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## Haoran Hu 2019-5-1

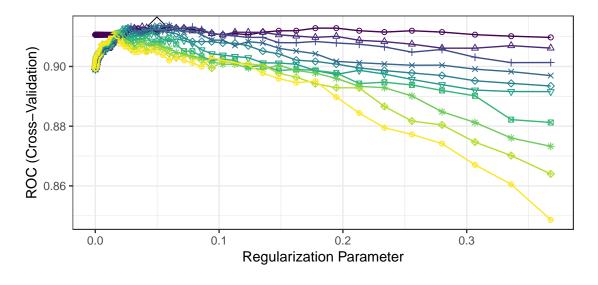
```
heart_disease = read_csv("..\\data\\heart.csv") %>%
    mutate(target=ifelse(target==1, "absence", "presence"))
## Parsed with column specification:
## cols(
##
     age = col_double(),
##
     sex = col_double(),
##
     cp = col_double(),
##
     trestbps = col_double(),
##
     chol = col_double(),
     fbs = col_double(),
##
     restecg = col_double(),
##
##
     thalach = col_double(),
##
     exang = col_double(),
     oldpeak = col_double(),
##
     slope = col double(),
##
##
     ca = col_double(),
##
     thal = col double(),
##
     target = col_double()
## )
set.seed(1)
#trRows = createDataPartition(heart_disease$target, p = .75, list = FALSE)
#train = heart_disease[trRows,]
#test = heart_disease[-trRows,]
heart_disease = heart_disease %>%
    mutate(sex=as.factor(sex),
           cp=as.factor(cp),
           fbs=as.factor(fbs),
           restecg=as.factor(restecg),
           exang=as.factor(exang),
           slope=as.factor(slope),
           thal=as.factor(thal))
model.x <- model.matrix(target~.,heart_disease)[,-1]</pre>
model.y <- heart_disease$target</pre>
# test = test %>%
#
      mutate(sex=as.factor(sex),
#
              cp=as.factor(cp),
#
             fbs=as.factor(fbs),
#
             restecq=as.factor(restecq),
              exang=as.factor(exang),
#
#
             slope=as.factor(slope),
             thal=as.factor(thal))
# test.x <- model.matrix(target~.,test)[,-1]</pre>
# test.y <- test$target</pre>
```

## Regularized logistic

```
ggplot(model.glm, highlight = T) +
    viridis::scale_color_viridis(discrete = TRUE) +
    scale_shape_manual(values = seq(1,10))
```

```
## Scale for 'colour' is already present. Adding another scale for
## 'colour', which will replace the existing scale.
```

## Scale for 'shape' is already present. Adding another scale for 'shape',
## which will replace the existing scale.



```
alpha  

→ 0.00000000  

→ 0.11111111  

→ 0.22222222  

→ 0.33333333  

→ 0.4444444

→ 0.055555556  

→ 0.16666667  

→ 0.27777778  

→ 0.38888889  

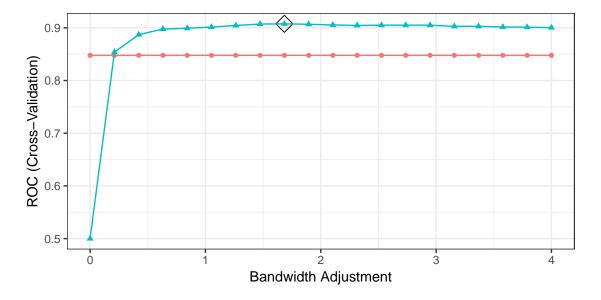
→ 0.5000000
```

```
model.glm$bestTune
```

```
## alpha lambda
## 378 0.1666667 0.04978707
```

## LDA

## Naive bayes



Distribution Type → Gaussian → Nonparametric

```
model.bayes$bestTune
```

```
## fL usekernel adjust
## 29 1 TRUE 1.684211
```