# Congestive Heart Failure Analysis

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# Is There a Connection Between Congestive Heart Failure and Socioeconomic Status (and/or) Urbanization Level?

## Loading packages

```
library(tidyverse)
library(ggplot2)
library(sjPlot)
library(formatR)
```

## **Importing Data**

```
wonder_data <- read_csv("CHF Data.csv")
education_data <- read_csv("Education Data.csv")
poverty_data <- read_csv("Poverty and Household Income Data.csv")
food_insecurity_data <- read_csv("Food Insecurity Data.csv")
unemployment_data <- read_csv("Unemployment Data.csv")</pre>
```

### **Merging Data**

```
full_data <- wonder_data %>%
    inner_join(education_data, by = c("County", "State")) %>%
    inner_join(poverty_data, by = c("County", "State")) %>%
    inner_join(food_insecurity_data, by = c("County", "State")) %>%
    inner_join(unemployment_data, by = c("County", "State"))
```

#### Cleaning the Data

Changing Some Variable Names For Simplicity

```
colnames(full_data)[colnames(full_data) == "Crude Rate"] <- "rate"
colnames(full_data)[colnames(full_data) == "Percent of adults with less than a high school diploma, 200
colnames(full_data)[colnames(full_data) == "Percent of adults with a high school diploma only, 2008-12"]
colnames(full_data)[colnames(full_data) == "Percent of adults with a bachelor's degree or higher, 2008-
colnames(full_data)[colnames(full_data) == "Poverty Percent, All Ages"] <- "poverty"
colnames(full_data)[colnames(full_data) == "Median Household Income"] <- "income"
colnames(full_data)[colnames(full_data) == "2014 Food Insecurity Rate"] <- "food_insecure"
colnames(full_data)[colnames(full_data) == "Rural_Urban_Continuum_Code_2013"] <- "urbanization"</pre>
```

#### Only Keeping Necessary Variables In the Dataframe

#### Removing Missing Values and Fixing Data Types

```
# Removing missing values

df <- na.omit(df)

# Removing percent sign from food_insecure

df$food_insecure <- as.numeric(gsub("[\\\\,]", "", df$food_insecure))

# Removing commas from income

df$income <- as.numeric(gsub(",", "", df$income))

# Changing poverty to a numeric

df$poverty <- as.numeric(df$poverty)

# Changing urbanization to a factor

df$urbanization <- as.factor(df$urbanization)</pre>
```

#### Exploratory Data Analysis and Data Visualization

Looking At the Summary Statistics For Crude Mortality Rate and Looking For Outliers

```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.00 19.20 27.60 32.27 39.58 309.60
```

```
summary(df$Deaths)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
```

141.0 10523.0

Removing Large Outliers Using IQR\*1.5

64.0

162.7

31.0

##

0.0

```
quartiles <- quantile(df$rate, probs = c(0.25, 0.75), na.rm = FALSE)
IQR <- IQR(df$rate)
lower <- quartiles[1] - 1.5 * IQR
upper <- quartiles[2] + 1.5 * IQR
df_new <- subset(df, df$rate > lower & df$rate < upper)</pre>
```

Attaching New Dataframe Without Outliers

```
attach(df_new)
```

New Summary Statistics After Removing Outliers

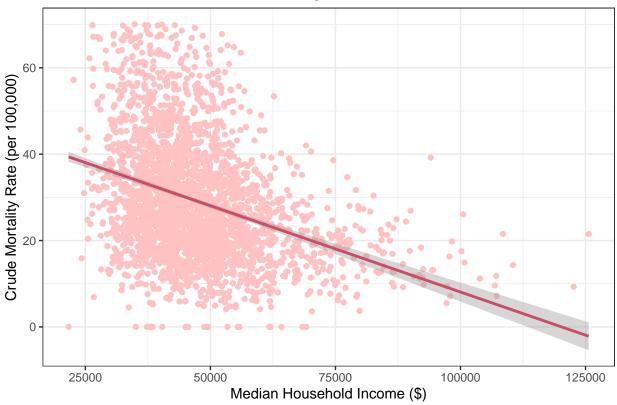
```
## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.00 18.80 26.80 29.09 37.40 70.10
```

#### **Data Visualization**

Crude Mortality Rate vs. Household Income

```
ggplot(df_new, aes(x = income, y = rate)) + geom_point(color = "rosybrown1") +
    geom_smooth(method = "lm", color = "#cc4b65") + labs(title = "Median Household Income vs. Congestive
    x = "Median Household Income ($)", y = "Crude Mortality Rate (per 100,000)") +
    theme_bw()
```

# Median Household Income vs. Congestive Heart Failure



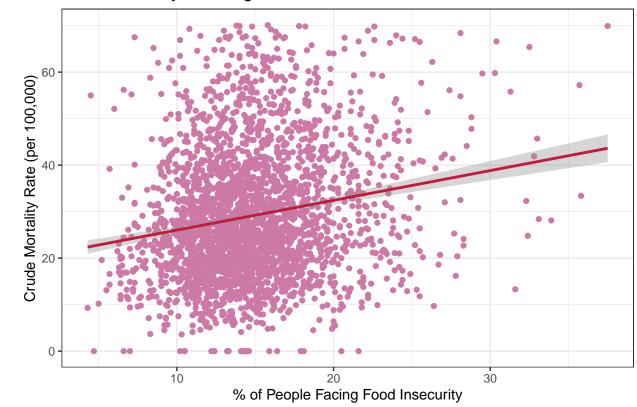
cor(income, rate)

## [1] -0.3493774

## Crude Mortality Rate vs Food Insecurity

```
ggplot(df_new, aes(x = food_insecure, y = rate)) + geom_point(color = "#CC79A7") +
    geom_smooth(method = "lm", color = "#c41b3d") + labs(title = "Food Insecurity vs. Congestive Heart to the state of the st
```

# Food Insecurity vs. Congestive Heart Failure



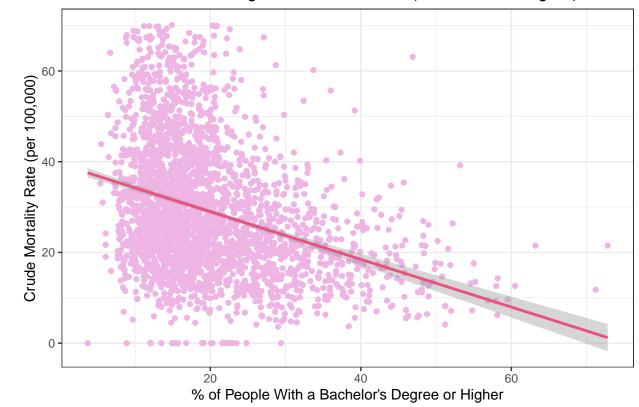
cor(food\_insecure, rate)

## [1] 0.1846869

## Crude Mortality Rate vs. Education Level

```
ggplot(df_new, aes(x = bs_higher, y = rate)) + geom_point(color = "#edb4e4") +
    geom_smooth(method = "lm", color = "#eb5081") + labs(title = "Education Level vs. Congestive Heart to x = "% of People With a Bachelor's Degree or Higher", y = "Crude Mortality Rate (per 100,000)") +
    theme_bw()
```

# Education Level vs. Congestive Heart Failure (Bachelor's or Higher)

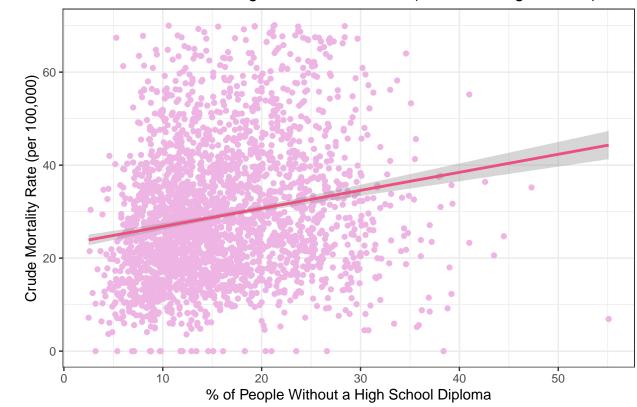


```
cor(bs_higher, rate)
```

```
## [1] -0.3374879
```

```
ggplot(df_new, aes(x = no_hs, y = rate)) + geom_point(color = "#edb4e4") +
    geom_smooth(method = "lm", color = "#eb5081") + labs(title = "Education Level vs. Congestive Heart :
    x = "% of People Without a High School Diploma", y = "Crude Mortality Rate (per 100,000)") +
    theme_bw()
```

# Education Level vs. Congestive Heart Failure (Less than High School)



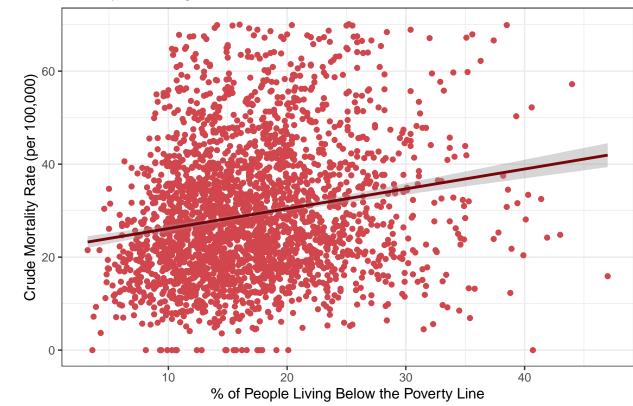
cor(no\_hs, rate)

## [1] 0.1897416

## Crude Mortality Rate vs. Poverty Level

```
ggplot(df_new, aes(x = poverty, y = rate)) + geom_point(color = "#d1454c") +
    geom_smooth(method = "lm", color = "#78050b") + labs(title = "Poverty vs. Congestive Heart Failure"
    x = "% of People Living Below the Poverty Line", y = "Crude Mortality Rate (per 100,000)") +
    theme_bw()
```

# Poverty vs. Congestive Heart Failure



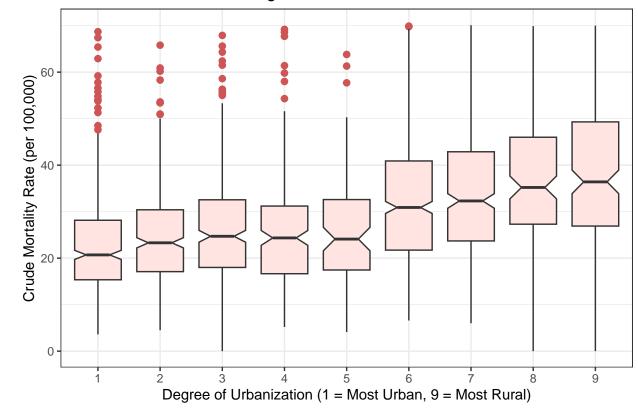
cor(poverty, rate)

## [1] 0.1903578

## Crude Mortality Rate vs. Urbanization Level

```
ggplot(df_new, aes(x = urbanization, y = rate)) + geom_boxplot(fill = "mistyrose",
    notch = TRUE, outlier.colour = "indianred3", outlier.size = 2) +
    labs(title = "Urbanization Level vs. Congestive Heart Failure",
    x = "Degree of Urbanization (1 = Most Urban, 9 = Most Rural)",
    y = "Crude Mortality Rate (per 100,000)") + theme_bw()
```

## Urbanization Level vs. Congestive Heart Failure



## Linear Regression Model

## food\_insecure 3.650e-01 1.001e-01

## bs\_higher
## poverty

```
model <- lm(rate ~ income + food_insecure + bs_higher + poverty +</pre>
    urbanization, data = df_new)
summary(model)
##
## Call:
## lm(formula = rate ~ income + food_insecure + bs_higher + poverty +
##
       urbanization, data = df_new)
##
## Residuals:
       Min
                1Q Median
                                3Q
                                       Max
## -39.557 -8.971 -1.270
                             7.023 43.731
##
## Coefficients:
##
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 5.387e+01 3.384e+00 15.921 < 2e-16 ***
## income
                 -3.961e-04 4.651e-05 -8.516 < 2e-16 ***
```

-1.614e-01 4.007e-02 -4.028 5.78e-05 \*\*\*

-5.536e-01 8.006e-02 -6.916 5.80e-12 \*\*\*

## urbanization2 -2.542e+00 9.523e-01 -2.670 0.00764 \*\*

3.647 0.00027 \*\*\*

```
## urbanization3 -1.368e+00 9.994e-01 -1.369 0.17124
## urbanization4 -2.702e+00 1.154e+00 -2.341 0.01930 *
## urbanization5 -2.146e+00 1.519e+00 -1.413 0.15778
## urbanization6 2.690e+00 9.600e-01 2.802 0.00512 **
## urbanization7 4.157e+00 1.024e+00 4.058 5.09e-05 ***
## urbanization8 6.249e+00 1.324e+00 4.721 2.47e-06 ***
## urbanization9 6.586e+00 1.220e+00 5.397 7.37e-08 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 12.62 on 2661 degrees of freedom
## Multiple R-squared: 0.1985, Adjusted R-squared: 0.1948
## F-statistic: 54.9 on 12 and 2661 DF, p-value: < 2.2e-16</pre>
```

#### Creating Table Displaying OLS Results