

QBS 181 Final Project Aim 2 (Say OK)

Say OK

2024-10-28

Load Libraries

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
library(ggplot2)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v forcats   1.0.0      v stringr   1.5.1
## v lubridate 1.9.3      v tibble   3.2.1
## v purrr     1.0.2      v tidyr    1.3.1
## v readr     2.1.5
```

```
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(tidyr)
library(readxl)
library(ppcor)
```

```
## Loading required package: MASS
##
## Attaching package: 'MASS'
##
## The following object is masked from 'package:dplyr':
##
##   select
```

```
library(lme4)
```

```
## Warning: package 'lme4' was built under R version 4.3.3
```

```
## Loading required package: Matrix
```

```
##
```

```
## Attaching package: 'Matrix'
```

```
##
```

```
## The following objects are masked from 'package:tidyr':
```

```
##
```

```
##      expand, pack, unpack
```

```
library(knitr)
```

```
## Warning: package 'knitr' was built under R version 4.3.3
```

I. Load and Merge Datasets

1. Load the cleaned agricultural dataset (from Excel worksheet)

Please refer to the excel file for details on data cleaning, calculations, and exploratory steps. We created a sheet named “R”, which includes all columns needed for input into R.

```
# Set working directory ("Session" --> "Set Working Directory")
```

```
# Input the sheet named "R" in the Excel file
```

```
ag_data <- read_excel("/Users/violet/Documents/QBS Fall 2024/QBS 181/Final Project/Data/Ag_Data_Clean.x
```

```
head(ag_data)
```

```
## # A tibble: 6 x 15
```

```
##   State_Fips_Code State_Name Geographic_Region Geographic_Division Total_Acres
```

```
##           <dbl> <chr>           <chr>           <chr>           <dbl>
```

```
## 1             1 ALABAMA        South            East South Central    8629101
```

```
## 2             2 ALASKA         West             Pacific              869852
```

```
## 3             4 ARIZONA        West             Mountain            25525087
```

```
## 4             5 ARKANSAS       South            West South Central    13722525
```

```
## 5             6 CALIFORNIA     West             Pacific              24190604
```

```
## 6             8 COLORADO       West             Mountain            30213899
```

```
## # i 10 more variables: Total_Chemical_Expenditure <dbl>,
```

```
## #   Chemical_Expenditure_per_Acres <dbl>,
```

```
## #   Insecticide_No_Nema_Treatment_Acres <dbl>,
```

```
## #   'Insecticide_No_Nema_Treatment_Percentage_('%)' <dbl>,
```

```
## #   Insecticide_Nema_Treatment_Acres <dbl>,
```

```
## #   'Insecticide_Nema_Treatment_Percentage_('%)' <dbl>,
```

```
## #   Herbicide_Treatment_Acres <dbl>, ...
```

```
dim(ag_data)
```

```
## [1] 50 15
```

2. Load 2022 Parkinson mortality rate data (downloaded from CDC)

Longitudinal data on Parkinson's disease mortality rate over years can be downloaded directly from CDC. All rates are age-adjusted mortality rates and represent the number of deaths per 100,000 total population.

```
# Set working directory ("Session" --> "Set Working Directory")
# Input the sheet named "R" in the Excel file
mortality <- read.csv("/Users/violet/Documents/QBS Fall 2024/QBS 181/Final Project/Data/Parkinson_Morta
# Remove the last column (unnecessary URL)
mortality <- mortality[, -ncol(mortality)]

head(mortality)
```

```
##   YEAR STATE RATE DEATHS
## 1 2022    AL 10.7    696
## 2 2022    AK 10.9     62
## 3 2022    AZ  9.9  1,007
## 4 2022    AR  9.8    382
## 5 2022    CA  9.4  4,289
## 6 2022    CO 10.7    665
```

```
dim(mortality)
```

```
## [1] 501  4
```

Filter to keep 2022

```
# Filter to keep year == 2022
mortality <- mortality %>%
  filter(YEAR == 2022)

dim(mortality)
```

```
## [1] 51  4
```

3. Combine agricultural dataset with 2022 Parkinson mortality rate data

Create a conversion table as state is presented as full name in ag_data whereas it is presented as abbreviation in the mortality.

```
# Create the conversion table as a data frame
state_abbreviations <- data.frame(
  State = toupper(c("Alabama", "Alaska", "Arizona", "Arkansas", "California",
    "Colorado", "Connecticut", "Delaware", "Florida", "Georgia",
    "Hawaii", "Idaho", "Illinois", "Indiana", "Iowa", "Kansas",
    "Kentucky", "Louisiana", "Maine", "Maryland",
    "Massachusetts", "Michigan", "Minnesota", "Mississippi",
    "Missouri", "Montana", "Nebraska", "Nevada",
    "New Hampshire", "New Jersey", "New Mexico", "New York",
    "North Carolina", "North Dakota", "Ohio", "Oklahoma",
    "Oregon", "Pennsylvania", "Rhode Island", "South Carolina",
```

```

        "South Dakota", "Tennessee", "Texas", "Utah", "Vermont",
        "Virginia", "Washington", "West Virginia", "Wisconsin",
        "Wyoming")),
State_Abbrev = c("AL", "AK", "AZ", "AR", "CA", "CO", "CT", "DE", "FL", "GA",
        "HI", "ID", "IL", "IN", "IA", "KS", "KY", "LA", "ME", "MD",
        "MA", "MI", "MN", "MS", "MO", "MT", "NE", "NV", "NH", "NJ",
        "NM", "NY", "NC", "ND", "OH", "OK", "OR", "PA", "RI", "SC",
        "SD", "TN", "TX", "UT", "VT", "VA", "WA", "WV", "WI", "WY")
)
head(state_abbreviations)

```

```

##      State State_Abbrev
## 1    ALABAMA          AL
## 2    ALASKA          AK
## 3    ARIZONA          AZ
## 4    ARKANSAS         AR
## 5 CALIFORNIA          CA
## 6    COLORADO         CO

```

Merge the datasets

```

ag_data <- merge(ag_data, state_abbreviations, by.x = "State_Name", by.y = "State", all.x = TRUE)
df <- merge(ag_data, mortality, by.x = "State_Abbrev", by.y = "STATE", all.x = TRUE)

head(df)

```

```

##   State_Abbrev State_Name State_Fips_Code Geographic_Region Geographic_Division
## 1           AK    ALASKA              2             West           Pacific
## 2           AL    ALABAMA              1             South East South Central
## 3           AR  ARKANSAS              5             South West South Central
## 4           AZ  ARIZONA              4             West           Mountain
## 5           CA CALIFORNIA              6             West           Pacific
## 6           CO  COLORADO              8             West           Mountain
##   Total_Acres Total_Chemical_Expenditure Chemical_Expenditure_per_Acres
## 1      869852              646000              0.7426551
## 2      8629101             188889000             21.8897658
## 3     13722525             761692000             55.5066943
## 4     25525087             198303000              7.7689451
## 5     24190604             2806374000            116.0109107
## 6     30213899             229485000              7.5953454
##   Insecticide_No_Nema_Treatment_Acres
## 1                      587
## 2                     852151
## 3                    2973520
## 4                     440869
## 5                    4811723
## 6                     864811
##   Insecticide_No_Nema_Treatment_Percentage_(%) Insecticide_Nema_Treatment_Acres
## 1                      0.06748274                      84
## 2                      9.87531610                     358745
## 3                     21.66889840                     546309
## 4                      1.72719881                     115041
## 5                     19.89087581                     1316284

```

```
## 6                2.86229526                111125
## Insecticide_Nema_Treatment_Percentage_(%) Herbicide_Treatment_Acres
## 1                0.009656815                10424
## 2                4.157385572                1967762
## 3                3.981111348                6395120
## 4                0.450697778                622607
## 5                5.441302747                5554203
## 6                0.367794306                4647801
## Herbicide_Treatment_Percentage_(%) Fungicide_Treatment_Acres
## 1                1.198365                168
## 2                22.803789                289571
## 3                46.603085                1671797
## 4                2.439196                96524
## 5                22.960167                2565494
## 6                15.382990                180797
## Fungicide_Treatment_Percentage_(%) YEAR RATE DEATHS
## 1                0.01931363 2022 10.9    62
## 2                3.35574934 2022 10.7   696
## 3                12.18286722 2022  9.8   382
## 4                0.37815346 2022  9.9  1,007
## 5                10.60533255 2022  9.4  4,289
## 6                0.59839016 2022 10.7   665
```

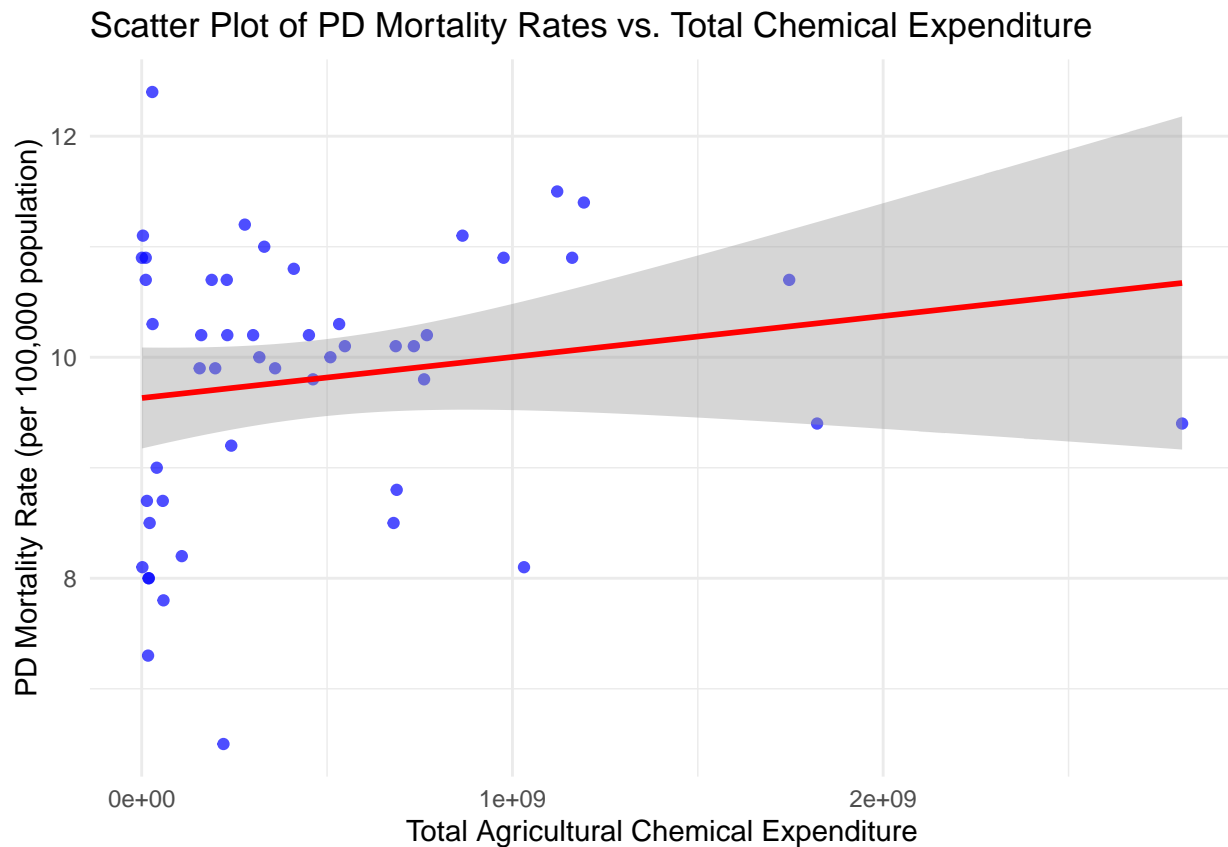
II. Aim 2a: Investigate the relationship between PD mortality rates and total agricultural chemical expenditure per state.

1. PD mortality rates vs. Total agricultural chemical expenditure per state

a. scatter plot visualization

```
ggplot(df, aes(x = Total_Chemical_Expenditure, y = RATE)) +
  geom_point(color = "blue", alpha = 0.7) +
  geom_smooth(method = "lm", color = "red", se = TRUE) +
  labs(title = "Scatter Plot of PD Mortality Rates vs. Total Chemical Expenditure",
       x = "Total Agricultural Chemical Expenditure",
       y = "PD Mortality Rate (per 100,000 population)") +
  theme_minimal()
```

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



b. Spearman's rank correlation for statistical testing

```
spearman_correlation <- cor(df$Total_Chemical_Expenditure, df$RATE, method = "spearman", use = "complete.obs")
spearman_correlation
```

```
## [1] 0.2019572
```

Result: There is a weak positive monotonic relationship between Total Chemical Expenditure and Mortality RATE.

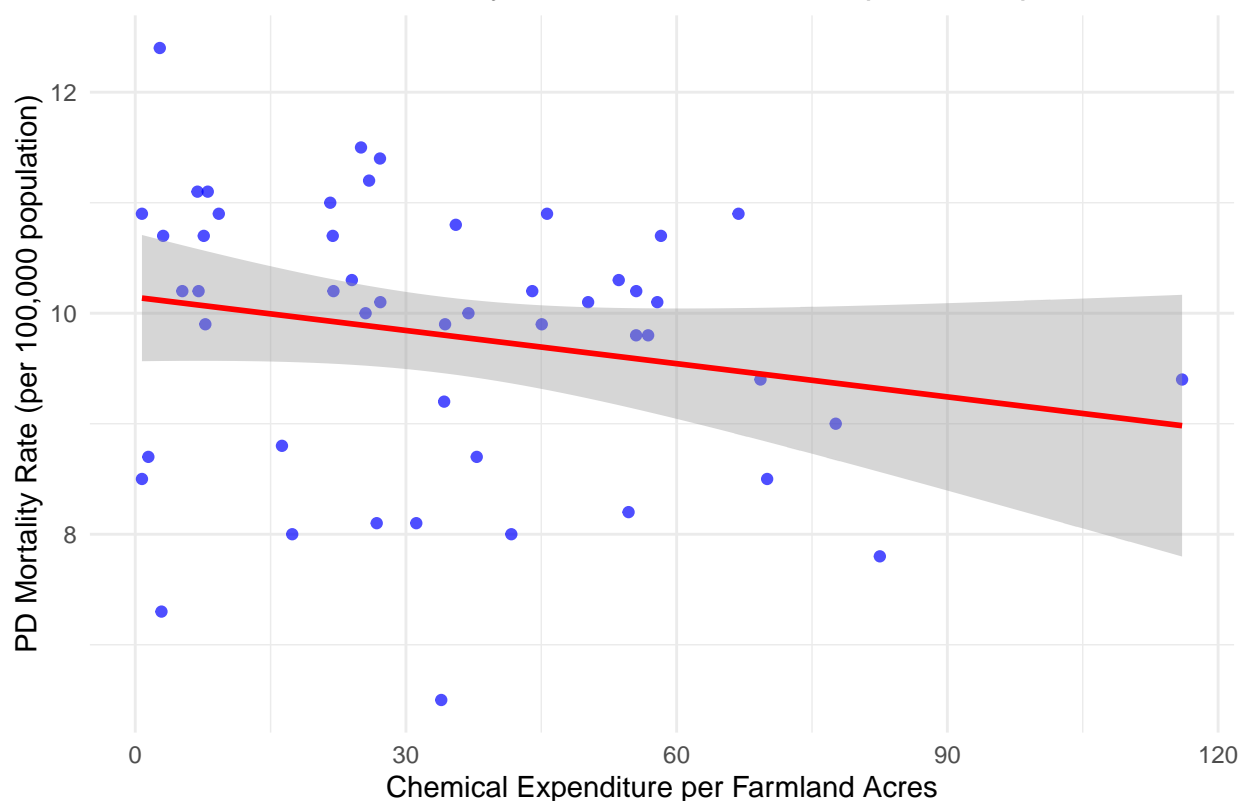
2. PD mortality rates vs. Chemical Expenditure per Farmland Acres per State

a. scatter plot visualization

```
ggplot(df, aes(x = Chemical_Expenditure_per_Acres, y = RATE)) +
  geom_point(color = "blue", alpha = 0.7) +
  geom_smooth(method = "lm", color = "red", se = TRUE) +
  labs(title = "Scatter Plot of PD Mortality Rates vs. Chemical Expenditure per Farmland Acres",
       x = "Chemical Expenditure per Farmland Acres",
       y = "PD Mortality Rate (per 100,000 population)") +
  theme_minimal()
```

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

Scatter Plot of PD Mortality Rates vs. Chemical Expenditure per Farmland Acres



b. Spearman's rank correlation for statistical testing

```
spearman_correlation <- cor(df$Chemical_Expenditure_per_Acres, df$RATE,
                             method = "spearman", use = "complete.obs")
spearman_correlation
```

```
## [1] -0.2528793
```

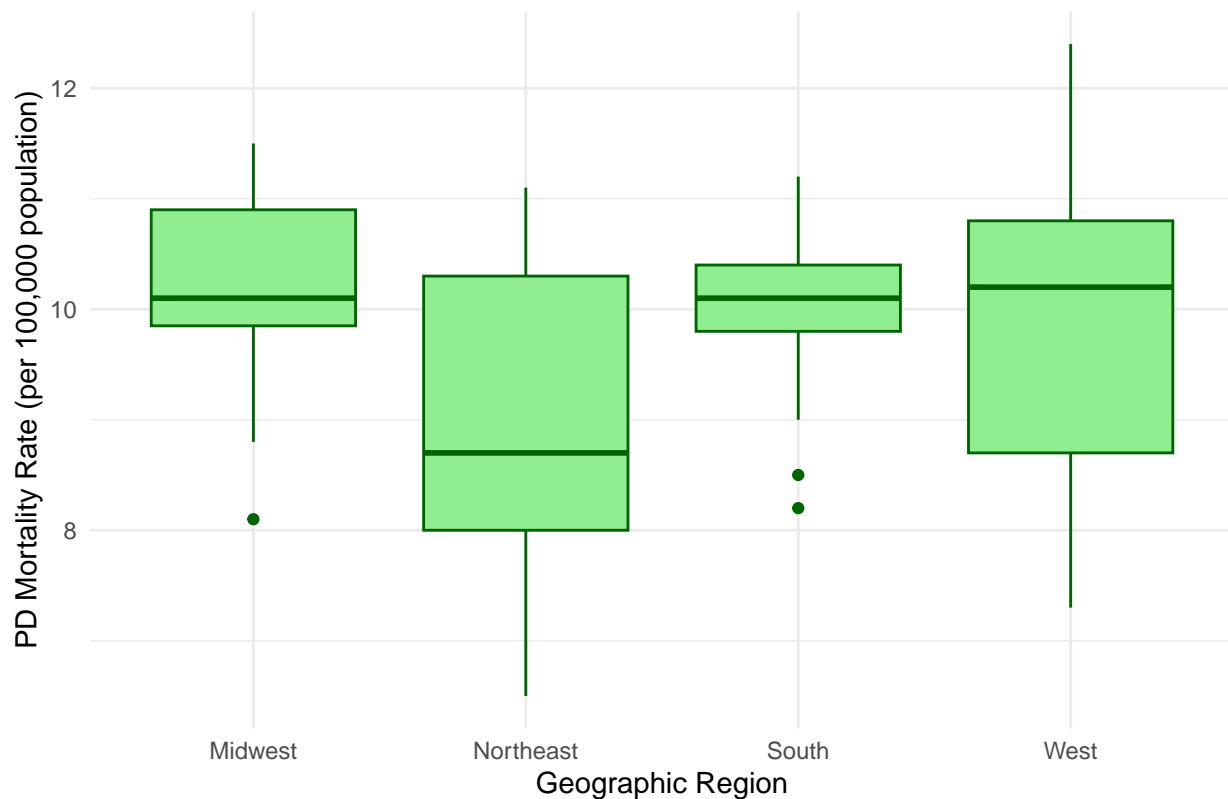
Result: There is a weak inverse relationship, meaning that as agricultural chemical expenditure rises, Parkinson's mortality rate slightly decreases.

3. PD mortality rates vs. Total Chemical Expenditure Considering Geographic Regions

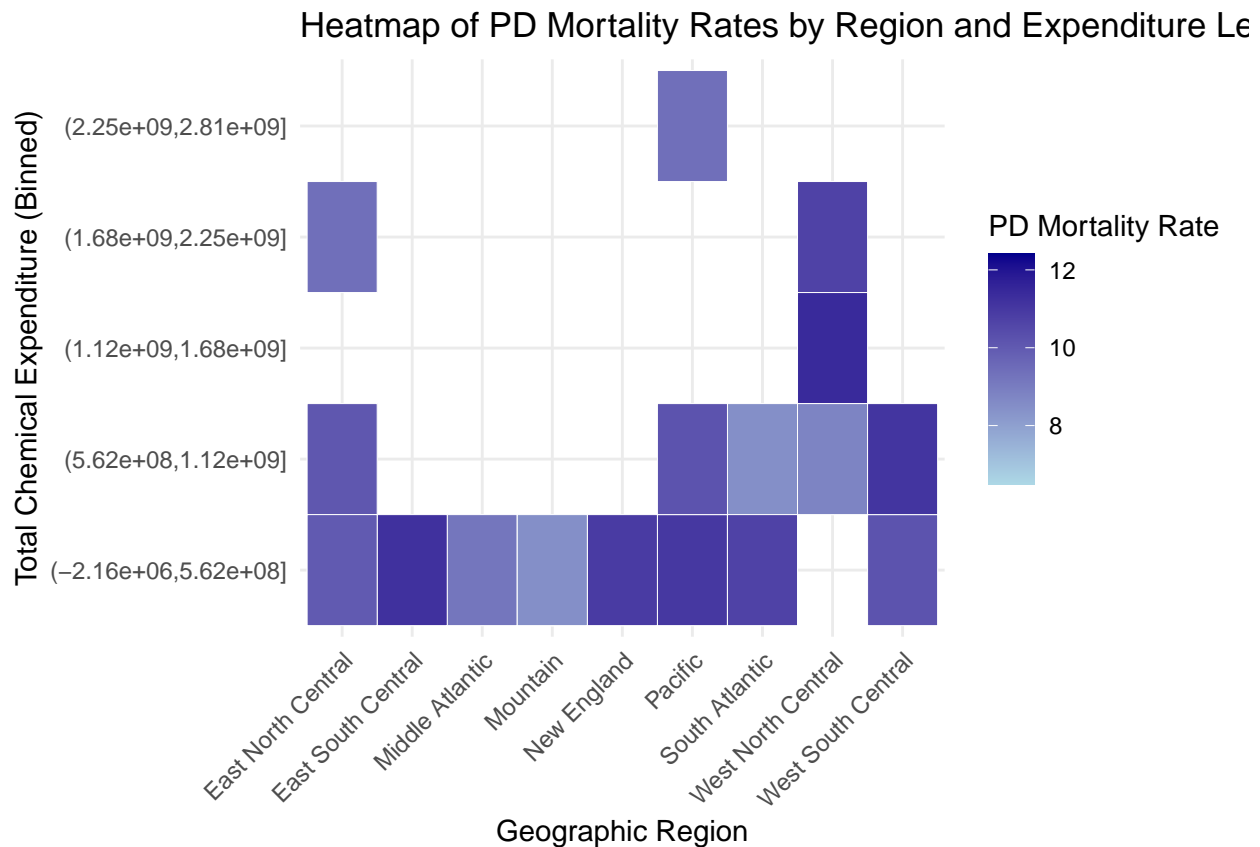
a. boxplot and heatmap

```
# Box Plot: PD mortality rates by Geographic Region
ggplot(df, aes(x = Geographic_Region, y = RATE)) +
  geom_boxplot(fill = "lightgreen", color = "darkgreen") +
  labs(title = "Box Plot of PD Mortality Rates by Geographic Region",
       x = "Geographic Region",
       y = "PD Mortality Rate (per 100,000 population)") +
  theme_minimal()
```

Box Plot of PD Mortality Rates by Geographic Region



```
# Heatmap: PD mortality rate by Geographic Divisions and Total Chemical Expenditures
## Divide the Chemical_Expenditure_per_Acres into quintile
df <- df %>%
  mutate(Expenditure_Bin = cut(Total_Chemical_Expenditure, breaks = 5))
## Plot
ggplot(df, aes(x = Geographic_Division, y = Expenditure_Bin, fill = RATE)) +
  geom_tile(color = "white") +
  scale_fill_gradient(low = "lightblue", high = "darkblue") +
  labs(title = "Heatmap of PD Mortality Rates by Region and Expenditure Levels",
       x = "Geographic Region",
       y = "Total Chemical Expenditure (Binned)",
       fill = "PD Mortality Rate") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

- b. Fit the mixed-effects model Since the Geographic Region is a higher-level above State, we can use a mixed-effects model to show a more accurate relationship

```
# Fit the mixed-effects model
# Assuming RATE is the dependent variable
# and Total_Chemical_Expenditure is the main predictor
model <- lmer(RATE ~ Total_Chemical_Expenditure + (1 | Geographic_Region)
, data = df)
```

```
## Warning: Some predictor variables are on very different scales: consider
## rescaling
```

```
# Summary of the model
summary(model)
```

```
## Linear mixed model fit by REML ['lmerMod']
## Formula: RATE ~ Total_Chemical_Expenditure + (1 | Geographic_Region)
## Data: df
##
## REML criterion at convergence: 203.5
##
## Scaled residuals:
## Min 1Q Median 3Q Max
## -2.5440 -0.7841 0.1603 0.7648 2.2556
##
```

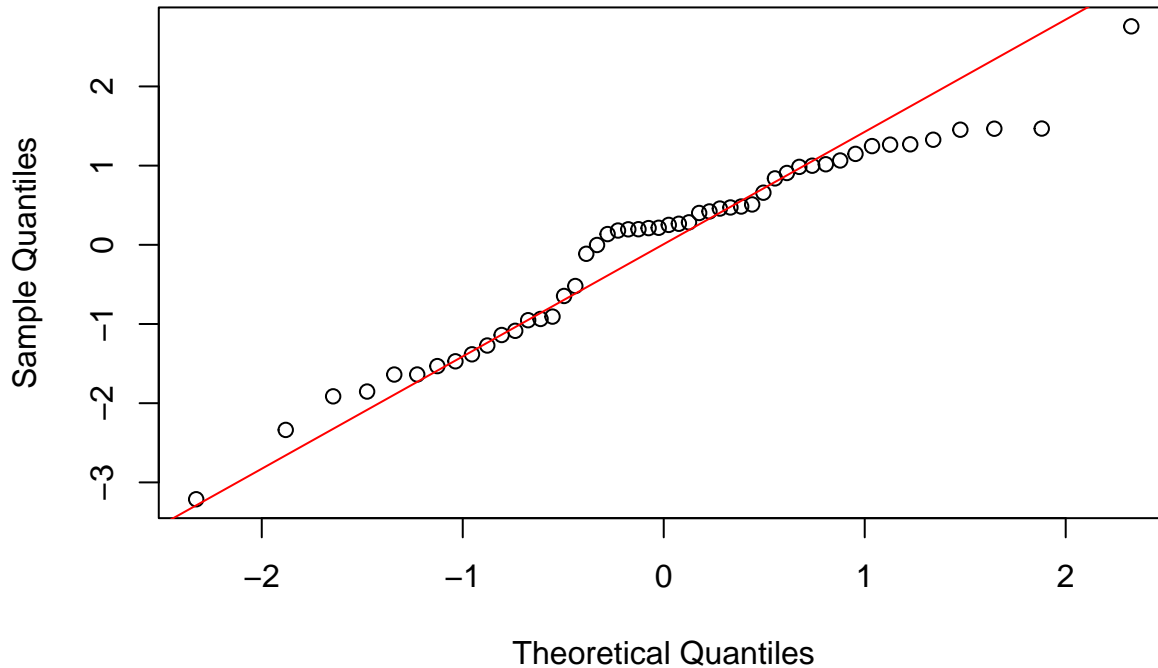
```
## Random effects:
##   Groups           Name          Variance Std.Dev.
##   Geographic_Region (Intercept) 0.03491  0.1868
##   Residual              1.47136  1.2130
## Number of obs: 50, groups:  Geographic_Region, 4
##
## Fixed effects:
##               Estimate Std. Error t value
## (Intercept)      9.636e+00  2.479e-01  38.875
## Total_Chemical_Expenditure 3.323e-10  3.222e-10   1.031
##
## Correlation of Fixed Effects:
##           (Intr)
## Ttl_Chmcl_E -0.612
## fit warnings:
## Some predictor variables are on very different scales: consider rescaling
```

```
# Check normality of residuals of the dependent variable
## Fit a simple linear regression model
model <- lm(RATE ~ Total_Chemical_Expenditure, data = df)
summary(model)
```

```
##
## Call:
## lm(formula = RATE ~ Total_Chemical_Expenditure, data = df)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.2127 -0.9480  0.2332  0.9647  2.7585
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)      9.631e+00  2.273e-01  42.379  <2e-16 ***
## Total_Chemical_Expenditure 3.709e-10  3.125e-10   1.187   0.241
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.222 on 48 degrees of freedom
## Multiple R-squared:  0.02851,    Adjusted R-squared:  0.008274
## F-statistic: 1.409 on 1 and 48 DF,  p-value: 0.2411
```

```
## Extract residuals
residuals <- resid(model)
qqnorm(residuals)
qqline(residuals, col = "red")
```

Normal Q-Q Plot



Interpretation: The coefficient indicates that for each dollar increase in Total Chemical Expenditure, the RATE (PD mortality rate) is expected to increase by 3.709×10^{-10} per 100,000 person. In other words, for each billion-dollar increase in Total Chemical Expenditure, the RATE (PD mortality rate) is expected to increase by 0.3709 deaths per 100,000 people. The p-value for Total_Chemical_Expenditure is 0.241, which is above the significance level (0.05). This means the effect of Total Chemical Expenditure on PD mortality rate is not statistically significant in this model.

III. Aim 2b: Evaluate the association between the amount of acreage treated with three types of pesticides and PD mortality rates.

Look at totals and averages

```
# create a summary table of the total acreage of treated farmland
# for different types of pesticides and the mortality rate
summary_table <- df %>%
  summarise(
    Insecticide_No_Nema_Sum_Acres =
      sum(Insecticide_No_Nema_Treatment_Acres, na.rm = TRUE),
    Insecticide_No_Nema_Avg_Percentage =
      mean(`Insecticide_No_Nema_Treatment_Percentage_(`, na.rm = TRUE),
    Insecticide_No_Nema_SD_Percentage =
      sd(`Insecticide_No_Nema_Treatment_Percentage_(`, na.rm = TRUE),

    Insecticide_Nema_Sum_Acres =
      sum(Insecticide_Nema_Treatment_Acres, na.rm = TRUE),
    Insecticide_Nema_Avg_Percentage =
```

```

    mean(`Insecticide_Nema_Treatment_Percentage_(`, na.rm = TRUE),
Insecticide_Nema_SD_Percentage =
    sd(`Insecticide_Nema_Treatment_Percentage_(`, na.rm = TRUE),

Herbicide_Sum_Acres =
    sum(Herbicide_Treatment_Acres, na.rm = TRUE),
Herbicide_Avg_Percentage =
    mean(`Herbicide_Treatment_Percentage_(`, na.rm = TRUE),
Herbicide_SD_Percentage =
    sd(`Herbicide_Treatment_Percentage_(`, na.rm = TRUE),

Fungicide_Sum_Acres =
    sum(Fungicide_Treatment_Acres, na.rm = TRUE),
Fungicide_Avg_Percentage =
    mean(`Fungicide_Treatment_Percentage_(`, na.rm = TRUE),
Fungicide_SD_Percentage =
    sd(`Fungicide_Treatment_Percentage_(`, na.rm = TRUE),

Mortality_Sum_Rate = sum(RATE, na.rm = TRUE),
Mortality_Avg_Rate = mean(RATE, na.rm = TRUE),
Mortality_SD_Rate = sd(RATE, na.rm = TRUE)
)

print(summary_table)

##   Insecticide_No_Nema_Sum_Acres Insecticide_No_Nema_Avg_Percentage
## 1                82860848                      11.33876
##   Insecticide_No_Nema_SD_Percentage Insecticide_Nema_Sum_Acres
## 1                9.365526                      21210813
##   Insecticide_Nema_Avg_Percentage Insecticide_Nema_SD_Percentage
## 1                2.873056                      2.998834
##   Herbicide_Sum_Acres Herbicide_Avg_Percentage Herbicide_SD_Percentage
## 1                264177083                30.04337                22.3878
##   Fungicide_Sum_Acres Fungicide_Avg_Percentage Fungicide_SD_Percentage
## 1                40710631                5.666355                4.813071
##   Mortality_Sum_Rate Mortality_Avg_Rate Mortality_SD_Rate
## 1                490.3                9.806                1.227476

```

```

(correlations <- data.frame(
  Treatment_Type = c("Insecticide No Nema Acres",
                    "Insecticide No Nema Percentage",
                    "Insecticide Nema Acres",
                    "Insecticide Nema Percentage",
                    "Herbicide Acres",
                    "Herbicide Percentage",
                    "Fungicide Acres",
                    "Fungicide Percentage"),
  Spearman_Correlation_with_Mortality = c(
    cor(df$Insecticide_No_Nema_Treatment_Acres, df$RATE,
        method = "spearman", use = "complete.obs"),
    cor(df`Insecticide_No_Nema_Treatment_Percentage_(`,
        df$RATE, method = "spearman", use = "complete.obs"),

```

```

cor(df$Insecticide_Nema_Treatment_Acres, df$RATE,
    method = "spearman", use = "complete.obs"),
cor(df$Insecticide_Nema_Treatment_Percentage_(`, df$RATE, method = "spearman", use = "complete.obs"),

cor(df$Herbicide_Treatment_Acres, df$RATE,
    method = "spearman", use = "complete.obs"),
cor(df$Herbicide_Treatment_Percentage_(`, df$RATE,
    method = "spearman", use = "complete.obs"),

cor(df$Fungicide_Treatment_Acres, df$RATE,
    method = "spearman", use = "complete.obs"),
cor(df$Fungicide_Treatment_Percentage_(`, df$RATE, method = "spearman", use = "complete.obs")
)
))

```

```

##           Treatment_Type Spearman_Correlation_with_Mortality
## 1 Insecticide No Nema Acres 0.22484574
## 2 Insecticide No Nema Percentage -0.13829263
## 3 Insecticide Nema Acres 0.22114319
## 4 Insecticide Nema Percentage -0.09285225
## 5 Herbicide Acres 0.27677760
## 6 Herbicide Percentage 0.00778978
## 7 Fungicide Acres 0.18844535
## 8 Fungicide Percentage -0.15233347

```

```

kable(correlations, caption = "Spearman Correlation Matrix")

```

Table 1: Spearman Correlation Matrix

| Treatment_Type | Spearman_Correlation_with_Mortality |
|--------------------------------|-------------------------------------|
| Insecticide No Nema Acres | 0.2248457 |
| Insecticide No Nema Percentage | -0.1382926 |
| Insecticide Nema Acres | 0.2211432 |
| Insecticide Nema Percentage | -0.0928522 |
| Herbicide Acres | 0.2767776 |
| Herbicide Percentage | 0.0077898 |
| Fungicide Acres | 0.1884453 |
| Fungicide Percentage | -0.1523335 |

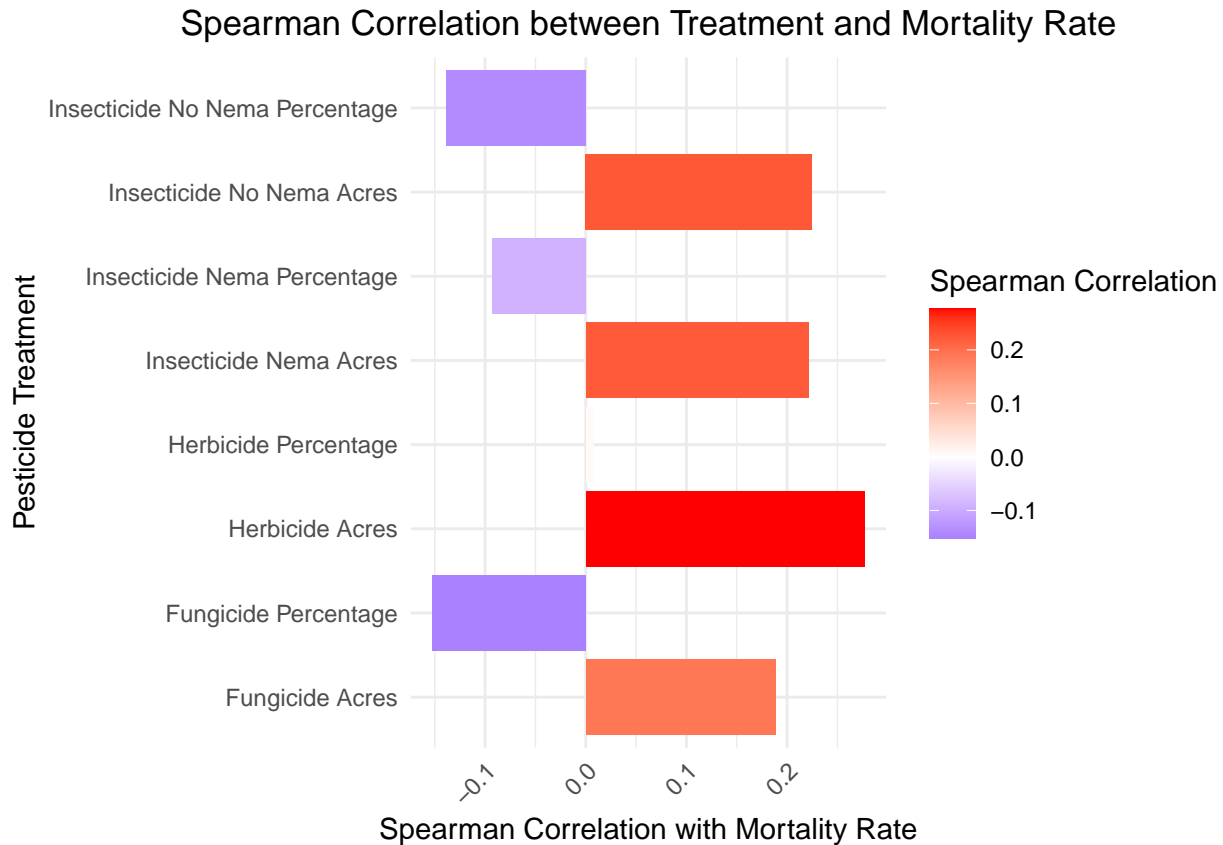
```

correlations_long <- correlations %>%
  pivot_longer(cols = Spearman_Correlation_with_Mortality,
    names_to = "Correlation_Type", values_to = "Correlation_Value")

ggplot(correlations_long, aes(x = Treatment_Type, y = Correlation_Value,
    fill = Correlation_Value)) +
  geom_bar(stat = "identity") +
  coord_flip() +
  scale_fill_gradient2(low = "blue", high = "red", mid = "white",
    midpoint = 0, name = "Spearman Correlation") +

```

```
labs(
  title = "Spearman Correlation between Treatment and Mortality Rate",
  x = "Pesticide Treatment",
  y = "Spearman Correlation with Mortality Rate"
) +
theme_minimal() +
theme(
  axis.text.x = element_text(angle = 45, hjust = 1),
  plot.title = element_text(hjust = 0.5)
)
```



Interpretation: There is no clear association between these pesticide treatments (measured as total acreage and percentage of farmland treated) with PD mortality rate.

```
cor_data <- df %>%
  dplyr::select(
    Total_Acres,
    Insecticide_No_Nema_Treatment_Acres,
    `Insecticide_No_Nema_Treatment_Percentage_(`,
    Insecticide_Nema_Treatment_Acres,
    `Insecticide_Nema_Treatment_Percentage_(`,
    Herbicide_Treatment_Acres,
```

```

  `Herbicide_Treatment_Percentage_(`,
  Fungicide_Treatment_Acres,
  `Fungicide_Treatment_Percentage_(`,
  DEATHS
)

cor_data <- cor_data %>% mutate(across(everything(), as.numeric))

```

```

## Warning: There was 1 warning in 'mutate()'.
## i In argument: 'across(everything(), as.numeric)'.
## Caused by warning:
## ! NAs introduced by coercion

```

```

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

# Display the correlation matrix
print(cor_matrix)

```

```

##                               Total_Acres
## Total_Acres                    1.00000000
## Insecticide_No_Nema_Treatment_Acres    0.73016741
## Insecticide_No_Nema_Treatment_Percentage_(%) -0.01827333
## Insecticide_Nema_Treatment_Acres        0.65882482
## Insecticide_Nema_Treatment_Percentage_(%) -0.05022431
## Herbicide_Treatment_Acres              0.85687712
## Herbicide_Treatment_Percentage_(%)      0.26578400
## Fungicide_Treatment_Acres              0.65794945
## Fungicide_Treatment_Percentage_(%)     -0.07429697
## DEATHS                                0.04847357
##                               Insecticide_No_Nema_Treatment_Acres
## Total_Acres                      0.7301674
## Insecticide_No_Nema_Treatment_Acres    1.0000000
## Insecticide_No_Nema_Treatment_Percentage_(%) 0.6124302
## Insecticide_Nema_Treatment_Acres        0.9501040
## Insecticide_Nema_Treatment_Percentage_(%) 0.5452457
## Herbicide_Treatment_Acres              0.9536054
## Herbicide_Treatment_Percentage_(%)      0.7724040
## Fungicide_Treatment_Acres              0.9610461
## Fungicide_Treatment_Percentage_(%)      0.5194223
## DEATHS                                0.3373454
##                               Insecticide_No_Nema_Treatment_Percentage_(%)
## Total_Acres                      -0.01827333
## Insecticide_No_Nema_Treatment_Acres    0.61243024
## Insecticide_No_Nema_Treatment_Percentage_(%) 1.00000000
## Insecticide_Nema_Treatment_Acres        0.65072765
## Insecticide_Nema_Treatment_Percentage_(%) 0.94594595
## Herbicide_Treatment_Acres              0.43669986
## Herbicide_Treatment_Percentage_(%)      0.91180654
## Fungicide_Treatment_Acres              0.66976693
## Fungicide_Treatment_Percentage_(%)      0.94857205
## DEATHS                                0.45201882

```

| | | |
|--|--|-------------|
| ## | Insecticide_Nema_Treatment_Acres | |
| ## Total_Acres | | 0.6588248 |
| ## Insecticide_No_Nema_Treatment_Acres | | 0.9501040 |
| ## Insecticide_No_Nema_Treatment_Percentage_ (%) | | 0.6507277 |
| ## Insecticide_Nema_Treatment_Acres | | 1.0000000 |
| ## Insecticide_Nema_Treatment_Percentage_ (%) | | 0.6577306 |
| ## Herbicide_Treatment_Acres | | 0.9087428 |
| ## Herbicide_Treatment_Percentage_ (%) | | 0.7824707 |
| ## Fungicide_Treatment_Acres | | 0.9610461 |
| ## Fungicide_Treatment_Percentage_ (%) | | 0.5721633 |
| ## DEATHS | | 0.4688697 |
| ## | Insecticide_Nema_Treatment_Percentage_ (%) | |
| ## Total_Acres | | -0.05022431 |
| ## Insecticide_No_Nema_Treatment_Acres | | 0.54524565 |
| ## Insecticide_No_Nema_Treatment_Percentage_ (%) | | 0.94594595 |
| ## Insecticide_Nema_Treatment_Acres | | 0.65773061 |
| ## Insecticide_Nema_Treatment_Percentage_ (%) | | 1.00000000 |
| ## Herbicide_Treatment_Acres | | 0.38220812 |
| ## Herbicide_Treatment_Percentage_ (%) | | 0.83608710 |
| ## Fungicide_Treatment_Acres | | 0.61855783 |
| ## Fungicide_Treatment_Percentage_ (%) | | 0.89232958 |
| ## DEATHS | | 0.55049787 |
| ## | Herbicide_Treatment_Acres | |
| ## Total_Acres | | 0.8568771 |
| ## Insecticide_No_Nema_Treatment_Acres | | 0.9536054 |
| ## Insecticide_No_Nema_Treatment_Percentage_ (%) | | 0.4366999 |
| ## Insecticide_Nema_Treatment_Acres | | 0.9087428 |
| ## Insecticide_Nema_Treatment_Percentage_ (%) | | 0.3822081 |
| ## Herbicide_Treatment_Acres | | 1.0000000 |
| ## Herbicide_Treatment_Percentage_ (%) | | 0.6874932 |
| ## Fungicide_Treatment_Acres | | 0.9067732 |
| ## Fungicide_Treatment_Percentage_ (%) | | 0.3557282 |
| ## DEATHS | | 0.2951089 |
| ## | Herbicide_Treatment_Percentage_ (%) | |
| ## Total_Acres | | 0.2657840 |
| ## Insecticide_No_Nema_Treatment_Acres | | 0.7724040 |
| ## Insecticide_No_Nema_Treatment_Percentage_ (%) | | 0.9118065 |
| ## Insecticide_Nema_Treatment_Acres | | 0.7824707 |
| ## Insecticide_Nema_Treatment_Percentage_ (%) | | 0.8360871 |
| ## Herbicide_Treatment_Acres | | 0.6874932 |
| ## Herbicide_Treatment_Percentage_ (%) | | 1.0000000 |
| ## Fungicide_Treatment_Acres | | 0.7951636 |
| ## Fungicide_Treatment_Percentage_ (%) | | 0.8441843 |
| ## DEATHS | | 0.4561768 |
| ## | Fungicide_Treatment_Acres | |
| ## Total_Acres | | 0.6579494 |
| ## Insecticide_No_Nema_Treatment_Acres | | 0.9610461 |
| ## Insecticide_No_Nema_Treatment_Percentage_ (%) | | 0.6697669 |
| ## Insecticide_Nema_Treatment_Acres | | 0.9610461 |
| ## Insecticide_Nema_Treatment_Percentage_ (%) | | 0.6185578 |
| ## Herbicide_Treatment_Acres | | 0.9067732 |
| ## Herbicide_Treatment_Percentage_ (%) | | 0.7951636 |
| ## Fungicide_Treatment_Acres | | 1.0000000 |
| ## Fungicide_Treatment_Percentage_ (%) | | 0.6319072 |


```
## DEATHS                                0.4049677
##                                     Fungicide_Treatment_Percentage_(%)
## Total_Acres                             -0.07429697
## Insecticide_No_Nema_Treatment_Acres      0.51942226
## Insecticide_No_Nema_Treatment_Percentage_(%) 0.94857205
## Insecticide_Nema_Treatment_Acres         0.57216326
## Insecticide_Nema_Treatment_Percentage_(%) 0.89232958
## Herbicide_Treatment_Acres               0.35572820
## Herbicide_Treatment_Percentage_(%)      0.84418427
## Fungicide_Treatment_Acres               0.63190721
## Fungicide_Treatment_Percentage_(%)      1.00000000
## DEATHS                                0.39205602
##                                     DEATHS
## Total_Acres                             0.04847357
## Insecticide_No_Nema_Treatment_Acres      0.33734544
## Insecticide_No_Nema_Treatment_Percentage_(%) 0.45201882
## Insecticide_Nema_Treatment_Acres         0.46886968
## Insecticide_Nema_Treatment_Percentage_(%) 0.55049787
## Herbicide_Treatment_Acres               0.29510887
## Herbicide_Treatment_Percentage_(%)      0.45617682
## Fungicide_Treatment_Acres               0.40496772
## Fungicide_Treatment_Percentage_(%)      0.39205602
## DEATHS                                1.00000000
```

```
colnames(cor_matrix) <- c(
  "Total Acres", "Insect No Nema Acres", "Insect No Nema %",
  "Insect Nema Acres", "Insect Nema %", "Herb Acres", "Herb %",
  "Fung Acres", "Fung %", "Deaths"
)
rownames(cor_matrix) <- c(
  "Total Acres", "Insect No Nema Acres", "Insect No Nema %",
  "Insect Nema Acres", "Insect Nema %", "Herb Acres", "Herb %",
  "Fung Acres", "Fung %", "Deaths"
)
kable(cor_matrix, caption = "Spearman Correlation Matrix")
```

Table 2: Spearman Correlation Matrix

| | Total Acres | Insect No Nema Acres | Insect No Nema % | Insect Nema Acres | Insect Nema % | Herb Acres | Herb % | Fung Acres | Fung % | Deaths |
|----------------------------|----------------|----------------------------|------------------------|-------------------------|---------------------|---------------|-----------|---------------|-----------|-----------|
| Total Acres | 1.00000000 | 0.7301674 | - | 0.6588248 | - | 0.8568770 | 0.2657840 | 0.6579494 | - | 0.0484736 |
| Insect No Nema Acres | 0.7301674 | 1.0000000 | 0.0182733 | 0.9501040 | 0.0502243 | 0.9536050 | 0.7724040 | 0.0961046 | 0.5194223 | 0.3373454 |
| Insect No Nema % | - | 0.6124302 | 1.0000000 | 0.6507277 | 0.9459459 | 0.4366990 | 0.9118065 | 0.6697660 | 0.9485720 | 0.4520188 |
| Insect Nema Acres | 0.6588248 | 0.9501040 | 0.6507277 | 1.0000000 | 0.6577306 | 0.9087428 | 0.7824707 | 0.9610460 | 0.5721633 | 0.4688697 |
| Insect Nema % | - | 0.5452457 | 0.9459459 | 0.6577306 | 1.0000000 | 0.3822080 | 0.8360870 | 0.1618557 | 0.8923296 | 0.5504979 |

| | Total Acres | Insect No Nema Acres | Insect No Nema % | Insect Nema Acres | Insect Nema % | Herb Acres | Herb % | Fung Acres | Fung % | Deaths |
|---------------|----------------|----------------------------|------------------------|-------------------------|---------------------|---------------|-----------|---------------|-----------|-----------|
| Herb Acres | 0.85687710 | 0.9536054 | 0.4366999 | 0.9087428 | 0.38220811 | 0.0000000 | 0.6874932 | 0.29067732 | 0.3557282 | 0.2951089 |
| Herb % | 0.26578400 | 0.7724040 | 0.9118065 | 0.7824707 | 0.83608710 | 0.6874932 | 0.0000000 | 0.7951636 | 0.8441843 | 0.4561768 |
| Fung Acres | 0.65794940 | 0.9610461 | 0.6697669 | 0.9610461 | 0.61855780 | 0.9067732 | 0.7951636 | 0.0000000 | 0.6319072 | 0.4049677 |
| Fung % | - | 0.5194223 | 0.9485721 | 0.5721633 | 0.89232960 | 0.3557282 | 0.8441843 | 0.6319072 | 0.0000000 | 0.3920560 |
| Deaths | 0.0742970 | 0.04847360 | 0.3373454 | 0.4520188 | 0.4688697 | 0.55049790 | 0.2951089 | 0.4561768 | 0.4049677 | 0.3920560 |
| | | | | | | | | | | 0.0000000 |

very weak to weak correlation of some treatment measurements and PD mortality