**XGBoost** 

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# Installing Packages

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```
rm(list=1s(a11=TRUE))
setwd('^/GitHub/IBD-EDA/aes/')
```

# Loading Data

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```
data <- read.csv('./data_processed/data_log.csv') %>%
  select(-X)
```

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```
set. seed(123)
train_ratio = 0.8

train_indices <- sample(1:nrow(data), size = floor(train_ratio * nrow(data)))
train_data <- data[train_indices, ]
test_data <- data[-train_indices, ]</pre>
```

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```
dtrain <- xgb. DMatrix(
  data = as. matrix(train_data[,-1]), label = train_data[,1]
)

dtest <- xgb. DMatrix(
  data = as. matrix(test_data[,-1]), label = test_data[,1]
)</pre>
```

### **XGBoost**

```
nrounds <- 50
params <- list(
  objective = "reg:squarederror",
  max_depth = 3,
  min_child_weight = 2,
  eta = 0.05,
  gamma = 0.1,
  subsample = 0.7,
  colsample_bytree = 0.8
final_model <- xgboost(</pre>
  data = dtrain,
  params = params,
  nrounds = nrounds,
  print every n = 10,
  early_stopping_rounds = 10,
  eval_metric = "rmse",
  evals = list(validation = dtest)
)
```

```
[16:26:05] WARNING: src/learner.cc:767:
Parameters: { "evals" } are not used.

[1] train-rmse:0.908634
Will train until train_rmse hasn't improved in 10 rounds.

[11] train-rmse:0.709320
[21] train-rmse:0.611662
[31] train-rmse:0.559718
[41] train-rmse:0.534238
[50] train-rmse:0.516972
```

### Model Evaluation

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```
actuals_test <- test_data[,1]
preds_test <- predict(final_model, newdata = as.matrix(test_data[,-1]))

results <- postResample(pred = preds_test, obs = actuals_test)

print(paste("RMSE on Test Set: ", results[1]))</pre>
```

```
[1] "RMSE on Test Set: 0.589553072948926"
```

```
print(paste("MAE on Test Set: ", results[2]))
```

```
[1] "MAE on Test Set: 0.0133120978388201"

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print(paste("R2 on Test Set: ", results[3]))

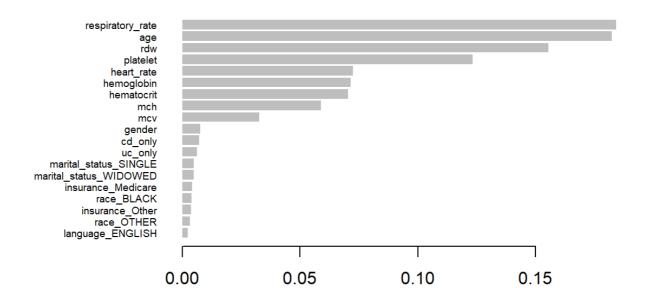
[1] "R2 on Test Set: 0.430741047965029"
```

### Model Results

## Importance Matrix

```
importance_matrix <- xgb.importance(
  feature_names = colnames(train_data[, -1]),
  model = final_model
)

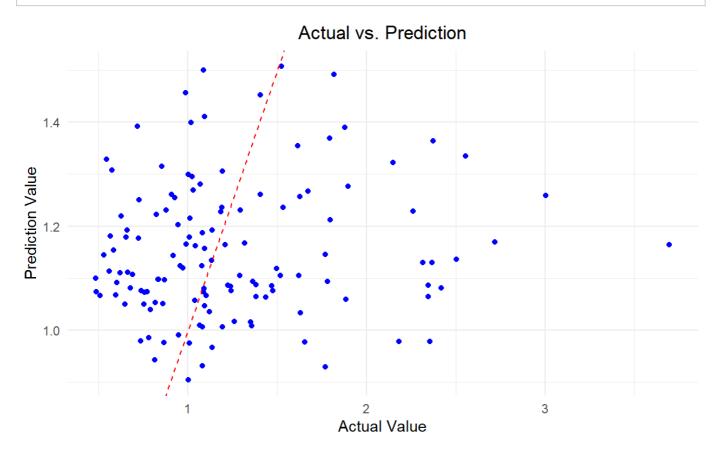
xgb.plot.importance(importance_matrix)</pre>
```



# 实际值与预测值比较

```
comparison_df <- data.frame(
    Actual = actuals_test,
    Prediction = preds_test
)

ggplot(comparison_df, aes(x = Actual, y = Prediction)) +
    geom_point(colour = "blue") + # 绘制散点
    geom_abline(intercept = 0, slope = 1, linetype = "dashed", color = "red") + # 添加等值线
    theme_minimal() + # 使用简洁主题
    labs(title = "Actual vs. Prediction", x = "Actual Value", y = "Prediction Value") + # 添加图
    标题和轴标题
    theme(plot.title = element_text(hjust = 0.5)) # 居中标题
```

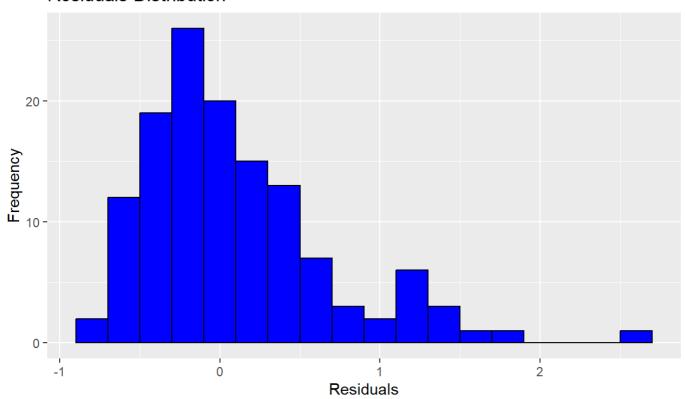


## Residual Analysis

```
residuals <- actuals_test - preds_test

ggplot() +
  geom_histogram(aes(x=residuals), binwidth = 0.2, fill="blue", color="black") +
  ggtitle("Residuals Distribution") +
  xlab("Residuals") +
  ylab("Frequency")</pre>
```

#### Residuals Distribution



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```
ggplot() +
  geom_point(aes(x=preds_test, y=residuals), color="red") +
  ggtitle("Residuals vs. Predicted Values") +
  xlab("Predicted Values") +
  ylab("Residuals") +
  geom_hline(yintercept=0, linetype="dashed", color = "blue")
```

#### Residuals vs. Predicted Values

