## report

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```
set.seed(100)
tea <- readRDS('nonglin_tea.RDS')</pre>
analyze_data <- readRDS('analyze.RDS')</pre>
library(tidyverse)
numeric_dat <- analyze_data %>% select(.,-c(key,level,observe_ys,observ
e_date:sample_label,FAA:total_catechins,cutting_height,tea_buds_cm,year,
season,temp differ))
train_idx <- sample(1:141,70)</pre>
test idx <- !(1:141 %in%train idx)</pre>
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)
```

## 28 28 27 20 5 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] 5.462277
lasso_mod <- glmnet(xtrain, ytrain, alpha = 1, lambda = best_lambda_las</pre>
so)
predict.glmnet(lasso_mod, type = 'coefficients')
## 31 x 1 sparse Matrix of class "dgCMatrix"
##
## (Intercept)
                      1.192076e+02
                      1.891797e-01
## duration_oc
## Total_leaf
                     -2.571372e-01
## GORatio
## OTRatio
## G1L
## G2L
## G3L
## G4L
## G5L
## G6L
## G7L
                     -1.094133e+00
## G8L
## G9L
## 01L
## 02L
```

```
## 03L
                    -5.807523e-02
## 04L
                    -9.718511e-01
## 05L
## 06L
## 07L
## 08L
                     6.790721e-01
## 09L
## leaf_number
                    -1.937634e+00
## avg_inter_node
## open_plane
## buds_weight_100
## acu_mean_temp 3.435788e-04
## rain
## growth_mean_temp .
## Growth_length
fit <- lm(polyphenol~leaf_number+duration_oc+Total_leaf+G8L+O3L+O4L+O8L</pre>
+acu_mean_temp,data = test)
lasso_pred <- predict(lasso_mod, newx = xtest)</pre>
lm_pred <- predict(fit)</pre>
mean((lasso_pred-ytest)^2)
## [1] 1089.807
mean((lm_pred-ytest)^2)
## [1] 738.1017
```