```
#libraries
library(dplyr)
library(tidyr)
library(caret)
# read CSV file
weather data <- read.csv("NOAAGISSWeatherDisasters.csv")</pre>
weather data <- read.csv("NOAAGISSWeatherDisasters.csv", header = TRUE, sep = ",")
names(weather data)
weather data$Year <- as.numeric(as.character(weather data$Year))
weather data$delta.tempYear <- with(weather data, delta.temp * Year)
# filter the data for drought
drought data <- weather data %>%
 filter(Drought.Count > 0) %>%
select(delta.temp, Year, delta.tempYear)
# filter the data for wildfire
wildfire data <- weather data %>%
 filter(Wildfire.Count > 0) %>%
select(delta.temp, Year, delta.tempYear)
# step 3 for drought data
# fit a binomial logit model with drought as the response and delta.temp, year, and #delta.temp*year as
predictors
drought model interaction <- glm(Drought.Count ~ delta.temp * Year, data = weather data, family =
"binomial")
# fit a second model without the interaction term
drought model additive <- glm(Drought.Count ~ delta.temp + Year, data = weather data, family =
"binomial")
# personal notes:
# An AIC score is a number used to determine which machine learning model
# is best for a given data set in situations where one can't easily test a dataset
# The lower the AIC score the better.
# compare the AIC values for the two models
AIC drought interaction <- AIC(drought model interaction)
AIC drought additive <- AIC(drought model additive)
```

```
cat("AIC for drought model with interaction:", AIC drought interaction, "\n")
#output: AIC for drought model with interaction: 54.52863
cat("AIC for drought model without interaction:", AIC drought additive, "\n")
#output: AIC for drought model without interaction: 52.99482
# compare deviance differences between the two models using ANOVA
drought anova <- anova(drought model additive, drought model interaction, test = "Chisq")
cat("Deviance difference for drought models:", drought anova$Deviance[2] -
drought anova$Deviance[1], "\n")
#output: Deviance difference for drought models: NA
cat("Deviance p-value for drought models:", drought anova$`Pr(>Chi)`[2], "\n")
#output: Deviance p-value for drought models: 0.4947468
# choose the best model based on AIC and deviance comparison
if (AIC drought interaction < AIC drought additive && drought anova$`Pr(>Chi)`[2] < 0.05) {
 drought best model <- drought model interaction
} else {
 drought best model <- drought model additive
# interpret coefficients of the best model
cat("Coefficients for drought model:", coef(drought best model), "\n")
#output: Coefficients for drought model: -87.14359 1.498534 0.04364329
# step 3 for wildfire
# fit a binomial logit model with wildfire as the response
# delta.temp, year, and delta.temp*year as predictors
wildfire model interaction <- glm(Wildfire.Count ~ delta.temp * Year, data = weather data, family =
"binomial")
# xit a second model without the interaction term
wildfire model additive <- glm(Wildfire.Count ~ delta.temp + Year, data = weather data, family =
"binomial")
# compare AIC values for two models
AIC wildfire interaction <- AIC(wildfire model interaction)
AIC wildfire additive <- AIC(wildfire model additive)
cat("AIC for wildfire model with interaction:", AIC wildfire interaction, "\n")
#output: AIC for wildfire model with interaction: 51.71748
cat("AIC for wildfire model without interaction:", AIC wildfire additive, "\n")
```

#output: AIC for wildfire model without interaction: 49.74829

```
# compare deviance differences between the two models using ANOVA
wildfire_anova <- anova(wildfire_model_additive, wildfire_model_interaction, test = "Chisq")

cat("Deviance difference for wildfire models:", wildfire_anova$Deviance[2] -
wildfire_anova$Deviance[1], "\n")

#output: Deviance difference for wildfire models: NA

cat("Deviance p-value for wildfire models:", wildfire_anova$`Pr(>Chi)`[2], "\n")

#output: Deviance p-value for wildfire models: 0.8606658

# choose the best model based on AIC and deviance comparison
if (AIC_wildfire_interaction < AIC_wildfire_additive && wildfire_anova$`Pr(>Chi)`[2] < 0.05) {
    wildfire_best_model <- wildfire_model_interaction
} else {
    wildfire_best_model <- wildfire_model_additive
}

# interpret coefficients of the best model
cat("Coefficients for wildfire model:", coef(wildfire_best_model), "\n")

#output: Coefficients for wildfire model: -212.8037 0.8126896 0.1061008
```

Code with Deviance Difference

```
library(dplyr)
library(lmtest)

# Read CSV file
weather_data <- read.csv("NOAAGISSWeatherDisasters.csv", header = TRUE, sep = ",")

# Convert Year to numeric
weather_data$Year <- as.numeric(as.character(weather_data$Year))
weather_data$delta.tempYear <- with(weather_data, delta.temp * Year)

# Filter the data for drought
drought_data <- weather_data %>%
filter(Drought.Count > 0) %>%
select(delta.temp, Year, delta.tempYear)

# Filter the data for wildfire
wildfire_data <- weather_data %>%
filter(Wildfire.Count > 0) %>%
```

```
select(delta.temp, Year, delta.tempYear)
# Fit drought models
drought model interaction <- glm(Drought.Count ~ delta.temp * Year, data = weather data, family =
"binomial")
drought model additive <- glm(Drought.Count ~ delta.temp + Year, data = weather data, family =
"binomial")
# Compare AIC values for drought models
AIC drought interaction <- AIC(drought model interaction)
AIC drought additive <- AIC(drought model additive)
cat("AIC for drought model with interaction:", AIC drought interaction, "\n")
cat("AIC for drought model without interaction:", AIC drought additive, "\n")
# Compare p-values for deviance difference using lrtest
deviance test drought <- lrtest(drought model additive, drought model interaction)
deviance p value drought <- deviance test drought$Pr[2]
cat("Deviance p-value for drought models:", deviance p value drought, "\n")
# Choose the best drought model based on AIC and deviance comparison
if (AIC drought interaction < AIC drought additive && deviance p value drought < 0.05) {
drought best model <- drought model interaction
} else {
drought best model <- drought model additive
# Fit wildfire models
wildfire model interaction <- glm(Wildfire.Count ~ delta.temp * Year, data = weather data, family =
"binomial")
wildfire model additive <- glm(Wildfire.Count ~ delta.temp + Year, data = weather data, family =
"binomial")
# Compare AIC values for wildfire models
AIC wildfire interaction <- AIC(wildfire model interaction)
AIC wildfire additive <- AIC(wildfire model additive)
cat("AIC for wildfire model with interaction:", AIC wildfire interaction, "\n")
cat("AIC for wildfire model without interaction:", AIC wildfire additive, "\n")
# Compare p-values for deviance difference using lrtest
deviance test wildfire <- lrtest(wildfire model additive, wildfire model interaction)
deviance p value wildfire <- deviance test wildfire $\Pr[2]
cat("Deviance p-value for wildfire models:", deviance p value wildfire, "\n")
```

Choose the best wildfire model based on AIC and deviance comparison

```
if (AIC wildfire interaction < AIC wildfire additive && deviance p value wildfire < 0.05) {
 wildfire best model <- wildfire model interaction
} else {
 wildfire best model <- wildfire model additive
# Interpret coefficients of the best models
cat("Coefficients for best drought model:", coef(drought best model), "\n")
cat("Coefficients for best wildfire model:", coef(wildfire best model), "\n")
# Graphs
plot drought \leftarrow ggplot(drought data, aes(x = Year, y = delta.temp)) +
 geom boxplot() +
 labs(title = "Drought Data",
    x = "Year",
    y = "Delta Temperature") +
 theme minimal()
print(plot drought)
plot wildfire \leq- ggplot(wildfire data, aes(x = "", y = delta.temp)) +
 geom boxplot() +
 labs(title = "Wildfire Data",
    x = "",
    y = "Delta Temperature") +
 theme minimal()
print(plot wildfire)
plot drought \leftarrow ggplot(drought data, aes(x = Year, y = delta.temp)) +
 geom point() +
labs(title = "Drought Data",
    x = "Year",
    y = "Delta Temperature") +
 theme minimal()
print(plot drought)
plot wildfire <- ggplot(wildfire_data, aes(x = "", y = delta.temp)) +
 geom point() +
 labs(title = "Wildfire Data",
    x = "",
    y = "Delta Temperature") +
 theme minimal()
```

print(plot_wildfire)