV_train = mse_results$mse_train,
  mse_CV_test = mse_results$mse_test
)

## # A tibble: 1 x 4
##   mse_model mse_LOOCV mse_CV_train mse_CV_test
##       <dbl>     <dbl>        <dbl>        <dbl>
## 1      431.      435.        430.        429.

```

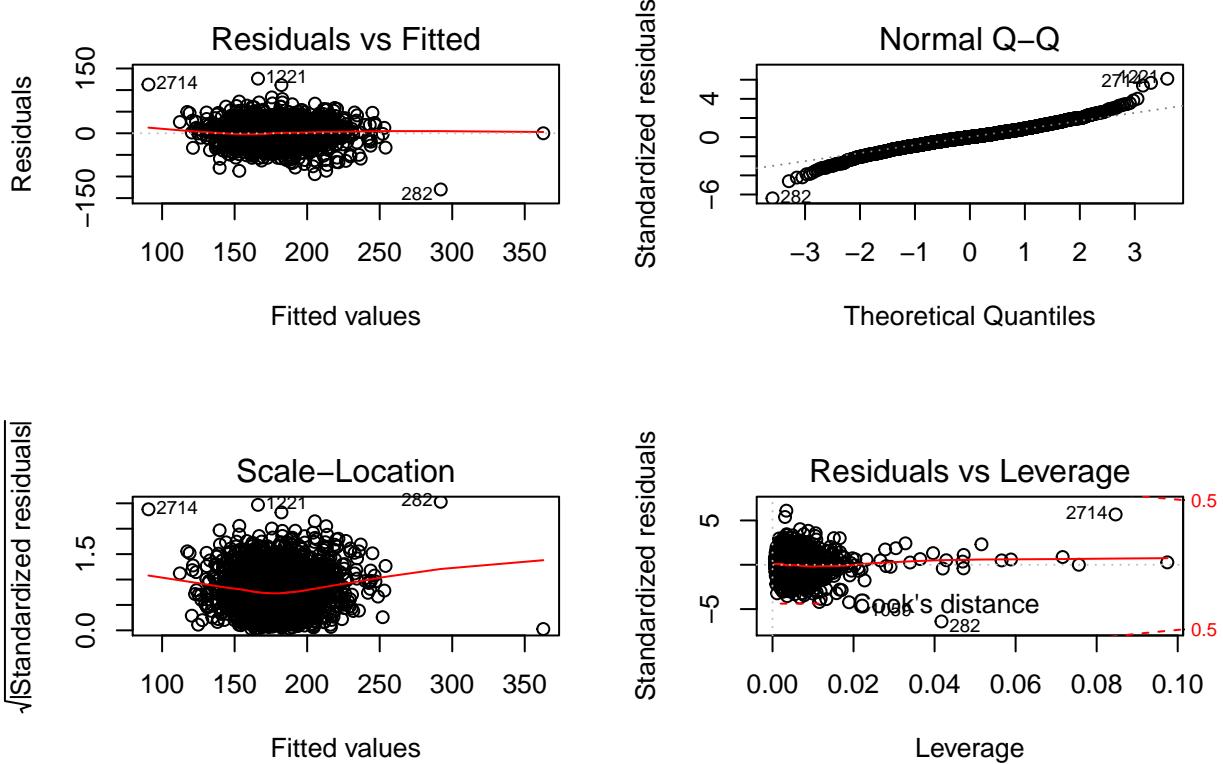
## Bootstrap

```
cancer_county %>%
  modelr::bootstrap(n = 1000) %>%
  mutate(models = map(strap, ~lm(target_death_rate ~ incidence_rate + poverty_percent +
    median_age_male + pct_unemployed16_over + pct_white + pct_asian +
    pct_other_race + pct_married_households + pct_upto_hs18_24 +
    pct_with_coverage + income_cat, data = .x)),
    results = map(models, broom::tidy)) %>%
  dplyr::select(-strap, -models) %>%
  unnest() %>%
  group_by(term) %>%
  summarize(boot_se = sd(estimate))

## # A tibble: 12 x 2
##   term          boot_se
##   <chr>        <dbl>
## 1 (Intercept) 15.3
## 2 incidence_rate 0.0124
## 3 income_cat    1.72
## 4 median_age_male 0.0963
## 5 pct_asian     0.268
## 6 pct_married_households 0.101
## 7 pct_other_race 0.151
## 8 pct_unemployed16_over 0.190
## 9 pct_upto_hs18_24 0.0430
## 10 pct_white    0.0411
## 11 pct_with_coverage 0.124
## 12 poverty_percent 0.143
```

## Outliers

```
par(mfrow = c(2,2))
plot(stepwise_model) # we observe the 282, 1059 are outliers
```



```
# Using the studentized residuals
```

```
stu_res_step <- rstandard(stepwise_model)
outliers_y_step = stu_res_step[abs(stu_res_step) > 2.5]
outliers_y_step
```

```
##      116      119      120      122      124      166      189
## 3.832977 -2.987984 -2.938723  3.451322 -3.218087  3.004467 -2.722478
##      209      250      254      256      264      282      522
## 2.634644  2.588483  3.557345 -2.852694 -3.008080 -6.396845  2.942781
##      524      621      627      666      775      783      803
## 2.568705 -2.758364  3.045997  2.920921  3.334436 -2.537021 -2.663472
##      812      845      912      921      979     1048     1058
## -3.536931 -3.028088 -3.843379 -3.048353  2.507593 -2.742014 -2.740598
##     1059     1066     1076     1094     1217     1221     1261
## -4.608218 -2.851923  2.589435 -3.878191  2.599136  6.099415  3.086750
##     1310     1331     1345     1366     1429     1497     1797
## 2.535483 -2.754403 -2.677687  5.386441 -3.247185  4.014689 -2.821788
##     1868     1942     1958     1965     2016     2036     2066
## -2.751254 -4.227918  3.000271 -2.799419  2.516370  2.505926 -3.205862
##     2174     2176     2318     2323     2444     2549     2563
## 2.961474  2.732626 -3.324763 -2.721648 -3.437022  2.598223  2.685328
##     2590     2600     2626     2637     2642     2646     2659
## 2.708745  3.313439 -3.205136  3.444507 -2.786091 -4.202767 -3.695226
##     2669     2674     2696     2714     2726     2727     2809
## -2.505761 -3.095519 -2.594155  5.675093  2.838797  3.252953 -2.983391
```

```
# various measure of influence
```

```
influence.measures(stepwise_model) %>%
  summary() #we observe the 282, 1059 are potential outliers
```

```

## Potentially influential observations of
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
## median_age_male + pct_une
## dfb.1_ dfb.incd_ dfb.pvr_ dfb.md_ dfb.p_16 dfb.pct_w dfb.pct_s
## 21 0.03 0.04 -0.07 0.04 0.01 -0.05 -0.05
## 116 0.19 0.04 -0.23 0.02 0.13 0.13 -0.04
## 119 -0.04 -0.01 0.12 0.09 -0.12 -0.04 0.01
## 120 0.00 -0.01 0.05 0.06 -0.10 -0.03 0.01
## 122 0.05 -0.01 -0.09 0.15 0.14 0.03 0.05
## 124 0.00 -0.01 0.01 0.02 -0.02 -0.03 0.00
## 140 0.00 0.00 0.00 0.00 0.00 0.00 -0.01
## 147 0.01 0.00 0.00 0.00 0.00 0.01 0.02
## 150 0.00 0.01 0.00 0.01 0.00 0.00 0.07
## 156 0.00 0.00 0.00 0.00 0.00 0.00 0.02
## 161 0.01 0.00 0.00 0.02 -0.01 0.00 0.00
## 165 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## 166 -0.05 0.05 -0.01 0.14 -0.09 0.02 -0.03
## 176 -0.11 0.07 0.01 0.03 0.01 0.12 0.04
## 177 0.00 0.00 0.00 0.03 -0.01 0.00 -0.02
## 180 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## 181 0.00 -0.02 0.02 -0.01 0.01 0.01 -0.03
## 184 0.00 0.00 0.00 0.06 -0.03 0.01 -0.04
## 188 0.00 0.01 0.01 -0.03 -0.01 0.00 0.01
## 189 0.11 -0.11 -0.13 -0.06 0.01 -0.03 -0.03
## 191 0.01 0.00 -0.01 0.00 0.00 0.00 0.00
## 194 0.02 -0.01 0.00 -0.01 0.00 0.00 0.01
## 209 0.02 -0.11 0.02 0.00 0.05 0.06 0.01
## 210 -0.01 -0.01 0.00 0.00 0.00 -0.01 0.06
## 217 0.00 0.01 -0.02 0.01 0.01 -0.01 -0.01
## 237 0.02 -0.06 0.01 0.02 0.01 0.03 0.00
## 245 -0.01 0.00 0.01 0.00 -0.01 -0.01 0.00
## 250 -0.01 -0.09 0.05 -0.02 0.05 0.05 0.00
## 253 0.01 -0.01 -0.01 0.00 0.00 0.00 0.00
## 254 0.12 -0.26 -0.10 -0.06 0.08 0.14 0.01
## 256 0.03 -0.24 -0.11 0.04 0.17 -0.05 -0.11
## 264 -0.02 -0.13 -0.12 0.10 0.10 -0.12 -0.07
## 270 0.01 0.00 -0.01 0.00 0.00 0.02 0.01
## 271 -0.02 -0.06 0.04 -0.01 0.01 0.05 0.02
## 274 0.00 0.00 0.00 0.01 0.00 -0.01 0.00
## 282 0.13 -1.22_* -0.06 0.14 0.16 -0.11 -0.15
## 284 0.01 0.00 0.00 0.01 -0.01 0.00 0.01
## 292 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## 294 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## 296 -0.01 0.01 -0.01 -0.01 0.02 0.00 -0.02
## 300 0.00 0.00 0.00 0.00 0.00 0.00 -0.02
## 321 0.00 0.00 -0.01 0.01 0.00 -0.01 0.00
## 394 -0.01 0.00 0.00 -0.01 0.01 -0.02 -0.04
## 469 0.10 -0.02 -0.05 0.00 -0.07 0.02 -0.02
## 479 0.00 0.00 0.00 0.00 0.00 0.00 0.00
## 492 0.04 0.02 0.02 0.00 -0.06 -0.06 -0.02
## 522 -0.01 -0.09 0.08 -0.02 -0.05 -0.10 -0.04
## 524 0.03 0.03 -0.05 0.02 -0.05 -0.08 0.01
## 537 0.03 0.01 -0.01 0.04 -0.01 -0.04 -0.06
## 564 0.01 0.02 0.02 -0.01 0.01 0.05 0.00

```

## 576	-0.03	0.00	0.02	0.01	0.01	0.00	-0.03
## 585	-0.01	0.00	0.01	0.02	0.00	0.00	0.12
## 591	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 594	0.02	0.00	-0.02	0.00	-0.01	-0.01	-0.02
## 597	0.03	0.02	-0.01	0.01	-0.10	-0.03	0.02
## 621	0.04	-0.09	0.01	-0.05	0.04	0.05	0.00
## 627	0.07	-0.06	-0.02	0.14	-0.06	-0.01	-0.01
## 650	-0.08	0.02	0.04	-0.05	0.02	-0.01	0.03
## 660	0.06	0.01	-0.06	-0.07	0.07	0.01	-0.04
## 666	0.02	0.02	0.01	0.08	-0.09	-0.03	-0.01
## 670	0.02	-0.07	-0.01	0.04	-0.02	0.00	-0.02
## 727	-0.11	-0.03	0.14	0.00	-0.06	0.00	0.03
## 749	-0.07	0.02	0.06	0.02	0.04	0.03	0.01
## 775	-0.01	-0.04	0.08	-0.02	-0.12	-0.06	0.01
## 780	-0.07	0.02	0.06	0.03	-0.08	-0.08	-0.02
## 783	-0.04	0.03	-0.02	0.04	0.08	-0.03	-0.02
## 796	-0.05	0.00	0.04	0.06	-0.07	-0.09	-0.01
## 803	0.09	-0.06	-0.03	0.02	0.00	0.03	0.00
## 809	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 812	-0.09	-0.05	0.03	0.05	0.03	-0.01	0.00
## 840	0.01	-0.01	-0.01	0.01	0.02	0.01	0.00
## 845	-0.19	0.04	0.11	-0.03	-0.02	0.03	0.02
## 853	-0.03	-0.01	0.03	0.01	0.00	-0.01	0.00
## 884	0.00	0.00	0.00	0.00	0.00	0.01	-0.03
## 886	0.00	0.01	0.01	-0.01	0.00	0.01	-0.04
## 900	-0.04	0.03	-0.09	0.06	0.06	-0.06	-0.03
## 905	-0.02	0.03	-0.06	0.06	0.02	-0.08	-0.04
## 912	-0.13	0.09	-0.02	0.22	-0.03	-0.20	-0.06
## 913	-0.10	0.08	0.02	0.09	-0.03	-0.14	0.00
## 921	-0.07	0.03	0.07	0.05	-0.06	-0.03	0.00
## 922	-0.01	0.00	0.00	0.00	0.00	0.00	0.00
## 925	-0.07	0.03	0.09	0.05	-0.01	-0.05	-0.01
## 927	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 937	-0.05	0.00	0.06	0.08	-0.01	0.00	0.01
## 961	0.01	-0.01	-0.01	0.00	0.00	0.00	0.01
## 979	0.00	-0.09	0.01	0.00	0.01	0.04	0.00
## 982	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1000	0.00	0.00	0.00	0.00	0.00	0.00	0.01
## 1005	-0.01	0.00	0.01	0.00	0.01	0.00	0.00
## 1007	-0.11	0.07	0.12	-0.01	-0.11	-0.04	0.03
## 1013	0.00	0.00	0.00	0.00	0.00	0.00	0.02
## 1016	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1017	0.00	0.00	0.00	0.00	0.00	0.00	0.01
## 1019	-0.01	0.00	0.00	0.00	0.01	0.00	0.01
## 1022	0.00	0.00	0.01	0.02	-0.01	0.02	0.14
## 1023	-0.01	0.00	0.01	0.00	0.01	0.00	0.02
## 1025	-0.01	0.00	0.00	0.01	0.00	0.00	0.04
## 1027	-0.02	0.00	0.01	0.02	0.00	0.00	0.11
## 1033	-0.01	0.01	0.00	0.01	0.02	0.00	0.06
## 1036	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1043	-0.01	0.00	0.01	0.00	0.00	0.00	0.02
## 1048	-0.02	0.01	0.03	-0.10	-0.02	0.01	0.00
## 1050	0.01	0.01	-0.02	0.00	0.00	0.00	0.00
## 1058	-0.05	0.12	0.01	-0.07	-0.07	-0.10	-0.04

## 1059	0.33	0.18	-0.24	-0.02	-0.46	-0.20	-0.03
## 1063	0.00	-0.01	0.00	0.00	0.00	0.00	0.00
## 1066	-0.12	0.07	0.12	0.06	0.00	-0.02	0.04
## 1074	-0.04	-0.01	0.08	-0.09	-0.08	-0.05	0.00
## 1076	0.07	-0.05	0.01	0.03	-0.07	-0.03	-0.03
## 1094	-0.26	0.22	0.28	0.02	-0.11	-0.15	0.03
## 1105	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1139	0.01	0.00	0.00	-0.01	0.00	-0.03	0.00
## 1174	0.07	-0.01	-0.09	-0.10	-0.01	0.02	-0.03
## 1202	0.00	-0.01	-0.01	0.00	0.00	0.02	0.01
## 1203	0.01	-0.03	0.01	0.01	-0.10	0.03	0.03
## 1204	-0.06	0.01	0.16	0.10	-0.09	-0.11	0.01
## 1217	0.00	0.06	0.02	-0.02	-0.03	-0.05	-0.03
## 1221	0.03	0.01	-0.01	0.01	-0.11	-0.30	-0.09
## 1236	0.04	-0.03	-0.03	0.02	0.13	-0.05	-0.04
## 1261	0.07	-0.08	-0.05	0.02	-0.03	0.05	0.01
## 1297	-0.04	0.05	-0.03	-0.02	0.02	-0.02	-0.06
## 1299	-0.03	0.02	0.03	0.07	-0.02	-0.05	-0.03
## 1310	-0.06	0.04	0.05	-0.05	-0.07	-0.06	-0.06
## 1311	-0.02	-0.01	0.03	0.01	0.00	-0.01	0.01
## 1331	0.03	-0.01	-0.05	-0.03	0.09	0.03	0.02
## 1345	-0.01	0.01	0.02	-0.07	0.01	-0.01	0.02
## 1350	0.00	0.01	-0.01	0.00	0.01	0.00	-0.01
## 1366	0.07	0.03	0.04	0.13	-0.14	0.07	0.02
## 1390	-0.04	0.06	0.09	0.03	-0.04	0.03	0.03
## 1422	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1429	-0.02	-0.01	0.04	-0.10	-0.10	0.19	0.08
## 1490	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1497	-0.07	0.08	0.09	0.08	-0.18	-0.01	-0.02
## 1542	0.01	-0.02	0.03	0.00	-0.01	-0.04	-0.01
## 1554	0.00	0.01	0.00	0.00	-0.02	0.00	-0.01
## 1568	0.06	0.00	-0.01	-0.04	-0.09	-0.03	-0.01
## 1601	0.00	0.00	-0.01	0.01	0.00	-0.01	0.00
## 1633	-0.06	0.01	0.03	0.04	0.01	-0.01	-0.01
## 1687	-0.04	-0.02	0.00	0.03	-0.03	-0.04	-0.05
## 1695	0.03	0.02	-0.03	-0.07	0.00	0.01	0.00
## 1723	-0.02	0.00	0.02	-0.01	-0.02	-0.01	0.00
## 1771	0.01	0.02	-0.01	-0.01	-0.02	0.02	0.01
## 1797	-0.07	0.02	-0.07	0.09	0.04	-0.12	-0.05
## 1813	0.03	0.01	-0.04	0.00	0.02	0.00	-0.01
## 1817	-0.02	0.01	-0.03	0.01	0.00	-0.03	-0.07
## 1818	-0.12	0.04	0.05	0.08	-0.08	-0.10	-0.11
## 1824	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1828	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1840	0.00	0.00	0.00	0.00	0.00	0.00	-0.01
## 1843	0.00	0.00	0.00	0.01	0.01	0.00	0.00
## 1844	0.02	0.00	-0.01	-0.02	-0.01	0.00	0.00
## 1846	-0.01	-0.04	0.06	0.01	-0.05	-0.01	0.01
## 1848	0.05	-0.01	-0.01	-0.04	-0.01	0.02	0.02
## 1849	-0.03	0.02	0.00	0.05	-0.03	-0.03	-0.07
## 1858	-0.01	-0.02	0.00	0.04	0.06	0.00	0.00
## 1868	-0.10	0.03	-0.05	0.09	0.03	-0.10	-0.05
## 1888	-0.03	0.01	0.03	0.00	-0.02	-0.01	-0.02
## 1898	0.00	0.02	-0.03	0.00	0.01	-0.02	0.01

## 1916	-0.09	0.01	0.04	0.07	-0.04	-0.07	-0.01
## 1918	0.01	-0.01	-0.01	0.00	0.00	0.00	0.00
## 1933	0.01	0.00	0.00	0.03	0.00	0.00	0.00
## 1942	-0.37	0.25	0.23	0.10	-0.03	-0.22	-0.17
## 1945	0.01	0.01	-0.01	0.01	-0.02	-0.01	-0.02
## 1946	0.01	-0.01	0.00	0.01	0.01	0.00	0.00
## 1958	0.06	0.08	-0.04	0.06	-0.05	-0.03	-0.09
## 1965	-0.15	0.01	-0.01	0.11	-0.03	-0.15	-0.05
## 1968	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 1979	-0.01	-0.01	-0.01	0.04	-0.04	-0.07	-0.11
## 2016	0.03	0.03	0.02	-0.02	-0.02	0.04	0.02
## 2022	0.04	-0.11	0.00	-0.01	0.01	0.00	0.00
## 2036	0.01	-0.06	-0.01	0.03	0.02	-0.02	-0.02
## 2066	-0.15	-0.02	0.04	0.05	-0.12	0.13	0.04
## 2087	0.00	0.00	0.00	-0.01	0.00	0.00	-0.04
## 2157	0.03	-0.11	-0.02	-0.01	0.03	-0.07	-0.06
## 2174	-0.02	-0.04	0.03	0.00	0.00	-0.01	-0.03
## 2176	-0.09	0.06	0.14	0.00	-0.05	0.08	0.06
## 2204	-0.05	-0.01	0.03	0.03	0.00	-0.01	0.00
## 2292	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 2305	-0.01	0.02	0.00	0.00	0.01	0.02	0.01
## 2313	0.01	0.00	-0.01	-0.01	0.00	0.01	-0.02
## 2314	-0.01	0.00	0.01	0.01	-0.01	0.00	0.08
## 2315	-0.04	-0.01	0.05	0.04	-0.03	-0.02	0.22
## 2316	0.02	0.00	-0.02	-0.01	0.01	0.01	-0.08
## 2318	-0.04	0.00	0.09	-0.08	-0.09	-0.06	-0.04
## 2323	-0.05	0.04	0.06	-0.02	0.03	-0.02	0.01
## 2328	-0.03	0.07	0.01	0.01	-0.01	-0.01	0.00
## 2347	0.12	0.05	-0.17	0.07	-0.03	0.01	0.00
## 2353	0.03	0.05	0.01	0.07	-0.05	-0.03	0.04
## 2355	-0.05	0.05	0.08	0.08	-0.05	-0.05	0.02
## 2395	0.04	0.09	-0.04	-0.02	0.01	0.09	0.02
## 2426	0.00	-0.05	-0.01	0.03	0.02	0.04	0.00
## 2444	0.00	0.09	-0.12	0.01	0.13	0.14	-0.03
## 2452	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 2510	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 2546	-0.01	0.04	-0.06	0.00	0.01	0.09	0.03
## 2549	0.04	-0.03	-0.02	-0.02	0.09	-0.10	-0.07
## 2563	-0.05	0.09	0.07	-0.02	-0.05	0.05	0.04
## 2573	-0.06	0.01	0.07	0.00	0.01	0.02	0.02
## 2579	-0.08	0.02	0.07	-0.02	0.05	0.05	0.03
## 2590	-0.03	0.11	0.03	-0.01	-0.02	0.04	0.02
## 2594	-0.09	0.02	0.15	0.02	-0.07	0.06	0.06
## 2596	-0.05	0.10	0.02	-0.01	0.05	0.08	0.03
## 2598	-0.04	0.13	0.01	-0.05	-0.05	0.10	0.04
## 2600	-0.18	0.08	0.18	0.09	-0.07	0.02	0.03
## 2626	-0.02	0.07	0.00	0.01	-0.06	0.08	0.02
## 2637	0.01	0.01	0.06	0.03	-0.02	-0.02	0.01
## 2642	-0.03	0.09	-0.01	-0.02	0.03	0.09	0.03
## 2646	-0.09	-0.26	0.03	-0.02	0.04	-0.08	0.00
## 2659	0.04	-0.08	-0.15	-0.09	0.16	0.18	0.00
## 2669	-0.01	-0.02	0.04	-0.01	-0.08	0.01	0.01
## 2671	0.00	-0.01	0.02	-0.01	-0.01	0.00	0.00
## 2674	0.02	-0.06	-0.13	-0.02	0.21	0.12	-0.01

## 2682	-0.06	0.02	0.03	-0.06	0.07	-0.02	0.00
## 2684	0.00	0.01	-0.09	0.05	0.07	-0.04	-0.04
## 2689	0.01	-0.02	0.00	0.01	0.01	-0.05	-0.03
## 2696	0.02	0.00	0.00	-0.01	-0.03	0.10	0.03
## 2699	0.02	-0.02	-0.05	-0.01	0.03	0.02	0.00
## 2706	0.02	-0.01	0.01	0.02	0.00	-0.03	-0.01
## 2714	0.22	-0.38	0.02	0.19	-0.31	0.07	1.54_*
## 2716	0.05	-0.08	-0.04	-0.06	0.06	-0.12	-0.04
## 2718	0.10	-0.01	-0.08	-0.03	0.01	-0.12	-0.03
## 2720	-0.15	-0.01	0.11	-0.03	0.06	0.00	-0.04
## 2724	0.09	0.02	-0.05	-0.01	-0.02	0.00	0.24
## 2726	0.09	0.04	-0.07	-0.07	0.04	-0.15	-0.05
## 2727	0.18	0.06	-0.21	-0.02	-0.02	-0.13	0.03
## 2729	0.01	-0.01	-0.01	-0.01	0.01	-0.01	-0.01
## 2733	-0.04	-0.01	0.03	0.00	-0.01	0.02	0.01
## 2747	0.01	0.02	-0.01	0.00	-0.01	-0.01	0.00
## 2783	0.00	0.00	0.00	-0.01	0.00	0.00	0.00
## 2785	0.03	0.05	0.00	0.04	-0.12	0.02	0.01
## 2789	0.04	0.06	-0.01	0.02	0.03	0.00	0.01
## 2790	-0.05	-0.02	0.06	0.02	-0.05	0.02	0.02
## 2809	-0.02	-0.03	-0.07	0.02	0.09	0.08	0.00
## 2811	0.02	0.15	-0.04	-0.07	0.00	-0.01	-0.01
## 2826	0.00	0.00	0.00	-0.01	0.01	0.00	0.00
## 2842	-0.01	0.07	-0.02	-0.01	0.02	0.01	-0.01
## 2858	-0.02	0.07	0.03	0.02	-0.04	0.00	-0.01
## 2922	-0.03	0.03	0.00	0.02	0.03	0.02	-0.01
## 3040	0.03	0.03	-0.01	0.05	-0.06	-0.02	-0.02
##	dfb.pct_t_	dfb.pct_m_	dfb.p_-1	dfb.pct_w_	dfb.incm_	dffit	cov.r
## 21	0.01	0.01	0.02	-0.05	-0.03	-0.13	1.01
## 116	0.12	-0.23	0.02	-0.18	-0.01	0.36_*	0.96_*
## 119	0.01	0.02	-0.10	0.03	0.02	-0.21_*	0.97_*
## 120	0.03	-0.05	-0.04	0.03	-0.01	-0.16	0.97_*
## 122	0.06	-0.16	-0.04	-0.06	0.05	0.31_*	0.97_*
## 124	-0.01	0.01	-0.05	0.02	-0.04	-0.11	0.96_*
## 140	0.00	0.00	0.00	0.00	0.00	-0.01	1.02_*
## 147	0.01	-0.01	0.00	-0.01	0.00	0.03	1.02_*
## 150	-0.01	0.01	0.00	-0.01	0.00	0.08	1.03_*
## 156	0.00	0.01	0.00	0.00	0.00	0.03	1.02_*
## 161	0.00	0.01	-0.02	-0.01	-0.01	0.04	1.02_*
## 165	0.00	0.00	0.00	0.00	0.00	0.00	1.01_*
## 166	0.27	-0.11	0.05	0.07	-0.16	0.37_*	0.98_*
## 176	0.05	-0.04	0.04	0.06	0.04	-0.24_*	0.99
## 177	0.07	0.00	-0.03	0.00	-0.02	0.08	1.05_*
## 180	-0.02	0.00	0.00	0.00	0.01	-0.02	1.02_*
## 181	0.08	0.00	-0.04	0.01	0.01	0.10	1.01_*
## 184	0.22	-0.06	-0.03	0.02	-0.06	0.25_*	1.05_*
## 188	-0.05	0.02	0.00	0.00	-0.01	-0.07	1.02_*
## 189	-0.04	-0.05	0.00	0.00	-0.10	-0.20_*	0.98_*
## 191	-0.02	0.00	0.00	-0.01	0.00	-0.02	1.01_*
## 194	-0.08	0.02	0.00	-0.03	0.02	-0.09	1.05_*
## 209	-0.02	0.03	-0.07	0.01	-0.10	0.21_*	0.98_*
## 210	-0.01	0.03	0.01	0.01	-0.01	0.08	1.02_*
## 217	0.00	0.00	0.01	-0.01	-0.01	-0.03	1.01_*
## 237	-0.02	-0.01	-0.02	-0.02	0.03	0.10	0.98_*

## 245	0.00	0.00	0.01	0.02	0.01	0.03	1.02_*
## 250	-0.03	-0.01	-0.02	0.03	0.05	0.14	0.98_*
## 253	0.00	-0.01	0.00	0.00	0.00	0.02	1.01_*
## 254	-0.02	-0.18	0.03	0.05	-0.19	0.37_*	0.97_*
## 256	0.00	0.09	0.02	0.02	-0.06	-0.38_*	0.99
## 264	0.01	0.08	0.08	0.04	-0.07	-0.33_*	0.98_*
## 270	0.00	-0.02	0.00	0.00	-0.02	0.03	1.01_*
## 271	0.00	0.03	-0.01	0.01	0.04	-0.11	1.01_*
## 274	0.00	0.01	0.00	-0.01	0.01	-0.02	1.02_*
## 282	-0.10	0.02	0.24	0.22	-0.02	-1.34_*	0.89_*
## 284	-0.04	-0.01	-0.01	-0.01	0.00	-0.05	1.03_*
## 292	-0.01	0.00	0.00	0.00	0.00	-0.01	1.02_*
## 294	0.02	0.00	0.00	0.00	0.00	0.02	1.02_*
## 296	0.08	0.00	0.01	0.01	0.00	0.09	1.01_*
## 300	0.00	0.00	0.00	0.00	0.00	-0.02	1.01_*
## 321	0.00	0.00	0.01	-0.01	-0.01	-0.03	1.01_*
## 394	-0.03	0.05	0.00	0.00	0.00	-0.08	1.02_*
## 469	-0.03	-0.07	-0.02	-0.06	-0.02	0.14	0.99_*
## 479	0.00	0.00	0.00	0.00	0.00	0.00	1.01_*
## 492	-0.03	0.01	-0.03	-0.04	0.03	0.11	1.02_*
## 522	-0.05	-0.03	0.04	0.08	0.10	0.22_*	0.98_*
## 524	-0.03	0.00	0.03	-0.01	-0.13	0.18	0.98_*
## 537	0.09	0.00	-0.06	-0.04	0.03	0.14	0.98_*
## 564	-0.01	-0.03	0.00	-0.03	0.04	0.10	0.98_*
## 576	0.15	0.01	0.00	0.02	0.02	0.16	1.01_*
## 585	0.00	0.01	0.01	0.00	0.00	0.12	1.04_*
## 591	0.00	0.00	0.00	0.00	0.00	0.00	1.02_*
## 594	-0.04	-0.01	0.00	-0.01	-0.01	-0.05	1.02_*
## 597	-0.14	-0.03	0.03	-0.01	-0.03	-0.19_*	1.01
## 621	0.00	-0.08	-0.03	0.02	0.02	-0.17	0.98_*
## 627	-0.02	-0.04	-0.07	-0.08	0.01	0.21_*	0.97_*
## 650	0.00	0.03	0.09	0.07	0.01	-0.14	0.99_*
## 660	0.01	-0.03	-0.07	-0.02	-0.03	-0.16	0.99_*
## 666	-0.01	0.02	-0.07	-0.04	0.00	0.15	0.97_*
## 670	-0.01	-0.03	-0.03	0.02	-0.01	0.12	0.98_*
## 727	0.01	0.03	0.07	0.11	0.00	0.19	1.01
## 749	0.01	0.04	0.00	0.02	0.05	0.11	0.98_*
## 775	-0.03	0.02	0.10	0.02	0.06	0.20_*	0.96_*
## 780	0.03	0.11	0.02	0.04	0.00	0.17	0.99_*
## 783	-0.08	0.00	-0.11	0.06	-0.03	-0.24_*	0.99_*
## 796	-0.05	0.03	0.01	0.05	0.05	-0.16	1.01_*
## 803	-0.01	-0.10	-0.05	-0.05	0.00	-0.16	0.98_*
## 809	0.00	0.00	0.00	0.00	0.00	-0.01	1.01_*
## 812	0.05	0.02	0.00	0.08	0.13	-0.21_*	0.96_*
## 840	0.00	0.00	-0.02	-0.01	0.01	-0.03	1.01_*
## 845	0.07	0.11	0.00	0.15	0.08	-0.28_*	0.98_*
## 853	0.01	0.01	0.00	0.03	0.02	-0.05	1.01_*
## 884	0.01	-0.01	0.00	0.00	0.00	-0.03	1.02_*
## 886	0.00	-0.01	0.00	0.00	0.01	-0.05	1.01_*
## 900	0.02	-0.03	0.03	0.07	-0.06	-0.18	1.01
## 905	0.02	-0.03	-0.01	0.05	0.02	-0.17	1.01
## 912	0.08	-0.05	-0.01	0.14	0.13	-0.42_*	0.96_*
## 913	-0.05	0.07	0.03	0.05	0.09	-0.23_*	0.99
## 921	0.06	-0.01	-0.01	0.08	0.01	-0.15	0.97_*

## 922	0.00	0.01	0.00	0.00	0.00	-0.01	1.01_*
## 925	0.04	0.08	-0.11	0.04	0.01	-0.17	0.99_*
## 927	0.00	-0.01	0.00	0.00	0.00	-0.01	1.01_*
## 937	0.04	-0.05	0.01	0.04	0.04	-0.14	0.99_*
## 961	0.00	-0.01	0.00	0.00	0.00	0.02	1.01_*
## 979	-0.01	0.01	-0.03	0.04	-0.13	0.19	0.98_*
## 982	0.00	0.00	0.00	0.00	0.00	-0.01	1.02_*
## 1000	0.01	0.00	0.00	0.00	0.00	0.02	1.02_*
## 1005	0.02	0.00	0.00	0.00	0.00	0.02	1.02_*
## 1007	-0.05	0.07	0.05	0.08	0.02	-0.20_*	1.00
## 1013	0.01	0.00	0.00	0.00	0.00	0.03	1.02_*
## 1016	-0.01	0.00	0.00	0.00	0.00	-0.01	1.01_*
## 1017	0.00	0.00	0.00	0.00	0.00	0.01	1.01_*
## 1019	0.02	0.01	0.00	0.00	0.00	0.03	1.01_*
## 1022	-0.01	-0.03	0.01	-0.01	0.00	0.15	1.07_*
## 1023	0.01	0.01	0.00	0.00	0.00	0.03	1.02_*
## 1025	0.00	0.00	0.00	0.00	0.00	0.05	1.04_*
## 1027	0.00	0.02	0.01	0.00	0.00	0.12	1.06_*
## 1033	0.01	0.01	0.01	0.01	0.00	0.07	1.01_*
## 1036	0.00	0.00	0.00	0.00	0.00	0.00	1.02_*
## 1043	0.01	0.00	0.00	0.00	0.00	0.03	1.01_*
## 1048	-0.07	-0.02	0.02	0.07	-0.02	-0.17	0.98_*
## 1050	0.00	0.00	-0.01	-0.01	-0.01	-0.02	1.02_*
## 1058	-0.01	0.05	0.04	0.04	0.09	-0.25_*	0.98_*
## 1059	-0.18	-0.15	0.09	-0.28	0.02	-0.70_*	0.94_*
## 1063	0.00	0.00	0.00	0.00	0.00	0.01	1.01_*
## 1066	-0.05	0.06	-0.01	0.05	0.05	-0.18	0.98_*
## 1074	-0.05	0.06	0.04	0.05	0.10	-0.20_*	0.99
## 1076	-0.04	0.02	-0.11	-0.06	0.00	0.16	0.98_*
## 1094	0.03	0.37	-0.02	0.07	0.07	-0.45_*	0.96_*
## 1105	0.00	-0.01	0.00	0.00	0.00	0.01	1.01_*
## 1139	-0.02	0.00	0.02	-0.01	0.02	0.05	1.02_*
## 1174	-0.02	-0.04	0.04	-0.02	-0.05	0.15	0.98_*
## 1202	0.01	0.00	0.00	-0.01	0.00	-0.04	1.01_*
## 1203	0.01	0.02	0.00	-0.02	0.01	-0.18	1.01
## 1204	-0.03	0.05	0.01	0.00	0.14	0.24_*	1.00
## 1217	-0.02	0.02	-0.02	0.00	0.03	0.11	0.98_*
## 1221	-0.04	0.11	-0.11	0.05	0.00	0.36_*	0.87_*
## 1236	-0.03	0.00	-0.07	-0.02	-0.03	0.21_*	1.00
## 1261	-0.03	-0.10	0.03	-0.03	0.00	0.16	0.97_*
## 1297	0.19	0.05	0.03	0.02	-0.02	0.23_*	1.01
## 1299	0.05	0.07	-0.09	0.00	0.00	0.13	1.02_*
## 1310	0.11	0.11	-0.02	0.06	-0.01	0.22_*	0.99_*
## 1311	0.02	0.01	-0.01	0.02	0.01	-0.07	0.99_*
## 1331	-0.01	-0.01	0.02	-0.03	-0.01	-0.13	0.98_*
## 1345	-0.01	0.04	0.04	0.00	0.00	-0.12	0.98_*
## 1350	0.08	-0.01	0.00	0.00	0.00	0.09	1.03_*
## 1366	-0.02	-0.14	-0.02	-0.10	0.07	0.30_*	0.90_*
## 1390	0.00	0.10	-0.09	-0.03	-0.06	0.20_*	0.99
## 1422	0.00	-0.01	0.00	0.00	0.00	0.01	1.01_*
## 1429	0.03	0.00	0.07	-0.03	0.06	-0.33_*	0.97_*
## 1490	0.00	0.00	0.00	0.00	0.00	0.00	1.09_*
## 1497	0.23	-0.02	0.09	0.01	0.04	0.33_*	0.95_*
## 1542	-0.03	0.03	-0.01	-0.02	0.05	0.10	0.98_*

## 1554	0.08	0.00	0.01	0.00	0.00	0.09	1.02_*
## 1568	-0.03	-0.02	-0.03	-0.03	-0.04	-0.13	0.99_*
## 1601	0.00	0.00	0.01	-0.01	0.01	-0.03	1.01_*
## 1633	0.04	-0.02	-0.01	0.07	0.00	-0.10	1.02_*
## 1687	0.16	0.03	0.03	0.05	-0.01	0.20_*	1.00
## 1695	-0.01	-0.03	0.05	-0.01	0.03	-0.11	1.01_*
## 1723	0.00	0.00	0.03	0.02	0.01	0.04	1.01_*
## 1771	0.00	0.00	0.01	-0.02	0.00	-0.05	1.01_*
## 1797	-0.01	-0.05	0.05	0.11	0.08	-0.29_*	0.98_*
## 1813	0.00	-0.02	-0.02	-0.02	-0.02	-0.05	1.02_*
## 1817	0.24	-0.01	0.02	0.03	-0.02	0.27_*	1.04_*
## 1818	0.49	0.09	0.02	0.09	-0.01	0.54_*	1.04_*
## 1824	0.00	0.00	0.00	0.00	0.00	0.00	1.01_*
## 1828	0.00	0.00	0.00	0.00	0.00	0.01	1.01_*
## 1840	0.00	0.00	0.00	0.00	0.00	-0.01	1.02_*
## 1843	-0.02	0.00	-0.01	0.00	0.01	-0.03	1.02_*
## 1844	-0.01	0.00	0.01	-0.02	0.00	0.03	1.02_*
## 1846	0.01	0.02	0.01	0.01	0.03	0.09	1.01_*
## 1848	-0.10	-0.02	0.00	-0.04	-0.01	-0.11	1.02_*
## 1849	0.31	-0.01	0.02	0.01	0.01	0.33_*	1.02_*
## 1858	0.00	-0.03	-0.04	0.02	0.01	-0.12	0.99_*
## 1868	0.00	-0.09	0.09	0.15	0.08	-0.32_*	0.99_*
## 1888	0.09	0.00	0.02	0.03	0.01	0.10	1.02_*
## 1898	-0.06	0.00	0.00	0.01	-0.02	-0.09	1.01_*
## 1916	0.01	-0.02	0.03	0.10	0.07	-0.18	1.01
## 1918	0.00	0.00	0.01	-0.01	-0.01	0.02	1.01_*
## 1933	0.01	-0.02	-0.02	-0.01	0.00	-0.04	1.01_*
## 1942	0.11	0.17	-0.11	0.30	0.23	-0.59_*	0.95_*
## 1945	0.09	0.01	0.00	-0.02	-0.01	0.11	1.05_*
## 1946	0.00	0.00	-0.01	-0.01	0.00	0.02	1.01_*
## 1958	0.32	0.04	-0.10	-0.11	-0.01	0.39_*	0.99_*
## 1965	0.03	-0.06	0.10	0.22	0.07	-0.36_*	0.99
## 1968	-0.01	0.00	0.00	0.00	0.00	-0.01	1.01_*
## 1979	0.40	0.04	-0.01	0.01	-0.01	0.45_*	1.01_*
## 2016	-0.03	-0.02	0.00	-0.05	-0.07	0.15	0.98_*
## 2022	-0.05	0.03	-0.04	-0.02	0.01	0.14	0.99_*
## 2036	-0.03	0.05	-0.04	-0.01	0.00	0.12	0.98_*
## 2066	0.14	-0.13	0.09	0.23	-0.13	-0.41_*	0.98_*
## 2087	0.00	0.00	0.00	0.00	0.00	-0.04	1.03_*
## 2157	0.00	0.00	-0.07	0.03	0.09	-0.20_*	0.99_*
## 2174	-0.02	0.05	-0.06	0.03	0.01	0.12	0.97_*
## 2176	0.01	0.05	0.00	0.04	-0.06	0.23_*	0.98_*
## 2204	0.03	-0.01	0.00	0.06	0.01	-0.08	1.03_*
## 2292	0.00	0.00	0.00	0.00	0.00	0.00	1.01_*
## 2305	0.01	0.00	0.01	-0.01	0.02	-0.04	1.01_*
## 2313	0.00	0.00	0.00	0.00	0.00	-0.03	1.03_*
## 2314	-0.02	0.01	0.01	0.00	0.00	0.09	1.11_*
## 2315	-0.05	0.04	0.04	0.01	0.01	0.24_*	1.08_*
## 2316	0.02	-0.01	-0.02	-0.01	0.00	-0.09	1.05_*
## 2318	0.02	0.06	-0.09	0.10	-0.03	-0.25_*	0.97_*
## 2323	-0.07	0.06	-0.02	0.02	0.02	-0.14	0.98_*
## 2328	0.03	0.00	-0.01	0.02	-0.01	-0.10	0.98_*
## 2347	0.02	-0.19	0.09	-0.08	-0.06	-0.27_*	1.01_*
## 2353	-0.09	0.01	-0.07	-0.07	0.00	-0.16	0.99_*

```

## 2355 -0.02      0.05      -0.07      0.01      0.02      -0.15      0.99_*
## 2395  0.01      0.02      -0.08      -0.10      0.02      -0.19_*   1.00
## 2426  0.02      0.00      -0.05      0.00      0.06      -0.16      0.99_*
## 2444  0.09      -0.03      -0.06      -0.05      0.05      -0.31_*   0.97_*
## 2452 -0.01      0.01      -0.01      0.00      0.00      -0.02      1.02_*
## 2510  0.00      0.00      0.00      0.00      0.00      0.00      1.01_*
## 2546  0.03      0.01      0.04      -0.04      0.00      -0.19_*   1.00
## 2549 -0.04      -0.02      -0.07      0.03      -0.06      0.25_*   0.99_*
## 2563  0.00      0.08      0.00      0.00      -0.11      0.21_*   0.98_*
## 2573  0.01      0.04      0.01      0.04      0.01      0.09      1.02_*
## 2579  0.01      0.05      0.02      0.05      -0.01      0.14      1.01_*
## 2590 -0.01      0.01      0.03      -0.04      0.04      0.15      0.98_*
## 2594  0.01      0.02      0.03      0.05      0.01      0.19      1.01_*
## 2596  0.01      0.03      0.00      -0.01      -0.09      0.19_*   0.99
## 2598  0.02      -0.06      0.09      0.02      -0.10      0.23_*   0.99
## 2600  0.03      0.05      0.03      0.09      0.09      0.23_*   0.97_*
## 2626  0.06      0.05      -0.10      -0.03      0.04      -0.28_*   0.97_*
## 2637 -0.05      0.14      -0.13      -0.08      0.03      0.20_*   0.96_*
## 2642  0.00      -0.01      -0.04      -0.02      0.08      -0.21_*   0.98_*
## 2646 -0.19      -0.09      0.12      0.25      0.18      -0.46_*   0.95_*
## 2659  0.05      -0.10      0.00      0.04      -0.14      -0.31_*   0.96_*
## 2669 -0.02      0.00      0.00      0.02      0.09      -0.16      0.98_*
## 2671  0.00      0.00      -0.01      0.00      0.01      0.03      1.01_*
## 2674  0.00      0.01      -0.10      -0.02      0.03      -0.33_*   0.98_*
## 2682  0.00      0.10      0.00      0.04      0.01      0.15      1.01_*
## 2684  0.00      0.03      0.04      -0.04      0.04      -0.20_*   1.00
## 2689 -0.01      -0.01      -0.03      0.03      -0.02      0.09      1.01_*
## 2696  0.03      0.03      -0.07      -0.07      0.04      -0.22_*   0.98_*
## 2699  0.00      -0.01      0.01      -0.02      -0.01      -0.07      1.02_*
## 2706 -0.01      0.00      -0.04      -0.01      -0.01      0.07      1.01_*
## 2714 -0.32      -0.18      0.36      -0.23      0.00      1.74_*   0.97_*
## 2716 -0.08      0.03      0.06      0.00      0.05      0.23_*   1.01
## 2718 -0.07      0.01      0.01      -0.08      0.02      0.18      1.01
## 2720  0.05      0.08      -0.03      0.15      0.01      -0.21_*   1.00
## 2724 -0.08      0.03      0.01      -0.13      -0.02      0.29_*   1.02_*
## 2726 -0.10      -0.01      0.07      -0.07      0.08      0.30_*   0.98_*
## 2727 -0.07      -0.13      0.09      -0.12      -0.02      0.34_*   0.97_*
## 2729 -0.01      0.00      0.01      -0.01      0.00      0.03      1.02_*
## 2733  0.02      0.01      0.00      0.05      -0.02      -0.07      1.02_*
## 2747 -0.03      -0.01      0.00      -0.01      -0.01      -0.04      1.01_*
## 2783  0.00      0.00      0.00      0.00      0.00      0.01      1.01_*
## 2785  0.02      -0.03      -0.03      -0.05      0.01      -0.17      1.02_*
## 2789  0.01      0.02      -0.09      -0.08      0.01      -0.13      0.99_*
## 2790  0.02      0.01      0.00      0.04      0.03      -0.09      1.02_*
## 2809  0.07      -0.02      -0.03      0.04      -0.08      -0.19      0.97_*
## 2811  0.02      0.04      -0.04      -0.06      -0.04      -0.19_*   0.99_*
## 2826  0.00      0.00      0.00      0.00      0.00      0.02      1.02_*
## 2842  0.01      0.03      -0.02      -0.03      0.00      0.10      0.99_*
## 2858 -0.01      0.05      -0.03      -0.02      0.01      0.11      0.98_*
## 2922  0.00      0.01      -0.02      0.01      0.01      0.08      0.99_*
## 3040  0.02      0.01      -0.05      -0.05      -0.01      0.11      0.99_*
##          cook.d hat
## 21     0.00      0.01_*
## 116    0.01      0.01

```

```
## 119 0.00 0.00
## 120 0.00 0.00
## 122 0.01 0.01
## 124 0.00 0.00
## 140 0.00 0.01_*
## 147 0.00 0.01_*
## 150 0.00 0.03_*
## 156 0.00 0.01_*
## 161 0.00 0.02_*
## 165 0.00 0.01
## 166 0.01 0.02_*
## 176 0.00 0.01
## 177 0.00 0.05_*
## 180 0.00 0.01_*
## 181 0.00 0.01_*
## 184 0.01 0.05_*
## 188 0.00 0.01_*
## 189 0.00 0.01
## 191 0.00 0.01
## 194 0.00 0.05_*
## 209 0.00 0.01
## 210 0.00 0.02_*
## 217 0.00 0.01
## 237 0.00 0.00
## 245 0.00 0.01_*
## 250 0.00 0.00
## 253 0.00 0.01
## 254 0.01 0.01
## 256 0.01 0.02_*
## 264 0.01 0.01_*
## 270 0.00 0.01
## 271 0.00 0.01_*
## 274 0.00 0.01_*
## 282 0.15 0.04_*
## 284 0.00 0.02_*
## 292 0.00 0.01_*
## 294 0.00 0.01_*
## 296 0.00 0.01_*
## 300 0.00 0.01
## 321 0.00 0.01
## 394 0.00 0.01_*
## 469 0.00 0.00
## 479 0.00 0.01
## 492 0.00 0.02_*
## 522 0.00 0.01
## 524 0.00 0.00
## 537 0.00 0.00
## 564 0.00 0.00
## 576 0.00 0.01_*
## 585 0.00 0.04_*
## 591 0.00 0.01
## 594 0.00 0.02_*
## 597 0.00 0.01_*
## 621 0.00 0.00
```

```
## 627 0.00 0.00
## 650 0.00 0.00
## 660 0.00 0.00
## 666 0.00 0.00
## 670 0.00 0.00
## 727 0.00 0.01_*
## 749 0.00 0.00
## 775 0.00 0.00
## 780 0.00 0.01
## 783 0.00 0.01
## 796 0.00 0.02_*
## 803 0.00 0.00
## 809 0.00 0.01
## 812 0.00 0.00
## 840 0.00 0.01
## 845 0.01 0.01
## 853 0.00 0.01
## 884 0.00 0.01_*
## 886 0.00 0.01
## 900 0.00 0.01_*
## 905 0.00 0.01_*
## 912 0.01 0.01_*
## 913 0.00 0.01
## 921 0.00 0.00
## 922 0.00 0.01
## 925 0.00 0.01
## 927 0.00 0.01
## 937 0.00 0.00
## 961 0.00 0.01
## 979 0.00 0.01
## 982 0.00 0.02_*
## 1000 0.00 0.01_*
## 1005 0.00 0.02_*
## 1007 0.00 0.01
## 1013 0.00 0.02_*
## 1016 0.00 0.01
## 1017 0.00 0.01
## 1019 0.00 0.01
## 1022 0.00 0.06_*
## 1023 0.00 0.01_*
## 1025 0.00 0.03_*
## 1027 0.00 0.06_*
## 1033 0.00 0.01
## 1036 0.00 0.01_*
## 1043 0.00 0.01
## 1048 0.00 0.00
## 1050 0.00 0.01_*
## 1058 0.01 0.01
## 1059 0.04 0.02_*
## 1063 0.00 0.01
## 1066 0.00 0.00
## 1074 0.00 0.01
## 1076 0.00 0.00
## 1094 0.02 0.01_*
```

```

## 1105 0.00 0.01
## 1139 0.00 0.01_*
## 1174 0.00 0.00
## 1202 0.00 0.01
## 1203 0.00 0.01_*
## 1204 0.00 0.01_*
## 1217 0.00 0.00
## 1221 0.01 0.00
## 1236 0.00 0.01
## 1261 0.00 0.00
## 1297 0.00 0.02_*
## 1299 0.00 0.02_*
## 1310 0.00 0.01
## 1311 0.00 0.00
## 1331 0.00 0.00
## 1345 0.00 0.00
## 1350 0.00 0.02_*
## 1366 0.01 0.00
## 1390 0.00 0.01
## 1422 0.00 0.01
## 1429 0.01 0.01
## 1490 0.00 0.08_*
## 1497 0.01 0.01
## 1542 0.00 0.00
## 1554 0.00 0.02_*
## 1568 0.00 0.00
## 1601 0.00 0.01
## 1633 0.00 0.02_*
## 1687 0.00 0.01
## 1695 0.00 0.01_*
## 1723 0.00 0.01
## 1771 0.00 0.01
## 1797 0.01 0.01
## 1813 0.00 0.01_*
## 1817 0.01 0.04_*
## 1818 0.02 0.05_*
## 1824 0.00 0.01
## 1828 0.00 0.01
## 1840 0.00 0.02_*
## 1843 0.00 0.02_*
## 1844 0.00 0.01
## 1846 0.00 0.01
## 1848 0.00 0.02_*
## 1849 0.01 0.03_*
## 1858 0.00 0.00
## 1868 0.01 0.01_*
## 1888 0.00 0.02_*
## 1898 0.00 0.01
## 1916 0.00 0.01_*
## 1918 0.00 0.01
## 1933 0.00 0.01
## 1942 0.03 0.02_*
## 1945 0.00 0.04_*
## 1946 0.00 0.01

```

```

## 1958 0.01 0.02_*
## 1965 0.01 0.02_*
## 1968 0.00 0.01
## 1979 0.02 0.03_*
## 2016 0.00 0.00
## 2022 0.00 0.00
## 2036 0.00 0.00
## 2066 0.01 0.02_*
## 2087 0.00 0.03_*
## 2157 0.00 0.01
## 2174 0.00 0.00
## 2176 0.00 0.01
## 2204 0.00 0.02_*
## 2292 0.00 0.01
## 2305 0.00 0.01
## 2313 0.00 0.03_*
## 2314 0.00 0.10_*
## 2315 0.00 0.07_*
## 2316 0.00 0.04_*
## 2318 0.01 0.01
## 2323 0.00 0.00
## 2328 0.00 0.00
## 2347 0.01 0.02_*
## 2353 0.00 0.00
## 2355 0.00 0.00
## 2395 0.00 0.01_*
## 2426 0.00 0.00
## 2444 0.01 0.01
## 2452 0.00 0.02_*
## 2510 0.00 0.01
## 2546 0.00 0.01
## 2549 0.01 0.01
## 2563 0.00 0.01
## 2573 0.00 0.02_*
## 2579 0.00 0.01_*
## 2590 0.00 0.00
## 2594 0.00 0.02_*
## 2596 0.00 0.01
## 2598 0.00 0.01
## 2600 0.00 0.00
## 2626 0.01 0.01
## 2637 0.00 0.00
## 2642 0.00 0.01
## 2646 0.02 0.01
## 2659 0.01 0.01
## 2669 0.00 0.00
## 2671 0.00 0.01
## 2674 0.01 0.01
## 2682 0.00 0.01_*
## 2684 0.00 0.01
## 2689 0.00 0.01_*
## 2696 0.00 0.01
## 2699 0.00 0.02_*
## 2706 0.00 0.01

```

```

## 2714 0.25 0.08_*
## 2716 0.00 0.02_*
## 2718 0.00 0.01_*
## 2720 0.00 0.01
## 2724 0.01 0.03_*
## 2726 0.01 0.01
## 2727 0.01 0.01
## 2729 0.00 0.02_*
## 2733 0.00 0.02_*
## 2747 0.00 0.01
## 2783 0.00 0.01
## 2785 0.00 0.02_*
## 2789 0.00 0.00
## 2790 0.00 0.02_*
## 2809 0.00 0.00
## 2811 0.00 0.01
## 2826 0.00 0.01_*
## 2842 0.00 0.00
## 2858 0.00 0.00
## 2922 0.00 0.00
## 3040 0.00 0.00

```

## Remove outliers:

```

# filter outlier 282
step_model_no_282 <- cancer_county %>%
  dplyr::select(-median_age, -pct_black, -median_age_female, avg_household_size) %>%
  tibble::rowid_to_column() %>%
  filter(rowid != 282) %>%
  na.omit()

step_no_282 <- update(stepwise_model, . ~ ., data = step_model_no_282)
summary(step_no_282)

##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##     median_age_male + pct_unemployed16_over + pct_white + pct_asian +
##     pct_other_race + pct_married_households + pct_upto_hs18_24 +
##     pct_with_coverage + income_cat, data = step_model_no_282)
##
## Residuals:
##      Min      1Q      Median      3Q      Max 
## -93.887 -11.689  -0.051   11.936  126.194 
##
## Coefficients:
##                               Estimate Std. Error t value Pr(>|t|)    
## (Intercept)             69.286035  11.623979  5.961 2.80e-09 ***  
## incidence_rate          0.228536   0.007592 30.102 < 2e-16 ***  
## poverty_percent         1.045635   0.118358  8.834 < 2e-16 ***  
## median_age_male        -0.214490   0.085989 -2.494  0.01267 *    
## pct_unemployed16_over   0.750476   0.151124  4.966 7.21e-07 *** 

```

```

## pct_white          0.084336  0.034622   2.436  0.01491 *
## pct_asian         -0.587738  0.166433  -3.531  0.00042 ***
## pct_other_race    -0.920930  0.118468  -7.774  1.04e-14 ***
## pct_married_households -0.224288  0.085471  -2.624  0.00873 **
## pct_upto_hs18_24   0.385467  0.037529  10.271 < 2e-16 ***
## pct_with_coverage -0.215951  0.098701  -2.188  0.02875 *
## income_cat        -4.039980  1.490114  -2.711  0.00674 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 20.62 on 3034 degrees of freedom
## Multiple R-squared:  0.4501, Adjusted R-squared:  0.4481
## F-statistic: 225.7 on 11 and 3034 DF,  p-value: < 2.2e-16

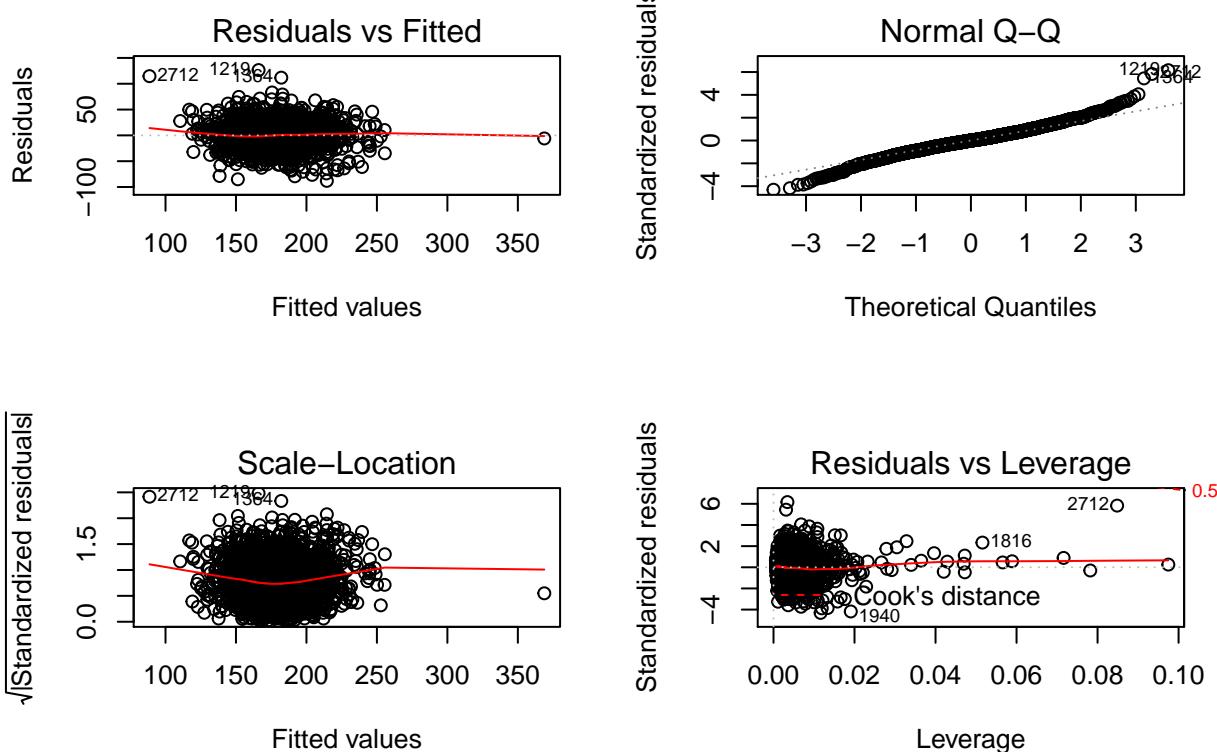
# filter #282, 1059
step_model_no_all <- cancer_county %>%
  dplyr::select(-median_age, -pct_black, -median_age_female, avg_household_size) %>%
  tibble::rowid_to_column() %>%
  filter(rowid != 282 & rowid != 1059) %>%
  na.omit()

step_no_all <- update(stepwise_model, . ~ ., data = step_model_no_all)
summary(step_no_all)

##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##     median_age_male + pct_unemployed16_over + pct_white + pct_asian +
##     pct_other_race + pct_married_households + pct_upto_hs18_24 +
##     pct_with_coverage + income_cat, data = step_model_no_all)
##
## Residuals:
##      Min       1Q   Median       3Q      Max 
## -88.069 -11.710 -0.021  11.854 126.460 
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 65.423198  11.615333  5.632 1.94e-08 ***
## incidence_rate 0.227187  0.007572 30.002 < 2e-16 ***
## poverty_percent 1.073476  0.118117  9.088 < 2e-16 ***
## median_age_male -0.212629  0.085703 -2.481 0.013155 *  
## pct_unemployed16_over 0.820282  0.151376  5.419 6.47e-08 ***
## pct_white        0.091233  0.034539  2.641 0.008297 ** 
## pct_asian        -0.583348  0.165881 -3.517 0.000443 *** 
## pct_other_race   -0.900209  0.118158 -7.619 3.40e-14 ***
## pct_married_households -0.211546  0.085231 -2.482 0.013117 *  
## pct_upto_hs18_24  0.382276  0.037410 10.218 < 2e-16 ***
## pct_with_coverage -0.188060  0.098557 -1.908 0.056468 .  
## income_cat       -4.075167  1.485162 -2.744 0.006107 ** 
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 20.55 on 3033 degrees of freedom
## Multiple R-squared:  0.4528, Adjusted R-squared:  0.4508
## F-statistic: 228.2 on 11 and 3033 DF,  p-value: < 2.2e-16

```

```
par(mfrow = c(2,2))
plot(step_no_all)
```



## Separate Income Level

```
income_low_data <- cancer_county %>% filter(income_cat == 0) %>% dplyr::select(-income_cat)
income_high_data <- cancer_county %>% filter(income_cat == 1) %>% dplyr::select(-income_cat)

model_low_income <- lm(target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + pct_unemployed16_over + pct_white + pct_asian +
  pct_other_race + pct_married_households + pct_upto_hs18_24 +
  pct_with_coverage, data = income_low_data)

model_high_income <- lm(target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + pct_unemployed16_over + pct_white + pct_asian +
  pct_other_race + pct_married_households + pct_upto_hs18_24 +
  pct_with_coverage, data = income_high_data)

summary_table = cbind(summary(model_low_income)$coefficients[,1],
                      summary(model_low_income)$coefficients[,4],
                      summary(model_high_income)$coefficients[,1],
                      summary(model_high_income)$coefficients[,4])

summary_table

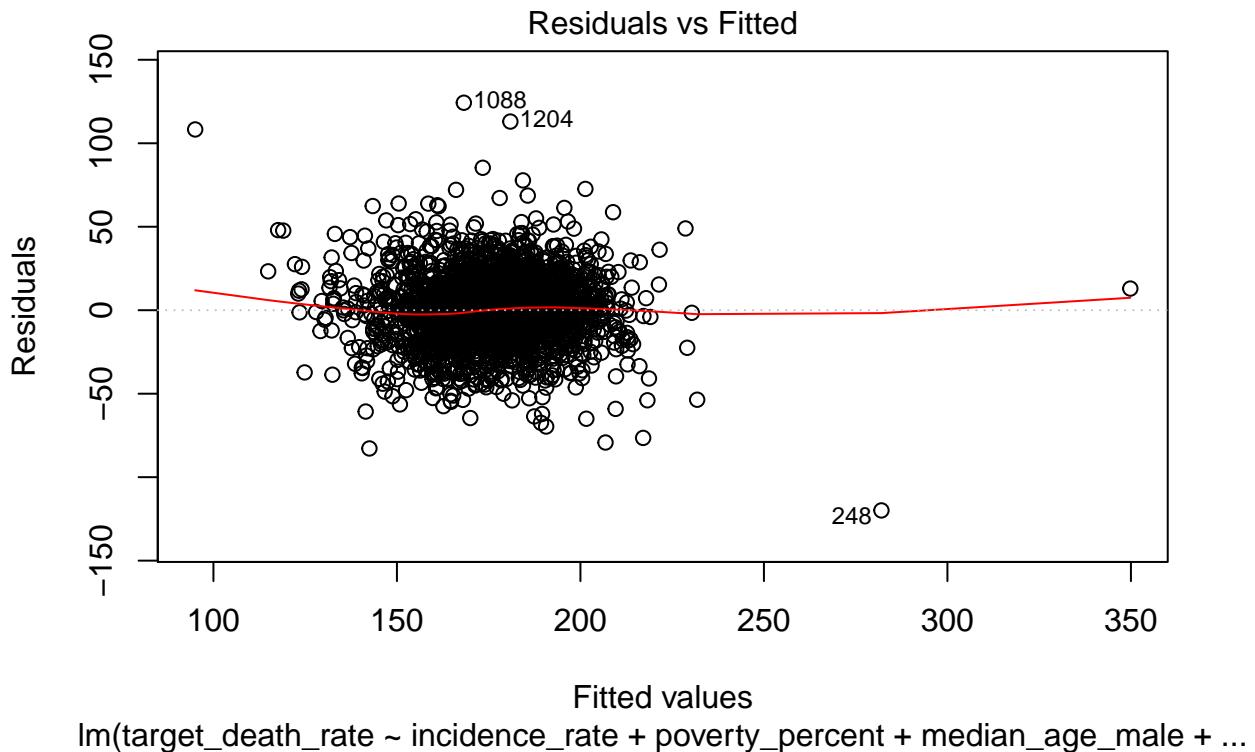
## [1] [2] [3] [4]
## (Intercept) -9.98041269 7.616353e-01 95.16691074 8.652255e-15
## incidence_rate 0.27773708 2.339449e-33 0.20015626 1.696636e-126
```

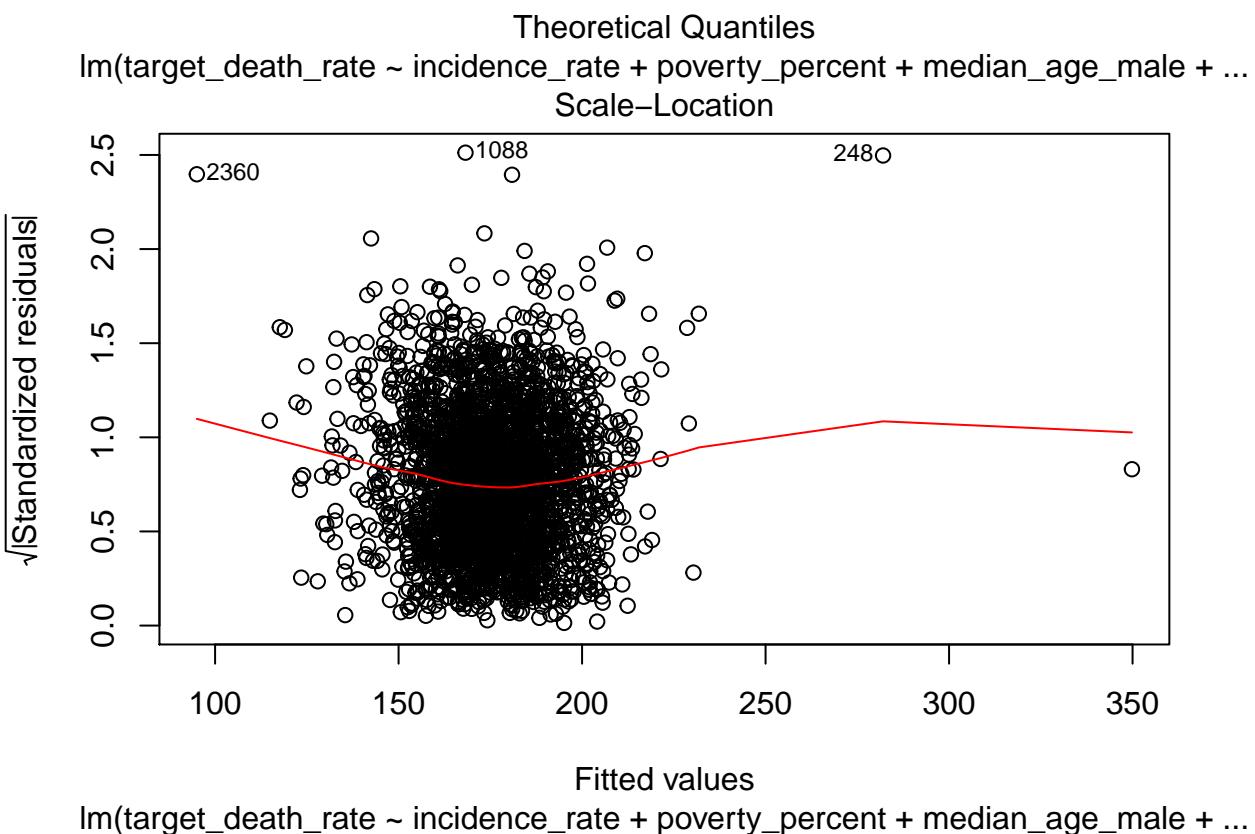
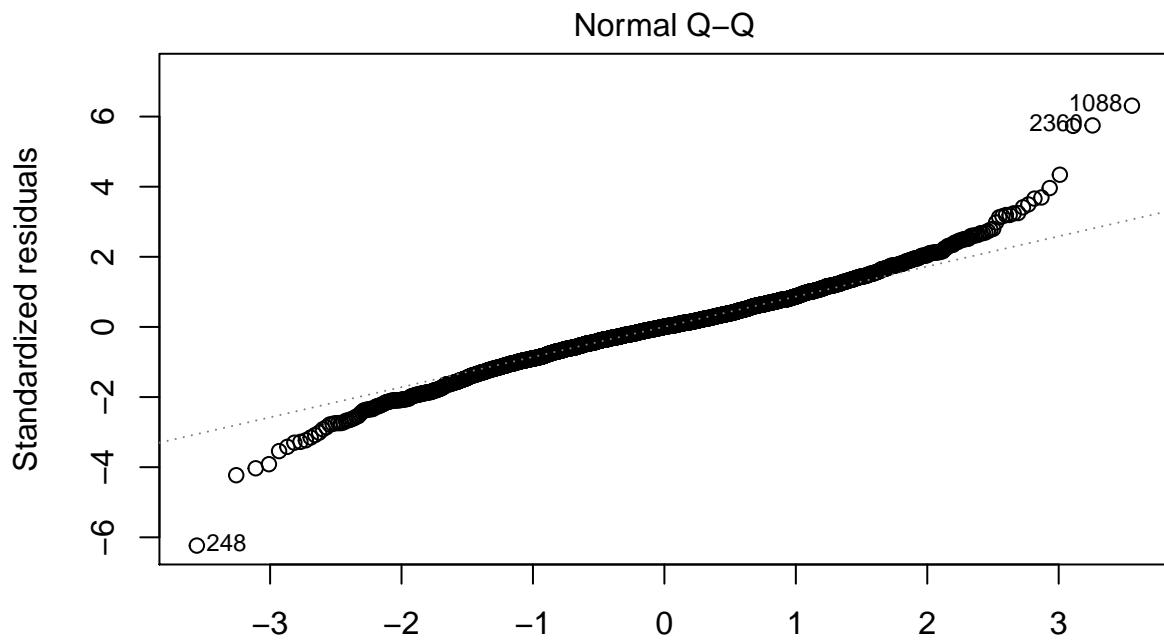
```

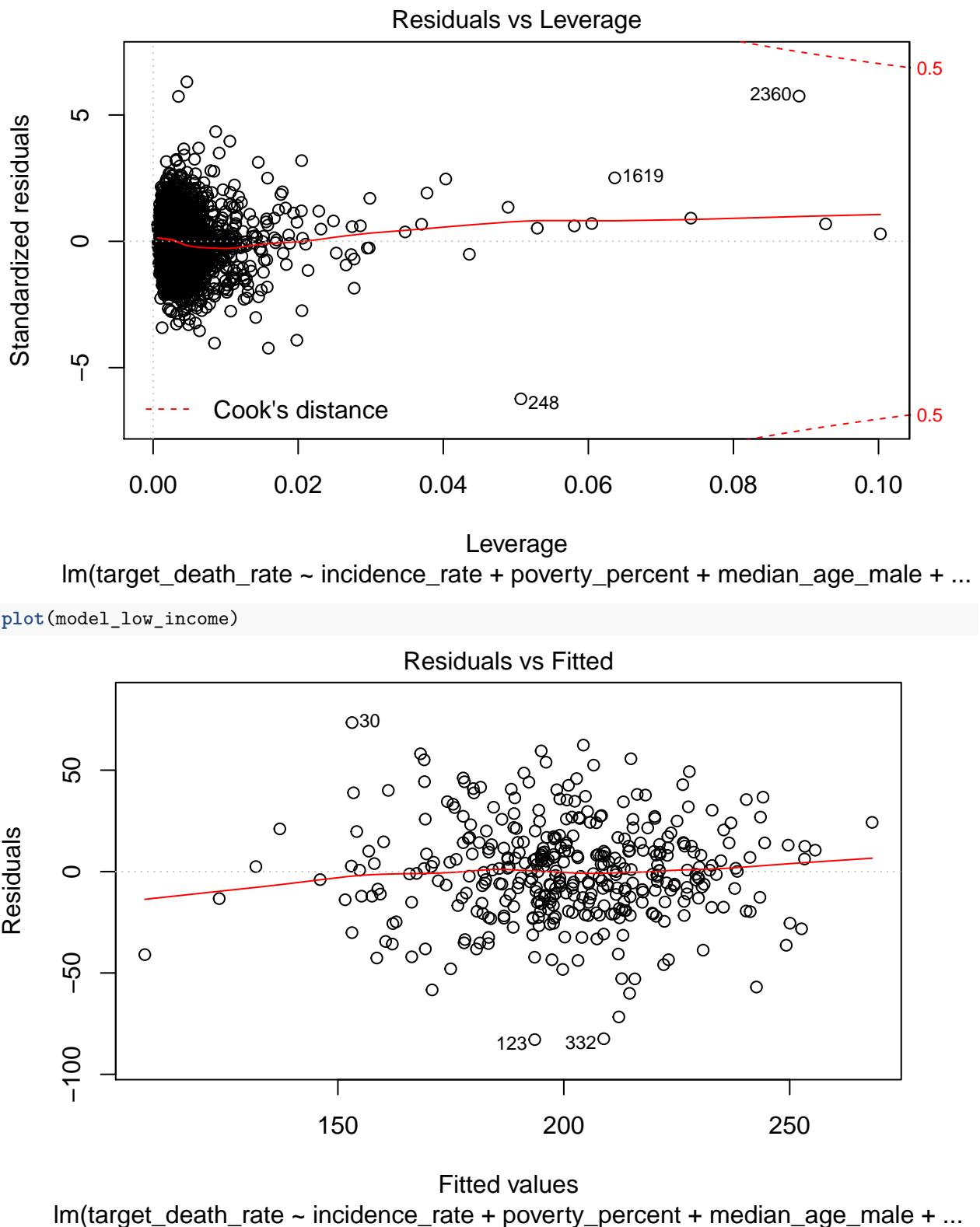
## poverty_percent      0.63536720 5.154360e-02  0.94031205 3.006044e-13
## median_age_male     -0.22206171 4.872790e-01 -0.27457885 1.840149e-03
## pct_unemployed16_over 0.01592747 9.670511e-01  0.98793425 1.865499e-09
## pct_white           0.20031237 3.292692e-02 -0.01723638 6.683679e-01
## pct_asian            -5.19387943 1.079811e-02 -0.77626937 3.481646e-06
## pct_other_race       -0.90270280 5.383448e-03 -1.07379851 6.805986e-17
## pct_married_households -0.45314232 1.701884e-01 -0.25796135 3.124664e-03
## pct_upto_hs18_24      0.28738237 1.442042e-02  0.40498578 7.434234e-25
## pct_with_coverage     0.92366136 2.082438e-03 -0.27491829 9.122972e-03

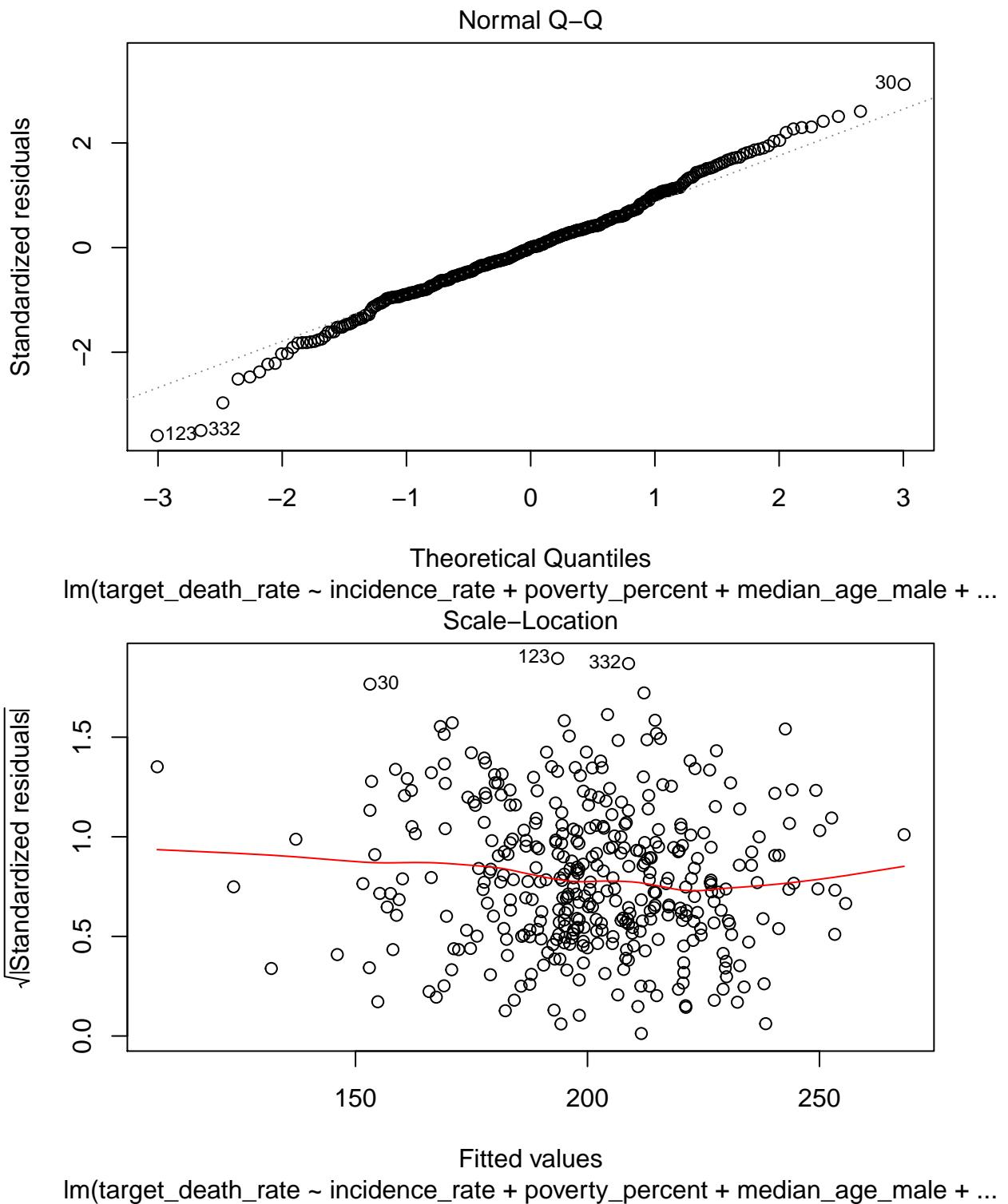
plot(model_high_income)

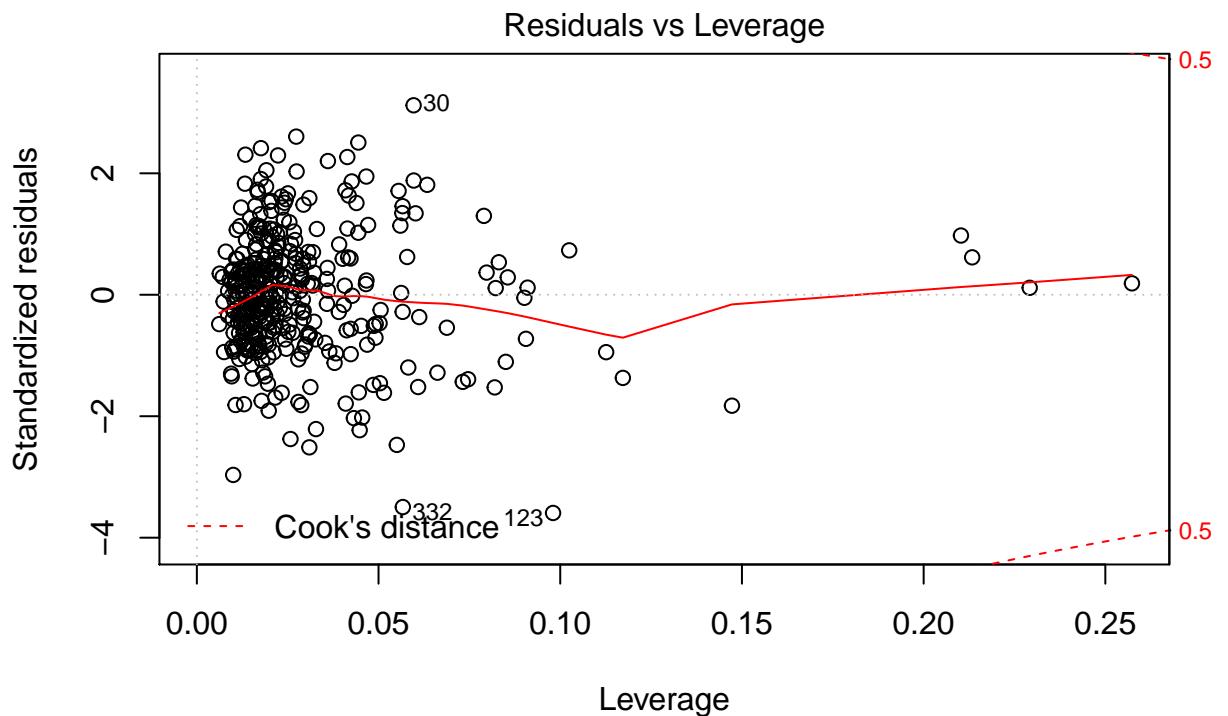
```











```
lm(target_death_rate ~ incidence_rate + poverty_percent + median_age_male + ...)
```

```
dim(income_high_data)
```

```
## [1] 2669 15
```

```
dim(income_low_data)
```

```
## [1] 378 15
```

### Stepwise by Income Level

```
# low income
full_model_low <- lm(target_death_rate ~ ., data = income_low_data)
summary(full_model_low)
```

```
##
## Call:
## lm(formula = target_death_rate ~ ., data = income_low_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -71.007 -13.685    0.306   13.741   70.463
##
## Coefficients:
##                               Estimate Std. Error t value Pr(>|t|)
## (Intercept)                 -5.621931  35.161833 -0.160  0.873059
## incidence_rate                0.284582   0.020753 13.713 < 2e-16 ***
## poverty_percent                0.699413   0.326259  2.144  0.032718 *
## median_age                  -0.001782   0.033246 -0.054  0.957295
## median_age_male              -1.604684   0.606440 -2.646  0.008498 **
## median_age_female              1.688740   0.597041  2.829  0.004936 **
```

```

## avg_household_size      -1.042002   3.058094  -0.341  0.733500
## pct_unemployed16_over  -0.009237   0.387431  -0.024  0.980992
## pct_white                -0.087916   0.135784  -0.647  0.517736
## pct_black                -0.374241   0.121281  -3.086  0.002186 **
## pct_asian                 -5.128157   2.034947  -2.520  0.012162 *
## pct_other_race            -1.134103   0.334695  -3.388  0.000780 ***
## pct_married_households   -0.531820   0.332391  -1.600  0.110472
## pct_upto_hs18_24           0.221861   0.117064   1.895  0.058858 .
## pct_with_coverage          1.095918   0.300486   3.647  0.000304 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 23.97 on 363 degrees of freedom
## Multiple R-squared:  0.4999, Adjusted R-squared:  0.4806
## F-statistic: 25.92 on 14 and 363 DF,  p-value: < 2.2e-16
step(full_model_low, direction = "backward")

## Start:  AIC=2416.34
## target_death_rate ~ incidence_rate + poverty_percent + median_age +
##   median_age_male + median_age_female + avg_household_size +
##   pct_unemployed16_over + pct_white + pct_black + pct_asian +
##   pct_other_race + pct_married_households + pct_upto_hs18_24 +
##   pct_with_coverage
##
##                                Df Sum of Sq    RSS     AIC
## - pct_unemployed16_over   1      0 208559 2414.3
## - median_age               1      2 208560 2414.3
## - avg_household_size       1     67 208625 2414.5
## - pct_white                 1    241 208799 2414.8
## <none>                      208559 2416.3
## - pct_married_households  1   1471 210029 2417.0
## - pct_upto_hs18_24          1   2064 210622 2418.1
## - poverty_percent           1   2640 211199 2419.1
## - pct_asian                  1   3649 212207 2420.9
## - median_age_male            1   4023 212581 2421.6
## - median_age_female          1   4597 213155 2422.6
## - pct_black                   1   5471 214029 2424.1
## - pct_other_race              1   6597 215155 2426.1
## - pct_with_coverage           1   7642 216201 2427.9
## - incidence_rate              1  108036 316594 2572.1
##
## Step:  AIC=2414.35
## target_death_rate ~ incidence_rate + poverty_percent + median_age +
##   median_age_male + median_age_female + avg_household_size +
##   pct_white + pct_black + pct_asian + pct_other_race + pct_married_households +
##   pct_upto_hs18_24 + pct_with_coverage
##
##                                Df Sum of Sq    RSS     AIC
## - median_age                  1      2 208561 2412.3
## - avg_household_size           1     67 208626 2412.5
## - pct_white                     1    245 208804 2412.8
## <none>                         208559 2414.3
## - pct_married_households      1   1478 210037 2415.0
## - pct_upto_hs18_24              1   2064 210623 2416.1

```

```

## - poverty_percent      1    2857 211416 2417.5
## - pct_asian            1    3663 212222 2418.9
## - median_age_male      1    4050 212609 2419.6
## - median_age_female     1    4621 213180 2420.6
## - pct_black             1    5494 214053 2422.2
## - pct_other_race        1    6872 215431 2424.6
## - pct_with_coverage     1    7671 216230 2426.0
## - incidence_rate        1   108727 317286 2570.9
##
## Step: AIC=2412.35
## target_death_rate ~ incidence_rate + poverty_percent + median_age_male +
##          median_age_female + avg_household_size + pct_white + pct_black +
##          pct_asian + pct_other_race + pct_married_households + pct_upto_hs18_24 +
##          pct_with_coverage
##
##                                     Df Sum of Sq   RSS   AIC
## - avg_household_size       1      68 208628 2410.5
## - pct_white                 1     243 208804 2410.8
## <none>                      208561 2412.3
## - pct_married_households   1    1479 210039 2413.0
## - pct_upto_hs18_24          1    2064 210625 2414.1
## - poverty_percent           1    2875 211436 2415.5
## - pct_asian                 1    3663 212223 2416.9
## - median_age_male           1    4058 212618 2417.6
## - median_age_female          1    4621 213182 2418.6
## - pct_black                  1    5517 214078 2420.2
## - pct_other_race             1    6873 215434 2422.6
## - pct_with_coverage          1    7806 216367 2424.2
## - incidence_rate             1   109612 318173 2570.0
##
## Step: AIC=2410.47
## target_death_rate ~ incidence_rate + poverty_percent + median_age_male +
##          median_age_female + pct_white + pct_black + pct_asian + pct_other_race +
##          pct_married_households + pct_upto_hs18_24 + pct_with_coverage
##
##                                     Df Sum of Sq   RSS   AIC
## - pct_white                  1     200 208828 2408.8
## <none>                      208628 2410.5
## - pct_married_households    1    1566 210195 2411.3
## - pct_upto_hs18_24            1    2130 210759 2412.3
## - poverty_percent             1    2808 211436 2413.5
## - pct_asian                   1    3596 212224 2414.9
## - median_age_male              1    4072 212701 2415.8
## - median_age_female             1    4847 213475 2417.2
## - pct_black                     1    5507 214136 2418.3
## - pct_other_race                1    6987 215616 2420.9
## - pct_with_coverage              1    7806 216435 2422.4
## - incidence_rate                 1   110176 318805 2568.8
##
## Step: AIC=2408.83
## target_death_rate ~ incidence_rate + poverty_percent + median_age_male +
##          median_age_female + pct_black + pct_asian + pct_other_race +
##          pct_married_households + pct_upto_hs18_24 + pct_with_coverage
##

```

```

##                                     Df Sum of Sq    RSS     AIC
## <none>                               208828 2408.8
## - pct_upto_hs18_24                  1      2393 211222 2411.1
## - pct_married_households            1      2520 211349 2411.4
## - poverty_percent                  1      2690 211518 2411.7
## - pct_asian                        1      3771 212599 2413.6
## - median_age_male                 1      4023 212852 2414.1
## - median_age_female                1      4647 213476 2415.2
## - pct_other_race                  1      6788 215617 2418.9
## - pct_with_coverage               1      7623 216452 2420.4
## - pct_black                        1      8384 217212 2421.7
## - incidence_rate                  1     110965 319793 2567.9

##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##      median_age_male + median_age_female + pct_black + pct_asian +
##      pct_other_race + pct_married_households + pct_upto_hs18_24 +
##      pct_with_coverage, data = income_low_data)
##
## Coefficients:
##             (Intercept)           incidence_rate       poverty_percent
##                   -6.7837              0.2831                  0.6631
## median_age_male      median_age_female       pct_black
##                   -1.5990              1.6474                 -0.3151
##          pct_asian          pct_other_race  pct_married_households
##                   -5.1184              -1.1093                 -0.6253
##      pct_upto_hs18_24      pct_with_coverage
##                   0.2354              1.0621

backward_model_low <- lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
                           median_age_male + median_age_female + pct_black + pct_asian +
                           pct_other_race + pct_married_households + pct_upto_hs18_24 +
                           pct_with_coverage, data = income_low_data)
summary(backward_model_low)

##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##      median_age_male + median_age_female + pct_black + pct_asian +
##      pct_other_race + pct_married_households + pct_upto_hs18_24 +
##      pct_with_coverage, data = income_low_data)
##
## Residuals:
##      Min    1Q   Median    3Q   Max
## -71.584 -13.813 -0.024  13.851  69.455
##
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)
## (Intercept) -6.78373  32.21806 -0.211 0.833350
## incidence_rate 0.28314  0.02028 13.965 < 2e-16 ***
## poverty_percent 0.66310  0.30497  2.174 0.030321 *
## median_age_male -1.59902  0.60135 -2.659 0.008180 **
## median_age_female 1.64742  0.57646  2.858 0.004509 **
## pct_black      -0.31512  0.08210 -3.838 0.000146 ***

```

```

## pct_asian           -5.11844   1.98835  -2.574 0.010438 *
## pct_other_race     -1.10933   0.32118  -3.454 0.000617 ***
## pct_married_households -0.62529   0.29711  -2.105 0.036009 *
## pct_upto_hs18_24    0.23543   0.11479   2.051 0.040992 *
## pct_with_coverage   1.06213   0.29018   3.660 0.000289 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 23.85 on 367 degrees of freedom
## Multiple R-squared:  0.4992, Adjusted R-squared:  0.4856
## F-statistic: 36.59 on 10 and 367 DF,  p-value: < 2.2e-16

# high income
full_model_high <- lm(target_death_rate ~ ., data = income_high_data)
summary(full_model_high)

##
## Call:
## lm(formula = target_death_rate ~ ., data = income_high_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -119.771  -11.264    0.223   11.410  123.636
##
## Coefficients:
##                               Estimate Std. Error t value Pr(>|t|)    
## (Intercept)             92.662611  13.217700  7.010 3.00e-12 ***
## incidence_rate          0.199461   0.007958 25.063 < 2e-16 ***
## poverty_percent         0.927587   0.129691  7.152 1.10e-12 ***
## median_age              -0.002868   0.008358 -0.343 0.731525  
## median_age_male         -0.247908   0.223519 -1.109 0.267482  
## median_age_female        -0.038208   0.219559 -0.174 0.861860  
## avg_household_size     -0.243315   1.044062 -0.233 0.815743  
## pct_unemployed16_over   0.988191   0.165310  5.978 2.57e-09 ***
## pct_white                0.038794   0.066375  0.584 0.558956  
## pct_black                0.073062   0.067584  1.081 0.279767  
## pct_asian                -0.699572   0.181421 -3.856 0.000118 *** 
## pct_other_race           -1.029068   0.134008 -7.679 2.24e-14 ***
## pct_married_households  -0.254272   0.089532 -2.840 0.004546 ** 
## pct_upto_hs18_24          0.409273   0.039506 10.360 < 2e-16 ***
## pct_with_coverage         -0.294225   0.108429 -2.714 0.006700 ** 
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 19.75 on 2654 degrees of freedom
## Multiple R-squared:  0.396, Adjusted R-squared:  0.3928
## F-statistic: 124.3 on 14 and 2654 DF,  p-value: < 2.2e-16

step(full_model_high, direction = "backward")

## Start:  AIC=15938.37
## target_death_rate ~ incidence_rate + poverty_percent + median_age +
##   median_age_male + median_age_female + avg_household_size +
##   pct_unemployed16_over + pct_white + pct_black + pct_asian +
##   pct_other_race + pct_married_households + pct_upto_hs18_24 +

```

```

##      pct_with_coverage
##
##                                Df Sum of Sq      RSS     AIC
## - median_age_female          1       12 1034983 15936
## - avg_household_size         1       21 1034992 15936
## - median_age                 1       46 1035017 15936
## - pct_white                  1      133 1035104 15937
## - pct_black                  1      456 1035427 15938
## - median_age_male            1      480 1035451 15938
## <none>                      1034971 15938
## - pct_with_coverage          1      2871 1037843 15944
## - pct_married_households    1      3145 1038116 15944
## - pct_asian                  1      5799 1040770 15951
## - pct_unemployed16_over      1      13935 1048906 15972
## - poverty_percent            1      19949 1054920 15987
## - pct_other_race             1      22996 1057967 15995
## - pct_upto_hs18_24           1      41853 1076824 16042
## - incidence_rate              1      244965 1279936 16503
##
## Step:  AIC=15936.4
## target_death_rate ~ incidence_rate + poverty_percent + median_age +
##      median_age_male + avg_household_size + pct_unemployed16_over +
##      pct_white + pct_black + pct_asian + pct_other_race + pct_married_households +
##      pct_upto_hs18_24 + pct_with_coverage
##
##                                Df Sum of Sq      RSS     AIC
## - avg_household_size         1       18 1035001 15934
## - median_age                 1       46 1035029 15934
## - pct_white                  1      128 1035111 15935
## - pct_black                  1      445 1035428 15936
## <none>                      1034983 15936
## - pct_with_coverage          1      2976 1037959 15942
## - pct_married_households    1      3135 1038118 15942
## - median_age_male            1      3633 1038616 15944
## - pct_asian                  1      5793 1040776 15949
## - pct_unemployed16_over      1      13989 1048972 15970
## - poverty_percent            1      20051 1055034 15986
## - pct_other_race             1      23007 1057990 15993
## - pct_upto_hs18_24           1      42543 1077526 16042
## - incidence_rate              1      245682 1280665 16503
##
## Step:  AIC=15934.45
## target_death_rate ~ incidence_rate + poverty_percent + median_age +
##      median_age_male + pct_unemployed16_over + pct_white + pct_black +
##      pct_asian + pct_other_race + pct_married_households + pct_upto_hs18_24 +
##      pct_with_coverage
##
##                                Df Sum of Sq      RSS     AIC
## - median_age                 1       47 1035048 15933
## - pct_white                  1      138 1035140 15933
## - pct_black                  1      458 1035459 15934
## <none>                      1035001 15934
## - pct_with_coverage          1      2959 1037960 15940
## - pct_married_households    1      3409 1038410 15941

```

```

## - median_age_male      1      3805 1038806 15942
## - pct_asian            1      5832 1040834 15947
## - pct_unemployed16_over 1     14062 1049064 15968
## - poverty_percent       1     20167 1055168 15984
## - pct_other_race        1     23093 1058095 15991
## - pct_upto_hs18_24       1     42542 1077544 16040
## - incidence_rate         1    246317 1281319 16502
##
## Step: AIC=15932.57
## target_death_rate ~ incidence_rate + poverty_percent + median_age_male +
##          pct_unemployed16_over + pct_white + pct_black + pct_asian +
##          pct_other_race + pct_married_households + pct_upto_hs18_24 +
##          pct_with_coverage
##
##                               Df Sum of Sq      RSS      AIC
## - pct_white              1      136 1035185 15931
## - pct_black              1      452 1035500 15932
## <none>                  1035048 15933
## - pct_with_coverage      1     2953 1038002 15938
## - pct_married_households 1     3398 1038446 15939
## - median_age_male        1     3944 1038992 15941
## - pct_asian              1     5827 1040875 15946
## - pct_unemployed16_over  1     14025 1049073 15966
## - poverty_percent         1     20229 1055277 15982
## - pct_other_race          1     23114 1058162 15990
## - pct_upto_hs18_24         1     42533 1077581 16038
## - incidence_rate          1    246274 1281322 16500
##
## Step: AIC=15930.92
## target_death_rate ~ incidence_rate + poverty_percent + median_age_male +
##          pct_unemployed16_over + pct_black + pct_asian + pct_other_race +
##          pct_married_households + pct_upto_hs18_24 + pct_with_coverage
##
##                               Df Sum of Sq      RSS      AIC
## - pct_black              1      387 1035572 15930
## <none>                  1035185 15931
## - pct_with_coverage      1     2817 1038002 15936
## - pct_married_households 1     3262 1038447 15937
## - median_age_male        1     3811 1038996 15939
## - pct_asian              1     9206 1044390 15953
## - pct_unemployed16_over  1     13889 1049073 15964
## - poverty_percent         1     20411 1055596 15981
## - pct_other_race          1     27135 1062320 15998
## - pct_upto_hs18_24         1     42605 1077789 16037
## - incidence_rate          1    247263 1282448 16501
##
## Step: AIC=15929.92
## target_death_rate ~ incidence_rate + poverty_percent + median_age_male +
##          pct_unemployed16_over + pct_asian + pct_other_race + pct_married_households +
##          pct_upto_hs18_24 + pct_with_coverage
##
##                               Df Sum of Sq      RSS      AIC
## <none>                  1035572 15930
## - pct_with_coverage      1     2790 1038362 15935

```

```

## - pct_married_households 1      4110 1039682 15938
## - median_age_male         1      4387 1039959 15939
## - pct_asian               1      9084 1044655 15951
## - pct_unemployed16_over   1      15086 1050658 15966
## - poverty_percent         1      21094 1056666 15982
## - pct_other_race          1      27742 1063313 15998
## - pct_upto_hs18_24        1      43134 1078706 16037
## - incidence_rate          1      249078 1284650 16503

##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##     median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
##     pct_married_households + pct_upto_hs18_24 + pct_with_coverage,
##     data = income_high_data)
## 

## Coefficients:
##             (Intercept)           incidence_rate           poverty_percent
##                   94.8143                  0.2002                  0.9428
##             median_age_male    pct_unemployed16_over           pct_asian
##                   -0.2847                  1.0012                 -0.7501
##             pct_other_race    pct_married_households    pct_upto_hs18_24
##                   -1.0655                 -0.2695                  0.4069
##             pct_with_coverage
##                   -0.2801

backward_model_high <- lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
  pct_married_households + pct_upto_hs18_24 + pct_with_coverage,
  data = income_high_data)
summary(backward_model_high)

##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##     median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
##     pct_married_households + pct_upto_hs18_24 + pct_with_coverage,
##     data = income_high_data)
## 

## Residuals:
##      Min       1Q     Median       3Q      Max
## -120.096  -11.354    0.142    11.441   124.769
## 

## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 94.814283  12.167303   7.793 9.34e-15 ***
## incidence_rate 0.200212  0.007917  25.289 < 2e-16 ***
## poverty_percent 0.942798  0.128106   7.360 2.45e-13 ***
## median_age_male -0.284695  0.084824  -3.356 0.000801 ***
## pct_unemployed16_over  1.001186  0.160862   6.224 5.62e-10 ***
## pct_asian      -0.750069  0.155312  -4.829 1.45e-06 ***
## pct_other_race -1.065515  0.126248  -8.440 < 2e-16 ***
## pct_married_households -0.269476  0.082954  -3.249 0.001175 ** 
## pct_upto_hs18_24      0.406944  0.038668  10.524 < 2e-16 ***
## pct_with_coverage     -0.280093  0.104649  -2.676 0.007485 **
```

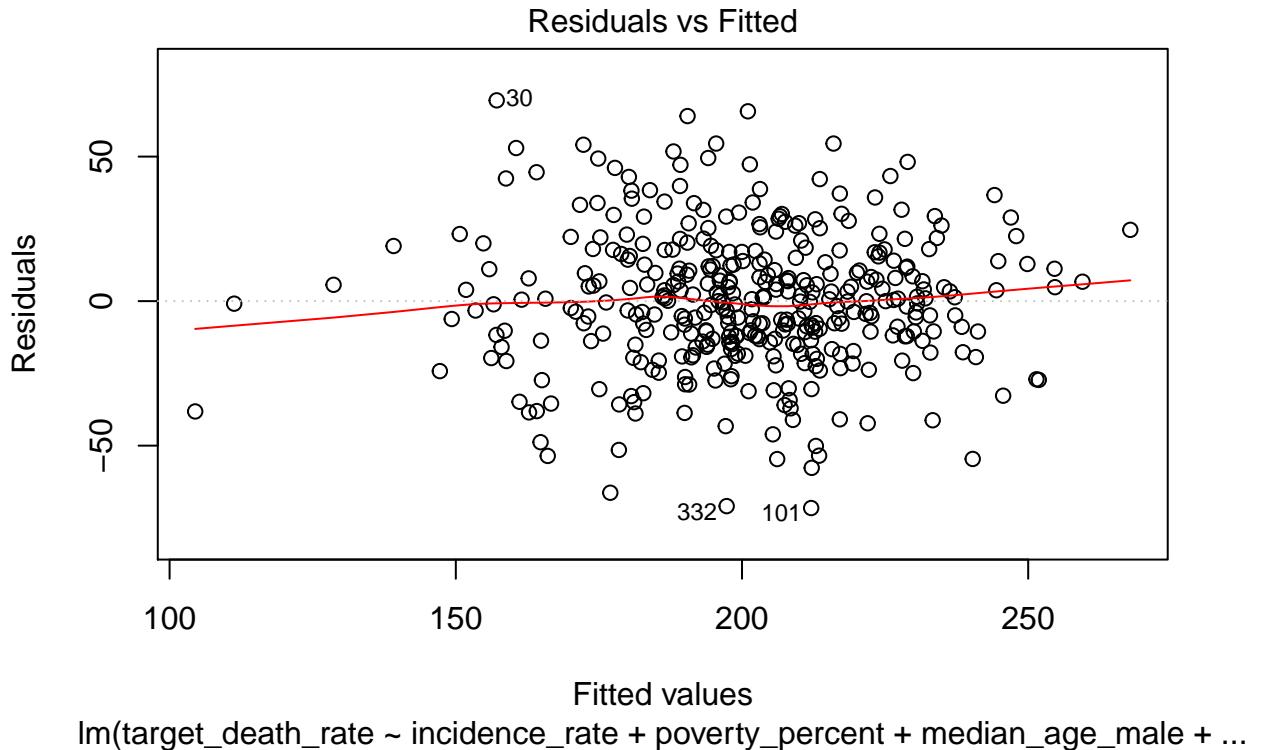
```

## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 19.73 on 2659 degrees of freedom
## Multiple R-squared: 0.3956, Adjusted R-squared: 0.3936
## F-statistic: 193.4 on 9 and 2659 DF, p-value: < 2.2e-16

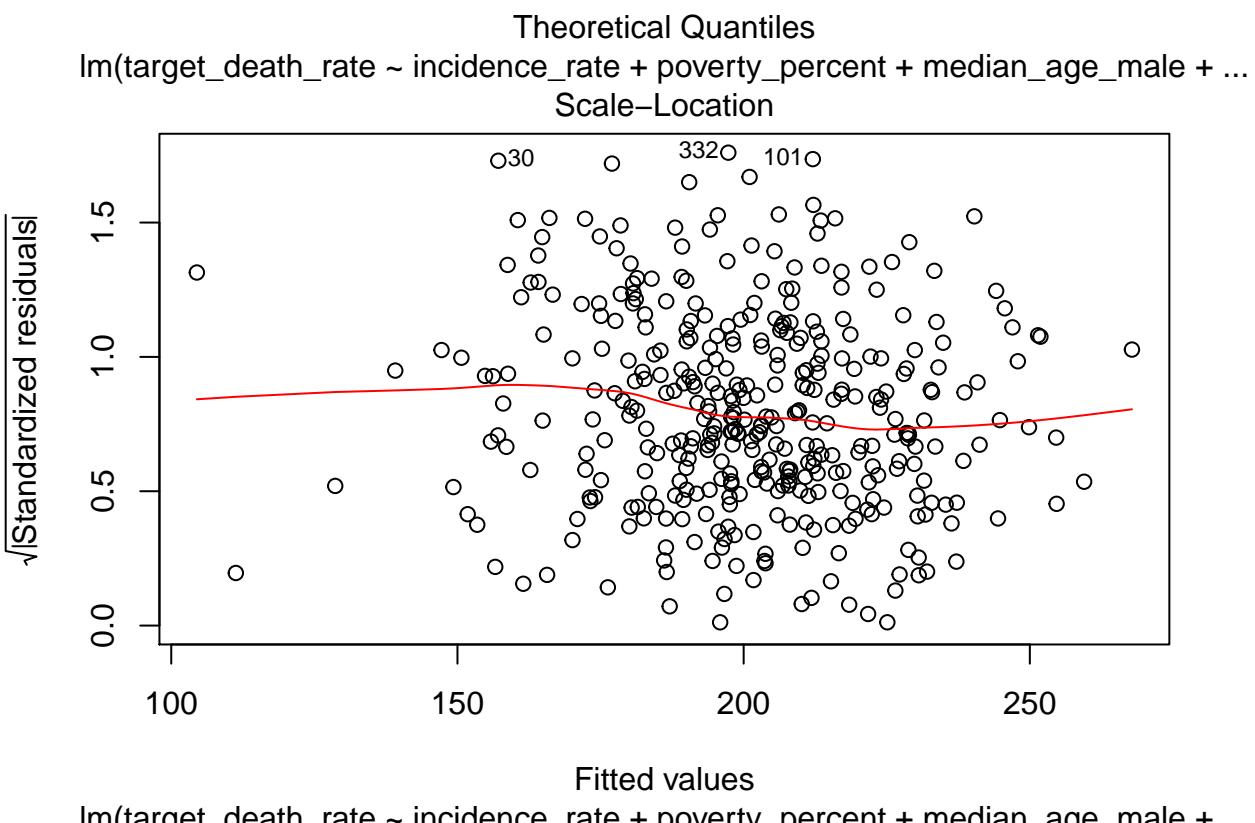
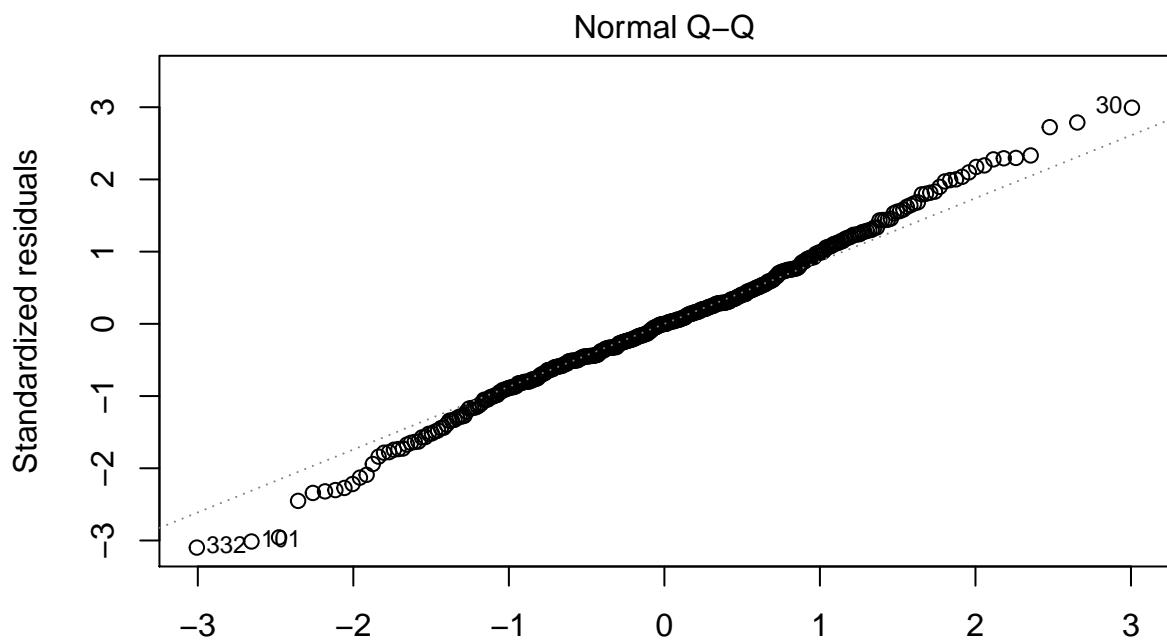
```

## Check Assumption

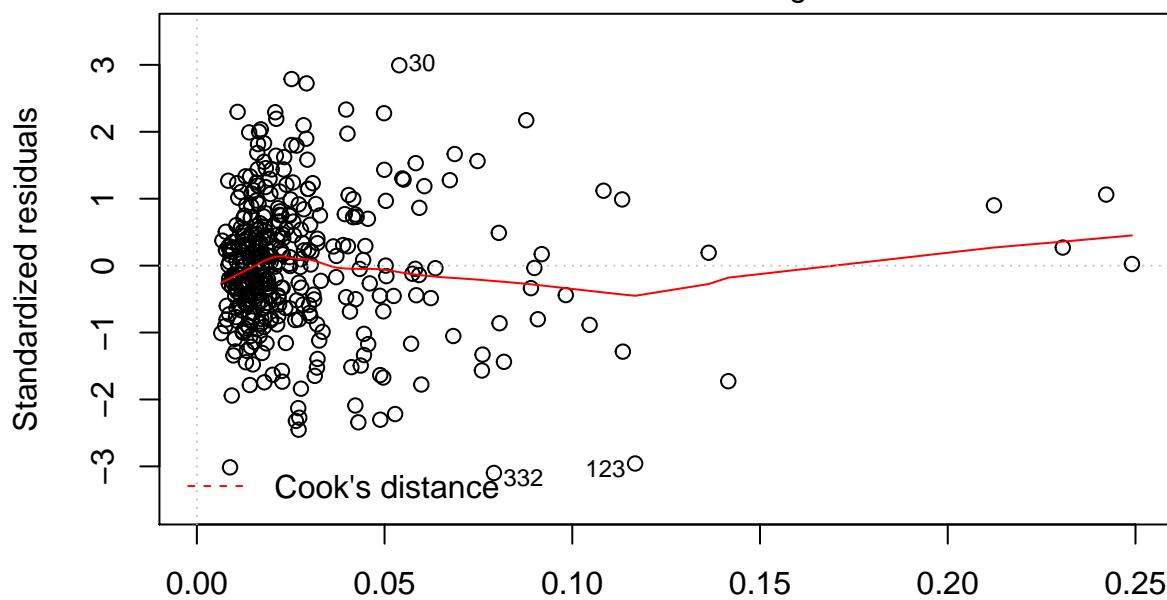
```
plot(backward_model_low)
```



Fitted values  
 $\text{lm}(\text{target\_death\_rate} \sim \text{incidence\_rate} + \text{poverty\_percent} + \text{median\_age\_male} + \dots)$



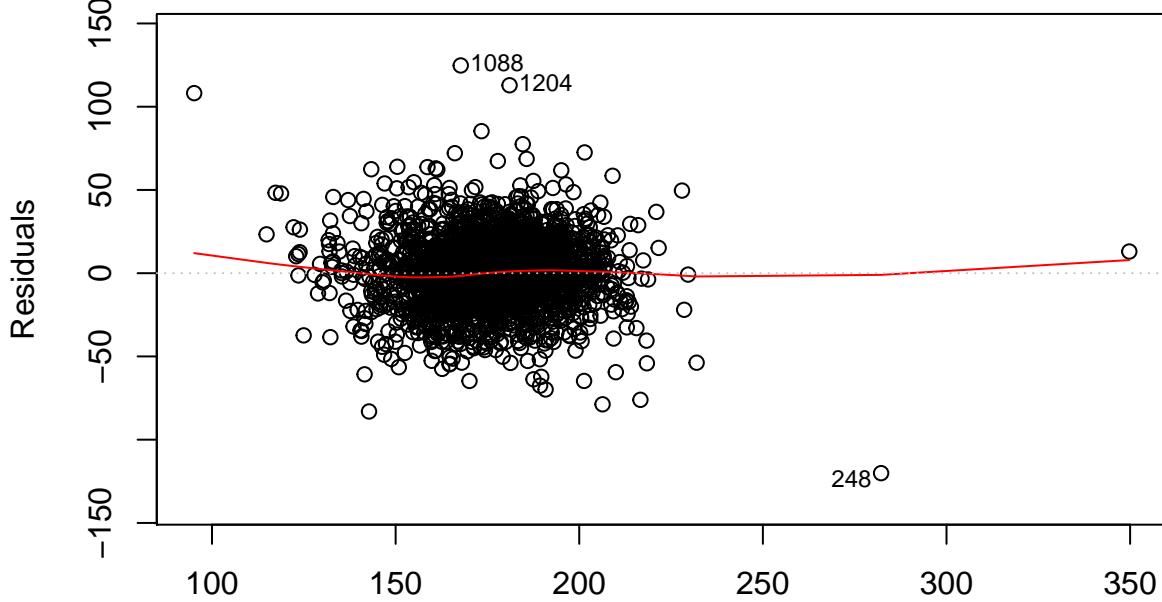
Residuals vs Leverage



Leverage

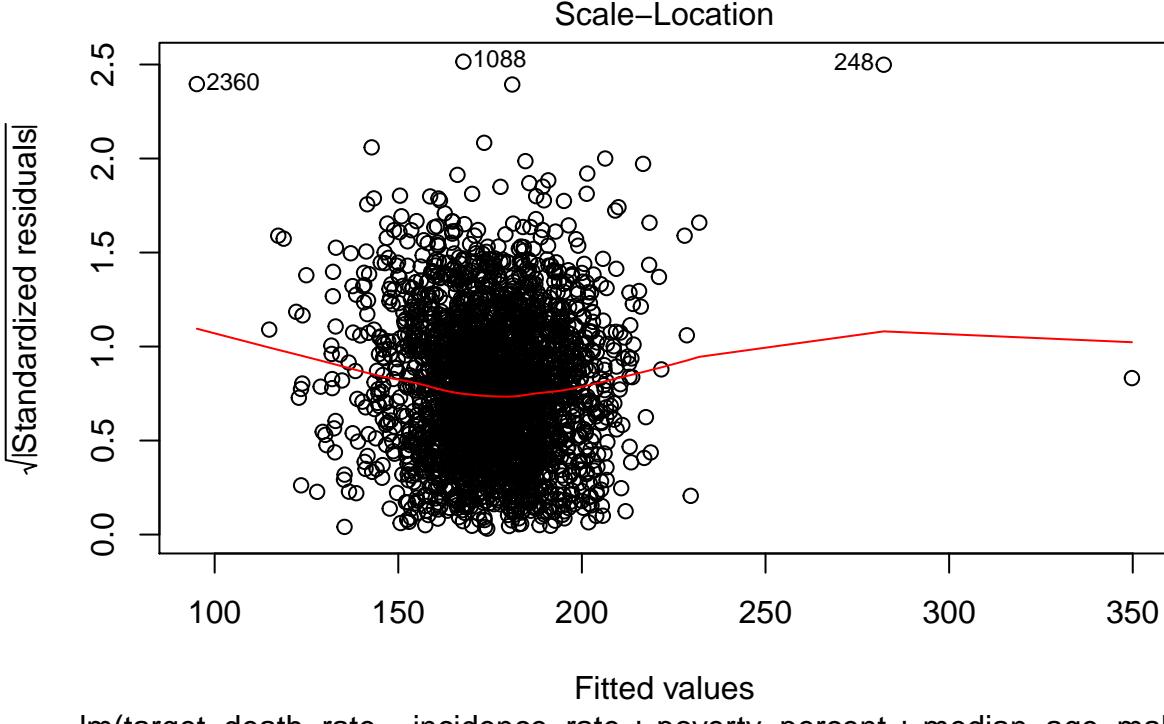
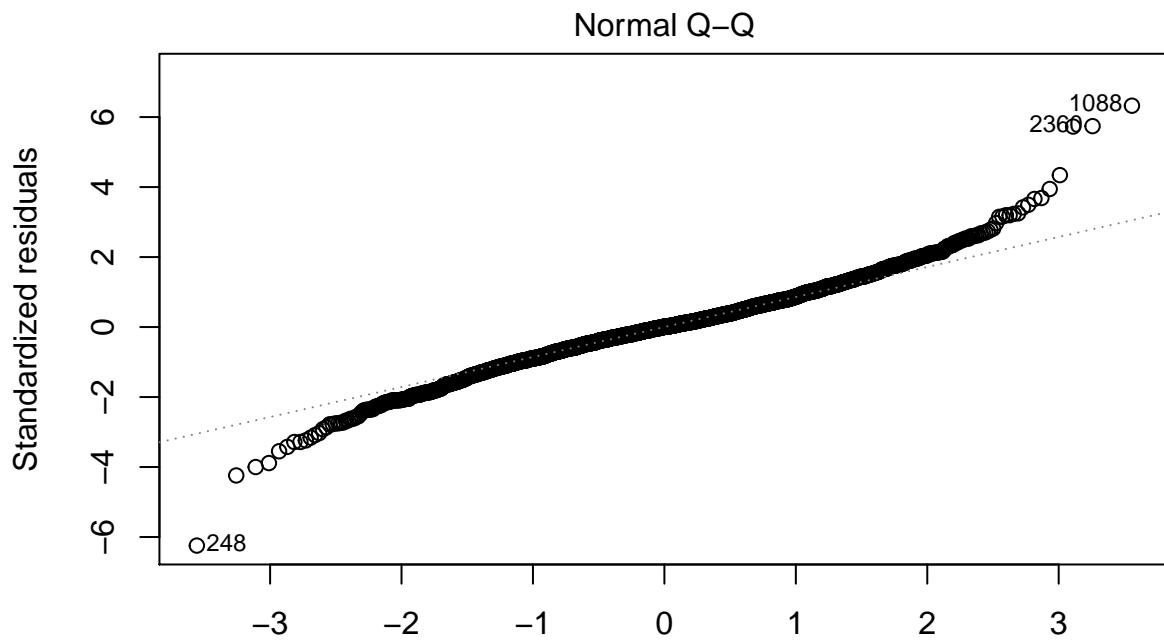
```
lm(target_death_rate ~ incidence_rate + poverty_percent + median_age_male + ...  
plot(backward_model_high)
```

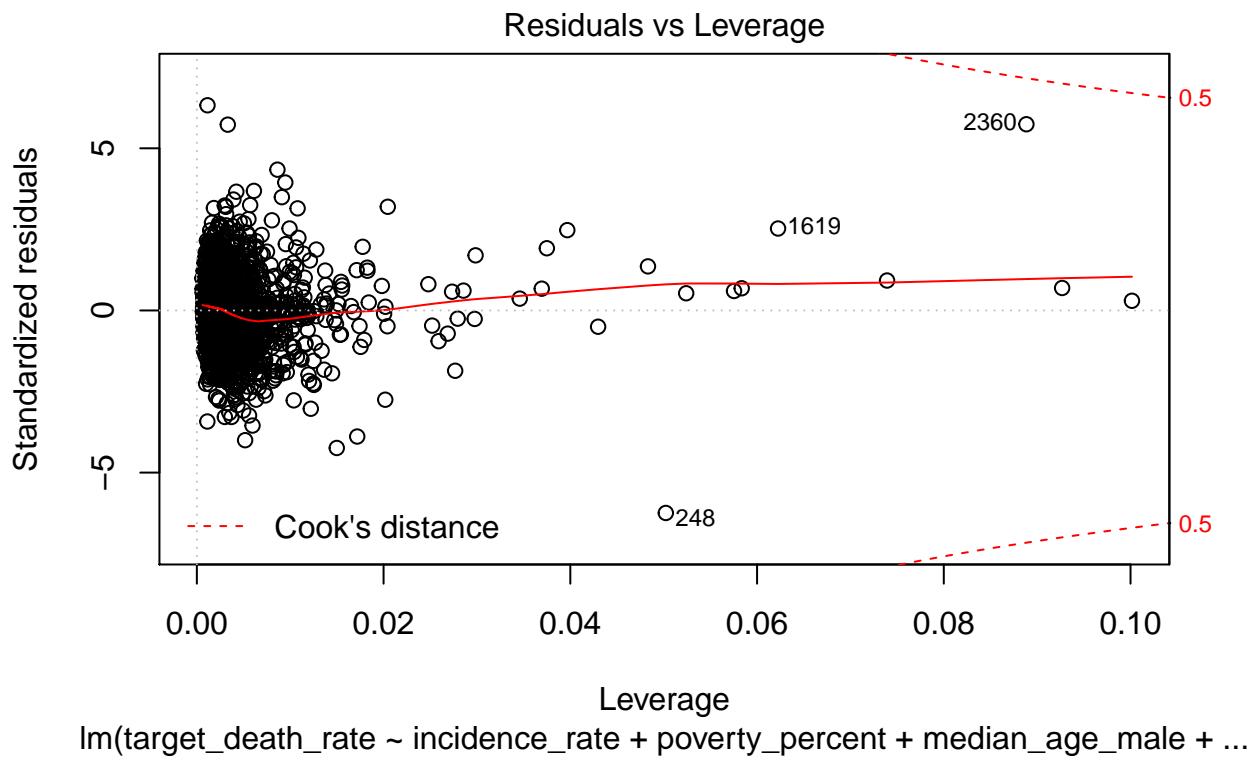
Residuals vs Fitted



Fitted values

```
lm(target_death_rate ~ incidence_rate + poverty_percent + median_age_male + ...
```





### Criterion-based Approach

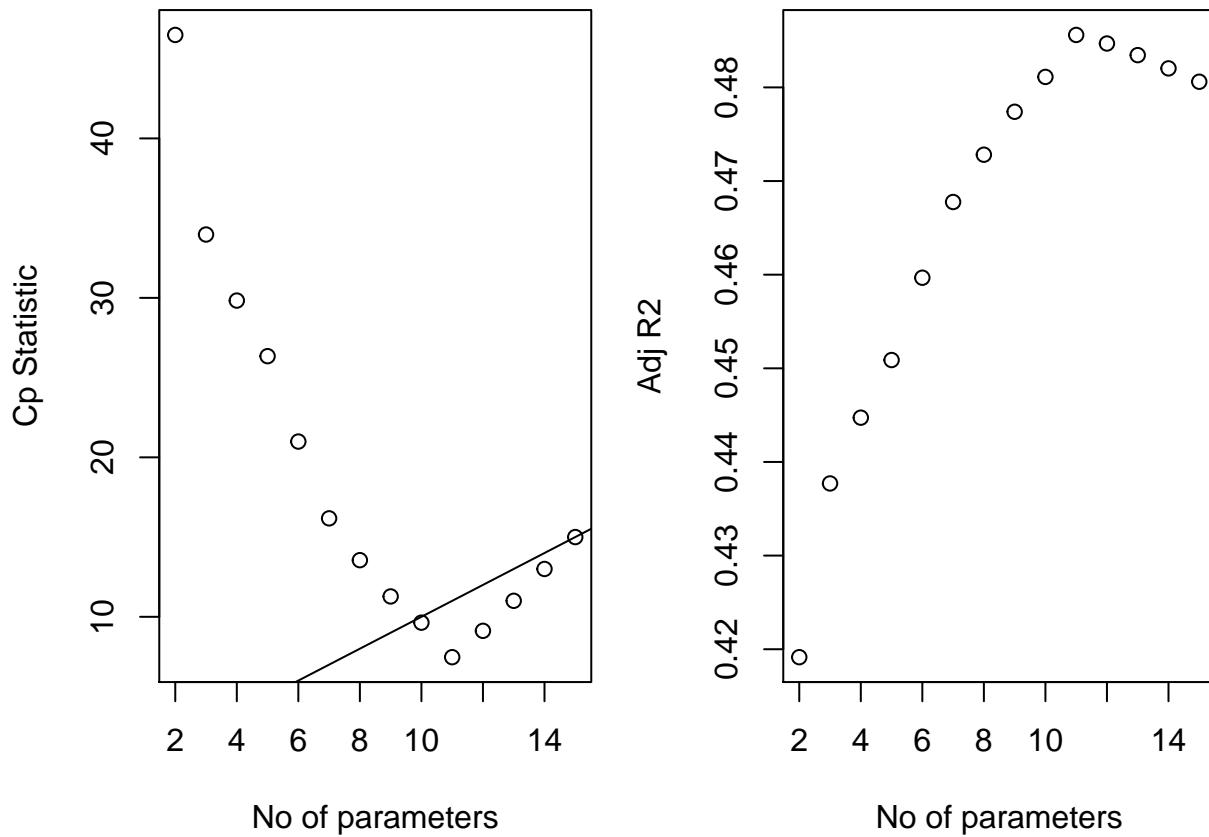
Low income

```
criterion_stats_low =
  regsubsets(target_death_rate ~ ., nvmax = 14, data = income_low_data) %>%
  summary()

par(mar = c(4,4,1,1))
par(mfrow = c(1,2))

plot(2:15, criterion_stats_low$cp, xlab = "No of parameters", ylab = "Cp Statistic")
abline(0,1)

plot(2:15, criterion_stats_low$adjr2, xlab = "No of parameters", ylab = "Adj R2")
```



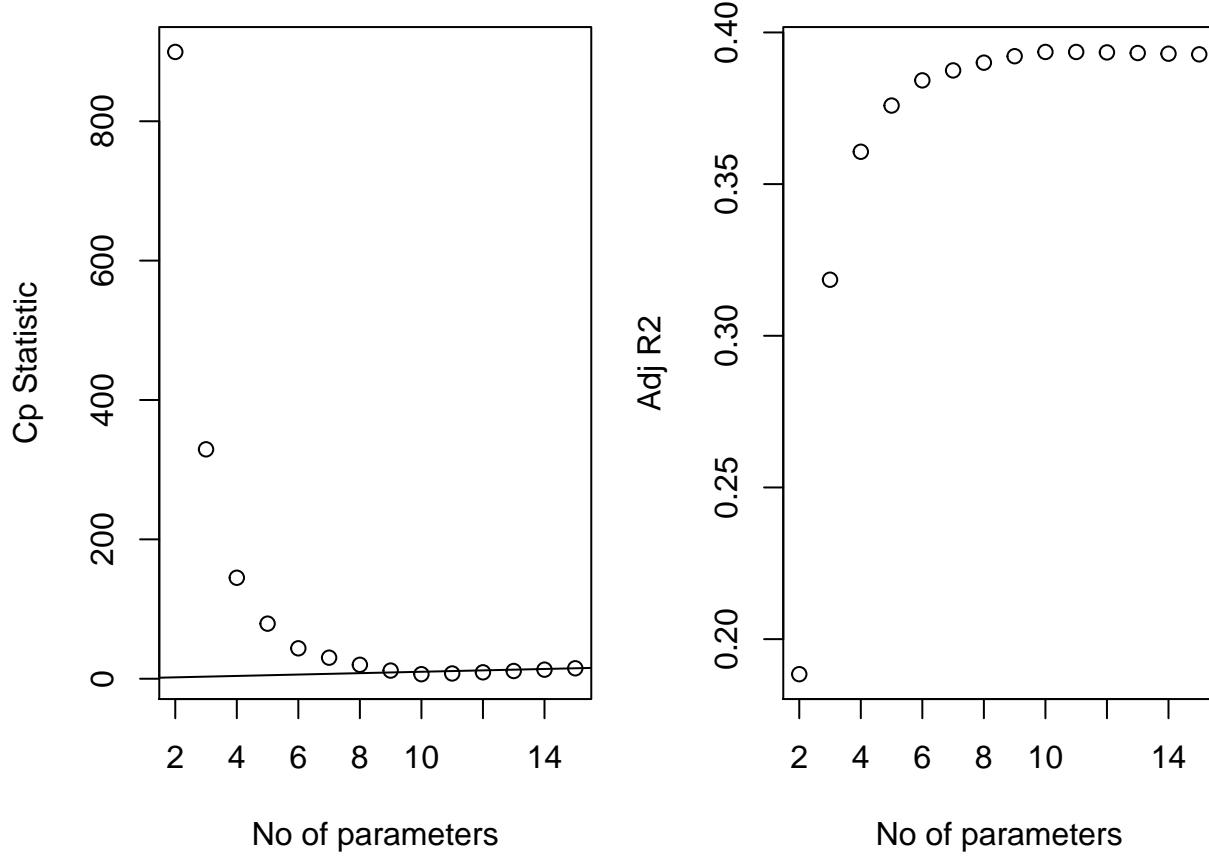
### High income

```
criterion_stats_high =
  regsubsets(target_death_rate ~ ., nvmax = 14, data = income_high_data) %>%
  summary()

par(mar = c(4,4,1,1))
par(mfrow = c(1,2))

plot(2:15, criterion_stats_high$cp, xlab = "No of parameters", ylab = "Cp Statistic")
abline(0,1)

plot(2:15, criterion_stats_high$adjr2, xlab = "No of parameters", ylab = "Adj R2")
```



## Cross Validation

```
# low
set.seed(1)
data_train <- trainControl(method = "repeatedcv", number = 10, repeats = 100)

low_cv <- train(target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + median_age_female + pct_black + pct_asian +
  pct_other_race + pct_married_households + pct_upto_hs18_24 +
  pct_with_coverage, data = income_low_data,
  trControl = data_train,
  method = 'lm',
  na.action = na.pass)

low_cv

## Linear Regression
##
## 378 samples
## 10 predictor
##
## No pre-processing
## Resampling: Cross-Validated (10 fold, repeated 100 times)
## Summary of sample sizes: 341, 340, 341, 338, 340, 341, ...
## Resampling results:
```

```

## 
##   RMSE      Rsquared     MAE
##   24.22195  0.4715015  18.7237
##
## Tuning parameter 'intercept' was held constant at a value of TRUE
high_cv <- train(target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
  pct_married_households + pct_upto_hs18_24 + pct_with_coverage,
  data = income_high_data,
  trControl = data_train,
  method = 'lm',
  na.action = na.pass)

high_cv

```

## Linear Regression

##

## 2669 samples

## 9 predictor

##

## No pre-processing

## Resampling: Cross-Validated (10 fold, repeated 100 times)

## Summary of sample sizes: 2401, 2403, 2402, 2402, 2401, 2402, ...

## Resampling results:

##

## RMSE Rsquared MAE

## 19.78867 0.3918871 14.65232

##

## Tuning parameter 'intercept' was held constant at a value of TRUE

## Cross Validation

```

cross_df_low = crossv_mc(income_low_data, n = 100, test = 0.2)
cross_df_high = crossv_mc(income_high_data, n = 100, test = 0.2)

cross_result_low =
  cross_df_low %>%
  mutate(
    step_mod = map(train, ~lm(target_death_rate ~ incidence_rate + poverty_percent +
      median_age_male + median_age_female + pct_black + pct_asian +
      pct_other_race + pct_married_households + pct_upto_hs18_24 +
      pct_with_coverage, data = .x)),
    rmse_train = map2_dbl(step_mod, train, ~rmse(model = .x, data = .y)),
    rmse_test = map2_dbl(step_mod, test, ~rmse(model = .x, data = .y))
  )

mse_results_low = cross_result_low %>%
  dplyr::select(rmse_train, rmse_test) %>%
  summarize(mse_train_low = (mean(rmse_train))^2,
            mse_test_low = (mean(rmse_test))^2) #mse results

```

```

cross_result_high =
  cross_df_high %>%
  mutate(
    step_mod = map(train, ~lm(target_death_rate ~ incidence_rate + poverty_percent +
    median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
    pct_married_households + pct_upto_hs18_24 + pct_with_coverage, data = .x)),
    rmse_train = map2_dbl(step_mod, train, ~rmse(model = .x, data = .y)),
    rmse_test = map2_dbl(step_mod, test, ~rmse(model = .x, data = .y))
  )

mse_results_high = cross_result_high %>%
  dplyr::select(rmse_train, rmse_test) %>%
  summarize(mse_train_high = (mean(rmse_train))^2,
            mse_test_high = (mean(rmse_test))^2)

#LOOCV
glm.fit_low = glm(target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + median_age_female + pct_black + pct_asian +
  pct_other_race + pct_married_households + pct_upto_hs18_24 +
  pct_with_coverage, data = income_low_data)

cv.err_low = cv.glm(income_low_data, glm.fit_low)

glm.fit_high = glm(target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
  pct_married_households + pct_upto_hs18_24 + pct_with_coverage, data = income_high_data)

cv.err_high = cv.glm(income_high_data, glm.fit_high)

# The two delta values should be similar: we use the first one
# The second value is bias corrected
cv.err_low$delta

## [1] 595.1885 595.1301
anova(backward_model_low)[11, 3] #MSE: 431

## [1] 569.0145
anova(backward_model_high)[10, 3]

## [1] 389.4591

mse_low =
  tibble(
    mse_model = anova(backward_model_low)[11, 3],
    mse_LOOCV = cv.err_low$delta[1],
    mse_CV_train = mse_results_low$mse_train_low,
    mse_CV_test = mse_results_low$mse_test_low
  )

mse_high =
  tibble(
    mse_model = anova(backward_model_high)[10, 3],
    mse_LOOCV = cv.err_high$delta[1],

```

```

    mse_CV_train = mse_results_high$mse_train_high,
    mse_CV_test = mse_results_high$mse_test_high
)

rbind(mse_low, mse_high) %>% mutate(dataset = c("low income", "high income")) %>% dplyr::select(dataset

```

dataset	mse_model	mse_LOOCV	mse_CV_train	mse_CV_test
low income	569.014	595.189	550.119	583.890
high income	389.459	393.070	386.918	394.509

## Influential points

### Low Income

```
influence.measures(backward_model_low) %>% summary()
```

```

## Potentially influential observations of
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent + median_age_male + median_a
##
##      dfb.1_ dfb.inc_ dfb.pvr_ dfb.mdn_g_m dfb.mdn_g_f dfb.pct_b dfb.pct_s
## 14   -0.01   0.00   -0.02   -0.01   -0.03    0.01    0.00
## 15   -0.05   0.05   -0.08    0.40   -0.34   -0.05   -0.13
## 20    0.00   0.01    0.01    0.03   -0.02    0.01   -0.04
## 22    0.00   0.00    0.00    0.00    0.00    0.00    0.00
## 23    0.02   0.00   -0.01    0.04   -0.01   -0.06   -0.07
## 26    0.00   0.00    0.00    0.00    0.00    0.00    0.00
## 30    0.26  -0.52   -0.23   -0.19    0.14   -0.36   -0.04
## 32    0.05   0.03   -0.06    0.25   -0.21   -0.07    0.46
## 34   -0.07   0.06    0.03    0.03    0.05    0.13    0.00
## 91    0.21  -0.09    0.00    0.01   -0.03   -0.14   -0.03
## 95   -0.04  -0.02    0.01    0.10   -0.10    0.05   -0.04
## 101   -0.14  -0.09    0.07    0.09   -0.07    0.02    0.08
## 112   -0.18   0.14   -0.04    0.12    0.03    0.00    0.09
## 123   0.27   0.24   -0.59   -0.77    0.82   -0.15    0.13
## 142   0.13  -0.03    0.14    0.15   -0.08    0.15   -0.12
## 173   0.01  -0.07   -0.07   -0.14    0.04   -0.17    0.05
## 187   0.22  -0.11   -0.10    0.03    0.01   -0.14    0.01
## 189   0.00   0.00   -0.01    0.03    0.01    0.06   -0.13
## 199   0.05  -0.02   -0.02   -0.24    0.28   -0.02   -0.03
## 200   0.00  -0.01   -0.01   -0.02    0.00    0.01   -0.01
## 204   0.00   0.01    0.00   -0.01    0.01    0.00    0.00
## 206  -0.02   0.00    0.01    0.01    0.00   -0.01    0.00
## 208  -0.23   0.19    0.13   -0.01   -0.02    0.01   -0.48
## 211  -0.11   0.00   -0.05    0.00    0.02   -0.01    0.02
## 216   0.08   0.07    0.05   -0.04    0.07   -0.09   -0.06
## 231  -0.02  -0.01    0.00    0.00    0.01    0.01    0.00
## 243   0.08  -0.23   -0.05   -0.14    0.09    0.05   -0.05
## 263   0.06   0.05   -0.06    0.06   -0.06   -0.01    0.00
## 275   0.14   0.10   -0.17    0.17   -0.22   -0.15   -0.35
## 277  -0.01   0.02    0.01   -0.03    0.04   -0.02    0.02
## 290   0.12  -0.04    0.02    0.01   -0.05    0.12   -0.08
## 302  -0.10   0.03    0.13    0.01    0.01    0.00    0.00

```

```

## 332 -0.15 -0.45 -0.01 -0.44 0.48 -0.23 0.10
## 336 0.15 0.05 -0.07 0.14 -0.21 0.15 -0.12
## 344 -0.01 0.00 -0.09 -0.04 0.06 0.05 -0.33
## 366 -0.01 0.00 0.00 0.00 0.00 0.00 0.00
## 369 0.16 -0.07 0.08 -0.01 -0.04 -0.26 -0.06
## dfb.pct_t_ dfb.pct_m_ dfb.p_-1 dfb.pct_w_ dffit cov.r cook.d
## 14 0.01 0.00 0.08 0.04 -0.10 1.13_* 0.00
## 15 0.36 -0.21 0.15 0.12 0.68_* 0.98 0.04
## 20 0.13 0.01 -0.05 -0.01 0.15 1.34_* 0.00
## 22 -0.01 0.00 0.00 0.00 -0.01 1.09_* 0.00
## 23 0.40 -0.10 -0.06 0.02 0.47 1.28_* 0.02
## 26 0.01 0.00 0.00 0.00 0.01 1.37_* 0.00
## 30 -0.21 -0.42 0.04 0.26 0.72_* 0.83_* 0.05
## 32 -0.13 -0.15 0.07 -0.03 0.60_* 1.31_* 0.03
## 34 0.07 0.12 0.06 -0.13 -0.30 1.12_* 0.01
## 91 -0.06 -0.25 -0.07 -0.05 -0.38 0.91_* 0.01
## 95 -0.03 0.01 0.02 0.04 -0.15 1.14_* 0.00
## 101 0.07 0.03 0.04 0.14 -0.29 0.79_* 0.01
## 112 0.04 -0.18 0.00 0.16 -0.52_* 0.92 0.02
## 123 -0.18 -0.21 0.07 -0.24 -1.09_* 0.89_* 0.11
## 142 -0.05 -0.02 -0.25 -0.21 0.48 0.85_* 0.02
## 173 0.02 0.05 0.13 0.09 -0.41 0.88_* 0.02
## 187 -0.12 -0.15 0.02 -0.17 0.34 0.90_* 0.01
## 189 0.05 0.02 0.07 -0.07 -0.25 1.11_* 0.01
## 199 0.02 0.06 -0.18 -0.09 0.35 1.13_* 0.01
## 200 0.04 0.02 0.01 0.01 0.05 1.13_* 0.00
## 204 0.00 0.00 0.00 0.00 -0.01 1.10_* 0.00
## 206 0.00 -0.01 0.01 0.02 -0.04 1.09_* 0.00
## 208 0.14 0.00 -0.20 0.30 -0.70_* 1.10_* 0.04
## 211 -0.01 -0.09 0.08 0.17 -0.25 1.10_* 0.01
## 216 -0.06 -0.07 -0.04 -0.12 0.24 0.89_* 0.01
## 231 0.01 0.01 0.01 0.02 -0.03 1.09_* 0.00
## 243 0.02 0.01 -0.16 0.08 -0.38 0.90_* 0.01
## 263 0.00 -0.04 -0.04 -0.04 -0.13 1.09_* 0.00
## 275 0.07 -0.16 -0.09 0.02 -0.53_* 0.94 0.02
## 277 0.03 -0.02 0.03 0.00 0.08 1.19_* 0.00
## 290 -0.07 -0.07 -0.21 0.00 0.45 0.84_* 0.02
## 302 0.02 0.06 0.00 0.04 0.14 1.11_* 0.00
## 332 -0.41 -0.30 0.14 0.50 -0.92_* 0.84_* 0.07
## 336 0.33 0.13 -0.02 -0.22 0.52_* 0.93 0.02
## 344 0.08 0.03 0.09 -0.03 -0.46 1.11_* 0.02
## 366 0.00 0.01 0.00 0.01 -0.01 1.13_* 0.00
## 369 -0.12 -0.23 0.06 -0.03 0.39 1.11_* 0.01
## hat
## 14 0.09_*
## 15 0.09_*
## 20 0.23_*
## 22 0.06
## 23 0.21_*
## 26 0.25_*
## 30 0.05
## 32 0.24_*
## 34 0.10_*
## 91 0.03

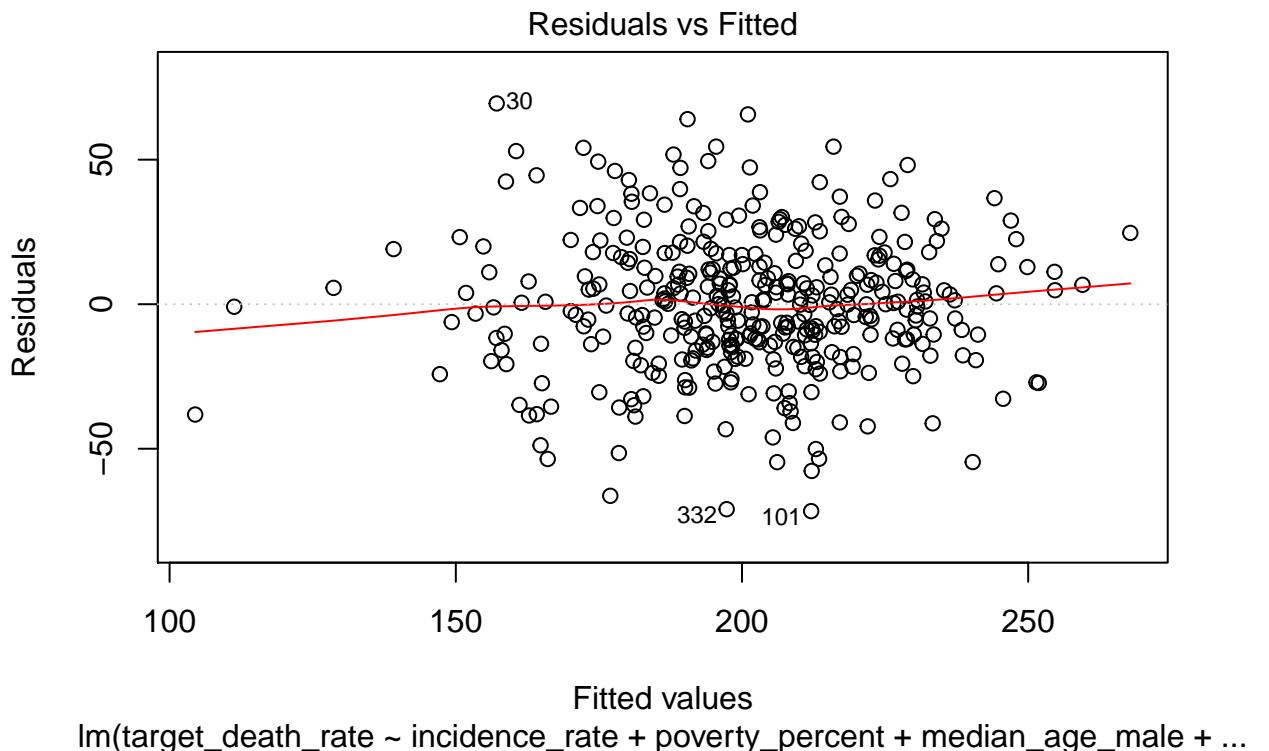
```

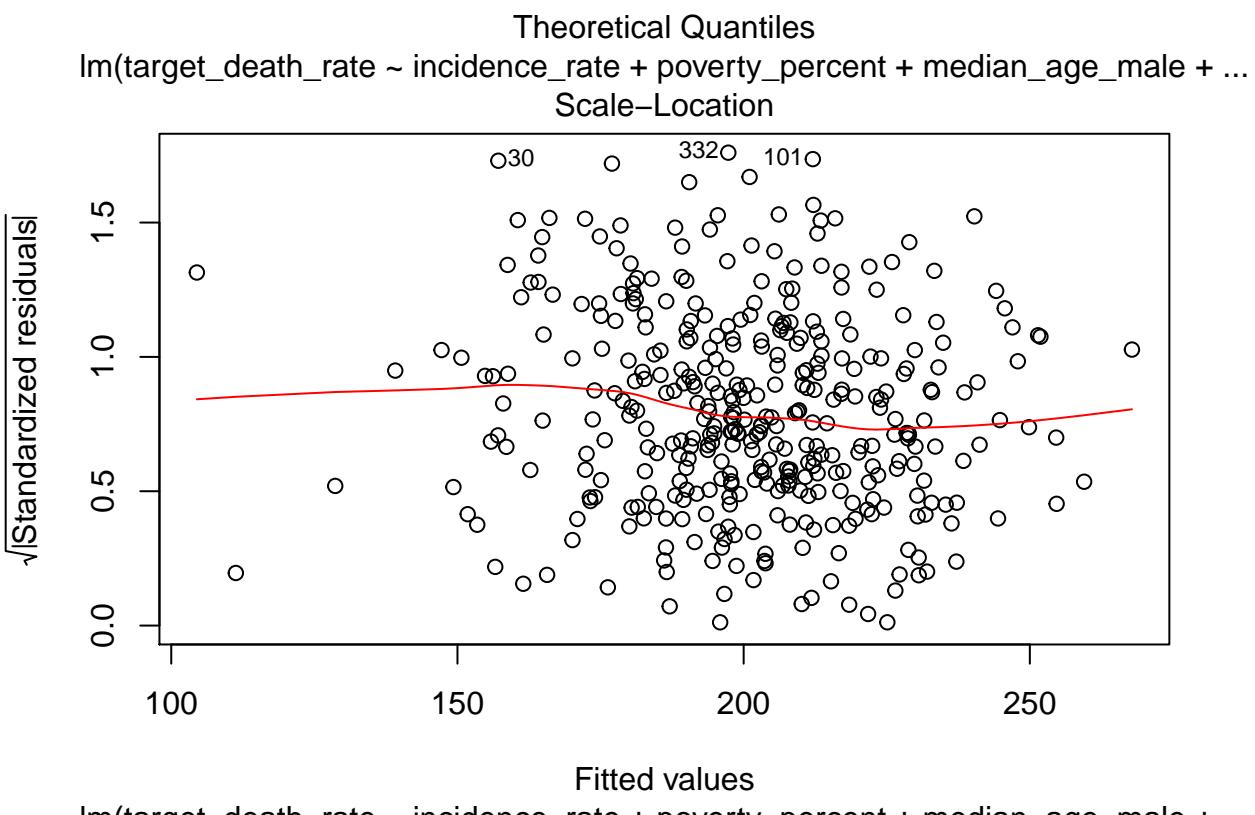
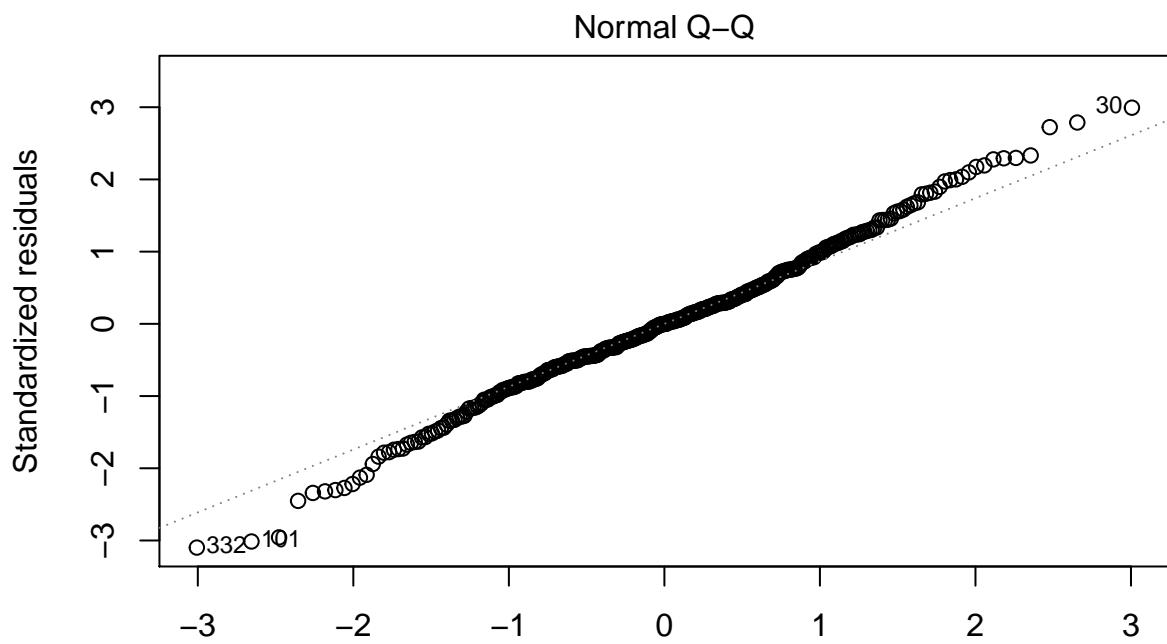
```

## 95  0.10_*
## 101 0.01
## 112 0.05
## 123 0.12_*
## 142 0.03
## 173 0.03
## 187 0.02
## 189 0.09_*
## 199 0.11_*
## 200 0.09_*
## 204 0.06
## 206 0.06
## 208 0.14_*
## 211 0.08
## 216 0.01
## 231 0.06
## 243 0.03
## 263 0.06
## 275 0.05
## 277 0.14_*
## 290 0.03
## 302 0.08
## 332 0.08
## 336 0.05
## 344 0.11_*
## 366 0.09_*
## 369 0.11_*

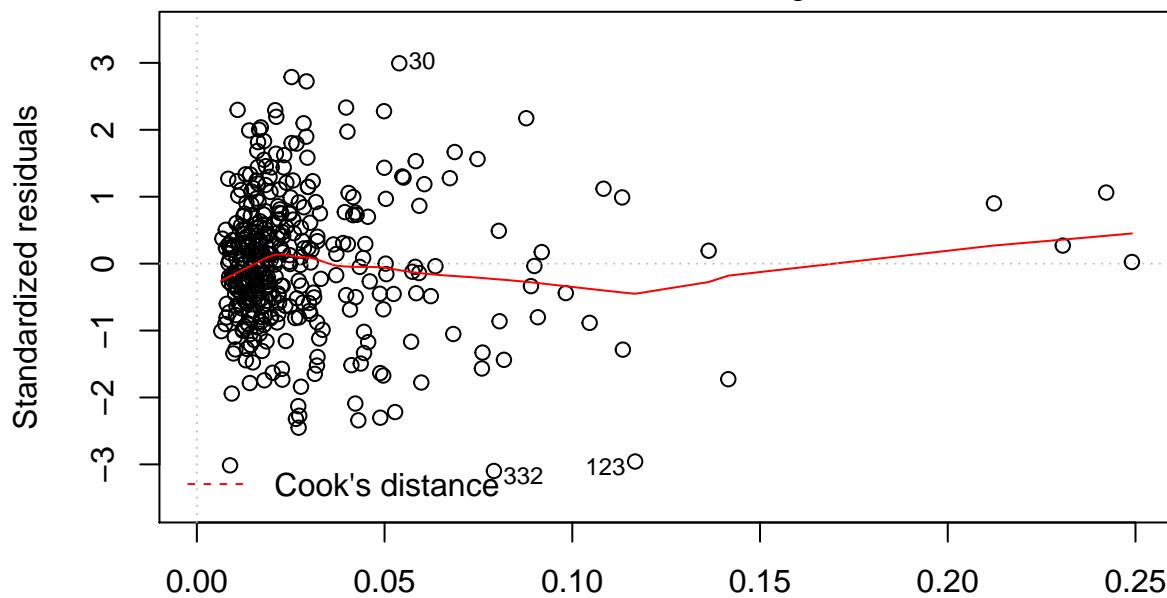
```

```
plot(backward_model_low)
```





### Residuals vs Leverage



### Leverage

```
lm(target_death_rate ~ incidence_rate + poverty_percent + median_age_male + ...)
```

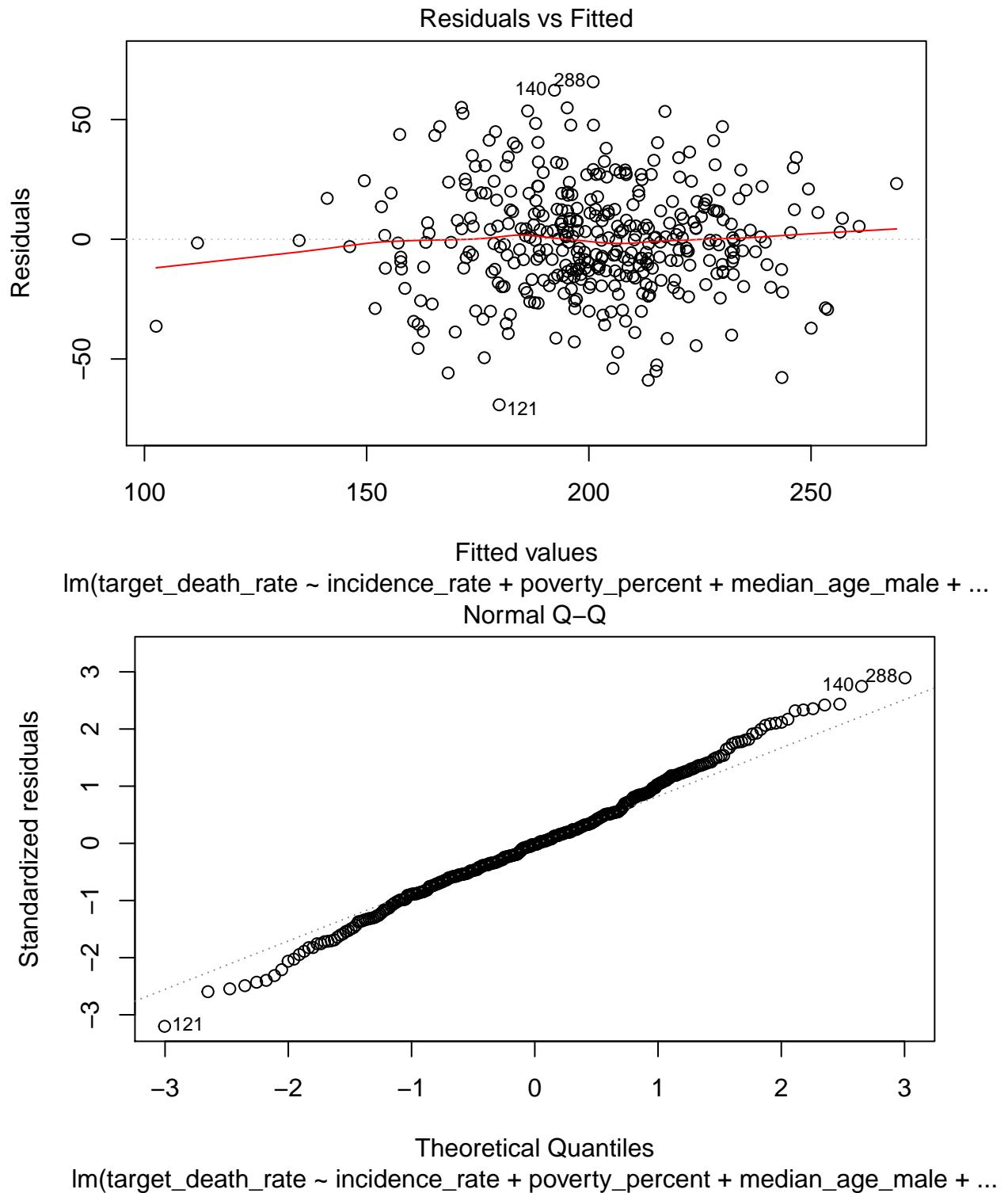
```
income_low_rm <- income_low_data[-c(332, 101, 30),]

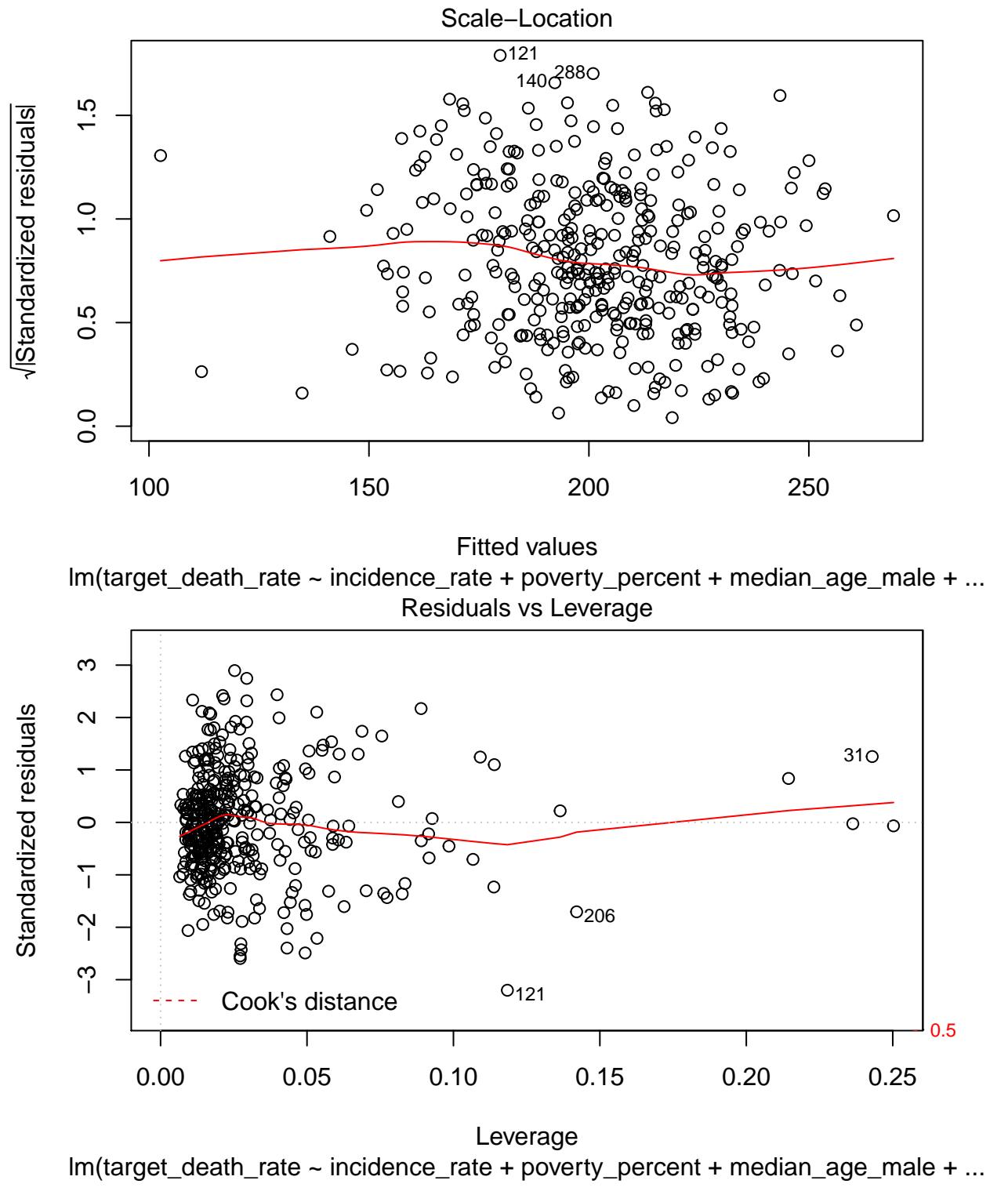
backward_model_low_rm <- lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + median_age_female + pct_black + pct_asian +
  pct_other_race + pct_married_households + pct_upto_hs18_24 +
  pct_with_coverage, data = income_low_rm)

cbind(summary(backward_model_low_rm) %>% tidy %>% dplyr::select(-statistic),
      summary(backward_model_low) %>% tidy %>% dplyr::select(-statistic)) %>%
  knitr::kable()
```

term	estimate	std.error	p.value	term	estimate	std.error
(Intercept)	-6.0167646	31.2444460	0.8474026	(Intercept)	-6.7837317	32.2180572
incidence_rate	0.3055177	0.0200670	0.0000000	incidence_rate	0.2831379	0.0202753
poverty_percent	0.7194054	0.2949757	0.0152108	poverty_percent	0.6630969	0.3049711
median_age_male	-1.2645667	0.5869468	0.0318580	median_age_male	-1.5990211	0.6013537
median_age_female	1.3151201	0.5631914	0.0200799	median_age_female	1.6474232	0.5764589
pct_black	-0.2665368	0.0799427	0.0009437	pct_black	-0.3151240	0.0820959
pct_asian	-5.4112789	1.9188836	0.0050650	pct_asian	-5.1184441	1.9883473
pct_other_race	-0.9219624	0.3132796	0.0034595	pct_other_race	-1.1093338	0.3211770
pct_married_households	-0.4126216	0.2907166	0.1566597	pct_married_households	-0.6252937	0.2971143
pct_upto_hs18_24	0.2094353	0.1108083	0.0595430	pct_upto_hs18_24	0.2354264	0.1147949
pct_with_coverage	0.7954130	0.2846906	0.0054822	pct_with_coverage	1.0621260	0.2901767

```
plot(backward_model_low_rm)
```





High Income

```
influence.measures(backward_model_high) %>% summary()
```

```
## Potentially influential observations of
##   lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##       median_age_male + pct_unem
```

```

##      dfb.1_ dfb.inc_ dfb.pvr_ dfb.md__ dfb.p_16 dfb.pct_ dfb.pct_t_
## 21    0.04   0.05  -0.09   0.04    0.02   -0.04    0.02
## 27    0.01  -0.15   0.01  -0.03   -0.01    0.01   -0.05
## 93    0.00   0.00  -0.01   0.01    0.00    0.00    0.01
## 103   0.22   0.03  -0.27   0.07    0.12   -0.10    0.12
## 106   -0.05   0.00   0.15   0.10   -0.15    0.02    0.03
## 107   -0.01  -0.01   0.06   0.07   -0.11    0.02    0.05
## 109   0.08  -0.02  -0.12   0.17    0.16    0.03    0.06
## 111   -0.01  -0.02   0.01   0.01   -0.02    0.01   -0.01
## 127    0.00   0.00   0.00   0.00    0.00    0.00    0.00
## 134    0.00   0.00   0.00   0.00    0.00    0.01    0.01
## 137    0.00   0.01   0.00   0.01    0.00    0.09   -0.01
## 143   -0.01   0.01   0.00   0.01    0.00    0.04    0.00
## 151    0.00   0.00   0.00   0.00    0.00    0.00   -0.01
## 159    0.00  -0.02   0.02   0.00    0.01   -0.03    0.08
## 163    0.00   0.02   0.01  -0.04   -0.01    0.02   -0.08
## 164    0.12  -0.13  -0.14  -0.07    0.02   -0.02   -0.04
## 165    0.03  -0.03   0.02   0.02  -0.05   -0.02    0.07
## 166    0.01   0.00  -0.01   0.00    0.00    0.00   -0.02
## 175   -0.03   0.05  -0.02  -0.13    0.07   -0.02    0.00
## 182   -0.01  -0.02   0.00   0.00    0.00    0.07   -0.01
## 189    0.00   0.01  -0.02   0.01    0.01   -0.01    0.00
## 208    0.03  -0.07   0.01   0.03    0.01   -0.01   -0.03
## 216    0.00   0.00   0.00   0.00    0.00    0.00    0.00
## 221   -0.01  -0.10   0.05  -0.01    0.05   -0.01   -0.04
## 224    0.00   0.00   0.00   0.00    0.00    0.00    0.00
## 226    0.04  -0.26  -0.13   0.03    0.20   -0.10    0.00
## 229    0.00   0.00   0.00   0.00    0.00    0.00    0.00
## 233   -0.01  -0.15  -0.14   0.08    0.14   -0.04    0.02
## 241    0.01   0.00   0.00   0.00    0.00    0.00    0.00
## 247    0.00  -0.01   0.00   0.00    0.00    0.00    0.00
## 248    0.18  -1.33_*  -0.10   0.13    0.21   -0.13   -0.10
## 250    0.01   0.01  -0.01   0.01    0.00    0.01   -0.07
## 258    0.00   0.00   0.00   0.00    0.00    0.00    0.00
## 260    0.00   0.00   0.00   0.00    0.00    0.00    0.02
## 262   -0.01   0.01  -0.01   0.00    0.02   -0.02    0.09
## 266    0.00   0.00   0.00   0.00    0.00   -0.02    0.00
## 285    0.00   0.01  -0.02   0.01    0.00    0.00    0.00
## 313    0.00   0.00   0.00  -0.01    0.00    0.00    0.00
## 356   -0.01   0.00   0.00  -0.02    0.01   -0.03   -0.04
## 409   -0.02   0.02  -0.02   0.01   -0.05   -0.02    0.03
## 420    0.11  -0.03  -0.06   0.00   -0.09   -0.03   -0.03
## 425    0.10   0.08  -0.11  -0.03    0.08   -0.02   -0.01
## 430   -0.02   0.02   0.01   0.01    0.00    0.01    0.01
## 435    0.02  -0.01   0.00   0.11   -0.10   -0.01    0.01
## 443    0.03   0.01   0.02  -0.01   -0.04    0.00   -0.02
## 462    0.03   0.07  -0.05  -0.04   -0.07    0.01    0.03
## 469   -0.01  -0.10   0.09  -0.06   -0.02   -0.01   -0.04
## 480    0.04   0.01  -0.01   0.03    0.00   -0.05    0.10
## 494    0.01   0.03   0.02   0.00    0.00   -0.01   -0.02
## 502   -0.03   0.00   0.02   0.02    0.00   -0.04    0.17
## 507   -0.01   0.00   0.02   0.02  -0.01    0.13    0.00
## 512    0.00   0.00   0.00   0.00    0.00    0.00    0.00
## 515    0.03   0.00  -0.03  -0.01   -0.02   -0.02   -0.05

```

## 518	0.03	0.02	-0.01	0.00	-0.12	0.05	-0.17
## 527	0.00	-0.04	-0.01	-0.03	0.03	-0.01	0.00
## 541	0.05	-0.10	0.01	-0.05	0.04	-0.02	-0.01
## 547	0.09	-0.07	-0.03	0.17	-0.07	-0.01	-0.01
## 567	-0.04	0.00	0.01	0.00	0.00	-0.01	0.16
## 570	-0.09	0.03	0.05	-0.07	0.03	0.04	0.00
## 580	0.06	0.01	-0.07	-0.07	0.09	-0.06	0.01
## 586	0.03	0.02	0.01	0.09	-0.10	0.00	-0.01
## 590	0.02	-0.08	-0.01	0.05	-0.02	-0.03	-0.01
## 631	0.00	0.00	0.00	0.00	-0.01	0.00	0.00
## 658	-0.08	0.02	0.07	0.04	0.04	0.00	0.00
## 664	-0.04	0.01	0.04	0.02	0.01	0.00	-0.01
## 667	-0.04	0.03	0.04	-0.01	-0.03	0.00	-0.01
## 681	-0.01	-0.04	0.09	-0.04	-0.13	0.03	-0.02
## 686	-0.09	0.03	0.07	0.02	-0.09	0.01	0.05
## 689	-0.05	0.03	-0.02	0.04	0.11	-0.01	-0.10
## 707	0.11	-0.07	-0.04	0.03	0.00	-0.01	-0.01
## 774	0.00	0.00	0.00	0.00	0.00	-0.03	0.01
## 776	0.00	0.01	0.01	-0.01	0.00	-0.05	-0.01
## 790	-0.05	0.04	-0.11	0.05	0.09	-0.02	0.04
## 803	0.02	0.05	-0.05	0.10	-0.01	0.01	0.01
## 808	-0.09	0.04	0.09	0.05	-0.07	0.01	0.08
## 809	-0.01	0.00	0.01	0.00	0.00	0.00	0.01
## 810	0.00	0.02	-0.06	0.08	-0.04	0.01	0.03
## 812	-0.09	0.04	0.11	0.04	0.00	0.01	0.05
## 814	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 817	-0.03	0.14	-0.10	0.09	0.04	-0.03	0.09
## 818	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 824	-0.05	0.00	0.06	0.09	-0.01	0.01	0.05
## 848	0.01	-0.01	-0.01	0.00	0.00	0.01	0.00
## 867	0.00	0.00	0.00	0.00	0.00	-0.01	0.00
## 884	0.00	0.00	0.00	0.00	0.00	0.01	0.01
## 889	-0.01	0.00	0.00	0.00	0.01	0.00	0.01
## 891	-0.15	0.09	0.17	-0.03	-0.14	0.06	-0.05
## 895	0.00	0.00	0.00	0.01	0.00	0.03	0.01
## 897	0.00	0.00	0.00	-0.01	-0.01	0.00	0.00
## 898	0.00	0.00	0.00	0.00	-0.01	0.00	-0.01
## 899	0.00	0.00	0.00	0.00	0.00	0.01	0.00
## 901	-0.01	0.00	0.00	0.00	0.01	0.00	0.02
## 904	0.01	0.00	0.01	0.04	-0.02	0.16	-0.01
## 905	-0.01	0.00	0.01	0.00	0.01	0.02	0.01
## 907	-0.01	0.00	0.00	0.01	0.00	0.06	0.00
## 909	-0.02	0.00	0.02	0.02	-0.01	0.14	0.00
## 912	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 915	-0.01	0.01	0.00	0.01	0.02	0.06	0.02
## 918	0.00	0.00	0.00	0.00	-0.01	-0.01	0.00
## 923	-0.01	0.00	0.01	0.00	0.00	0.02	0.01
## 928	-0.04	0.01	0.04	-0.12	-0.02	0.00	-0.09
## 930	0.02	0.01	-0.03	0.00	0.01	-0.01	0.00
## 937	-0.01	-0.01	0.02	-0.08	-0.05	0.00	0.00
## 940	0.00	0.01	0.00	0.00	0.00	0.00	0.00
## 943	-0.14	0.08	0.14	0.06	0.00	0.06	-0.05
## 950	-0.12	0.14	0.06	0.04	-0.03	0.05	0.05
## 952	0.08	-0.05	0.01	0.04	-0.08	-0.02	-0.03

##	969	-0.33	0.27	0.36	-0.03	-0.11	0.10	0.05
##	980	0.00	0.00	0.00	0.00	0.00	0.00	0.00
##	1014	0.00	0.00	0.00	0.00	0.00	0.00	0.00
##	1049	0.08	-0.01	-0.10	-0.11	-0.01	-0.04	-0.03
##	1055	0.00	0.00	0.00	0.00	0.00	0.00	0.00
##	1076	-0.05	0.01	0.16	0.06	-0.07	0.04	-0.02
##	1085	0.00	0.06	0.03	-0.04	-0.02	-0.01	-0.02
##	1088	0.03	0.01	-0.01	-0.08	-0.05	-0.01	-0.01
##	1104	0.01	0.01	-0.03	0.01	0.01	0.00	0.01
##	1106	0.09	-0.09	-0.06	0.04	-0.05	0.00	-0.04
##	1123	0.02	0.07	0.02	0.04	-0.06	0.00	-0.01
##	1135	-0.04	0.06	-0.03	-0.02	0.02	-0.06	0.22
##	1137	-0.04	0.02	0.05	0.09	-0.02	-0.02	0.08
##	1142	0.00	-0.05	0.02	0.07	0.04	0.01	-0.08
##	1148	-0.08	0.05	0.07	-0.07	-0.08	-0.05	0.15
##	1149	-0.02	-0.01	0.04	0.00	0.00	0.02	0.02
##	1154	-0.04	-0.01	0.02	0.02	0.02	-0.01	0.00
##	1169	0.03	-0.01	-0.06	-0.03	0.10	0.01	-0.02
##	1183	-0.01	0.01	0.02	-0.09	0.01	0.03	-0.01
##	1188	0.00	0.02	-0.01	0.00	0.00	-0.02	0.10
##	1194	0.03	-0.02	-0.02	-0.01	0.03	0.02	-0.10
##	1195	0.01	-0.02	0.00	0.01	0.01	0.01	-0.03
##	1204	0.09	0.03	0.04	0.18	-0.19	0.01	-0.02
##	1231	-0.14	0.11	0.12	-0.01	-0.02	0.05	0.03
##	1233	-0.14	0.12	0.10	0.02	0.00	0.04	0.00
##	1250	0.00	0.00	0.00	0.00	0.00	0.00	0.00
##	1312	-0.06	0.22	0.03	0.02	-0.03	0.02	0.03
##	1318	-0.08	0.10	0.11	0.10	-0.24	-0.01	0.29
##	1356	0.02	-0.02	0.03	-0.02	0.00	0.00	-0.03
##	1367	0.00	0.01	0.00	0.00	-0.02	-0.01	0.10
##	1381	0.07	0.00	-0.01	-0.06	-0.10	0.00	-0.03
##	1444	-0.09	0.01	0.05	0.06	0.02	-0.01	0.06
##	1469	-0.03	-0.01	0.00	-0.01	0.01	-0.01	0.01
##	1497	-0.05	-0.03	0.01	0.04	-0.04	-0.04	0.20
##	1532	-0.02	0.00	0.02	-0.01	-0.02	0.01	0.00
##	1556	0.00	0.00	0.00	0.00	0.00	0.01	0.00
##	1597	-0.01	0.00	0.01	0.00	0.01	0.01	-0.02
##	1603	-0.02	0.00	0.00	0.00	0.01	0.00	0.00
##	1615	0.03	0.01	-0.04	0.00	0.02	-0.01	0.00
##	1618	-0.02	0.01	-0.04	0.02	-0.01	-0.08	0.29
##	1619	-0.15	0.05	0.08	0.10	-0.11	-0.11	0.62
##	1625	0.00	0.00	0.00	0.00	0.00	0.00	0.00
##	1629	0.00	0.00	0.00	0.00	0.00	0.00	0.00
##	1641	0.00	0.00	0.00	0.00	0.00	-0.01	0.00
##	1644	0.01	0.00	0.00	-0.01	0.00	0.00	0.00
##	1646	-0.01	-0.04	0.06	0.01	-0.05	0.01	0.01
##	1648	0.05	-0.01	-0.02	-0.04	-0.01	0.02	-0.12
##	1649	-0.03	0.02	0.00	0.07	-0.04	-0.07	0.36
##	1657	-0.01	-0.03	-0.01	0.04	0.08	0.00	0.00
##	1665	0.10	0.04	-0.07	-0.01	-0.02	-0.03	0.05
##	1678	0.01	-0.03	-0.01	0.02	0.00	0.01	0.01
##	1684	-0.04	0.01	0.03	0.01	-0.03	-0.02	0.12
##	1694	0.00	0.03	-0.04	-0.01	0.02	0.02	-0.08
##	1708	0.01	-0.01	-0.01	-0.01	-0.01	-0.01	0.01

## 1726	0.01	0.00	-0.01	0.04	0.00	0.00	0.01
## 1737	0.01	0.02	-0.01	0.02	-0.02	-0.02	0.11
## 1738	0.00	0.00	0.00	0.01	0.00	0.00	0.00
## 1748	0.07	0.09	-0.05	0.08	-0.08	-0.10	0.39
## 1752	0.09	-0.10	-0.01	0.01	-0.03	-0.01	-0.01
## 1757	0.00	0.00	0.00	0.00	0.00	0.00	-0.01
## 1768	-0.02	-0.01	-0.01	0.05	-0.05	-0.11	0.47
## 1795	0.01	0.02	0.00	0.00	-0.01	0.00	0.02
## 1815	0.02	-0.07	-0.01	0.03	0.02	-0.01	-0.03
## 1834	0.01	0.00	0.00	0.00	0.00	0.00	0.00
## 1836	-0.23	-0.01	0.07	0.13	-0.23	0.01	0.19
## 1837	-0.12	-0.05	0.04	0.02	-0.03	0.02	0.06
## 1848	-0.02	0.10	0.08	0.13	-0.10	0.04	0.00
## 1856	0.00	0.00	0.00	-0.01	0.00	-0.04	-0.01
## 1870	0.00	0.00	-0.01	0.01	0.01	-0.02	0.01
## 1926	-0.02	-0.04	0.04	0.00	0.00	-0.03	-0.02
## 1955	-0.08	-0.02	0.05	0.04	0.01	0.00	0.04
## 2018	-0.02	0.01	0.01	0.00	-0.06	-0.01	0.00
## 2036	0.00	0.01	0.00	0.01	0.00	0.00	0.00
## 2047	0.01	0.01	-0.01	-0.01	0.00	-0.04	0.01
## 2048	-0.02	0.00	0.02	0.01	-0.01	0.10	-0.02
## 2049	-0.05	-0.01	0.06	0.04	-0.04	0.26	-0.05
## 2050	0.02	0.01	-0.02	-0.01	0.01	-0.10	0.02
## 2052	-0.06	0.00	0.11	-0.11	-0.09	-0.02	0.04
## 2057	-0.06	0.05	0.07	-0.03	0.04	0.02	-0.08
## 2062	-0.04	0.09	0.01	0.01	-0.01	0.00	0.04
## 2081	0.14	0.05	-0.20	0.08	-0.03	-0.01	0.03
## 2085	-0.03	-0.03	-0.01	0.01	-0.02	0.00	0.00
## 2087	0.04	0.06	0.01	0.07	-0.05	0.06	-0.10
## 2089	-0.07	0.06	0.10	0.08	-0.05	0.04	-0.01
## 2145	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 2154	0.01	-0.01	0.02	0.01	-0.06	0.01	0.02
## 2232	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 2282	-0.03	0.13	0.04	0.00	-0.04	0.01	-0.02
## 2287	-0.22	0.10	0.23	0.12	-0.10	0.04	0.04
## 2311	0.02	0.01	0.07	0.03	-0.03	0.03	-0.06
## 2322	0.03	-0.10	-0.18	-0.04	0.16	-0.07	0.04
## 2332	0.00	0.00	0.01	-0.01	0.00	0.00	0.00
## 2340	-0.07	0.02	0.03	-0.08	0.09	0.01	-0.01
## 2358	-0.01	0.00	-0.01	-0.01	0.05	0.00	-0.01
## 2360	0.27	-0.43	0.01	0.21	-0.38	1.63_*	-0.35
## 2362	0.04	-0.06	-0.04	-0.08	0.07	-0.01	-0.05
## 2366	-0.19	-0.02	0.14	-0.03	0.08	-0.05	0.06
## 2370	0.10	0.02	-0.06	-0.01	-0.02	0.25	-0.09
## 2372	0.10	0.03	-0.07	-0.13	0.09	-0.01	-0.09
## 2373	0.20	0.06	-0.23	-0.07	0.02	0.06	-0.06
## 2374	-0.03	0.02	0.03	0.03	-0.05	0.01	0.02
## 2375	0.03	-0.02	-0.02	-0.01	-0.01	0.01	-0.01
## 2378	-0.10	-0.02	0.07	0.02	-0.05	0.00	0.05
## 2387	0.04	0.12	-0.04	0.04	-0.15	0.02	0.01
## 2390	0.01	0.03	-0.02	0.00	-0.01	0.01	-0.05
## 2419	0.00	0.00	0.00	0.00	0.00	0.00	0.00
## 2424	0.05	0.06	-0.01	0.02	0.04	0.01	0.01
## 2425	0.00	0.00	0.00	0.00	0.00	0.00	0.00

	##	2442	-0.03	-0.04	-0.08	0.06	0.10	-0.03	0.07
	##	2443	-0.01	0.00	0.01	0.00	-0.05	-0.02	0.00
	##	2444	0.02	0.18	-0.04	-0.08	0.00	-0.01	0.02
	##	2469	-0.01	0.09	-0.02	0.00	0.02	-0.02	0.01
	##	2482	-0.03	0.08	0.04	0.02	-0.05	0.00	-0.01
	##	2493	0.00	0.00	0.00	0.00	0.00	0.00	0.02
	##	2544	-0.03	0.04	0.00	0.03	0.04	-0.01	0.00
	##	2626	0.01	0.01	-0.02	0.02	0.01	-0.01	0.01
	##	2662	0.04	0.04	-0.02	0.06	-0.07	-0.01	0.04
	##		dfb.pct_m_	dfb.p_1	dfb.pct_w_	dffit	cov.r	cook.d	hat
	##	21	-0.01	0.02	-0.06	-0.14	1.01_*	0.00	0.01_*
	##	27	-0.03	0.11	0.05	0.19_*	1.00	0.00	0.01_*
	##	93	0.00	0.00	0.00	-0.02	1.01_*	0.00	0.01
	##	103	-0.23	0.00	-0.19	0.39_*	0.96_*	0.01	0.01
	##	106	0.01	-0.11	0.04	-0.24_*	0.97_*	0.01	0.01
	##	107	-0.07	-0.04	0.04	-0.19_*	0.97_*	0.00	0.00
	##	109	-0.18	-0.05	-0.05	0.34_*	0.97_*	0.01	0.01
	##	111	0.00	-0.05	0.02	-0.12	0.96_*	0.00	0.00
	##	127	0.00	0.00	0.00	0.00	1.02_*	0.00	0.01_*
	##	134	-0.01	0.00	0.00	0.02	1.02_*	0.00	0.01_*
	##	137	0.02	-0.01	-0.02	0.10	1.03_*	0.00	0.03_*
	##	143	0.01	0.00	0.00	0.04	1.02_*	0.00	0.01_*
	##	151	0.00	0.00	0.00	-0.01	1.01_*	0.00	0.01
	##	159	0.00	-0.04	0.02	0.11	1.02_*	0.00	0.02_*
	##	163	0.03	0.00	-0.01	-0.09	1.02_*	0.00	0.02_*
	##	164	-0.09	0.01	-0.01	-0.22_*	0.98_*	0.00	0.01
	##	165	-0.02	-0.03	-0.01	0.11	1.01_*	0.00	0.01_*
	##	166	0.00	0.00	-0.01	-0.03	1.02_*	0.00	0.01_*
	##	175	0.01	0.06	0.05	-0.17	1.01	0.00	0.01_*
	##	182	0.03	0.01	0.00	0.08	1.02_*	0.00	0.02_*
	##	189	0.00	0.01	-0.01	-0.03	1.01_*	0.00	0.01
	##	208	0.01	-0.03	-0.02	0.11	0.98_*	0.00	0.00
	##	216	0.00	0.01	0.01	0.01	1.02_*	0.00	0.01_*
	##	221	0.02	-0.02	0.04	0.15	0.98_*	0.00	0.00
	##	224	0.00	0.00	0.00	0.00	1.02_*	0.00	0.01_*
	##	226	0.07	0.01	0.00	-0.40_*	1.00	0.02	0.02_*
	##	229	0.00	0.00	0.00	0.00	1.01_*	0.00	0.01
	##	233	0.03	0.10	0.02	-0.34_*	0.98_*	0.01	0.01_*
	##	241	-0.02	0.00	0.00	0.02	1.01_*	0.00	0.01
	##	247	0.01	-0.01	0.00	-0.01	1.01_*	0.00	0.01
	##	248	-0.06	0.25	0.20	-1.45_*	0.91_*	0.21	0.05_*
	##	250	-0.01	-0.01	-0.01	-0.08	1.03_*	0.00	0.03_*
	##	258	0.00	0.00	0.00	0.00	1.02_*	0.00	0.01_*
	##	260	0.00	0.00	0.00	0.02	1.02_*	0.00	0.02_*
	##	262	0.00	0.01	0.01	0.10	1.02_*	0.00	0.02_*
	##	266	0.00	0.00	0.00	-0.02	1.01_*	0.00	0.01
	##	285	0.00	0.01	-0.01	-0.03	1.01_*	0.00	0.01
	##	313	0.00	0.00	0.00	-0.01	1.01_*	0.00	0.01
	##	356	0.06	0.00	-0.01	-0.09	1.02_*	0.00	0.02_*
	##	409	-0.03	0.04	0.03	0.12	0.99_*	0.00	0.00
	##	420	-0.08	-0.02	-0.07	0.16	0.99_*	0.00	0.00
	##	425	-0.07	-0.07	-0.10	-0.20_*	0.99	0.00	0.01
	##	430	0.01	0.00	0.01	-0.03	1.01_*	0.00	0.01
	##	435	-0.06	-0.04	-0.02	0.18	0.99_*	0.00	0.01

## 443	-0.01	-0.01	-0.03	0.06	1.02_*	0.00	0.01_*
## 462	0.01	-0.01	-0.04	-0.17	0.98_*	0.00	0.00
## 469	-0.08	0.06	0.09	0.21_*	0.98_*	0.00	0.01
## 480	-0.02	-0.07	-0.04	0.15	0.98_*	0.00	0.00
## 494	-0.01	0.00	-0.03	0.09	0.98_*	0.00	0.00
## 502	0.00	0.00	0.03	0.18	1.02_*	0.00	0.02_*
## 507	0.01	0.01	0.00	0.13	1.04_*	0.00	0.04_*
## 512	0.00	0.00	0.00	0.00	1.02_*	0.00	0.01
## 515	-0.01	0.00	-0.02	-0.06	1.02_*	0.00	0.02_*
## 518	-0.04	0.04	-0.02	-0.24_*	1.00	0.01	0.01_*
## 527	0.00	-0.01	0.02	-0.06	1.01_*	0.00	0.01
## 541	-0.07	-0.04	0.03	-0.18	0.98_*	0.00	0.00
## 547	-0.05	-0.09	-0.09	0.25_*	0.97_*	0.01	0.01
## 567	0.02	0.01	0.04	0.17	1.01	0.00	0.01_*
## 570	0.03	0.11	0.08	-0.16	0.99_*	0.00	0.00
## 580	-0.03	-0.08	-0.02	-0.17	0.99_*	0.00	0.01
## 586	0.02	-0.08	-0.06	0.18	0.97_*	0.00	0.00
## 590	-0.04	-0.04	0.02	0.13	0.98_*	0.00	0.00
## 631	0.00	-0.01	-0.01	-0.02	1.01_*	0.00	0.01
## 658	0.07	-0.01	0.03	0.12	0.98_*	0.00	0.00
## 664	0.05	-0.01	0.01	0.08	0.99_*	0.00	0.00
## 667	0.04	0.02	0.02	0.08	0.99_*	0.00	0.00
## 681	0.00	0.12	0.02	0.21_*	0.96_*	0.00	0.00
## 686	0.11	0.04	0.04	0.19_*	0.98_*	0.00	0.01
## 689	-0.02	-0.12	0.07	-0.28_*	0.99_*	0.01	0.01
## 707	-0.11	-0.06	-0.05	-0.17	0.98_*	0.00	0.00
## 774	-0.01	0.00	0.00	-0.03	1.02_*	0.00	0.01_*
## 776	0.00	0.00	0.00	-0.06	1.01_*	0.00	0.01
## 790	-0.07	0.04	0.07	-0.21_*	1.00	0.00	0.01_*
## 803	-0.11	0.06	-0.03	-0.19_*	1.00	0.00	0.01
## 808	-0.03	0.00	0.09	-0.18	0.97_*	0.00	0.00
## 809	0.01	-0.01	0.01	-0.02	1.01_*	0.00	0.01
## 810	-0.12	0.07	0.02	-0.18_*	0.99	0.00	0.01
## 812	0.09	-0.13	0.05	-0.19_*	0.99_*	0.00	0.01
## 814	-0.01	0.00	0.00	-0.01	1.02_*	0.00	0.01_*
## 817	-0.06	-0.03	0.00	-0.24_*	1.00	0.01	0.01_*
## 818	0.00	0.00	0.00	0.01	1.01_*	0.00	0.01
## 824	-0.06	0.01	0.05	-0.16	0.99_*	0.00	0.00
## 848	-0.01	0.00	0.00	0.02	1.01_*	0.00	0.01
## 867	0.00	0.00	0.00	-0.01	1.02_*	0.00	0.02_*
## 884	0.00	0.00	0.00	0.02	1.02_*	0.00	0.02_*
## 889	0.00	0.00	0.00	0.02	1.02_*	0.00	0.02_*
## 891	0.08	0.08	0.11	-0.25_*	1.00	0.01	0.01_*
## 895	0.01	0.00	0.00	0.03	1.02_*	0.00	0.02_*
## 897	0.00	0.01	0.00	-0.01	1.01_*	0.00	0.01
## 898	0.00	0.00	0.00	-0.02	1.01_*	0.00	0.01
## 899	0.00	0.00	0.00	0.01	1.01_*	0.00	0.01
## 901	0.00	0.00	0.00	0.02	1.01_*	0.00	0.01
## 904	-0.02	0.00	-0.01	0.17	1.06_*	0.00	0.06_*
## 905	0.01	0.00	0.00	0.03	1.02_*	0.00	0.01_*
## 907	0.01	0.00	0.00	0.07	1.04_*	0.00	0.03_*
## 909	0.03	0.01	0.00	0.15	1.06_*	0.00	0.06_*
## 912	0.00	0.00	0.00	0.00	1.01_*	0.00	0.01
## 915	0.00	0.01	0.01	0.08	1.01_*	0.00	0.01_*

## 918	0.00	0.00	0.00	-0.02	1.02_*	0.00	0.01_*
## 923	0.00	0.00	0.01	0.03	1.01_*	0.00	0.01
## 928	-0.02	0.03	0.09	-0.19_*	0.98_*	0.00	0.00
## 930	-0.01	-0.02	-0.02	-0.04	1.02_*	0.00	0.01_*
## 937	-0.03	0.04	0.05	-0.11	1.01_*	0.00	0.01_*
## 940	0.00	0.00	0.00	-0.01	1.01_*	0.00	0.01
## 943	0.07	0.00	0.07	-0.22_*	0.97_*	0.00	0.00
## 950	0.09	0.05	0.02	-0.19_*	1.00	0.00	0.01
## 952	0.02	-0.12	-0.07	0.19_*	0.98_*	0.00	0.00
## 969	0.41	0.00	0.08	-0.52_*	0.95_*	0.03	0.01_*
## 980	0.00	0.00	0.00	0.01	1.01_*	0.00	0.01
## 1014	0.00	0.00	0.00	0.00	1.02_*	0.00	0.01_*
## 1049	-0.04	0.05	-0.03	0.16	0.98_*	0.00	0.00
## 1055	0.00	0.00	0.00	-0.01	1.01_*	0.00	0.01
## 1076	0.01	0.03	0.01	0.21_*	1.00	0.00	0.01_*
## 1085	0.00	-0.01	-0.01	0.11	0.98_*	0.00	0.00
## 1088	-0.02	-0.07	0.02	0.22_*	0.86_*	0.00	0.00
## 1104	-0.01	0.02	-0.01	-0.05	1.01_*	0.00	0.01
## 1106	-0.10	0.03	-0.03	0.18	0.97_*	0.00	0.00
## 1123	0.04	-0.07	-0.07	0.13	0.99_*	0.00	0.00
## 1135	0.05	0.03	0.02	0.26_*	1.01	0.01	0.02_*
## 1137	0.07	-0.12	-0.01	0.17	1.02_*	0.00	0.02_*
## 1142	-0.05	-0.08	0.02	-0.19_*	0.99	0.00	0.01
## 1148	0.11	-0.01	0.06	0.25_*	0.98_*	0.01	0.01
## 1149	0.01	-0.01	0.02	-0.07	0.99_*	0.00	0.00
## 1154	0.02	0.01	0.04	0.07	0.99_*	0.00	0.00
## 1169	-0.01	0.02	-0.03	-0.14	0.98_*	0.00	0.00
## 1183	0.05	0.05	0.00	-0.13	0.98_*	0.00	0.00
## 1188	-0.01	0.00	0.01	0.10	1.03_*	0.00	0.03_*
## 1194	-0.02	0.01	-0.02	-0.11	1.01_*	0.00	0.01_*
## 1195	-0.03	-0.01	0.00	-0.06	1.01_*	0.00	0.01
## 1204	-0.14	-0.03	-0.11	0.33_*	0.89_*	0.01	0.00
## 1231	0.15	0.05	0.04	-0.21_*	0.99_*	0.00	0.01
## 1233	0.07	0.05	0.08	-0.19_*	1.00	0.00	0.01
## 1250	0.00	0.00	0.00	-0.01	1.02_*	0.00	0.01_*
## 1312	0.04	0.01	-0.03	0.22_*	1.10_*	0.00	0.09_*
## 1318	-0.03	0.11	0.02	0.41_*	0.94_*	0.02	0.01
## 1356	0.01	0.00	-0.02	0.09	0.98_*	0.00	0.00
## 1367	-0.01	0.01	0.00	0.11	1.02_*	0.00	0.02_*
## 1381	-0.04	-0.03	-0.04	-0.15	0.99_*	0.00	0.00
## 1444	-0.04	-0.01	0.11	-0.15	1.03_*	0.00	0.03_*
## 1469	0.01	0.03	0.04	0.07	0.99_*	0.00	0.00
## 1497	0.01	0.03	0.05	0.23_*	1.00	0.01	0.01
## 1532	0.00	0.04	0.03	0.04	1.01_*	0.00	0.01_*
## 1556	-0.01	0.00	0.00	0.02	1.01_*	0.00	0.01
## 1597	0.01	-0.01	0.01	-0.04	1.01_*	0.00	0.01
## 1603	-0.01	0.00	0.02	-0.03	1.01_*	0.00	0.01
## 1615	-0.02	-0.02	-0.02	-0.05	1.02_*	0.00	0.01_*
## 1618	-0.03	0.03	0.04	0.31_*	1.05_*	0.01	0.05_*
## 1619	0.05	0.03	0.10	0.65_*	1.04_*	0.04	0.06_*
## 1625	0.00	0.00	0.00	0.00	1.01_*	0.00	0.01
## 1629	0.00	0.00	0.00	-0.01	1.01_*	0.00	0.01
## 1641	-0.01	0.00	0.00	-0.01	1.02_*	0.00	0.02_*
## 1644	0.00	0.00	-0.01	0.02	1.02_*	0.00	0.01_*

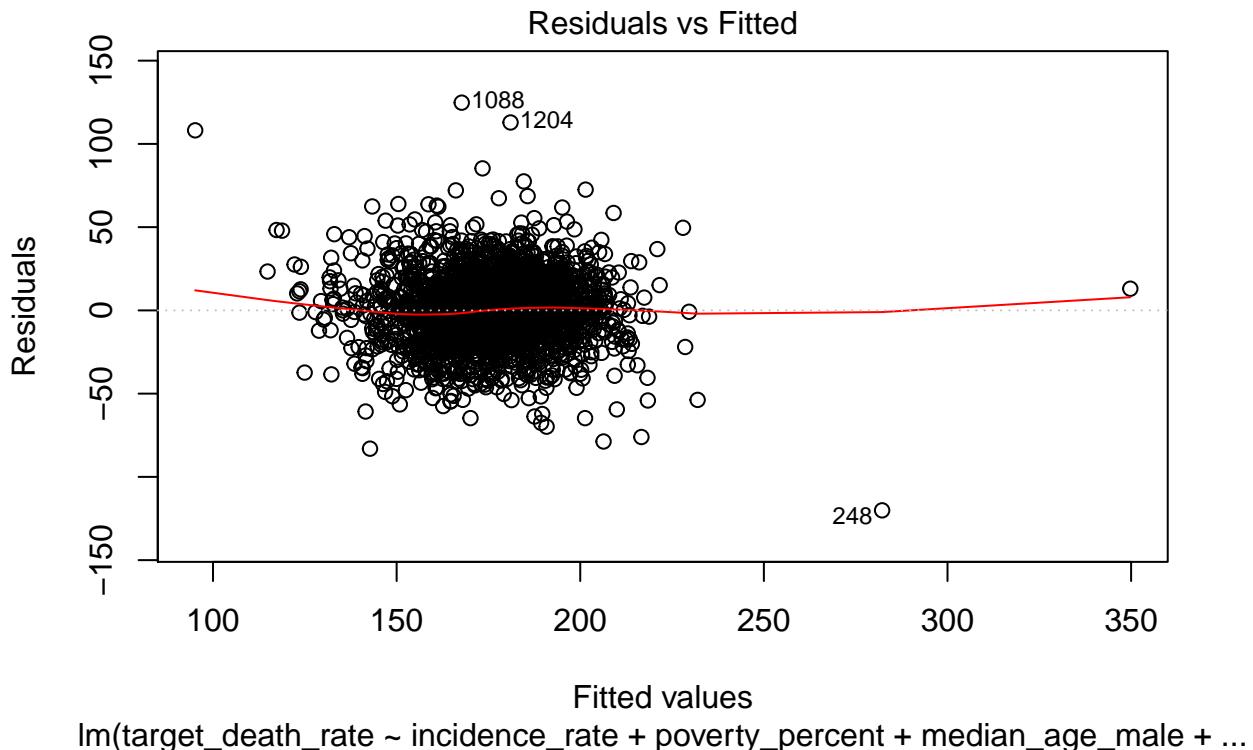
## 1646	0.02	0.01	0.01	0.09	1.02_*	0.00	0.01_*
## 1648	-0.01	-0.01	-0.04	-0.12	1.02_*	0.00	0.02_*
## 1649	-0.03	0.02	0.01	0.38_*	1.03_*	0.01	0.04_*
## 1657	-0.03	-0.05	0.03	-0.13	0.99_*	0.00	0.00
## 1665	0.07	-0.09	-0.14	0.20_*	1.00	0.00	0.01
## 1678	-0.02	0.03	-0.01	0.07	1.01_*	0.00	0.01_*
## 1684	-0.01	0.03	0.04	0.13	1.03_*	0.00	0.02_*
## 1694	-0.01	0.01	0.00	-0.11	1.01_*	0.00	0.01_*
## 1708	-0.01	0.01	-0.01	0.02	1.01_*	0.00	0.01
## 1726	-0.02	-0.02	-0.01	-0.05	1.01_*	0.00	0.01
## 1737	0.00	0.00	-0.02	0.12	1.06_*	0.00	0.05_*
## 1738	0.00	0.00	-0.01	0.01	1.01_*	0.00	0.01
## 1748	0.03	-0.12	-0.13	0.46_*	0.99_*	0.02	0.02_*
## 1752	0.06	-0.07	-0.09	0.19_*	0.99	0.00	0.01
## 1757	-0.01	0.00	0.01	-0.02	1.02_*	0.00	0.01_*
## 1768	0.00	-0.01	0.01	0.50_*	1.02_*	0.03	0.04_*
## 1795	0.02	-0.05	-0.02	-0.07	0.99_*	0.00	0.00
## 1815	0.06	-0.05	-0.02	0.13	0.98_*	0.00	0.00
## 1834	-0.03	0.00	0.00	0.04	1.01_*	0.00	0.01
## 1836	-0.11	0.10	0.31	-0.52_*	0.96_*	0.03	0.02_*
## 1837	0.00	0.07	0.15	-0.18_*	1.00	0.00	0.01
## 1848	0.07	-0.06	-0.08	0.20_*	1.00	0.00	0.01
## 1856	0.00	0.00	0.01	-0.05	1.03_*	0.00	0.03_*
## 1870	0.00	0.01	-0.01	-0.04	1.01_*	0.00	0.01
## 1926	0.07	-0.07	0.03	0.14	0.97_*	0.00	0.00
## 1955	-0.02	0.00	0.10	-0.12	1.03_*	0.00	0.03_*
## 2018	-0.04	0.05	0.04	0.10	0.99_*	0.00	0.00
## 2036	0.01	-0.01	-0.01	0.02	1.02_*	0.00	0.01_*
## 2047	0.00	-0.01	-0.01	-0.04	1.03_*	0.00	0.03_*
## 2048	0.01	0.02	0.00	0.10	1.12_*	0.00	0.10_*
## 2049	0.04	0.04	0.01	0.26_*	1.08_*	0.01	0.07_*
## 2050	-0.01	-0.03	-0.01	-0.11	1.05_*	0.00	0.04_*
## 2052	0.04	-0.09	0.11	-0.28_*	0.96_*	0.01	0.01
## 2057	0.07	-0.02	0.03	-0.16	0.98_*	0.00	0.00
## 2062	-0.01	-0.01	0.02	-0.12	0.98_*	0.00	0.00
## 2081	-0.24	0.10	-0.10	-0.31_*	1.02_*	0.01	0.03_*
## 2085	-0.03	-0.02	0.08	-0.15	0.99_*	0.00	0.00
## 2087	0.01	-0.07	-0.09	-0.18	0.98_*	0.00	0.00
## 2089	0.05	-0.08	0.01	-0.17	0.98_*	0.00	0.00
## 2145	0.00	0.00	0.00	0.00	1.01_*	0.00	0.01
## 2154	-0.03	-0.01	0.00	-0.09	0.99_*	0.00	0.00
## 2232	0.00	0.00	0.00	0.00	1.01_*	0.00	0.01
## 2282	0.05	0.03	-0.04	0.17	0.97_*	0.00	0.00
## 2287	0.09	0.03	0.12	0.29_*	0.96_*	0.01	0.01
## 2311	0.18	-0.15	-0.10	0.24_*	0.96_*	0.01	0.00
## 2322	-0.04	-0.04	0.06	-0.29_*	0.95_*	0.01	0.01
## 2332	0.00	-0.01	0.00	0.02	1.01_*	0.00	0.01
## 2340	0.11	0.00	0.04	0.16	1.02_*	0.00	0.02_*
## 2358	0.01	0.00	0.00	0.05	1.01_*	0.00	0.01
## 2360	-0.19	0.38	-0.25	1.80_*	0.97_*	0.32	0.09_*
## 2362	-0.03	0.06	0.00	0.15	1.01_*	0.00	0.01_*
## 2366	0.11	-0.03	0.19	-0.26_*	1.00	0.01	0.01_*
## 2370	0.03	0.00	-0.14	0.30_*	1.02_*	0.01	0.03_*
## 2372	-0.08	0.09	-0.07	0.25_*	0.99	0.01	0.01

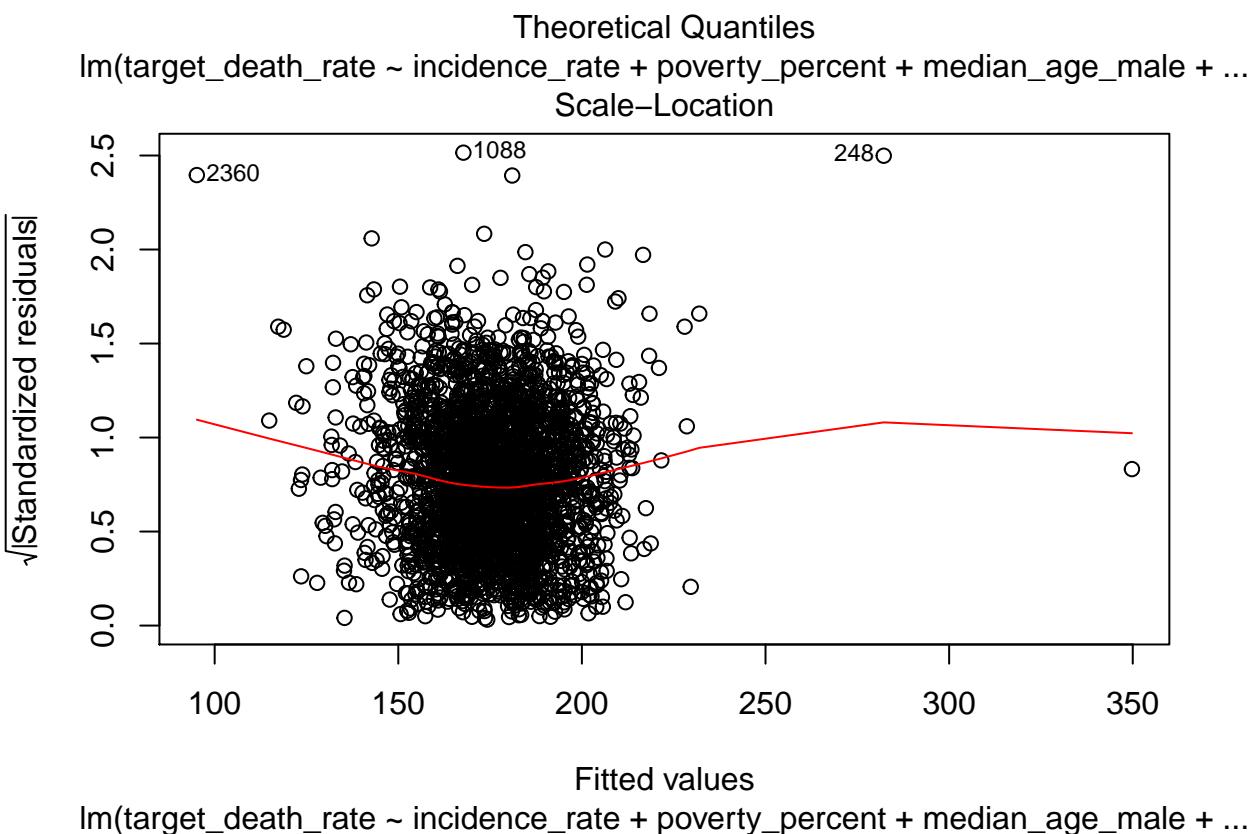
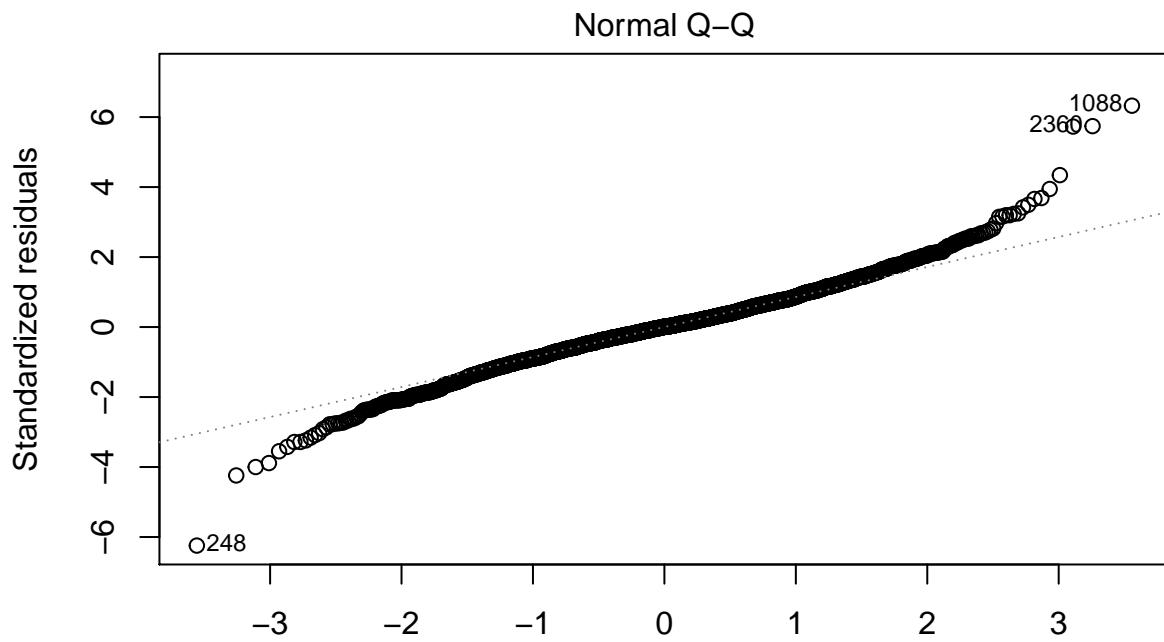
```

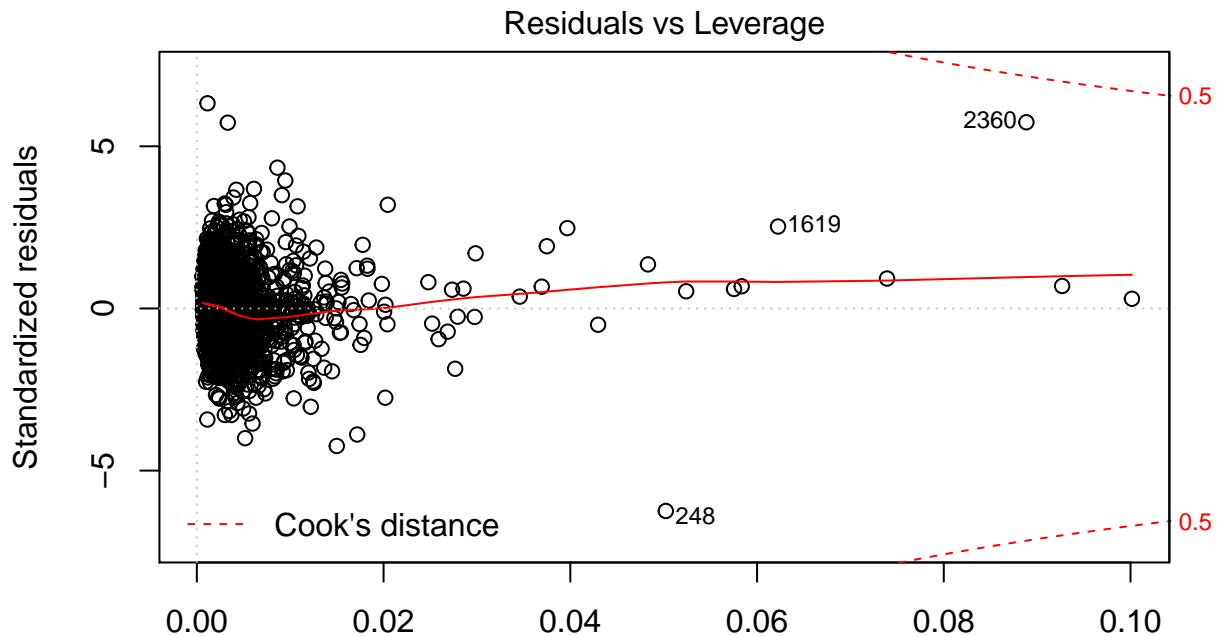
## 2373 -0.22      0.12    -0.15     0.33_*   0.98_*   0.01    0.01
## 2374  0.01     -0.02     0.02    -0.07   1.02_*   0.00    0.02_*
## 2375 -0.02      0.02    -0.01     0.04   1.01_*   0.00    0.01
## 2378  0.05     -0.01     0.10    -0.15   1.02_*   0.00    0.02_*
## 2387 -0.01     -0.04    -0.08   -0.22_*   1.00   0.00    0.01_*
## 2390 -0.01      0.01    -0.02    -0.07   1.01_*   0.00    0.01_*
## 2419  0.00      0.00     0.00     0.01   1.01_*   0.00    0.01
## 2424  0.02     -0.10    -0.09   -0.15   0.99_*   0.00    0.00
## 2425  0.00      0.00     0.00     0.00   1.01_*   0.00    0.01
## 2442  0.01     -0.06     0.04   -0.20_*   0.97_*   0.00    0.00
## 2443  0.01     -0.01     0.02     0.09   0.99_*   0.00    0.00
## 2444  0.05     -0.05    -0.08   -0.23_*   0.99_*   0.01    0.01
## 2469  0.04     -0.03    -0.03     0.12   0.98_*   0.00    0.00
## 2482  0.07     -0.04    -0.02     0.13   0.98_*   0.00    0.00
## 2493  0.00      0.00     0.00     0.02   1.01_*   0.00    0.01
## 2544  0.03     -0.03     0.01     0.09   0.98_*   0.00    0.00
## 2626  0.00      0.01    -0.02   -0.05   1.01_*   0.00    0.01
## 2662  0.00     -0.06   -0.07     0.14   0.99_*   0.00    0.00

```

```
plot(backward_model_high)
```







Leverage

```
lm(target_death_rate ~ incidence_rate + poverty_percent + median_age_male + ...)
```

```
income_high_rm <- income_high_data[-c(1088, 2360, 248),]

backward_model_high_rm <- lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
  median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
  pct_married_households + pct_upto_hs18_24 + pct_with_coverage,
  data = income_high_rm)
summary(backward_model_high_rm)
```

```
##
## Call:
## lm(formula = target_death_rate ~ incidence_rate + poverty_percent +
##   median_age_male + pct_unemployed16_over + pct_asian + pct_other_race +
##   pct_married_households + pct_upto_hs18_24 + pct_with_coverage,
##   data = income_high_rm)
##
## Residuals:
##    Min      1Q  Median      3Q     Max 
## -80.779 -11.178   0.048  11.373 113.085 
## 
## Coefficients:
##             Estimate Std. Error t value Pr(>|t|)    
## (Intercept) 88.984965  11.931213  7.458 1.18e-13 ***
## incidence_rate 0.214122  0.007945 26.949 < 2e-16 ***
## poverty_percent 0.955402  0.125446  7.616 3.61e-14 ***
## median_age_male -0.306979  0.083133 -3.693 0.000226 ***
## pct_unemployed16_over 1.035570  0.157925  6.557 6.56e-11 ***
## pct_asian      -0.983436  0.157993 -6.225 5.59e-10 ***
## pct_other_race -1.006896  0.123860 -8.129 6.55e-16 ***
## pct_married_households -0.247956  0.081267 -3.051 0.002302 ** 
## pct_upto_hs18_24   0.385275  0.037974 10.146 < 2e-16 ***
```

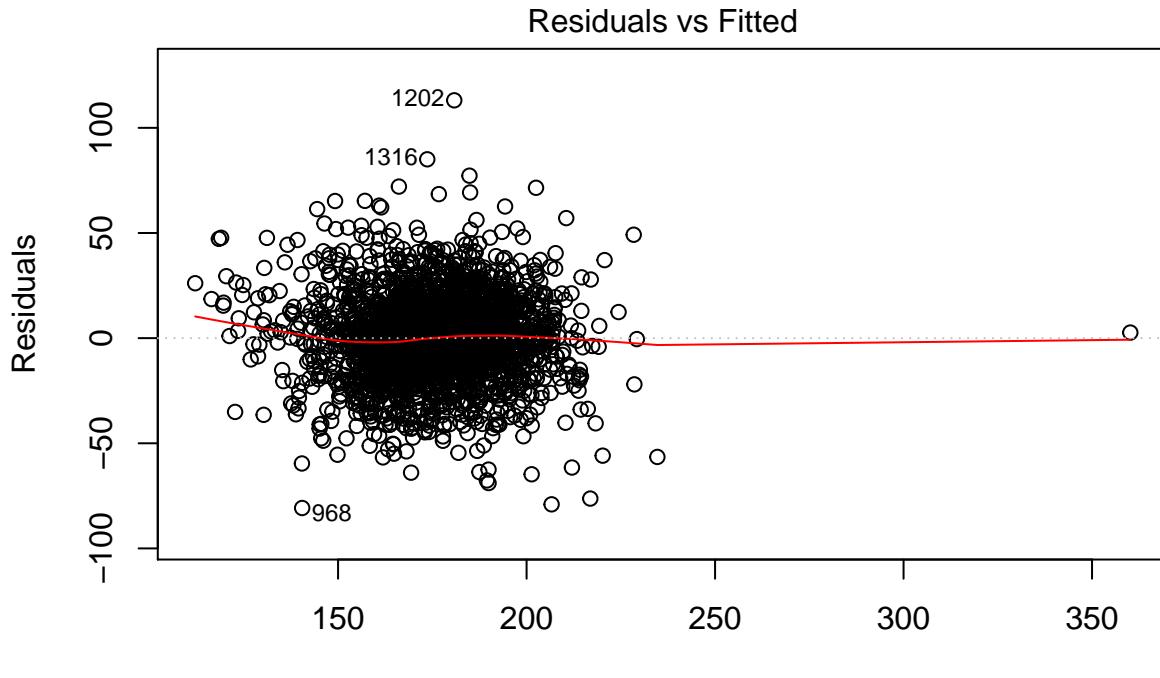
```

## pct_with_coverage      -0.277761   0.102610  -2.707 0.006834 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 19.32 on 2656 degrees of freedom
## Multiple R-squared:  0.4163, Adjusted R-squared:  0.4143
## F-statistic: 210.4 on 9 and 2656 DF,  p-value: < 2.2e-16
cbind(summary(backward_model_high_rm) %>% tidy %>% dplyr::select(-statistic),
       summary(backward_model_high) %>% tidy %>% dplyr::select(-statistic)) %>%
knitr::kable()

```

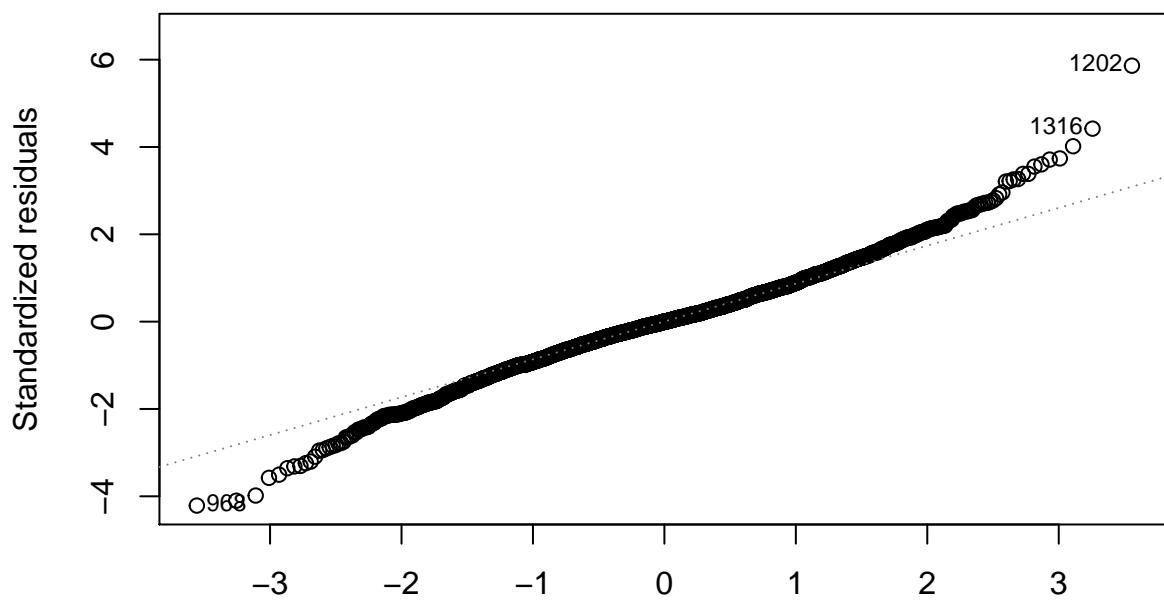
term	estimate	std.error	p.value	term	estimate	std.error
(Intercept)	88.9849652	11.9312133	0.0000000	(Intercept)	94.8142834	12.1673029
incidence_rate	0.2141222	0.0079454	0.0000000	incidence_rate	0.2002120	0.0079169
poverty_percent	0.9554022	0.1254455	0.0000000	poverty_percent	0.9427978	0.1281059
median_age_male	-0.3069793	0.0831330	0.0002264	median_age_male	-0.2846952	0.0848240
pct_unemployed16_over	1.0355703	0.1579250	0.0000000	pct_unemployed16_over	1.0011861	0.1608617
pct_asian	-0.9834361	0.1579927	0.0000000	pct_asian	-0.7500694	0.1553122
pct_other_race	-1.0068963	0.1238602	0.0000000	pct_other_race	-1.0655150	0.1262476
pct_married_households	-0.2479565	0.0812670	0.0023023	pct_married_households	-0.2694759	0.0829536
pct_upto_hs18_24	0.3852755	0.0379742	0.0000000	pct_upto_hs18_24	0.4069440	0.0386684
pct_with_coverage	-0.2777608	0.1026097	0.0068336	pct_with_coverage	-0.2800928	0.1046493

```
plot(backward_model_high_rm)
```



```
lm(target_death_rate ~ incidence_rate + poverty_percent + median_age_male + ...)
```

Normal Q–Q



Theoretical Quantiles

lm(target\_death\_rate ~ incidence\_rate + poverty\_percent + median\_age\_male + ...)

Scale–Location

