

Census Regression Project

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```
library(psych)
```

```
##  
## Attaching package: 'psych'  
## The following objects are masked from 'package:ggplot2':  
##  
##    %+%, alpha
```

```
library(xtable)  
library(gridExtra)
```

```
##  
## Attaching package: 'gridExtra'  
## The following object is masked from 'package:dplyr':  
##  
##    combine
```

```
library(grid)
```

```
# Compute summary statistics  
table <- describe(tract_data)
```

```
# Convert the summary statistics to a data frame  
table_df <- as.data.frame(table)
```

```
# Remove unnecessary columns and rename them  
table_df <- table_df[, c("mean", "sd", "median", "range", "skew", "kurtosis")]  
names(table_df) <- c("Mean", "SD", "Median", "Range", "Skew", "Kurtosis")
```

```
# Add a column for variable names  
table_df$Variable <- rownames(table_df)
```

```
# Reorder columns  
describe(table_df)
```

```
##      vars  n      mean      sd median trimmed  mad   min     max  
## Mean      1 22 3631.24 13683.04 28.63  419.63 40.97 -81.85 64266.14  
## SD        2 22 1858.37  5491.86  9.66  276.97 14.13  0.00 23202.54  
## Median    3 22 3125.11 12531.72 27.39  340.68 39.40 -81.86 59015.75  
## Range     4 22 11369.91 33647.15 53.16 1293.36 77.81  0.00 131764.75  
## Skew      5 20   0.96   1.71  0.89   0.94  1.84 -2.34   3.74  
## Kurtosis  6 20   4.75   6.81  1.38   3.74  2.58 -0.79  20.61  
## Variable* 7 22   11.50   6.49 11.50  11.50  8.15  1.00  22.00
```

```
##           range skew kurtosis      se
## Mean      64347.99 3.94    14.53 2917.23
## SD        23202.54 2.99     8.07 1170.87
## Median    59097.61 4.02    14.99 2671.77
## Range     131764.75 2.75     6.20 7173.60
## Skew       6.09 0.18    -1.01   0.38
## Kurtosis   21.41 1.06    -0.45   1.52
## Variable*  21.00 0.00    -1.36   1.38

# Generate LaTeX table
#latex_table <- xtable(table_df, caption = "Summary Statistics", label = "tab:summary_statistics")

# Print LaTeX code
#print(latex_table, type = "latex", include.rownames = FALSE, booktabs = TRUE)
```

VISUALS

```
# Load necessary libraries
# Select the relevant columns for correlation analysis
cor_data <- tract_data[, c("med_householdincome", "poverty_total", "employment_total",
                          "white_prct", "black_prct", "asian_prct", "other_prct",
                          "college_prct", "avg_workhours", "distance_from_closest_stop_miles", "total_pop")]

# Shorten column names
colnames(cor_data) <- c("Med. Income", "Poverty %", "Employment",
                      "White %", "Black %", "Asian %", "Other %",
                      "College %", "Avg. Work Hours", "Distance to Stop (miles)", "total_pop")

# Compute the correlation matrix
cor_matrix <- cor(cor_data, use = "complete.obs")

# Print the correlation matrix
print(cor_matrix)
```

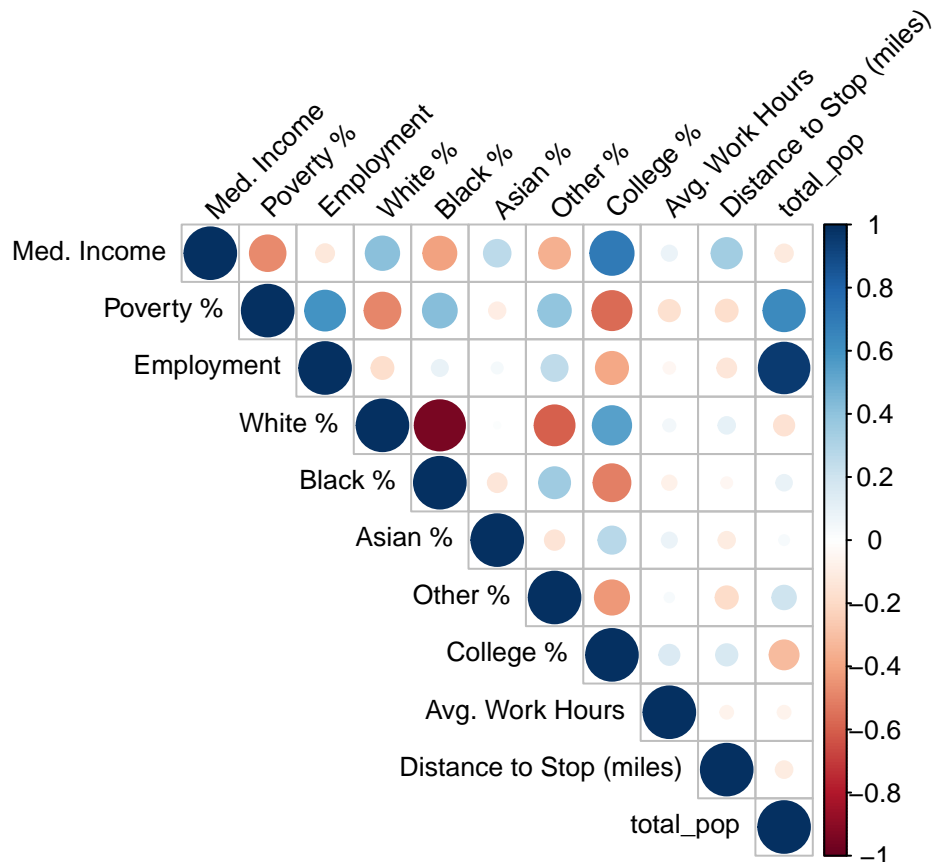
```
##           Med. Income Poverty % Employment White %
## Med. Income      1.00000000 -0.4740038 -0.12077539 0.41733408
## Poverty %        -0.47400378  1.00000000  0.59409866 -0.48040160
## Employment       -0.12077539  0.5940987  1.00000000 -0.17973704
## White %           0.41733408 -0.4804016 -0.17973704  1.00000000
## Black %          -0.40634404  0.4282422  0.09630535 -0.94347791
## Asian %           0.26468248 -0.0999783  0.04512880  0.01425825
## Other %          -0.35629846  0.3939630  0.25271781 -0.59663592
## College %         0.70026247 -0.5695919 -0.38640962  0.54947634
## Avg. Work Hours   0.08667611 -0.1682983 -0.04969641  0.05569615
## Distance to Stop (miles) 0.34314730 -0.1784229 -0.13305873 0.10285681
## total_pop        -0.11372220  0.6351349  0.95683452 -0.15333452
##           Black % Asian % Other % College %
## Med. Income      -0.40634404 0.26468248 -0.35629846 0.7002625
## Poverty %         0.42824219 -0.09997830 0.39396303 -0.5695919
## Employment        0.09630535 0.04512880 0.25271781 -0.3864096
## White %          -0.94347791 0.01425825 -0.59663592 0.5494763
## Black %           1.00000000 -0.13124352 0.35908669 -0.5087550
## Asian %          -0.13124352  1.00000000 -0.14080189 0.2747686
## Other %           0.35908669 -0.14080189  1.00000000 -0.4369937
## College %         -0.50875504 0.27476860 -0.43699372  1.0000000
## Avg. Work Hours   -0.07735642 0.08506950 0.03295925 0.1510042
```

```
## Distance to Stop (miles) -0.04431720 -0.10093935 -0.18518514 0.1691971
## total_pop                0.09097806 0.03651585 0.20697278 -0.3133628
##                           Avg. Work Hours Distance to Stop (miles) total_pop
## Med. Income              0.08667611                0.34314730 -0.11372220
## Poverty %                -0.16829831                -0.17842288 0.63513492
## Employment               -0.04969641                -0.13305873 0.95683452
## White %                   0.05569615                0.10285681 -0.15333452
## Black %                  -0.07735642                -0.04431720 0.09097806
## Asian %                   0.08506950                -0.10093935 0.03651585
## Other %                   0.03295925                -0.18518514 0.20697278
## College %                 0.15100415                0.16919708 -0.31336284
## Avg. Work Hours           1.00000000                -0.06086333 -0.06146735
## Distance to Stop (miles) -0.06086333                1.00000000 -0.10410035
## total_pop                -0.06146735                -0.10410035 1.00000000
```

```
# Visualize the correlation matrix with shortened labels
library(corrplot)
```

```
## corrplot 0.95 loaded
```

```
corrplot(cor_matrix, method = "circle", type = "upper",
         tl.col = "black", tl.srt = 45,
         tl.cex = 0.8) # Optionally adjust text size with tl.cex
```



```
# Load necessary library
library(ggplot2)
# Load necessary package
library(gridExtra)
```

```

# Create individual plots
plot1 <- ggplot(tract_data, aes(x = distance_from_closest_stop_miles, y = med_householdincome)) +
  geom_point(color = "blue") +
  geom_smooth(method = "lm", se = FALSE, color = "red") +
  labs(
    x = "Distance to Stop (miles)",
    y = "Median Household Income") +
  theme_minimal()

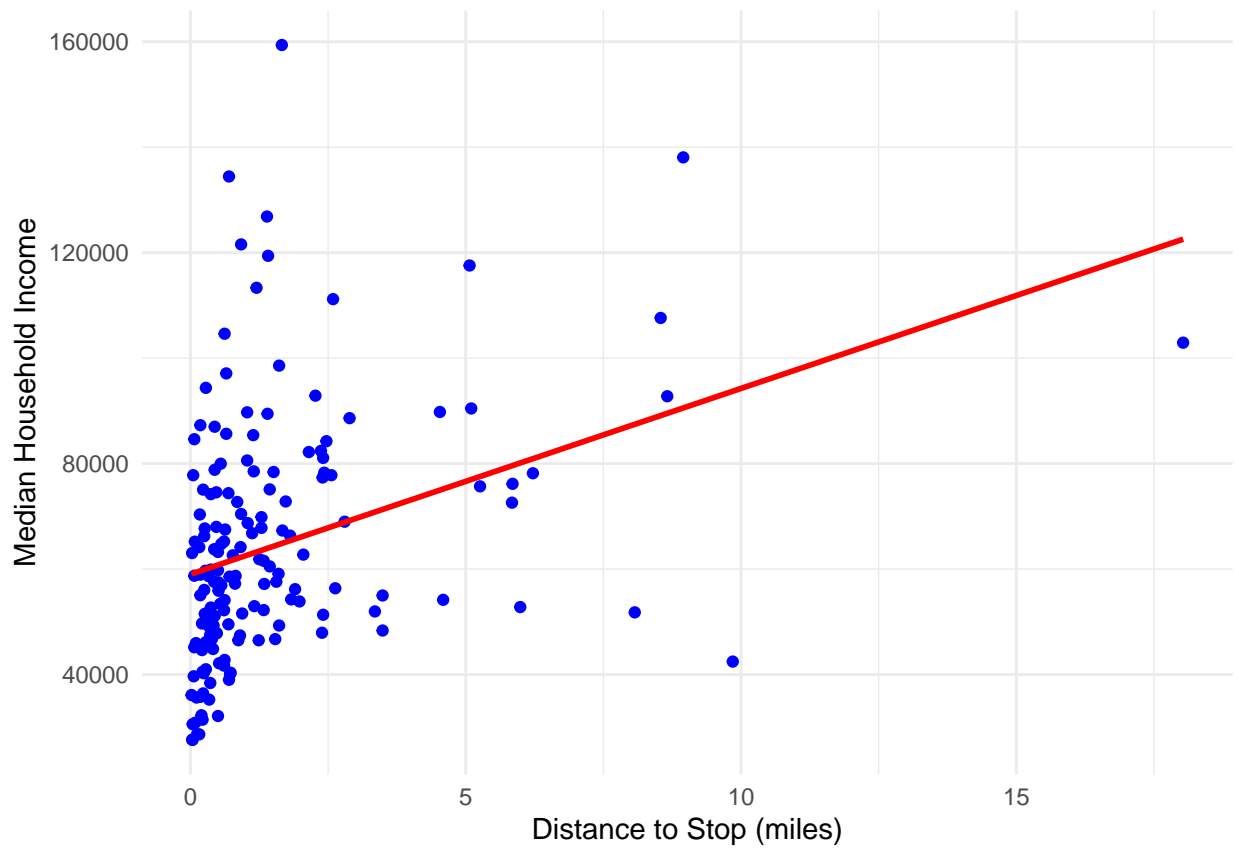
plot2 <- ggplot(tract_data, aes(x = distance_from_closest_stop_miles, y = poverty_prct)) +
  geom_point(color = "green") +
  geom_smooth(method = "lm", se = FALSE, color = "red") +
  labs(
    x = "Distance to Stop (miles)",
    y = "Poverty Percentage") +
  theme_minimal()

plot3 <- ggplot(tract_data, aes(x = distance_from_closest_stop_miles, y = employment_prct)) +
  geom_point(color = "purple") +
  geom_smooth(method = "lm", se = FALSE, color = "red") +
  labs(
    x = "Distance to Stop (miles)",
    y = "Employment Percentage") +
  theme_minimal()

print(plot1)

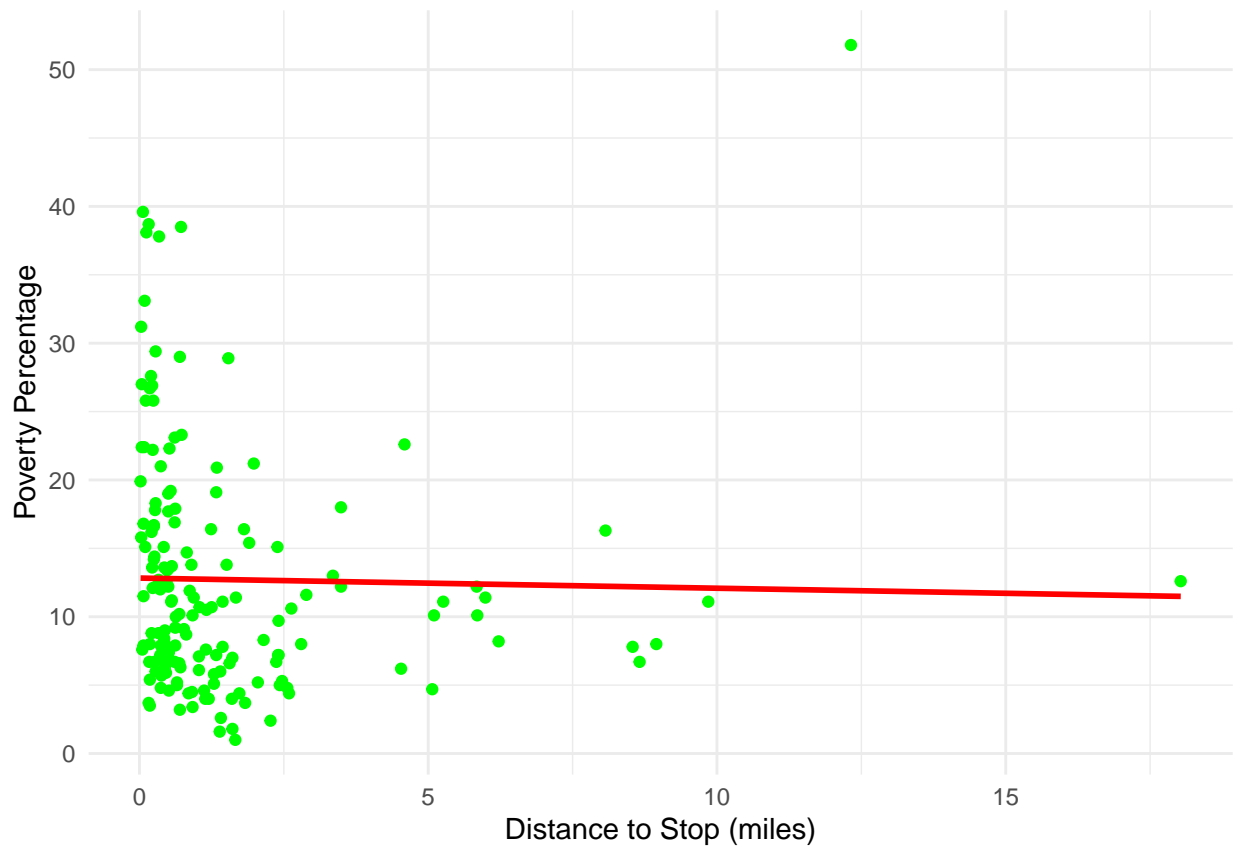
## `geom_smooth()` using formula = 'y ~ x'
## Warning: Removed 1 row containing non-finite outside the scale range
## (`stat_smooth()`).
## Warning: Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).

```



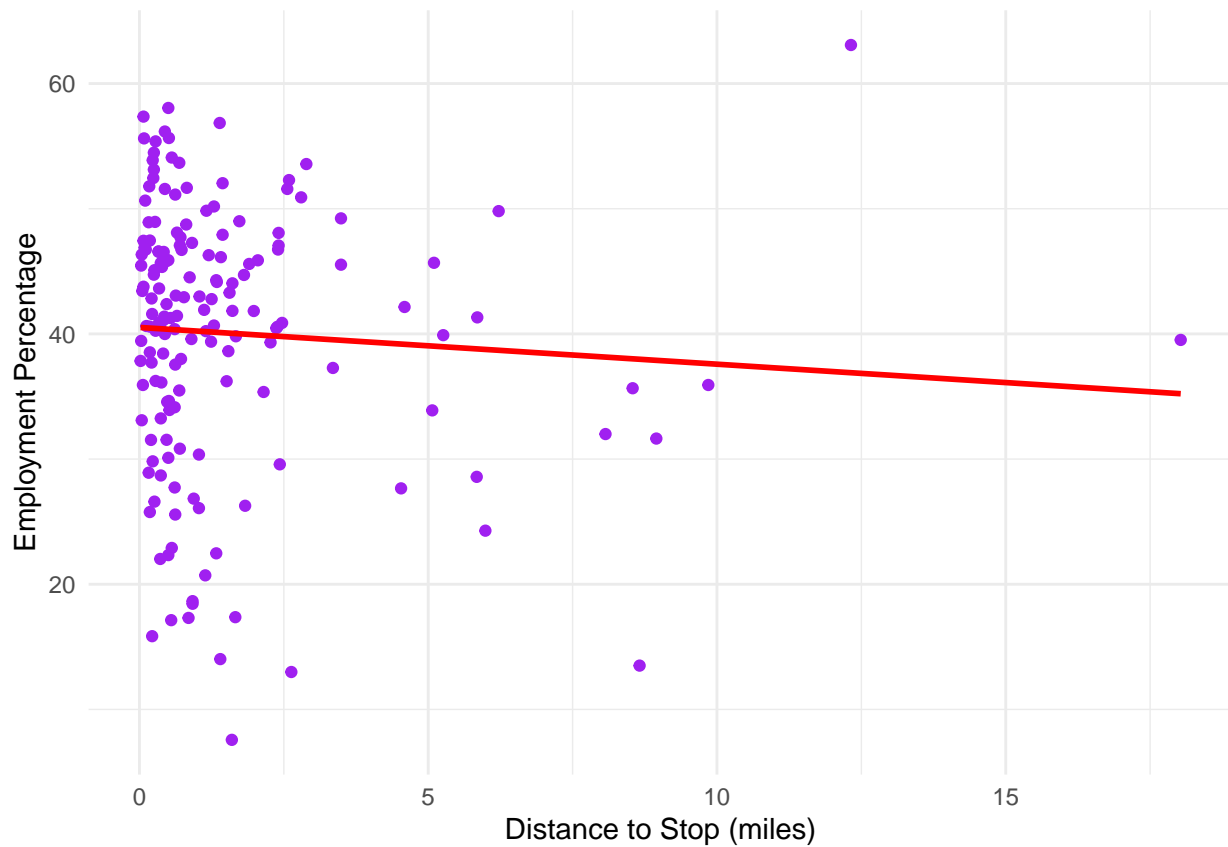
```
print(plot2)
```

```
## `geom_smooth()` using formula = 'y ~ x'
```



```
print(plot3)
```

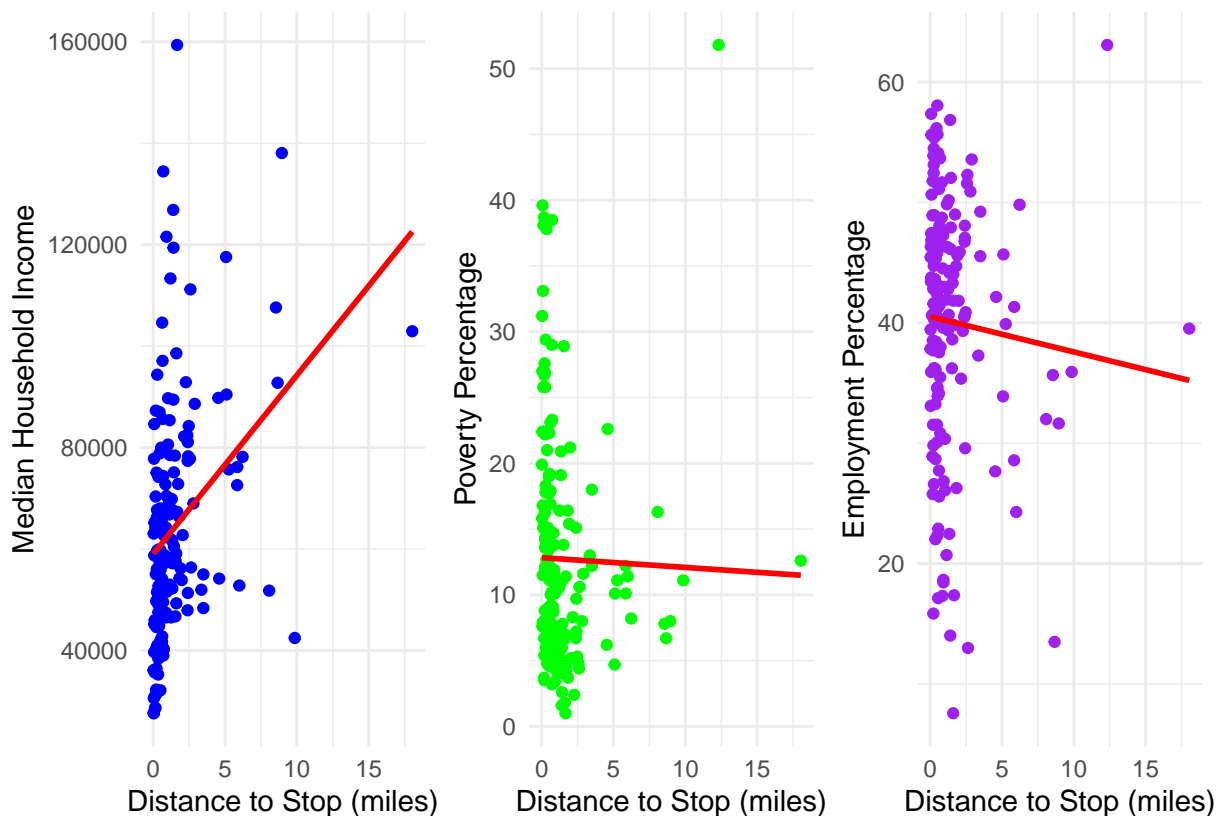
```
## `geom_smooth()` using formula = 'y ~ x'
```



```
library(patchwork)

# Combine plots side by side
plot1 + plot2 + plot3

## `geom_smooth()` using formula = 'y ~ x'
## Warning: Removed 1 row containing non-finite outside the scale range (`stat_smooth()`).
## Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).
## `geom_smooth()` using formula = 'y ~ x'
## `geom_smooth()` using formula = 'y ~ x'
```



3 Different Models

```
# Create non_white_prct variable
tract_data$non_white_prct <- 100 - tract_data$white_prct

# Model 1: Median Household Income (using non_white_prct)
model1 <- lm(med_householdincome ~ distance_from_closest_stop_miles + total_pop +
             non_white_prct +
             college_prct + avg_workhours, data = tract_data)

summary(model1)

##
## Call:
## lm(formula = med_householdincome ~ distance_from_closest_stop_miles +
##     total_pop + non_white_prct + college_prct + avg_workhours,
##     data = tract_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -44141  -8346   -350    7243   57433
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    22099.0034  23759.8687   0.930   0.354
## distance_from_closest_stop_miles  2445.8372   553.7963   4.416 1.85e-05 ***
## total_pop           1.0983     0.4712   2.331   0.021 *
```



```
## non_white_prct          -61.9856    98.2576  -0.631    0.529
## college_prct           1055.6924   104.2820  10.123 < 2e-16 ***
## avg_workhours          51.5489    609.6063   0.085    0.933
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15650 on 158 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.5588, Adjusted R-squared:  0.5449
## F-statistic: 40.03 on 5 and 158 DF,  p-value: < 2.2e-16
```

```
# Model 2: Total Poverty (using non_white_prct)
model2 <- lm(poverty_prct ~ distance_from_closest_stop_miles + total_pop +
             non_white_prct +
             college_prct + avg_workhours, data = tract_data)

summary(model2)
```

```
##
## Call:
## lm(formula = poverty_prct ~ distance_from_closest_stop_miles +
##     total_pop + non_white_prct + college_prct + avg_workhours,
##     data = tract_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -14.2627  -3.8989  -0.2931   2.9885  27.8206
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    43.3777465   9.7666835   4.441 1.67e-05 ***
## distance_from_closest_stop_miles  0.0985276   0.2160275   0.456  0.6489
## total_pop      -0.0003239   0.0001940  -1.670  0.0970 .
## non_white_prct   0.2131452   0.0405839   5.252 4.76e-07 ***
## college_prct    -0.2343747   0.0431465  -5.432 2.05e-07 ***
## avg_workhours   -0.6526934   0.2507031  -2.603  0.0101 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.486 on 159 degrees of freedom
## Multiple R-squared:  0.4712, Adjusted R-squared:  0.4546
## F-statistic: 28.34 on 5 and 159 DF,  p-value: < 2.2e-16
```

```
# Model 3: Employment Percentage (using non_white_prct)
model3 <- lm(employment_prct ~ distance_from_closest_stop_miles + total_pop +
             non_white_prct +
             college_prct + avg_workhours, data = tract_data)

summary(model3)
```

```
##
## Call:
## lm(formula = employment_prct ~ distance_from_closest_stop_miles +
##     total_pop + non_white_prct + college_prct + avg_workhours,
##     data = tract_data)
```

```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -28.1439  -5.5340   0.4477   5.3537  22.7078
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      35.8248220  13.7207297   2.611  0.00989 **
## distance_from_closest_stop_miles  0.1208936  0.3034863   0.398  0.69091
## total_pop         0.0005568  0.0002725   2.043  0.04271 *
## non_white_prct    -0.0204344  0.0570143  -0.358  0.72051
## college_prct      -0.3454503  0.0606144  -5.699  5.7e-08 ***
## avg_workhours     0.3221758  0.3522004   0.915  0.36171
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.112 on 159 degrees of freedom
## Multiple R-squared:  0.2756, Adjusted R-squared:  0.2528
## F-statistic: 12.1 on 5 and 159 DF, p-value: 6.137e-10

# Model 1: Median Household Income
modell1 <- lm(med_householdincome ~ distance_from_closest_stop_miles + total_pop +
             black_prct + asian_prct + other_prct +
             college_prct + avg_workhours, data = tract_data)

summary(modell1)

##
## Call:
## lm(formula = med_householdincome ~ distance_from_closest_stop_miles +
##     total_pop + black_prct + asian_prct + other_prct + college_prct +
##     avg_workhours, data = tract_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -44628  -9085   -245   6555  58664
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      25039.9521 23630.7821   1.060  0.2909
## distance_from_closest_stop_miles  2603.4141  560.8072   4.642 7.26e-06 ***
## total_pop         0.9651    0.4754   2.030  0.0441 *
## black_prct       -126.8181  115.9915  -1.093  0.2759
## asian_prct       1079.9039  582.6794   1.853  0.0657 .
## other_prct       -124.1290  322.4062  -0.385  0.7008
## college_prct      964.5357  109.4693   8.811 2.27e-15 ***
## avg_workhours     25.6805   608.2672   0.042  0.9664
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 15530 on 156 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.5714, Adjusted R-squared:  0.5522
## F-statistic: 29.71 on 7 and 156 DF, p-value: < 2.2e-16
```

```

# Model 2: Total Poverty
model2 <- lm(poverty_prct ~ distance_from_closest_stop_miles + total_pop +
             black_prct + asian_prct + other_prct +
             college_prct + avg_workhours, data = tract_data)

summary(model2)

##
## Call:
## lm(formula = poverty_prct ~ distance_from_closest_stop_miles +
##     total_pop + black_prct + asian_prct + other_prct + college_prct +
##     avg_workhours, data = tract_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -13.342  -3.989  -0.432   2.928  27.118
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    43.7784625   9.7657109   4.483 1.41e-05 ***
## distance_from_closest_stop_miles  0.0925207   0.2189339   0.423  0.6732
## total_pop      -0.0003109   0.0001964  -1.583  0.1154
## black_prct      0.2337364   0.0482741   4.842 3.06e-06 ***
## asian_prct      0.2050642   0.2361388   0.868  0.3865
## other_prct      0.1914674   0.1344255   1.424  0.1563
## college_prct    -0.2351021   0.0452911  -5.191 6.39e-07 ***
## avg_workhours   -0.6555436   0.2516768  -2.605  0.0101 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 6.474 on 157 degrees of freedom
## Multiple R-squared:  0.4798, Adjusted R-squared:  0.4566
## F-statistic: 20.69 on 7 and 157 DF,  p-value: < 2.2e-16

# Model 3: Employment Percentage
model3 <- lm(employment_prct ~ distance_from_closest_stop_miles + total_pop +
             black_prct + asian_prct + other_prct +
             college_prct + avg_workhours, data = tract_data)

summary(model3)

##
## Call:
## lm(formula = employment_prct ~ distance_from_closest_stop_miles +
##     total_pop + black_prct + asian_prct + other_prct + college_prct +
##     avg_workhours, data = tract_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -27.8135  -5.1871   0.0835   5.3736  22.3175
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    39.9143249  12.8882698   3.097  0.00232 **
## distance_from_closest_stop_miles  0.3131106   0.2889374   1.084  0.28018

```

```

## total_pop                0.0003596  0.0002592   1.387  0.16731
## black_prct               -0.1551475  0.0637097  -2.435  0.01600 *
## asian_prct               1.3266567  0.3116436   4.257  3.55e-05 ***
## other_prct               0.3035306  0.1774077   1.711  0.08907 .
## college_prct            -0.4289035  0.0597729  -7.176  2.69e-11 ***
## avg_workhours            0.2451623  0.3321498   0.738  0.46155
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 8.544 on 157 degrees of freedom
## Multiple R-squared:  0.3711, Adjusted R-squared:  0.343
## F-statistic: 13.23 on 7 and 157 DF,  p-value: 2.379e-13

combined_model <- lm(distance_from_closest_stop_miles ~ med_householdincome + poverty_prct + employment_prct +
                    black_prct + asian_prct + other_prct + college_prct + avg_workhours, data = tract_data)
summary(combined_model)

##
## Call:
## lm(formula = distance_from_closest_stop_miles ~ med_householdincome +
##     poverty_prct + employment_prct + total_pop + black_prct +
##     asian_prct + other_prct + college_prct + avg_workhours, data = tract_data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.2300 -0.9998 -0.2933  0.3633 14.1261
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    3.188e+00  3.476e+00   0.917  0.3605
## med_householdincome  5.452e-05  1.148e-05   4.750 4.61e-06 ***
## poverty_prct      2.688e-03  2.975e-02   0.090  0.9281
## employment_prct  -3.666e-02  2.161e-02  -1.697  0.0918 .
## total_pop        -4.937e-05  6.430e-05  -0.768  0.4438
## black_prct        1.334e-02  1.675e-02   0.796  0.4271
## asian_prct       -1.542e-01  8.027e-02  -1.921  0.0566 .
## other_prct       -4.779e-02  4.390e-02  -1.089  0.2780
## college_prct     -4.163e-02  2.263e-02  -1.840  0.0678 .
## avg_workhours    -5.153e-02  8.256e-02  -0.624  0.5335
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.071 on 154 degrees of freedom
## (1 observation deleted due to missingness)
## Multiple R-squared:  0.2055, Adjusted R-squared:  0.1591
## F-statistic: 4.427 on 9 and 154 DF,  p-value: 3.624e-05

```