# STA 141A Project: Food Insecurity

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### Introduction

# **Data Wrangling**

#### Libraries

library(dplyr) library(tidyr) library(ggplot2) library(hrbrthemes) library(gganimate) library(png) library(gifski) library(ggridges) library(tidyverse) library(tibble) library(mapview) library(sp) library(janitor) library(GGally) library(RColorBrewer) library(MASS) library(knitr) library(tidyverse) library(matlib) library(lubridate) library(pdftools) library(stringr) library(ggmap) library(ggsci) library(patchwork) library(ddpcr) library(caret) library(car) library(paletteer)

#### Import All Data Sets

```
rawdata2010 <- readr::read_csv("FeedAmerica/FeedAmerica2010.csv")</pre>
rawdata2011 <- readr::read_csv("FeedAmerica/FeedAmerica2011.csv")</pre>
rawdata2012 <- readr::read_csv("FeedAmerica/FeedAmerica2012.csv")</pre>
rawdata2013 <- readr::read_csv("FeedAmerica/FeedAmerica2013.csv")</pre>
rawdata2014 <- readr::read_csv("FeedAmerica/FeedAmerica2014.csv")</pre>
rawdata2015 <- readr::read_csv("FeedAmerica/FeedAmerica2015.csv")</pre>
rawdata2016 <- readr::read_csv("FeedAmerica/FeedAmerica2016.csv")</pre>
rawdata2017 <- readr::read_csv("FeedAmerica/FeedAmerica2017.csv")</pre>
rawdata2018 <- readr::read csv("FeedAmerica/FeedAmerica2018.csv")</pre>
rawdata2019 2021 <- readr::read csv("FeedAmerica/FeedAmerica2019 2021.csv")
AgeData2019 <- readr::read_csv("DisabilityData/DisabilityData2019.csv")</pre>
unemploymentdataraw <- readr::read csv("Local Area Unemployment Statistics LAUS .csv")
unemploymentdataraw <- janitor::clean names(unemploymentdataraw)</pre>
incomedataraw <- readr::read_csv("incomedata.csv")</pre>
incomedataraw <- janitor::clean_names(incomedataraw)</pre>
DisData2021 <- readr::read_csv("DisabilityData/DisabilityData2021.csv")</pre>
DisData2020 <- readr::read_csv("DisabilityData/DisabilityData2020.csv")</pre>
DisData2019 <- readr::read_csv("DisabilityData/DisabilityData2019.csv")</pre>
DisData2018 <- readr::read_csv("DisabilityData/DisabilityData2018.csv")</pre>
DisData2017 <- readr::read_csv("DisabilityData/DisabilityData2017.csv")</pre>
DisData2016 <- readr::read_csv("DisabilityData/DisabilityData2016.csv")</pre>
DisData2015 <- readr::read csv("DisabilityData/DisabilityData2015.csv")</pre>
DisData2014 <- readr::read_csv("DisabilityData/DisabilityData2014.csv")</pre>
DisData2013 <- readr::read csv("DisabilityData/DisabilityData2013.csv")</pre>
DisData2012 <- readr::read_csv("DisabilityData/DisabilityData2012.csv")</pre>
DisData2011 <- readr::read csv("DisabilityData/DisabilityData2011.csv")</pre>
DisData2010 <- readr::read_csv("DisabilityData/DisabilityData2010.csv")</pre>
```

### Clean Up Feed America 2019-2020

```
FeedData2019_2021 <- rawdata2019_2021%>%
  filter(State=="CA")

FeedData2019_2021 <- FeedData2019_2021[,c(3:5)]

FeedData2019_2021 <- janitor::clean_names(FeedData2019_2021)

FeedData2019_2021[,3] <- sapply(FeedData2019_2021[,3],function(x) as.numeric(gsub("%","",x)))

FeedData2019_2021 <- FeedData2019_2021[c("year","county_state","overall_food_insecurity_rate")]</pre>
```

#### Clean Up Feed America 2010-2018

```
cleanFeed <- function(data){</pre>
  dataCA <- data%>%
    filter(State=="CA")
  dataCA \leftarrow dataCA[,c(3,4)]
  dataCA[,2] <- sapply(dataCA[,2],function(x) as.numeric(gsub("%","",x)))</pre>
  colnames(dataCA) <- c("county_state","overall_food_insecurity_rate")</pre>
 return (dataCA)
}
FeedData2018 <- cbind("year"=rep(2018,nrow(cleanFeed(rawdata2018))),cleanFeed(rawdata2018))
FeedData2017 <- cbind("year"=rep(2017,nrow(cleanFeed(rawdata2017))),cleanFeed(rawdata2017)))
FeedData2016 <- cbind("year"=rep(2016,nrow(cleanFeed(rawdata2016))),cleanFeed(rawdata2016)))
FeedData2015 <- cbind("year"=rep(2015,nrow(cleanFeed(rawdata2015))),cleanFeed(rawdata2015)))
FeedData2014 <- cbind("year"=rep(2014,nrow(cleanFeed(rawdata2014))),cleanFeed(rawdata2014))
FeedData2013 <- cbind("year"=rep(2013,nrow(cleanFeed(rawdata2013))),cleanFeed(rawdata2013))
FeedData2012 <- cbind("year"=rep(2012,nrow(cleanFeed(rawdata2012))),cleanFeed(rawdata2012))
FeedData2011 <- cbind("year"=rep(2011,nrow(cleanFeed(rawdata2011))),cleanFeed(rawdata2011))
FeedData2010 <- cbind("year"=rep(2010,nrow(cleanFeed(rawdata2010))),cleanFeed(rawdata2010))
```

### Food Insecurity 2010-2021

```
FeedData <- rbind(FeedData2010,FeedData2011,FeedData2012,FeedData2013,FeedData2014,FeedData2015,FeedData
countyNameDisable <- FeedData$county_state

FeedData <- FeedData%>%
    mutate(county_state = gsub(", California", "", county_state))%>%
    rename(county = county_state)%>%
    arrange(county)

countyName <- FeedData$county</pre>
```

### Cleaning Age Data for 2019 Correlation

```
AgeData2019CA <- AgeData2019 %>%
  filter(NAME %in% c(countyNameDisable, 'Geographic Area Name')) %>%
  row_to_names(row_number = 1)

colnames(AgeData2019CA)[3] <- "Total Population"

Age2019 <- AgeData2019CA %>%
  .[,!grepl("Margin of Error", colnames(.))] %>%
```

```
.[,!grepl("Annotation", colnames(.))] %>%
  .[,!grepl("Percent", colnames(.))] %>%
  .[, grepl("Geographic Area Name|Total Population|Population under 18 years|Population 65 years and ov
  .[,!grepl("years!!|over!!", colnames(.))]
Age2019$'Geographic Area Name' <-gsub(", California", "", Age2019$'Geographic Area Name')
colnames(Age2019)[1] <- "county"</pre>
colnames(Age2019) <- gsub(".*\\Estimate!!", "", colnames(Age2019))</pre>
colnames(Age2019) <- gsub("Total civilian noninstitutionalized population!!",</pre>
                                         "",colnames(Age2019))
colnames(Age2019) <-gsub("DISABILITY TYPE BY DETAILED AGE!!", "", colnames(Age2019))</pre>
colnames(Age2019) <- gsub("!!", ": ", colnames(Age2019))</pre>
colnames(Age2019) <- gsub(": :", ":", colnames(Age2019))</pre>
Age2019 <- Age2019 %>% mutate_at(-1, as.numeric)
TailAge2019 <- Age2019[,c(1:4)]</pre>
for (i in c(3:4)) {
  TailAge2019[, i] <- Age2019[, i] / Age2019[, 2]</pre>
colnames(TailAge2019) <- c("county", "total_population", "population_under_18", "population_over_65")</pre>
```

### Clean Up Disability 2013-2021

```
cleanDis1 <- function(data){
  dataCA <- data %>%
    filter(NAME %in% c(countyNameDisable, 'Geographic Area Name')) %>%
    row_to_names(row_number = 1)

colnames(dataCA)[3] <- "Total Population"

dataCA1 <- dataCA %>%
        [,!grepl("Margin of Error", colnames(.))] %>%
        [,!grepl("Annotation", colnames(.))] %>%
        [, grepl("Geographic Area Name|Percent with a disability",colnames(.))] %>%
        [,!grepl("population!",colnames(.))]

dataCA1$'Geographic Area Name' <-gsub(", California", "", dataCA1$'Geographic Area Name')
        colnames(dataCA1)[1] <- "county"
        colnames(dataCA1)[2] <- "Percent with a disability"

dataCA1 <- dataCA1 %>% mutate_at(-1, as.numeric)
    return (dataCA1)
}
```

```
Dis2021 <- cbind("year"=rep(2021,nrow(cleanDis1(DisData2021))),cleanDis1(DisData2021))
Dis2020 <- cbind("year"=rep(2020,nrow(cleanDis1(DisData2020))),cleanDis1(DisData2020))
Dis2019 <- cbind("year"=rep(2019,nrow(cleanDis1(DisData2019))),cleanDis1(DisData2019))
Dis2018 <- cbind("year"=rep(2018,nrow(cleanDis1(DisData2018)[,c(1,2)])),cleanDis1(DisData2018)[,c(1,2)]
Dis2017 <- cbind("year"=rep(2017,nrow(cleanDis1(DisData2017)[,c(1,2)])),cleanDis1(DisData2017)[,c(1,2)]
Dis2016 <- cbind("year"=rep(2016,nrow(cleanDis1(DisData2016)[,c(1,2)])),cleanDis1(DisData2016)[,c(1,2)]
Dis2015 <- cbind("year"=rep(2015,nrow(cleanDis1(DisData2015)[,c(1,2)])),cleanDis1(DisData2015)[,c(1,2)]
Dis2014 <- cbind("year"=rep(2014,nrow(cleanDis1(DisData2014)[,c(1,2)])),cleanDis1(DisData2014)[,c(1,2)]
Dis2013 <- cbind("year"=rep(2013,nrow(cleanDis1(DisData2013)[,c(1,2)])),cleanDis1(DisData2013)[,c(1,2)]
```

#### Clean Up Disability 2010-2012

```
cleanDis2 <- function(data){</pre>
  dataCA <- data %>%
    filter(NAME %in% c(countyNameDisable, 'Geographic Area Name')) %>%
    row_to_names(row_number = 1)
  colnames(dataCA)[3] <- "Total Population"</pre>
  dataCA1 <- dataCA %>%
    .[,!grepl("Margin of Error", colnames(.))] %>%
    .[,!grepl("Annotation", colnames(.))] %>%
    .[, grepl("Geographic Area Name|Percent with a disability",colnames(.))] %%
    .[,!grepl("population!",colnames(.))]
  dataCA1$'Geographic Area Name' <-gsub(", California", "", dataCA1$'Geographic Area Name')
  colnames(dataCA1)[1] <- "county"</pre>
  colnames(dataCA1)[2] <- "Percent with a disability"</pre>
  dataCA1 <- (dataCA1[,c(1,2)] %>% mutate_at(-1, as.numeric))
 return (dataCA1)
}
Dis2012 <- cbind("year"=rep(2012,nrow(cleanDis2(DisData2012))),cleanDis2(DisData2012))
Dis2011 <- cbind("year"=rep(2011,nrow(cleanDis2(DisData2011))),cleanDis2(DisData2011))
Dis2010 <- cbind("year"=rep(2010,nrow(cleanDis2(DisData2010))),cleanDis2(DisData2010))
```

### Total Disability 2010-2021

```
totalDisability <- rbind(Dis2010,Dis2011,Dis2012,Dis2013,Dis2014,Dis2015,Dis2016,Dis2017,Dis2018,Dis201
totalDisability <- totalDisability %>% arrange(county)
colnames(totalDisability) <- c("year", "county", "percent_disabled")</pre>
```

### Clean Up Unemployment 2010-2021

```
UnemploymentData <- unemploymentdataraw%>%
    filter(area_type=="County", status_preliminary_final=="Final")%>%
    filter(year>=2010 & year<2022)%>%
    filter(!area_name %in% c("Non Residential County", "Resident Out of State County", "Unallocated County"
    group_by(year, area_name)%>%
    summarise("unemployment_rate_avg"=mean(unemployment_rate))%>%
    distinct(.)%>%
    ungroup() %>%
    rename("county"="area_name")
```

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

#### Clean Up Income 2010-2020

```
IncomeData <- incomedataraw%>%
  filter(taxable_year >= 2010 & taxable_year <= 2021)%>%
  filter(!county %in% c("Nonresident", "Resident Out of State County", "Unallocated", "Resident Out of State rename("year"="taxable_year")%>%
  mutate("county"=paste(.$county, "County"))%>%
  arrange(year,county)
IncomeData <- IncomeData[,c(1,2,6)]
```

#### Final CSV

```
majorDF <- FeedData %>% full_join(totalDisability)
majorDF <- majorDF %>% full_join(UnemploymentData)
majorDF <- majorDF %>% full_join(IncomeData)

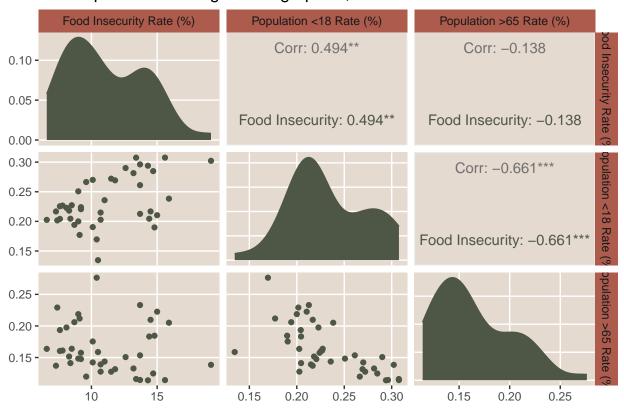
majorDF <- majorDF %>% filter(!county %in% c("Resident Out of State County", "Nonresident20 County", "Res
majorDF2019 <- majorDF %>% filter(year==2019)
majorDF2019 <- TailAge2019 %>% inner_join(majorDF2019)
```

# Analysis

### Data Visualization

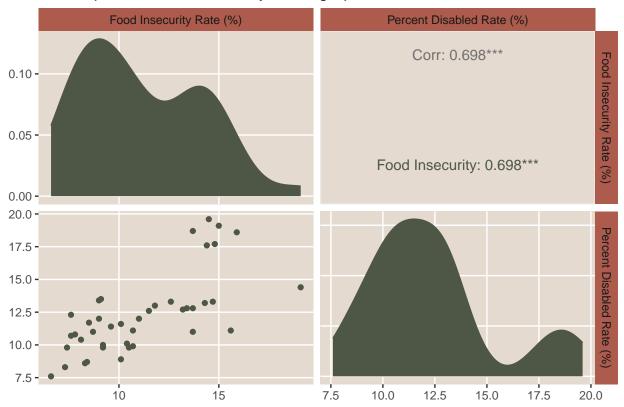
### Age Demographics 2019 GGpair

## Scatterplot Matrix of Age Demographics, 2019



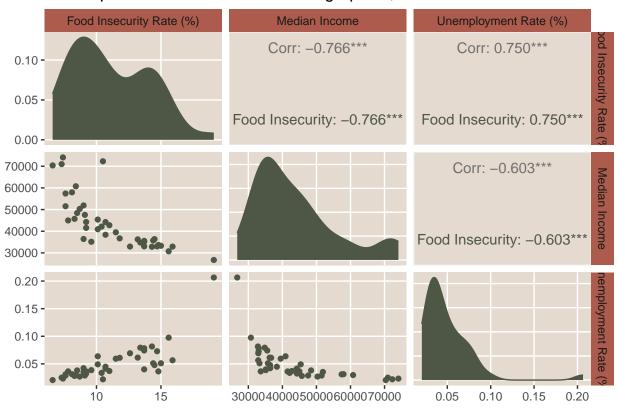
#### Disability Demographics 2019 GGpair

# Scatterplot Matrix of Disability Demographics, 2019



### Income Demographics 2019 GGpair

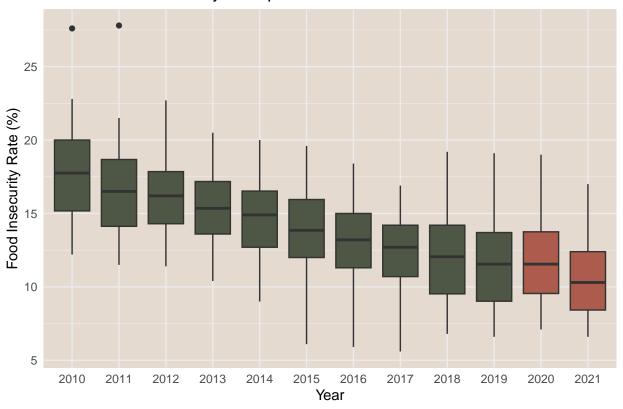
# Scatterplot Matrix of Financial Demographics, 2019



### Food Insecurity Boxplot

```
majorDF[,c(1,3)] %>%
mutate(year=as.factor(year))%>%
ggplot(aes(x=year, y=overall_food_insecurity_rate, fill=(year==c(2020,2021)))) +
geom_boxplot(show.legend = F)+
scale_fill_manual(values=c("#4E5745","#AC5B4C"))+
theme_minimal()+
theme(panel.background=element_rect(fill="#E5DACF",color="#E5DACF",size=0.5,linetype="solid"))+
labs(title="Overall Food Insecurity Rate per Year",x="Year",y="Food Insecurity Rate (%)")
```

# Overall Food Insecurity Rate per Year

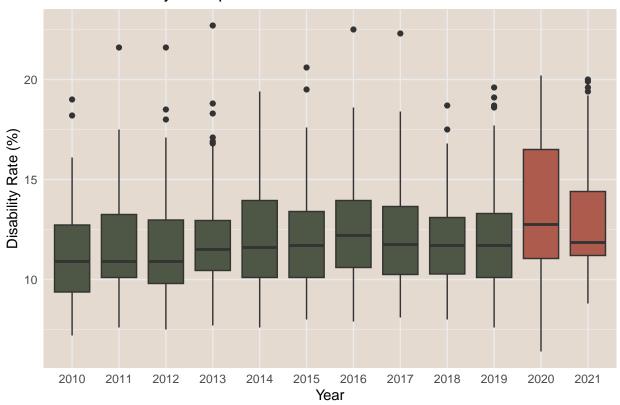


### Disability Boxplot

```
majorDF[,c(1,4)] %>%
  mutate(year=as.factor(year))%>%
  ggplot(aes(x=year, y=percent_disabled, fill=(year==2020|year==2021))) +
  geom_boxplot(show.legend = F)+
  scale_fill_manual(values=c("#4E5745","#AC5B4C"))+
  theme_minimal()+
  theme(panel.background=element_rect(fill="#E5DACF",color="#E5DACF",size=0.5,linetype="solid"))+
  labs(title="Overall Disability Rate per Year",x="Year",y="Disability Rate (%)")
```

## Warning: Removed 195 rows containing non-finite values ('stat\_boxplot()').

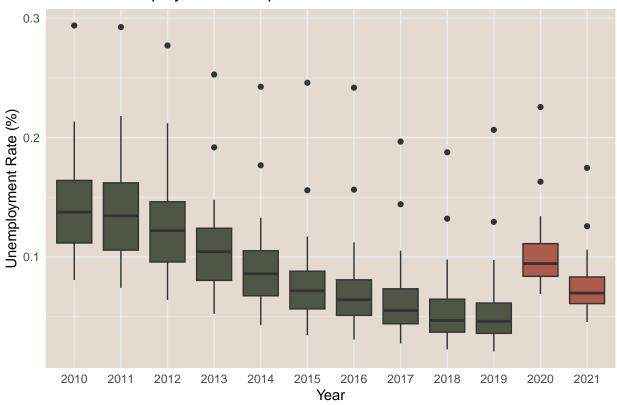
# Overall Disability Rate per Year



### Unemployment Boxplot

```
majorDF[,c(1,5)] %>%
mutate(year=as.factor(year))%>%
ggplot(aes(x=year, y=unemployment_rate_avg, fill=(year==2020 | year==2021))) +
geom_boxplot(show.legend = F)+
scale_fill_manual(values=c("#4E5745","#AC5B4C"))+
theme_minimal()+
theme(panel.background=element_rect(fill="#E5DACF",color="#E5DACF",size=0.5,linetype="solid"))+
labs(title="Overall Unemployment Rate per Year",x="Year",y="Unemployment Rate (%)")
```

# Overall Unemployment Rate per Year

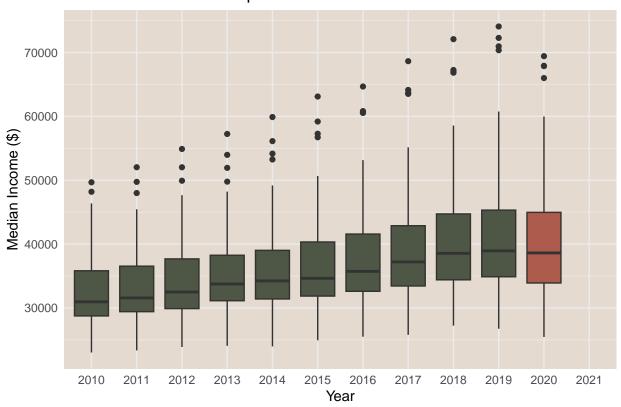


### **Income Boxplot**

```
majorDF[,c(1,6)]%>%
  mutate(year=as.factor(year))%>%
  ggplot(aes(x=year, y=median_income, fill=(year==2020))) +
  geom_boxplot(show.legend = F)+
  scale_fill_manual(values=c("#4E5745","#AC5B4C"))+
  theme_minimal()+
  theme(panel.background=element_rect(fill="#E5DACF",color="#E5DACF",size=0.5,linetype="solid"))+
  labs(title="Overall Median Income per Year",x="Year",y="Median Income ($)")
```

## Warning: Removed 58 rows containing non-finite values ('stat\_boxplot()').

## Overall Median Income per Year



### 2019 Map Set-up

```
shape <- sf::read_sf(dsn = "CA_Counties_ShapeFile", layer = "CA_Counties_TIGER2016")
counties <- shape['NAME'] %>% arrange(NAME)
colnames(counties)[1] <- "county"

majorDF2019$county <- gsub(" County","",majorDF2019$county)
counties <- counties %>% full_join(majorDF2019)
```

### 2019 Map Execution

```
map <- mapview(
   counties,
   zcol = "overall_food_insecurity_rate",
   layer.name = "Food Insecurity Rate (%)",
   map.types = "CartoDB.Positron",
   na.color = "#AC5B4C",
   col.regions = brewer.pal(100, "Blues"),
   alpha.regions = 1
) +
   mapview(
   counties,
   zcol = "unemployment_rate_avg",</pre>
```

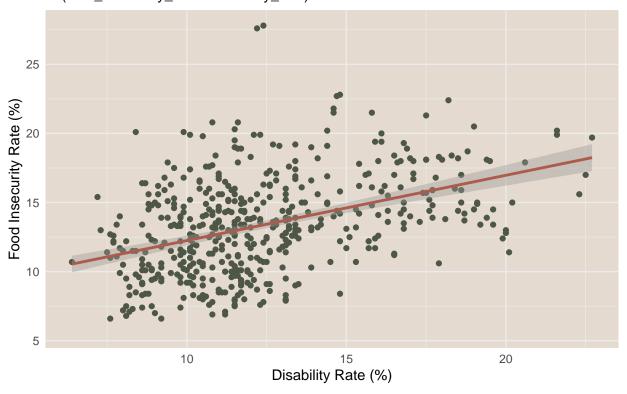
```
layer.name = "Unemployment Rate (%)",
  map.types = "CartoDB.Positron",
  na.color = "#AC5B4C",
  col.regions = brewer.pal(100, "Blues"),
  alpha.regions = 1
) +
mapview(
  counties,
  zcol = "median_income",
  layer.name = "Median Income",
  map.types = "CartoDB.Positron",
 na.color = "#AC5B4C",
  col.regions = brewer.pal(100, "Blues"),
  alpha.regions = 1
) +
mapview(
  counties,
  zcol = "percent_disabled",
  layer.name = "Disability Rate (%)",
  map.types = "CartoDB.Positron",
 na.color = "#AC5B4C",
 col.regions = brewer.pal(100, "Blues"),
  alpha.regions = 1
```

### Modeling

Food Insecurity vs. Disability Linear Model

## Relationship Between FI and Disability Rate

Im(food\_insecurity\_rate ~ disability\_rate)

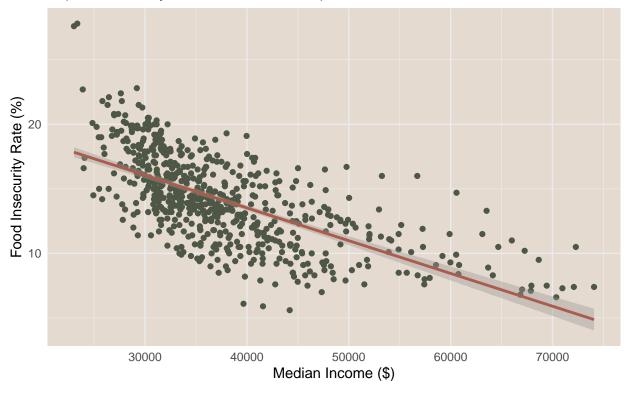


### Food Insecurity vs. Median Income Linear Model

## Warning: Removed 58 rows containing missing values ('geom\_point()').

## Relationship Between FI and Median Income

Im(food\_insecurity\_rate ~ median\_income)

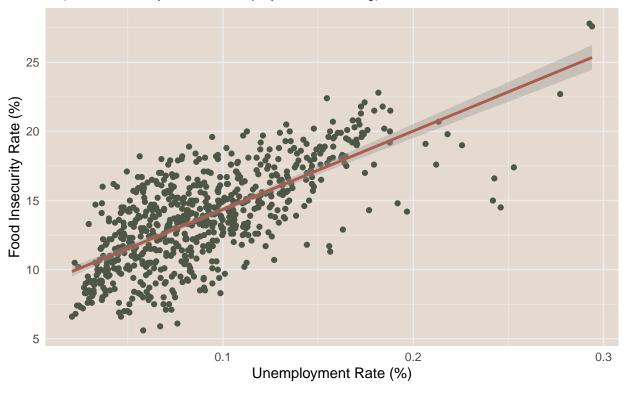


### Food Insecurity vs. Unemployment Linear Model

## 'geom\_smooth()' using formula = 'y ~ x'

# Relationship Between FI and Unemployment Rate

Im(food\_insecurity\_rate ~ unemployment\_rate\_avg)



#### **AIC**

```
none_mod <- lm(overall_food_insecurity_rate~1,data=na.omit(majorDF)) ##model with only intercept
full_mod <- lm(overall_food_insecurity_rate~unemployment_rate_avg+median_income+percent_disabled,data=n
stepAIC(none_mod, scope=list(upper=full_mod, lower = ~1), direction="both", k=2)</pre>
```

```
## Start: AIC=1137.21
## overall_food_insecurity_rate ~ 1
##
##
                           Df Sum of Sq
                                           RSS
                                                   AIC
## + unemployment_rate_avg 1
                                 2881.1 2562.9 793.40
## + median_income
                                 2393.8 3050.2 873.31
                            1
## + percent_disabled
                           1
                                 1047.3 4396.7 1041.13
## <none>
                                        5444.0 1137.21
##
## Step: AIC=793.4
## overall_food_insecurity_rate ~ unemployment_rate_avg
##
##
                           Df Sum of Sq
                                           RSS
                                                   AIC
                                 511.02 2051.8 693.32
## + percent_disabled
                            1
## + median_income
                                 355.04 2207.8 726.96
## <none>
                                        2562.9 793.40
## - unemployment_rate_avg 1
                                2881.14 5444.0 1137.21
```

```
##
## Step: AIC=693.32
## overall_food_insecurity_rate ~ unemployment_rate_avg + percent_disabled
##
                                                   AIC
                           Df Sum of Sq
                                           RSS
## + median income
                                  96.57 1955.3 673.20
## <none>
                                        2051.8 693.32
                                 511.02 2562.9 793.40
## - percent_disabled
                           1
## - unemployment_rate_avg 1
                                2344.83 4396.7 1041.13
##
## Step: AIC=673.2
## overall_food_insecurity_rate ~ unemployment_rate_avg + percent_disabled +
       median_income
##
##
                           Df Sum of Sq
                                           RSS
                                                  AIC
## <none>
                                        1955.3 673.20
## - median_income
                                  96.57 2051.8 693.32
                            1
## - percent_disabled
                            1
                                 252.55 2207.8 726.96
## - unemployment_rate_avg 1
                                 967.15 2922.4 855.66
##
## Call:
## lm(formula = overall_food_insecurity_rate ~ unemployment_rate_avg +
       percent_disabled + median_income, data = na.omit(majorDF))
##
##
## Coefficients:
##
             (Intercept)
                          unemployment_rate_avg
                                                      percent disabled
               9.159e+00
                                      4.261e+01
                                                              2.716e-01
##
##
           median income
              -6.762e-05
##
```

[Overall against 2020] Linear Model: Food Insecurity  ${\scriptstyle \sim}$  Unemployment Rate + Disability Rate + Median Income

```
# Define Training Data (2010-2019) and Test Data (2020)
majorDFtrain <- na.omit(majorDF)%>%
    filter(year<2020)

majorDFtest <- na.omit(majorDF)%>%
    filter(year==2020)

model <- lm(overall_food_insecurity_rate ~ unemployment_rate_avg + percent_disabled + median_income, da

lm_results <- data.frame(
    "county" = majorDFtest$county,
    "predicted_food_insecurity_rate" = as.numeric(model$coefficients[1]) + as.numeric(model$coefficients[
    majorDFtest$unemployment_rate_avg + as.numeric(model$coefficients[3]) *
    majorDFtest$percent_disabled + as.numeric(model$coefficients[4]) * majorDFtest$median_income,
    "actual_food_insecurity_rate" = majorDFtest$voverall_food_insecurity_rate
)

lm_overall_acc20 <- 1 - mean((abs(lm_results$predicted_food_insecurity_rate-lm_results$actual_food_insecurity_rate)</pre>
```

[Overall against 2019] Linear Model: Food Insecurity  $\sim$  Unemployment Rate + Disability Rate + Median Income

[Overall against 2020] Random Forest Model: Food Insecurity  $\sim$  Unemployment Rate + Disability Rate + Median Income

[Overall against 2019] Random Forest Model: Food Insecurity  ${\scriptstyle\sim}$  Unemployment Rate + Disability Rate + Median Income

```
# Define Training Data (2010-2019) and Test Data (2020)
majorDFtrain <- na.omit(majorDF)%>%
```

[By county against 2019,2020] Linear Model: Food Insecurity  $\sim$  Unemployment Rate + Disability Rate + Median Income

```
lm_trainr <- function(c,y) {</pre>
  # Define Training Data (2010-2019) and Test Data (2020)
  majorDFtrain <- na.omit(majorDF) %>%
   filter(year < y)%>%
   filter(county==c)
  majorDFtest <- na.omit(majorDF) %>%
   filter(year < y)%>%
   filter(county==c)
  if (nrow(majorDFtrain) == 0 | nrow(majorDFtest) == 0){
   rf_results <- data.frame("county"=c,
           "predicted_food_insecurity_rate"=NA,
           "actual_food_insecurity_rate"=NA)
   rf_acc <- NA
   return(list(results=rf_results, accuracy=rf_acc))
  }
  model <-
      overall_food_insecurity_rate ~ unemployment_rate_avg + percent_disabled + median_income,
      data = majorDFtrain
   )
  lm_results <- data.frame(</pre>
    "county" = majorDFtrain$county,
    "predicted_food_insecurity_rate" = as.numeric(model$coefficients[1]) + as.numeric(model$coefficient
      majorDFtrain$unemployment_rate_avg + as.numeric(model$coefficients[3]) *
```

```
majorDFtrain$percent_disabled + as.numeric(model$coefficients[4]) * majorDFtrain$median_income,
    "actual_food_insecurity_rate" = majorDFtrain$overall_food_insecurity_rate
  lm_acc <-</pre>
    1 - mean(
      abs(
        lm_results$predicted_food_insecurity_rate - lm_results$actual_food_insecurity_rate
      ) / lm_results$actual_food_insecurity_rate
  return(list(results = lm_results, accuracy = lm_acc))
results20 <- data.frame(county=c(), predicted_food_insecurity_rate=c(), actual_food_insecurity=c())
accuracy20 <- c()</pre>
for (c in unique(majorDF$county)){
  trainr_data <- lm_trainr(c,2020)</pre>
  #results20 <- rbind(results20, trainr_data[[1]])</pre>
  accuracy20 <- c(accuracy20, trainr_data[[2]])</pre>
#results20$accuracy20 <- accuracy20</pre>
lm_county_acc20 <- mean(accuracy20, na.rm=T)</pre>
results19 <- data.frame(county=c(), predicted_food_insecurity_rate=c(), actual_food_insecurity=c())
accuracy19 <- c()
for (c in unique(majorDF$county)){
  trainr_data <- lm_trainr(c,2019)</pre>
  results19 <- rbind(results19, trainr_data[[1]])</pre>
  accuracy19 <- c(accuracy19, trainr_data[[2]])</pre>
}
#results19$accuracy19 <- accuracy19</pre>
lm_county_acc19 <- mean(accuracy19, na.rm=T)</pre>
```

[By county against 2019,2020] Linear Model: Food Insecurity  ${\scriptstyle\sim}$  Unemployment Rate + Disability Rate + Median Income

```
rf_trainr <- function(c,y) {
    # Define Training Data (2010-2019) and Test Data (2020)
majorDFtrain <- na.omit(majorDF) %>%
    filter(year < y)%>%
    filter(county==c)

majorDFtest <- na.omit(majorDF) %>%
    filter(year < y)%>%
    filter(county==c)
```

```
if (nrow(majorDFtrain) <= 1 | nrow(majorDFtest) == 0){</pre>
    rf_results <- data.frame("county"=c,</pre>
            "predicted_food_insecurity_rate"=NA,
            "actual_food_insecurity_rate"=NA)
    rf_acc <- NA
    return(list(results=rf_results, accuracy=rf_acc))
  train_rf <-
    train(
      overall_food_insecurity_rate ~ unemployment_rate_avg + percent_disabled + median_income,
      method = "rf",
      data = majorDFtrain,
      tuneGrid = data.frame(mtry = 1:2),
      trControl = trainControl(method = "cv", number = 5)
    )
  pred <- as.numeric(predict(train_rf, newdata=majorDFtest))</pre>
  rf_results <- data.frame("county"=majorDFtest$county,</pre>
                              "predicted_food_insecurity_rate"= pred,
                              "actual food insecurity rate" = majorDFtest$overall food insecurity rate)
  rf acc <-
    1 - mean((
      rf_results$predicted_food_insecurity_rate - rf_results$actual_food_insecurity_rate
    ) / rf_results$actual_food_insecurity_rate
  return(list(results=rf_results, accuracy=rf_acc))
results20 <- data.frame(county=c(), predicted_food_insecurity_rate=c(), actual_food_insecurity=c())</pre>
accuracy20 <- c()</pre>
for (c in unique(majorDF$county)){
 trainr_data <- rf_trainr(c,2020)</pre>
  #results20 <- rbind(results20, trainr_data[[1]])</pre>
  accuracy20 <- c(accuracy20, trainr_data[[2]])</pre>
}
#results20$accuracy20 <- accuracy20</pre>
rf_county_acc20 <- mean(accuracy20, na.rm=T)</pre>
results19 <- data.frame(county=c(), predicted_food_insecurity_rate=c(), actual_food_insecurity=c())</pre>
accuracy19 <- c()</pre>
for (c in unique(majorDF$county)){
  trainr_data <- rf_trainr(c,2019)</pre>
  results19 <- rbind(results19, trainr_data[[1]])</pre>
```

```
accuracy19 <- c(accuracy19, trainr_data[[2]])
}
#results19$accuracy19 <- accuracy19
rf_county_acc19 <- mean(accuracy19, na.rm=T)</pre>
```

### **Predictive Model Results**

Model.Type	Test.Accuracy.2019	Test.Accuracy.2020
Overall Linear Regression	0.8671502	0.6799830
Overall Random Forest	0.9332404	0.6671922
By County Linear Model	0.9758715	0.9738838
By County Random Forest	0.9948065	0.9953549

# Conclusion