

Counterfactual Adjusted Causal Model

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```
# clear the working space
rm(list = ls())

library(here)
library(stats) # need this to calculate Mahalanobis Distance
library(parallel) # parallelize
library(dplyr)
library(FNN)
library(cluster)
library(ggplot2)
library(rpart)
library(caret)
```

Counterfactual Data Input

```
# we need the renaming function for cleaning
melor_2015 <- read.csv(here("data", "clustered_M15_CF_data2.csv"))
```

Counterfactual predictions

Importing trained models

```
# Import trained BASE models
# From folder: adjusted SCM/new base models

base_models_list <- list()

# base models file path
base_file_path <- here("adjusted SCM/new base models")
base_wind_model <- readRDS(file.path(base_file_path,
                                     "dec_base_wind_model_tuned.rds"))

base_class_full_model <- readRDS(file.path(base_file_path,
                                           "damage_fit_class_full.rds"))
base_reg_model <- readRDS(file.path(base_file_path,
                                    "base_reg_model.rds"))
```

```

base_models_list <- list("base_wind_model" = base_wind_model,
                        "base_class_full_model" = base_class_full_model,
                        "base_reg_model" = base_reg_model)

# Import trained Truncated models
# From folder: adjusted SCM/new trunc models

# empty list
trunc_models_list <- list()

trunc_file_path <- here("adjusted SCM/new trunc models")

trunc_wind_model <- readRDS(file.path(trunc_file_path,
                                     "trunc_wind_model_tuned.rds"))

trunc_reg_model <- readRDS(file.path(trunc_file_path,
                                     "trunc_reg_model.rds"))

trunc_models_list <- list("trunc_wind_model" = trunc_wind_model,
                        "trunc_reg_model" = trunc_reg_model)

# calling hurdle function
source(here("R", "adj_hurdle_function.R"))

adj_counterfactual_hurdle_preds <- adj_hurdle_function(df = melor_2015,
                                                    scm_models_base = base_models_list,
                                                    scm_models_high = trunc_models_list,
                                                    threshold = 0.3 # threshold in train/test models is 0.35
                                                    )

# append the results to the counterfactual dataset
melor_2015 <- melor_2015 %>%
  mutate(damage_preds = adj_counterfactual_hurdle_preds)

```

Counteractual Results

```

# convert the Cluster column to factor
melor_2015$Cluster <- as.factor(melor_2015$Cluster)

# extract cluster_levels
cluster_levels <- levels(melor_2015$Cluster)

# Source function: counterfactual_results
source(here("R", "counterfactual_results.R"))

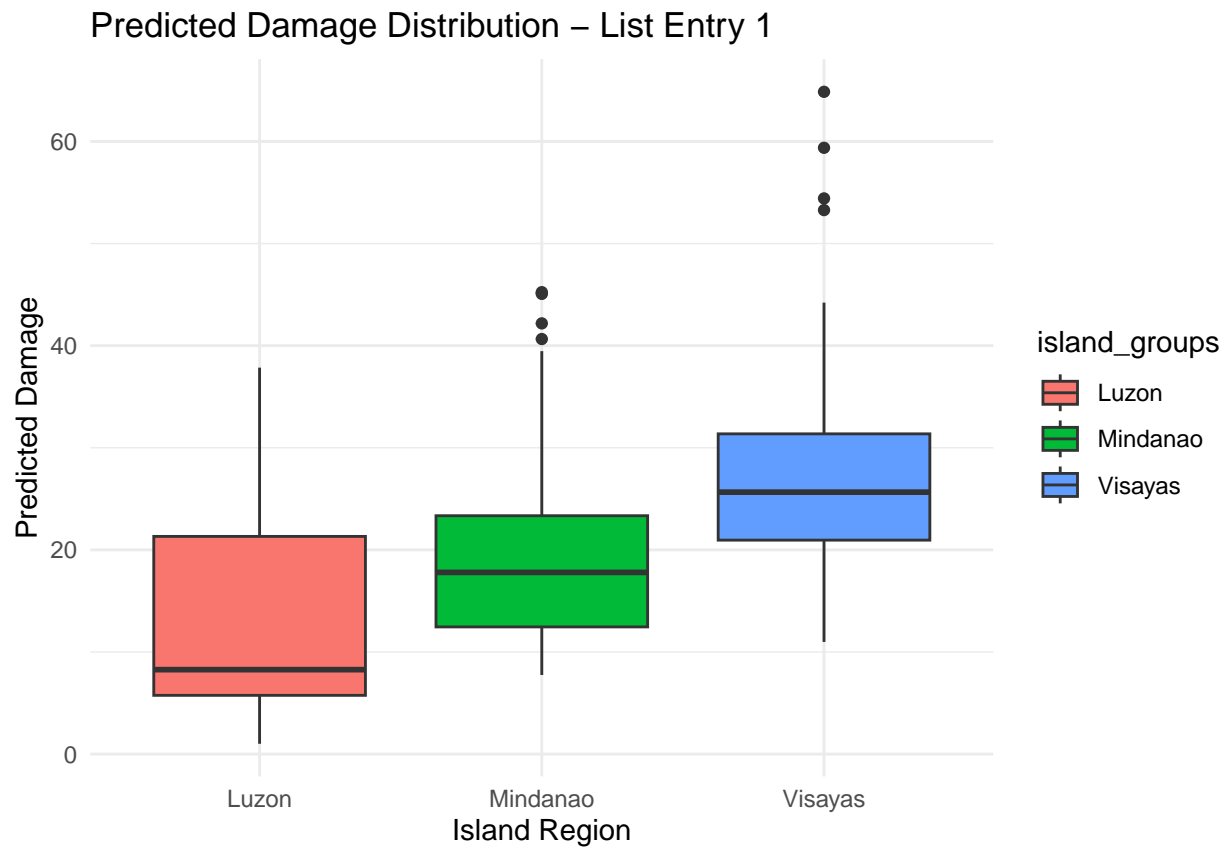
cf_results <- counterfactual_results(cf_data = melor_2015,
                                    cluster_levels = cluster_levels)

```

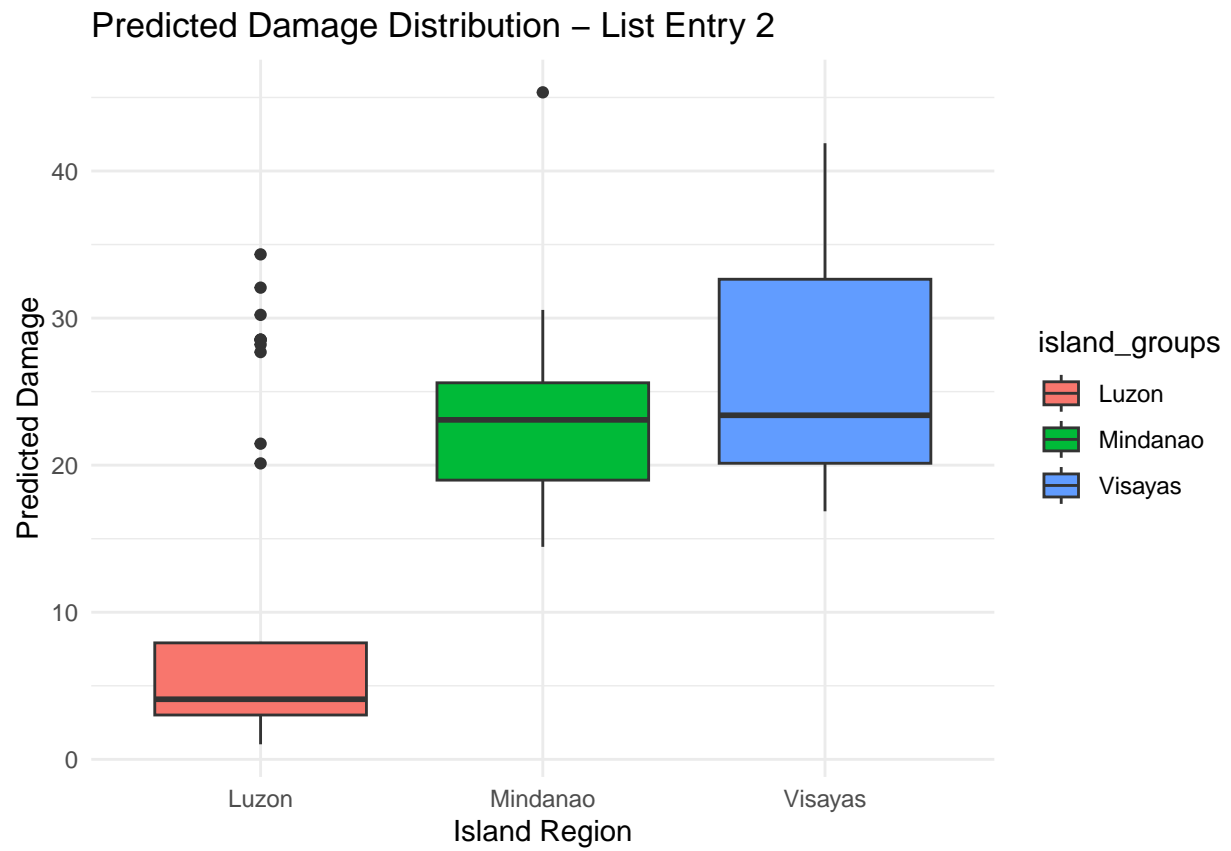
```
## currently evaluating cluster: 1currently evaluating cluster: 2currently evaluating cluster: 3current
```

```
# Check the list to confirm plots are stored  
print(cf_results$plots)
```

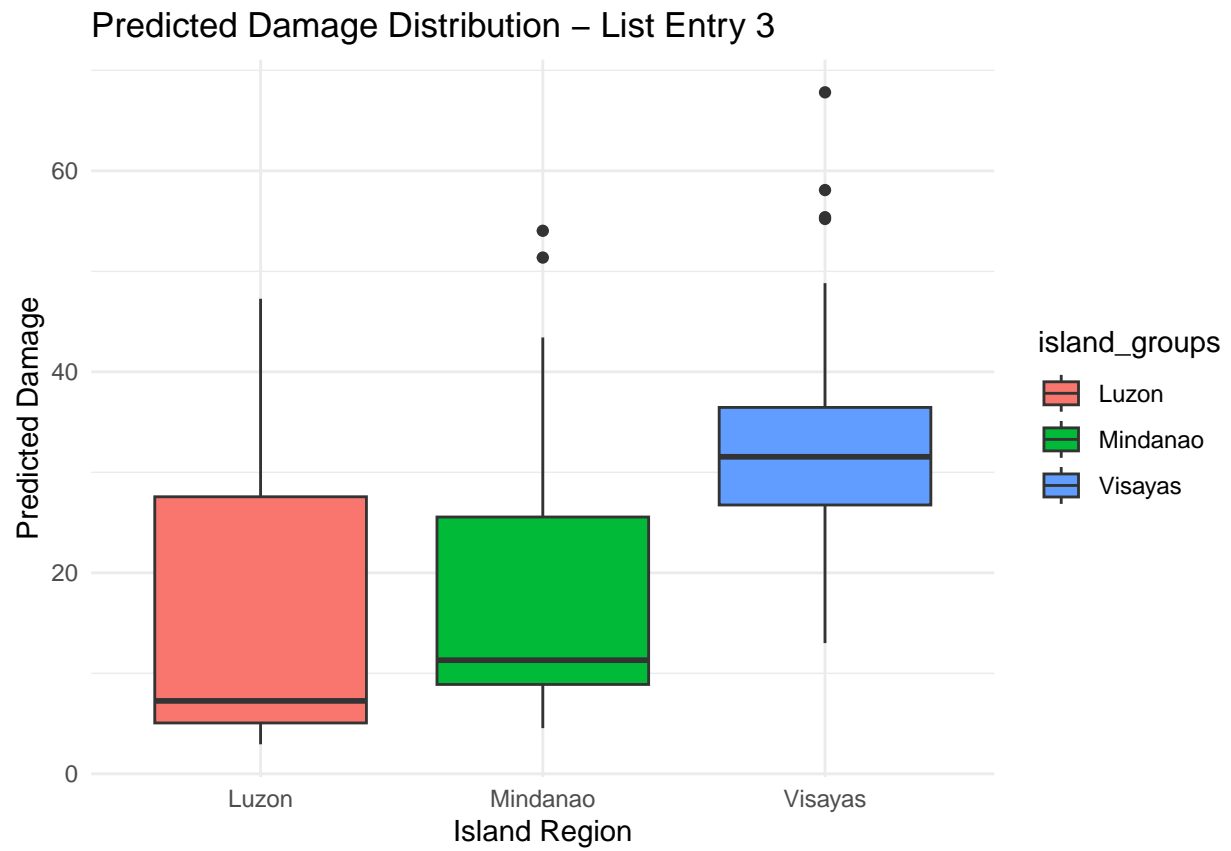
```
## [[1]]
```



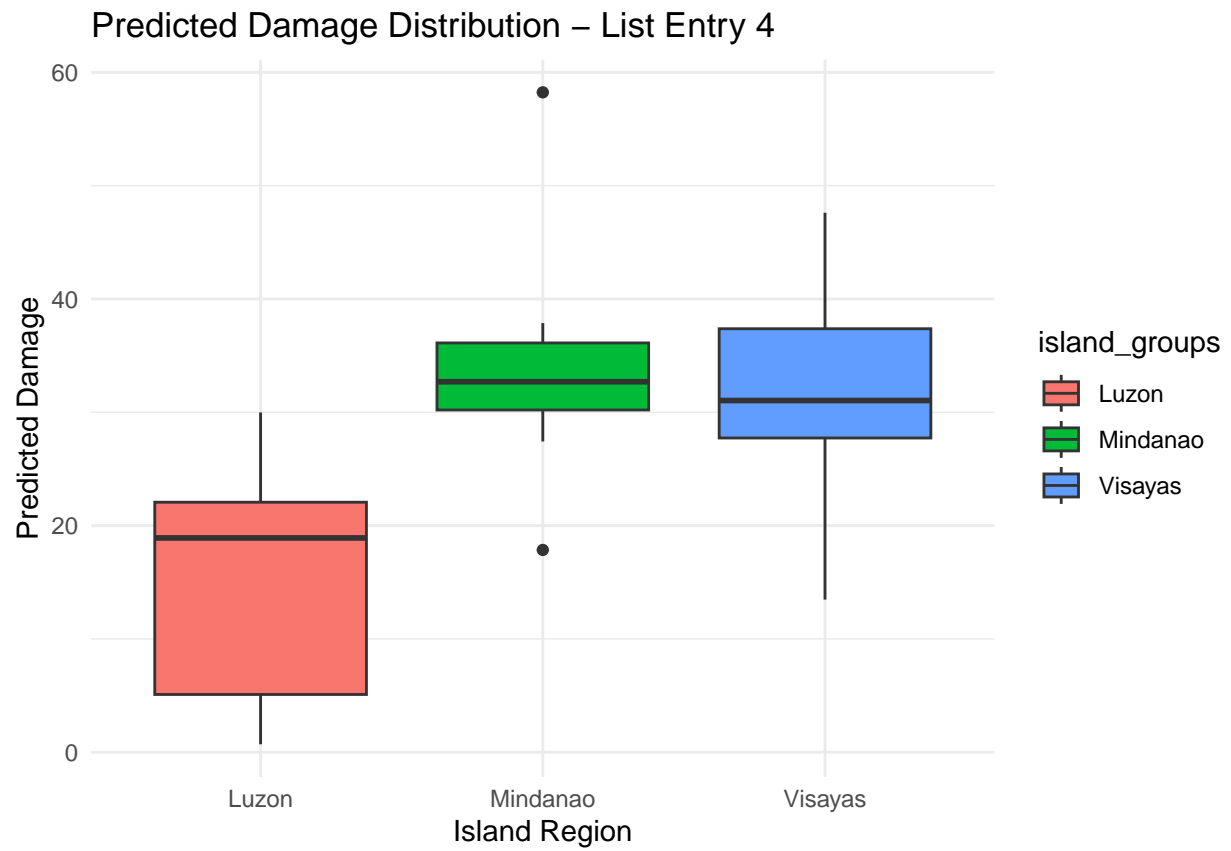
```
##  
## [[2]]
```



```
##  
## [[3]]
```

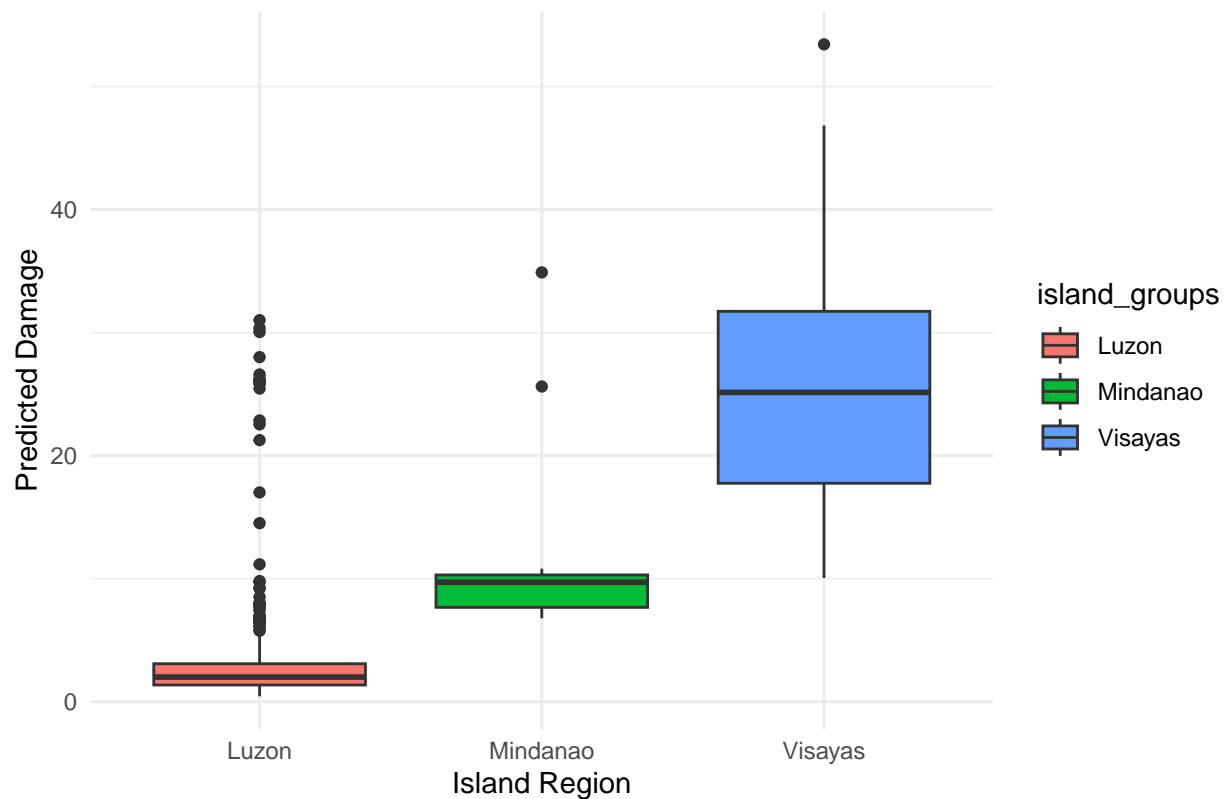


```
##  
## [[4]]
```



```
##  
## [[5]]
```

Predicted Damage Distribution – List Entry 5



```
print(cf_results$median)
```

```
## [[1]]
## # A tibble: 3 x 2
##   island_groups median_damage
##   <chr>          <dbl>
## 1 Luzon          8.27
## 2 Mindanao       17.8
## 3 Visayas       25.7
##
## [[2]]
## # A tibble: 3 x 2
##   island_groups median_damage
##   <chr>          <dbl>
## 1 Luzon          4.08
## 2 Mindanao       23.1
## 3 Visayas       23.4
##
## [[3]]
## # A tibble: 3 x 2
##   island_groups median_damage
##   <chr>          <dbl>
## 1 Luzon          7.26
## 2 Mindanao       11.3
## 3 Visayas       31.5
```

```
##
## [[4]]
## # A tibble: 3 x 2
##   island_groups median_damage
##   <chr>           <dbl>
## 1 Luzon           18.9
## 2 Mindanao        32.7
## 3 Visayas         31.0
##
## [[5]]
## # A tibble: 3 x 2
##   island_groups median_damage
##   <chr>           <dbl>
## 1 Luzon           2.00
## 2 Mindanao        9.71
## 3 Visayas        25.1
```

```
print(cf_results$averages)
```

```
## [[1]]
## # A tibble: 3 x 2
##   island_groups mean_damage
##   <chr>           <dbl>
## 1 Luzon          13.1
## 2 Mindanao       19.5
## 3 Visayas        28.7
##
## [[2]]
## # A tibble: 3 x 2
##   island_groups mean_damage
##   <chr>           <dbl>
## 1 Luzon           9.46
## 2 Mindanao       24.4
## 3 Visayas        27.4
##
## [[3]]
## # A tibble: 3 x 2
##   island_groups mean_damage
##   <chr>           <dbl>
## 1 Luzon          15.0
## 2 Mindanao       16.7
## 3 Visayas        32.1
##
## [[4]]
## # A tibble: 3 x 2
##   island_groups mean_damage
##   <chr>           <dbl>
## 1 Luzon          15.4
## 2 Mindanao       34.2
## 3 Visayas        32.5
##
## [[5]]
## # A tibble: 3 x 2
##   island_groups mean_damage
```



```
##      <chr>                <dbl>
## 1 Luzon                   3.27
## 2 Mindanao                11.2
## 3 Visayas                 25.0
```

Output the counterfactual predictions

Saving the counterfactual predictions for mapping differences between this adjusted causal model and the associational XGBOOST model in QGIS.

```
CF_output <- melor_2015 %>%
  select(Mun_Code, Municipality, Cluster, damage_preds) %>%
  rename(adj_scm_CF_M15 = damage_preds)

write.csv2(CF_output, file = here("adjusted SCM/outputs", "scm_CF_M15.csv"))
```