report

ZJH

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```
set.seed(100)
tea <- readRDS('nonglin_tea.RDS')</pre>
analyze_data <- readRDS('analyze.RDS')</pre>
numeric_dat <- analyze_data %>% select(.,-c(key,level,observe_ys,observ
e_date:sample_label,FAA:total_catechins,G1L:09L,cutting_height,tea_buds
_cm,year,season,temp_differ))
train_idx <- sample(1:141,70)</pre>
test idx <- !(1:141 %in%train idx)
train <- numeric_dat[train_idx,]</pre>
test <- numeric_dat[test_idx,]</pre>
xtrain <- model.matrix(polyphenol~., train)[,-1]</pre>
ytrain <- train$polyphenol</pre>
ytest <- test$polyphenol</pre>
xtest <- model.matrix(polyphenol~., test)[,-1]</pre>
lambdas_to_try <- 10^seq(-3, 7, length.out = 100)</pre>
lasso_cv <- cv.glmnet(xtrain, ytrain, alpha = 1, lambda = lambdas_to_tr</pre>
y)
plot(lasso_cv)
```

12 12 12 9 7 0 0 0 0 0 0 0 0 0

```
-5 0 5 10 15 Log(λ)
```

```
best_lambda_lasso <- lasso_cv$lambda.min</pre>
best_lambda_lasso
## [1] 0.8497534
lasso_mod <- glmnet(xtrain, ytrain, alpha = 1, lambda = best_lambda_las</pre>
so)
predict.glmnet(lasso_mod, type = 'coefficients')
## 13 x 1 sparse Matrix of class "dgCMatrix"
##
## (Intercept)
                     123.248609229
## duration_oc
                       0.283093707
## Total_leaf
                      -0.773099465
## GORatio
## OTRatio
                     -36.272840050
## leaf_number
                     -14.668985328
## avg_inter_node
                     -10.065369534
## open_plane
## buds_weight_100
                       0.201941770
## acu mean temp
                       0.002401389
## rain
## growth_mean_temp
                       2.704565774
## Growth length
fit <- lm(polyphenol~leaf_number+duration_oc+Total_leaf+acu_mean_temp,d</pre>
ata = test)
```

```
lasso_pred <- predict(lasso_mod, newx = xtest)
lm_pred <- predict(fit)
mean((lasso_pred-ytest)^2)
## [1] 917.5125
mean((lm_pred-ytest)^2)
## [1] 913.1562</pre>
效果仍糟
```