# 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"))</pre>
```

## Counterfactual predictions

#### Importing trained models

```
base_models_list <- list("base_wind_model" = base_wind_model,</pre>
                          "base_rain_model" = base_rain_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()</pre>
trunc_file_path <- here("adjusted SCM/new trunc models")</pre>
trunc_wind_model <- readRDS(file.path(trunc_file_path,</pre>
                                       "trunc_wind_model_tuned.rds"))
trunc rain model <- readRDS(here("adjusted SCM/new trunc models",
                                 "dec_trunc_rain_model_tuned.rds"))
trunc_reg_model <- readRDS(file.path(trunc_file_path,</pre>
                                       "trunc_reg_model.rds"))
trunc_models_list <- list("trunc_wind_model" = trunc_wind_model,</pre>
                            "trunc_rain_model" = trunc_rain_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)

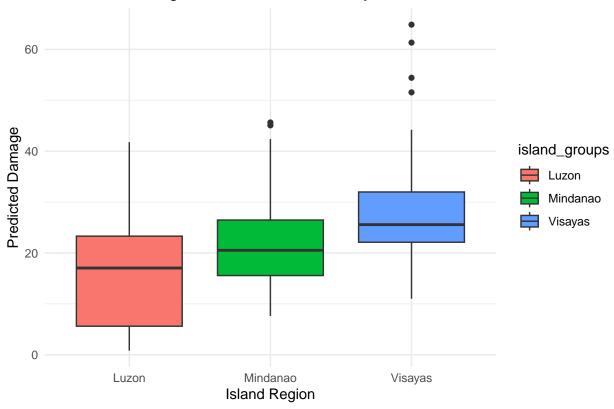
# Source function: counterfactual_results
source(here("R", "counterfactual_results.R"))</pre>
```

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

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

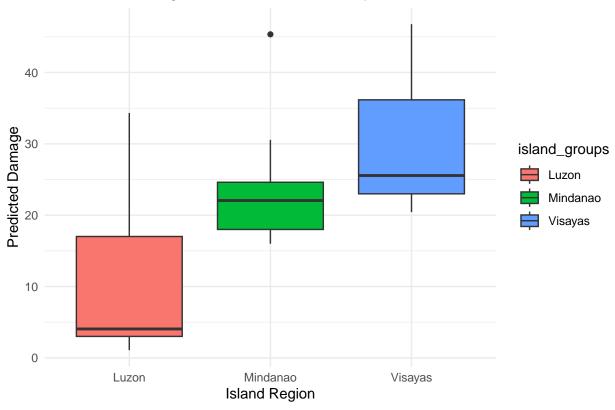
## [[1]]

# Predicted Damage Distribution – List Entry 1



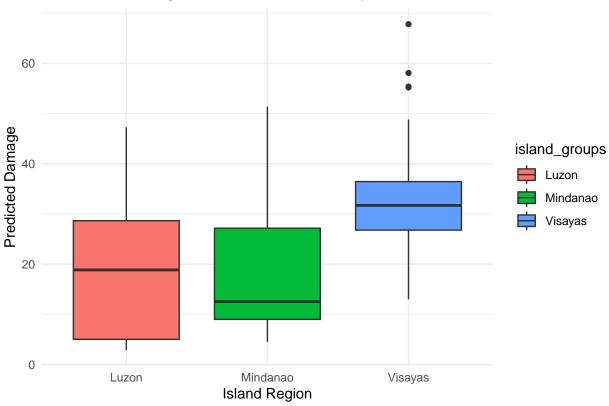
## ## [[2]]





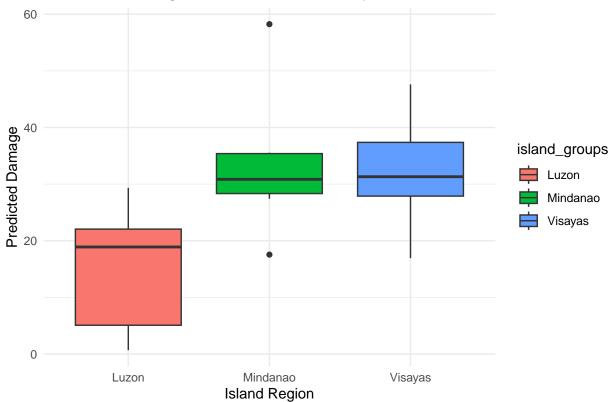
## ## [[3]]





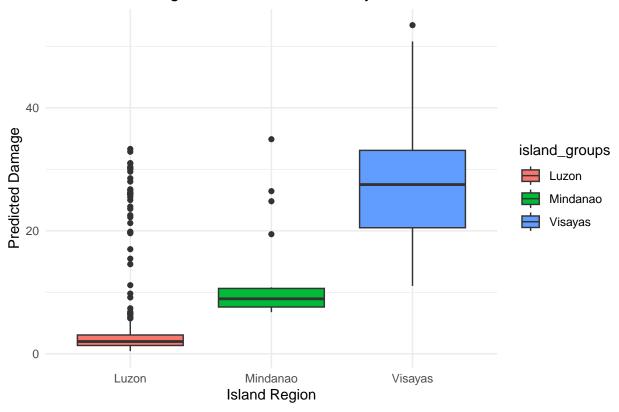
## ## [[4]]





## ## [[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
                             17.0
## 2 Mindanao
                             20.5
## 3 Visayas
                             25.6
##
## [[2]]
## # A tibble: 3 x 2
##
     island_groups median_damage
##
## 1 Luzon
                             4.05
## 2 Mindanao
                            22.1
## 3 Visayas
                            25.6
##
## [[3]]
## # A tibble: 3 \times 2
     island_groups median_damage
##
     <chr>
                            <dbl>
                             18.9
## 1 Luzon
## 2 Mindanao
                             12.6
## 3 Visayas
                             31.7
```

```
##
## [[4]]
## # A tibble: 3 x 2
     island_groups median_damage
     <chr>
## 1 Luzon
                             18.9
## 2 Mindanao
                             30.8
## 3 Visayas
                             31.3
##
## [[5]]
## # A tibble: 3 x 2
##
     island_groups median_damage
##
## 1 Luzon
                             2.00
## 2 Mindanao
                             8.95
## 3 Visayas
                            27.5
print(cf_results$averages)
     island_groups mean_damage
```

```
## [[1]]
## # A tibble: 3 x 2
##
     <chr>
                   <dbl>
                          15.8
## 1 Luzon
## 2 Mindanao
                          21.5
## 3 Visayas
                          29.0
##
## [[2]]
## # A tibble: 3 x 2
     island_groups mean_damage
##
     <chr>
                         <dbl>
## 1 Luzon
                          9.80
                         23.7
## 2 Mindanao
## 3 Visayas
                         30.9
##
## [[3]]
## # A tibble: 3 x 2
     island_groups mean_damage
##
     <chr>
                        <dbl>
## 1 Luzon
                          17.8
## 2 Mindanao
                         18.2
## 3 Visayas
                          32.2
##
## [[4]]
## # A tibble: 3 x 2
     island_groups mean_damage
##
     <chr>>
                         <dbl>
## 1 Luzon
                          15.2
## 2 Mindanao
                          33.1
## 3 Visayas
                          33.0
##
## [[5]]
## # A tibble: 3 x 2
   island_groups mean_damage
```

## 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"))
```