# Prediction

Youlan Shen

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# Task 4: Prediction Using Estimated Model Parameters

```
library(tidyverse)
library(dplyr)
library(cowplot)
```

### load data

```
load("hurricane.RData")
data <- hurricane %>% ungroup()
load("test.RData")
B <- test$B
gamma <- test$gamma</pre>
```

```
# get XYD
Y <- as.matrix(data$Wind.kt)</pre>
D <- data %>% mutate(intercept = 1) %>% dplyr::select(intercept, Wind_prev,
                                                       Lat_change, Long_change,
                                                       Wind_change) %>%
  as.matrix()
X <- data %>%
  mutate(index = 1:nrow(data)) %>%
  group_by(index) %>%
  dplyr::select(Month, Season, Nature) %>%
  mutate(n = 1) \%
  pivot_wider(names_from = Month, values_from = n, values_fill = 0) %>%
  mutate(n = 1) \%
  pivot_wider(names_from = Nature, values_from = n, values_fill = 0) %>%
  ungroup() %>%
  dplyr::select(April, May, June, July, August, September, October, November,
                December, Season, ET,NR,SS,TS) %>%
  as.matrix()
```

## Adding missing grouping variables: 'index'

## Get Estimated Model Parameters Using MCMC Train

```
# get parameter estimates function
get_estimate <- function(null_matrix, iter, MCMCchain){
   for (i in iter) {
      null_matrix <- null_matrix + MCMCchain[[i]]
   }
   estimate <- null_matrix / length(iter)
   estimate
}

# set iter, we only use the last 3000 of MCMC to get posterior mean
iter <- 3001:6000
# set Null matrix
beta_estimate <- matrix(data = 0, nrow = 5, ncol = 699)
gamma_estimate <- matrix(data = 0, nrow = 14, ncol = 1)
beta_estimate <- get_estimate(beta_estimate, iter, B)
gamma_estimate <- get_estimate(gamma_estimate, iter, gamma)</pre>
```

#### Get Prediction

```
X_with_id <- cbind(as.numeric(factor(data$ID)), X)
#number of hurricanes
H <- max(X_with_id[,1])
ith_hurricane_idx <- 1:H |>
    map(\(i) which(X_with_id[,1] == i))
# calculate Y estimates
N <- length(Y)
mu_H <- rep(NA, N)
for (i in 1:H) {
    curr_hurricane_idx <- ith_hurricane_idx[[i]]
    mu_i <- D[curr_hurricane_idx,,drop = FALSE] %*%
        beta_estimate[,i,drop = FALSE]
    mu_H[curr_hurricane_idx] <- mu_i
}
mu_H <- as.matrix(mu_H)
Wind_pred <- mu_H + X %*% gamma_estimate</pre>
```

```
# get the posterior mean
# colnames(beta_estimate) <- unique(data$ID)
# colnames(gamma_estimate) <- "gamma_estimate"</pre>
```

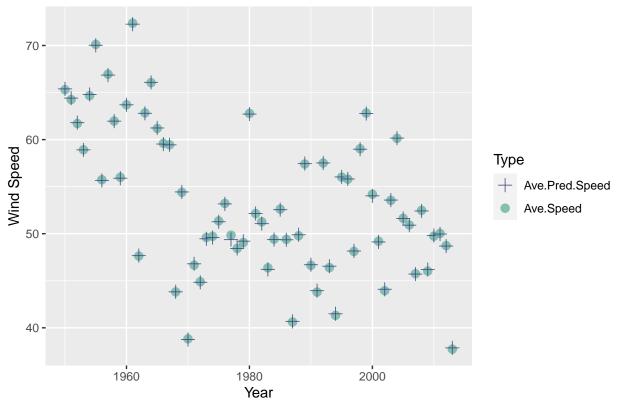
### **Analysis on Prediction**

```
hurricane["Wind_pred"] <- Wind_pred
hurricane <- hurricane %>%
   dplyr::select(ID, Season, Month, Nature, Date, Wind.kt, Wind_pred)
hurricane %>%
   mutate(Pred_E = Wind.kt - Wind_pred,
```

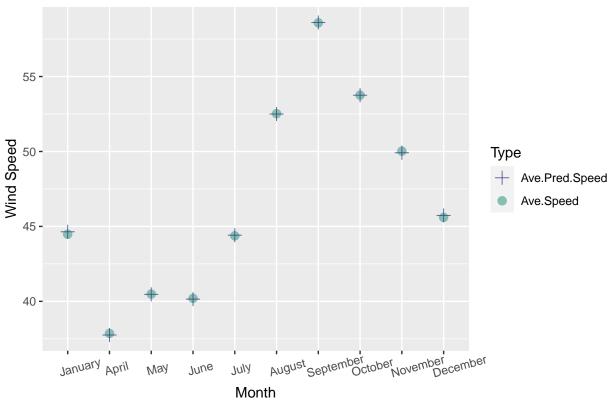
Table 1: Prediction Standard Error of Wind Speed in Ascending Order

Hurricane ID	Number of Observation	Prediction RMSE
SIXTEEN.2008	6	1.095
FABIAN.1991	9	1.245
FOUR:UNNAMED.1988	5	1.297
TEN.2005	19	1.420
LORENZO.2001	14	1.471
FIFTEEN:UNNAMED.1988	4	1.857
FABIAN.1997	15	1.933
ANDREW.1986	14	1.947
CINDY.1987	20	1.984
GUSTAV.1996	27	2.000

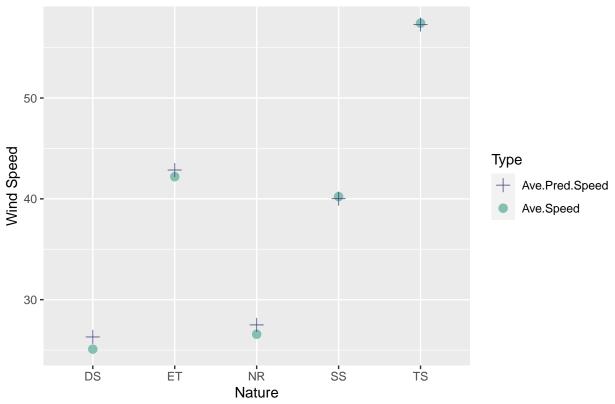


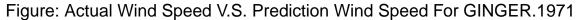


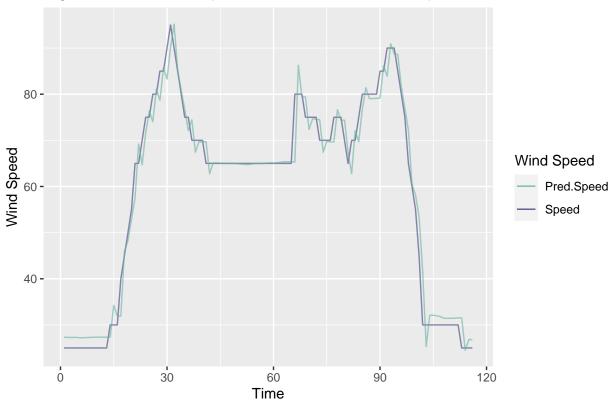


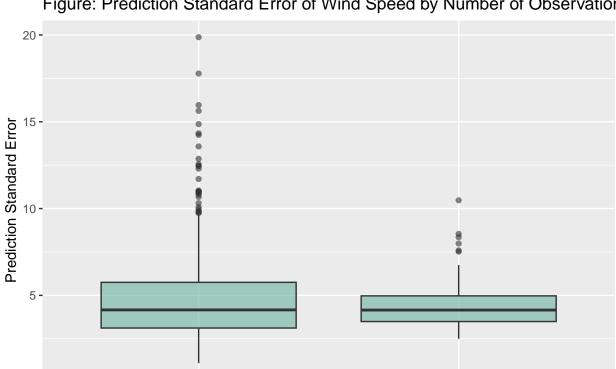












Observation

More Observation

Fewer Observation

Figure: Prediction Standard Error of Wind Speed by Number of Observation