

2_ksvm__support_vector_machines.R

win10

2021-05-14

```
# #####
rm(list = ls())
options(digits = 5)
# if (!is.null(dev.list())){dev.off()}
# #####

library(kernlab) # Kernel-based machine learning methods
library(ggplot2)

##
## Attaching package: 'ggplot2'

## The following object is masked from 'package:kernlab':
##
##      alpha

library(reshape2)
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##      filter, lag

## The following objects are masked from 'package:base':
##
##      intersect, setdiff, setequal, union

# read file. Use credit_card_data.txt
my_data <- read.table(file.choose(), header=FALSE, sep = "", dec=".")

#split data into train-test

# create function for scaling
scale_func = function(a_var) {
  (a_var - min(a_var)) / (max(a_var) - min(a_var))
}

# use lapply to apply function to data
# cast the output of lapply to a data frame
my_data_scaled = as.data.frame(lapply(my_data[, 1:11], scale_func))

# to ensure repeatable results despite random selection
```

```

set.seed(123)

# determine test:train split ratio.
split_ratio = 0.7 # e.g. train:test = 0.7 : (1-0.7)

#split data into test and train
random_sampling = sample(
  1:nrow(my_data_scaled),
  size = nrow(my_data_scaled) * split_ratio,
  replace = FALSE
) #randomly select data.

# capture training and testing, predictors/factors/features
train_data = my_data_scaled[random_sampling, ]
test_data = my_data_scaled[-random_sampling, ]

# capture training and testing, responses
train_results = train_data[, 11]
test_results = test_data[, 11]
# capturing "known" responses is essential for SVM; this is the "supervised" part
# because SVM is a classifier OR supervised type of machine learning

# predictors/features/variables
x_train <- as.matrix(train_data)
x_test <- as.matrix(test_data)

# target/response
y_train <- as.factor(train_results)
y_test <- as.factor(test_results)

#results=list()
results <- vector(mode = "list", length = 1)
C_set_values <- vector(mode = "list", length = 1)
Accuracy_set_values <- vector(mode = "list", length = 1)
Kernel_set_values <- vector(mode = "list", length = 1)

i_index <- 1

df_of_values <- NA # initialize results df

# runs sum, returns result
func_C_val_test <- function(kernel_name, C_value, x, y, data_set_type)
{
  model <- ksvm(y~x, scaled=TRUE, type="C-svc", kernel= kernel_name, C=C_value, kpar="automatic")
  a <- colSums(model@xmatrix[[1]] * model@coef[[1]])
  a0 <- model@b * (-1)
  predict_y <- predict(model, x)
  qty_predict_is_actual <- sum(predict_y == y)
  total_observations <- nrow(my_data)
  match_predict <- qty_predict_is_actual / total_observations
  C_set_values[i_index] <- C_value
  Accuracy_set_values[i_index] <- match_predict
  Kernel_set_values[i_index] <- kernel_name
}

```


[illegible]

```
## Setting default kernel parameters
## Setting default kernel parameters
## Setting default kernel parameters
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```

```
df_of_values
```

##	kernel_name	C_value	match_predict	dataset
## 1	<NA>	NA	NA	<NA>
## 2	rbfdot	1e-06	0.37920	training
## 3	rbfdot	1e-05	0.37920	training
## 4	rbfdot	1e-04	0.37920	training
## 5	rbfdot	1e-03	0.37920	training
## 6	rbfdot	1e-01	0.68960	training
## 7	rbfdot	1e+00	0.69878	training
## 8	rbfdot	1e+01	0.69878	training
## 9	rbfdot	1e+02	0.69878	training
## 10	rbfdot	1e+03	0.69878	training
## 11	rbfdot	1e+04	0.69878	training
## 12	polydot	1e-06	0.37920	training
## 13	polydot	1e-05	0.37920	training
## 14	polydot	1e-04	0.37920	training
## 15	polydot	1e-03	0.67584	training
## 16	polydot	1e-01	0.69878	training
## 17	polydot	1e+00	0.69878	training
## 18	polydot	1e+01	0.69878	training
## 19	polydot	1e+02	0.69878	training
## 20	polydot	1e+03	0.69878	training
## 21	polydot	1e+04	0.69878	training
## 22	vanilladot	1e-06	0.37920	training
## 23	vanilladot	1e-05	0.37920	training
## 24	vanilladot	1e-04	0.37920	training
## 25	vanilladot	1e-03	0.67584	training
## 26	vanilladot	1e-01	0.69878	training
## 27	vanilladot	1e+00	0.69878	training
## 28	vanilladot	1e+01	0.69878	training
## 29	vanilladot	1e+02	0.69878	training
## 30	vanilladot	1e+03	0.69878	training
## 31	vanilladot	1e+04	0.69878	training
## 32	tanhdot	1e-06	0.37920	training
## 33	tanhdot	1e-05	0.37920	training
## 34	tanhdot	1e-04	0.37920	training
## 35	tanhdot	1e-03	0.37920	training
## 36	tanhdot	1e-01	0.64526	training
## 37	tanhdot	1e+00	0.65443	training
## 38	tanhdot	1e+01	0.64985	training
## 39	tanhdot	1e+02	0.63761	training
## 40	tanhdot	1e+03	0.64679	training
## 41	tanhdot	1e+04	0.64679	training
## 42	laplacedot	1e-06	0.37920	training
## 43	laplacedot	1e-05	0.37920	training
## 44	laplacedot	1e-04	0.37920	training
## 45	laplacedot	1e-03	0.37920	training

## 46	laplacedot	1e-01	0.68807	training
## 47	laplacedot	1e+00	0.69878	training
## 48	laplacedot	1e+01	0.69878	training
## 49	laplacedot	1e+02	0.69878	training
## 50	laplacedot	1e+03	0.69878	training
## 51	laplacedot	1e+04	0.69878	training
## 52	besseldot	1e-06	0.37920	training
## 53	besseldot	1e-05	0.37920	training
## 54	besseldot	1e-04	0.37920	training
## 55	besseldot	1e-03	0.37920	training
## 56	besseldot	1e-01	0.68349	training
## 57	besseldot	1e+00	0.69419	training
## 58	besseldot	1e+01	0.69878	training
## 59	besseldot	1e+02	0.69878	training
## 60	besseldot	1e+03	0.69878	training
## 61	besseldot	1e+04	0.69878	training
## 62	anovadot	1e-06	0.37920	training
## 63	anovadot	1e-05	0.37920	training
## 64	anovadot	1e-04	0.37920	training
## 65	anovadot	1e-03	0.50153	training
## 66	anovadot	1e-01	0.69878	training
## 67	anovadot	1e+00	0.69878	training
## 68	anovadot	1e+01	0.69878	training
## 69	anovadot	1e+02	0.69878	training
## 70	anovadot	1e+03	0.69878	training
## 71	anovadot	1e+04	0.69878	training
## 72	splinedot	1e-06	0.38685	training
## 73	splinedot	1e-05	0.40367	training
## 74	splinedot	1e-04	0.44648	training
## 75	splinedot	1e-03	0.69878	training
## 76	splinedot	1e-01	0.69878	training
## 77	splinedot	1e+00	0.69878	training
## 78	splinedot	1e+01	0.69878	training
## 79	splinedot	1e+02	0.69878	training
## 80	splinedot	1e+03	0.69878	training
## 81	splinedot	1e+04	0.69878	training
## 82	<NA>	NA	NA	<NA>
## 83	rbfdot	1e-06	0.16820	Test
## 84	rbfdot	1e-05	0.16820	Test
## 85	rbfdot	1e-04	0.16820	Test
## 86	rbfdot	1e-03	0.16820	Test
## 87	rbfdot	1e-01	0.29511	Test
## 88	rbfdot	1e+00	0.30122	Test
## 89	rbfdot	1e+01	0.30122	Test
## 90	rbfdot	1e+02	0.30122	Test
## 91	rbfdot	1e+03	0.30122	Test
## 92	rbfdot	1e+04	0.30122	Test
## 93	polydot	1e-06	0.16820	Test
## 94	polydot	1e-05	0.16820	Test
## 95	polydot	1e-04	0.16820	Test
## 96	polydot	1e-03	0.20183	Test
## 97	polydot	1e-01	0.30122	Test
## 98	polydot	1e+00	0.30122	Test
## 99	polydot	1e+01	0.30122	Test

## 100	polydot	1e+02	0.30122	Test
## 101	polydot	1e+03	0.30122	Test
## 102	polydot	1e+04	0.30122	Test
## 103	vanilladot	1e-06	0.16820	Test
## 104	vanilladot	1e-05	0.16820	Test
## 105	vanilladot	1e-04	0.16820	Test
## 106	vanilladot	1e-03	0.20183	Test
## 107	vanilladot	1e-01	0.30122	Test
## 108	vanilladot	1e+00	0.30122	Test
## 109	vanilladot	1e+01	0.30122	Test
## 110	vanilladot	1e+02	0.30122	Test
## 111	vanilladot	1e+03	0.30122	Test
## 112	vanilladot	1e+04	0.30122	Test
## 113	tanhdot	1e-06	0.16820	Test
## 114	tanhdot	1e-05	0.16820	Test
## 115	tanhdot	1e-04	0.16820	Test
## 116	tanhdot	1e-03	0.16820	Test
## 117	tanhdot	1e-01	0.27829	Test
## 118	tanhdot	1e+00	0.26147	Test
## 119	tanhdot	1e+01	0.27217	Test
## 120	tanhdot	1e+02	0.27064	Test
## 121	tanhdot	1e+03	0.26758	Test
## 122	tanhdot	1e+04	0.26758	Test
## 123	laplacedot	1e-06	0.16820	Test
## 124	laplacedot	1e-05	0.16820	Test
## 125	laplacedot	1e-04	0.16820	Test
## 126	laplacedot	1e-03	0.16820	Test
## 127	laplacedot	1e-01	0.29205	Test
## 128	laplacedot	1e+00	0.30122	Test
## 129	laplacedot	1e+01	0.30122	Test
## 130	laplacedot	1e+02	0.30122	Test
## 131	laplacedot	1e+03	0.30122	Test
## 132	laplacedot	1e+04	0.30122	Test
## 133	besseldot	1e-06	0.16820	Test
## 134	besseldot	1e-05	0.16820	Test
## 135	besseldot	1e-04	0.16820	Test
## 136	besseldot	1e-03	0.16820	Test
## 137	besseldot	1e-01	0.29205	Test
## 138	besseldot	1e+00	0.29969	Test
## 139	besseldot	1e+01	0.30122	Test
## 140	besseldot	1e+02	0.30122	Test
## 141	besseldot	1e+03	0.30122	Test
## 142	besseldot	1e+04	0.30122	Test
## 143	anovadot	1e-06	0.16820	Test
## 144	anovadot	1e-05	0.16820	Test
## 145	anovadot	1e-04	0.16820	Test
## 146	anovadot	1e-03	0.16820	Test
## 147	anovadot	1e-01	0.30122	Test
## 148	anovadot	1e+00	0.30122	Test
## 149	anovadot	1e+01	0.30122	Test
## 150	anovadot	1e+02	0.30122	Test
## 151	anovadot	1e+03	0.30122	Test
## 152	anovadot	1e+04	0.30122	Test
## 153	splinedot	1e-06	0.17737	Test

## 154	splinedot	1e-05	0.17737	Test
## 155	splinedot	1e-04	0.19878	Test
## 156	splinedot	1e-03	0.29664	Test
## 157	splinedot	1e-01	0.30122	Test
## 158	splinedot	1e+00	0.30122	Test
## 159	splinedot	1e+01	0.30122	Test
## 160	splinedot	1e+02	0.30122	Test
## 161	splinedot	1e+03	0.30122	Test
## 162	splinedot	1e+04	0.30122	Test