

analysis for erosion

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目录

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

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.1      v readr      2.1.4
## v forcats    1.0.0      v stringr   1.5.0
## v lubridate  1.9.2      v tibble    3.2.1
## v purrr      1.0.2      v tidyr     1.3.0
## -- 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
```

```
library(readr)
library(gridExtra)
```

```
##
## 载入程辑包: 'gridExtra'
##
## The following object is masked from 'package:dplyr':
##
##      combine
```

```
library(car)
```

```
## 载入需要的程辑包: carData
##
## 载入程辑包: 'car'
##
## The following object is masked from 'package:dplyr':
##
##      recode
##
## The following object is masked from 'package:purrr':
##
##      some
```

```
library(grid)
```

```
erosion<-read.csv("erosion.csv")
```

linear regression for all variables(MAX waveheight)

```
stan <- function(x) {
  return((x - min(x)) / (max(x) - min(x)))
}
data <- read_csv("erosion.csv")
```

```
## Rows: 31 Columns: 8
## -- Column specification -----
## Delimiter: ","
## chr (1): Bluff
## dbl (7): Orientation (deg), RR (m/yr), Max Wave Height (m), Mud (%), BaseEl ...
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

```

names(data) <- gsub(" ", "_", names(data))
names(data) <- gsub("\\(", "", names(data))
names(data) <- gsub("\\)", "", names(data))
names(data) <- gsub("/", "_per_", names(data))
names(data) <- gsub("%", "percent", names(data))
if ("RR (m/yr)" %in% names(data)) {
  names(data)[names(data) == "RR (m/yr)"] <- "RR_m_per_yr"
}
numeric_columns <- sapply(data, is.numeric) & names(data) != "RR_m_per_yr"
data[numeric_columns] <- lapply(data[numeric_columns], stan)
predictors <- setdiff(names(data), c("Bluff", "RR_m_per_yr"))
target <- "RR_m_per_yr"
model_formula <- as.formula(paste(target, "~", paste(predictors, collapse = " + ")))
model1 <- lm(model_formula, data = data)
print(summary(model1))

```

```

##
## Call:
## lm(formula = model_formula, data = data)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -0.45124 -0.12697 -0.03711  0.09424  0.72960
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.22946    0.16714   1.373  0.18249
## Orientation_deg  0.06428    0.16130   0.399  0.69376
## Max_Wave_Height_m 0.79835    0.24809   3.218  0.00368 **
## Mud_percent    -0.21316    0.18391  -1.159  0.25785
## BaseEl_m       -0.08859    0.31426  -0.282  0.78043
## BluffEl_m      -0.48522    0.22160  -2.190  0.03851 *
## Seawall        -0.05905    0.15774  -0.374  0.71143

```

```
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1  
##  
## Residual standard error: 0.2641 on 24 degrees of freedom  
## Multiple R-squared:  0.478, Adjusted R-squared:  0.3475  
## F-statistic: 3.662 on 6 and 24 DF, p-value: 0.01006
```

```
predictions <- predict(model1, data)  
mse <- mean((data[[target]] - predictions)^2)  
rsquared <- summary(model1)$r.squared  
cat("(MSE):", mse, "\n")
```

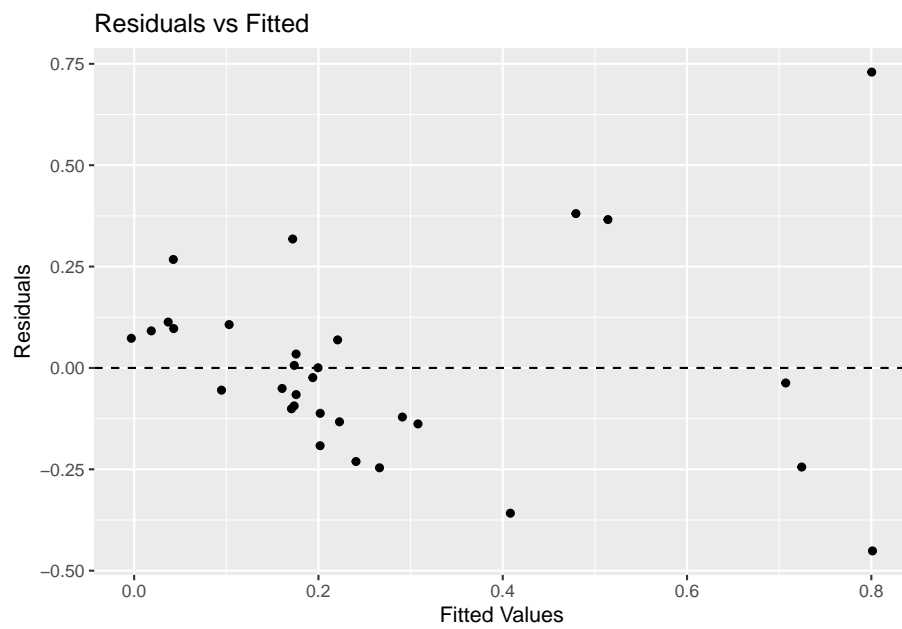
```
## (MSE): 0.05400199
```

```
cat("R square:", rsquared, "\n")
```

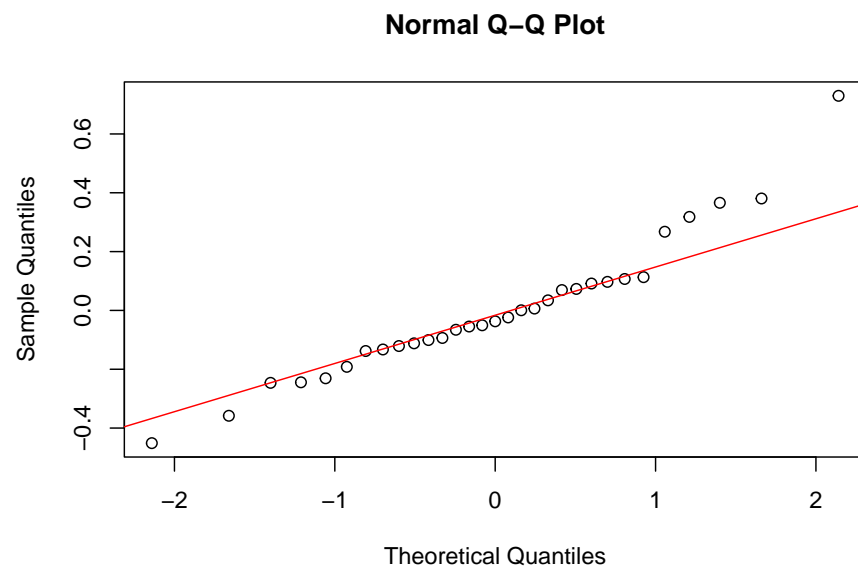
```
## R square: 0.4779682
```

check

```
residuals <- residuals(model1)  
fitted_values <- fitted(model1)  
  
ggplot() +  
  geom_point(aes(x = fitted_values, y = residuals)) +  
  geom_hline(yintercept = 0, linetype = "dashed") +  
  xlab("Fitted Values") +  
  ylab("Residuals") +  
  ggtitle("Residuals vs Fitted")
```

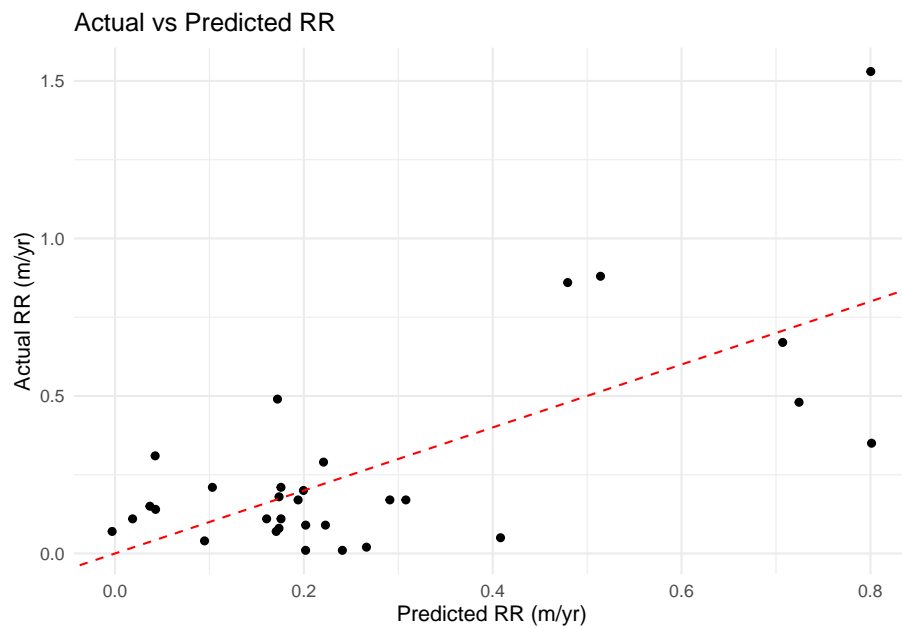


```
qqnorm(residuals)
qqline(residuals, col = "red")
```



```
data$Predicted_RR <- predictions

ggplot(data, aes(x = Predicted_RR, y = RR_m_per_yr)) +
  geom_point() + # Add points
  geom_abline(intercept = 0, slope = 1, color = "red", linetype = "dashed") +
  labs(x = "Predicted RR (m/yr)", y = "Actual RR (m/yr)", title = "Actual vs Predicted")
  theme_minimal()
```



linear regression for all variables(MAX waveheight nne 15m/s)

```
stan <- function(x) {
  return((x - min(x)) / (max(x) - min(x)))
}
data <- read_csv("erosionnne15.csv")
```

```
## Rows: 31 Columns: 8
```

```
## -- Column specification -----
## Delimiter: ","
## chr (1): Bluff
## dbl (7): Orientation (deg), RR (m/yr), Wave Height for NNE wind 15 m/s (m), ...
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

```
names(data) <- gsub(" ", "_", names(data))
names(data) <- gsub("\\(", "", names(data))
names(data) <- gsub("\\)", "", names(data))
names(data) <- gsub("/", "_per_", names(data))
names(data) <- gsub("%", "percent", names(data))
if ("RR (m/yr)" %in% names(data)) {
  names(data)[names(data) == "RR (m/yr)"] <- "RR_m_per_yr"
}
numeric_columns <- sapply(data, is.numeric) & names(data) != "RR_m_per_yr"
data[numeric_columns] <- lapply(data[numeric_columns], stan)
predictors <- setdiff(names(data), c("Bluff", "RR_m_per_yr"))
target <- "RR_m_per_yr"
model_formula <- as.formula(paste(target, "~", paste(predictors, collapse = " + ")))
model2 <- lm(model_formula, data = data)
print(summary(model2))
```

```
##
## Call:
## lm(formula = model_formula, data = data)
##
## Residuals:
```

	Min	1Q	Median	3Q	Max
##	-0.46536	-0.10867	-0.05099	0.06523	0.56911

```
##
## Coefficients:
```

	Estimate	Std. Error	t value	Pr(> t)
##				

```
## (Intercept)                0.10887    0.15424    0.706 0.487101
## Orientation_deg            0.05753    0.14458    0.398 0.694197
## Wave_Height_for_NNE_wind_15_m_per_s_m  1.10008    0.25469    4.319 0.000234 ***
## Mud_percent                -0.10007    0.16959   -0.590 0.560675
## BaseEl_m                   -0.15821    0.28083   -0.563 0.578423
## BluffEl_m                  -0.60463    0.20486   -2.951 0.006962 **
## Seawall                    -0.09229    0.14080   -0.655 0.518401
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.237 on 24 degrees of freedom
## Multiple R-squared:  0.5796, Adjusted R-squared:  0.4744
## F-statistic: 5.514 on 6 and 24 DF,  p-value: 0.001042
```

```
predictions <- predict(model2, data)
mse <- mean((data[[target]] - predictions)^2)
rsquared <- summary(model2)$r.squared
cat("(MSE):", mse, "\n")
```

```
## (MSE): 0.04349345
```

```
cat("R square:", rsquared, "\n")
```

```
## R square: 0.5795532
```

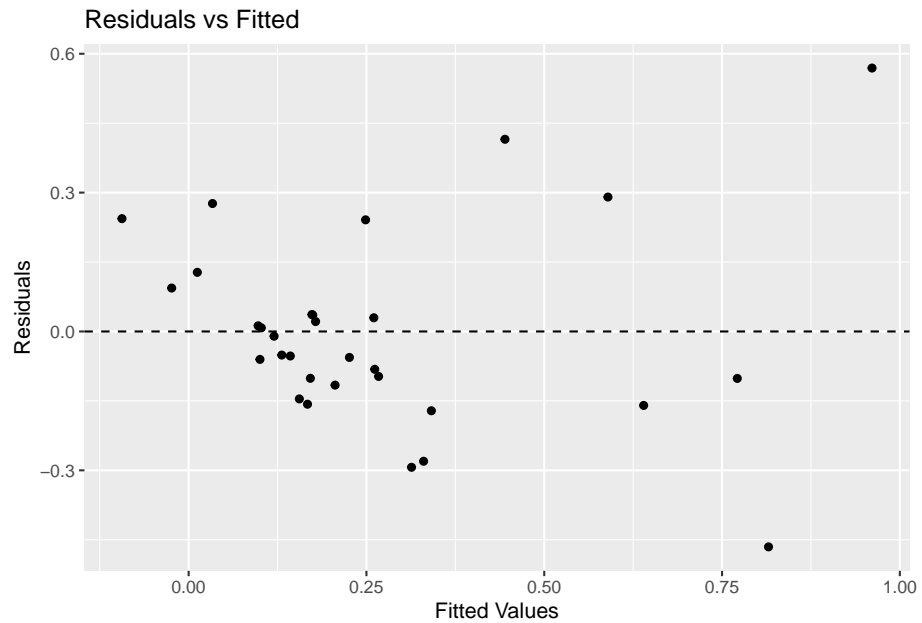
check

```
# 1.residual vs fitted
residuals <- residuals(model2)
fitted_values <- fitted(model2)

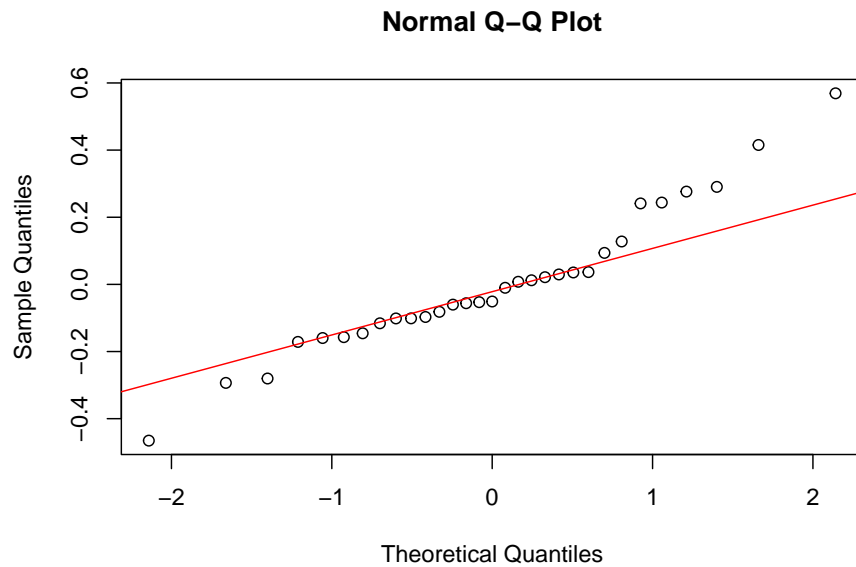
ggplot() +
  geom_point(aes(x = fitted_values, y = residuals)) +
```



```
geom_hline(yintercept = 0, linetype = "dashed") +  
xlab("Fitted Values") +  
ylab("Residuals") +  
ggtitle("Residuals vs Fitted")
```



```
# 2. Q-Q Plot for Normal Distribution of Residuals  
qqnorm(residuals)  
qqline(residuals, col = "red")
```



```
# 3.actual with fitted
data$Predicted_RR <- predictions

ggplot(data, aes(x = Predicted_RR, y = RR_m_per_yr)) +
  geom_point() + # Add points
  geom_abline(intercept = 0, slope = 1, color = "red", linetype = "dashed") + # Add a
  labs(x = "Predicted RR (m/yr)", y = "Actual RR (m/yr)", title = "Actual vs Predicted
  theme_minimal()
```

