## Universidad de Santiago de Chile Facultad de Ingeniería Depto. de Ingeniería Informática



# Taller de minería de datos avanzada Capítulo V "Máquinas de vectores soporte"

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### **Ejemplos**

- SVM en clasificación
- Estimación con SVM
- Tunning con SVM
  - Kernel Lineal
  - Kernel RBF



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#### e1071

 Librería de procesamiento y análisis estadístico que entre otras funciones permite trabajar con maquinas de vectores soporte

```
require(e1071)
data(iris)
attach(iris)
                                                  > summary(model)
formula=Species ~ .
model <- svm(formula, data = iris)
                                                  Call:
                                                  svm(formula = formula, data = iris)
> print(model)
                                                  Parameters:
                                                     SVM-Type: C-classification
Call:
                                                   SVM-Kernel: radial
svm(formula = formula, data = iris)
                                                         cost:
                                                        gamma: 0.25
Parameters:
                                                  Number of Support Vectors:
                                                                               51
   SVM-Type: C-classification
             radial
 SVM-Kernel:
                                                   (8 22 21)
       cost:
              0.25
      gamma:
                                                  Number of Classes: 3
Number of Support Vectors: 51
                                                  Levels:
                                                   setosa versicolor virginica
```

- $\rightarrow$  x <- subset(iris, select = -Species)
- > y <- Species
- $\triangleright$  model <- svm(x, y)
- # test with train data
- pred <- predict(model, x)</pre>
- > table(pred, y)

# compute decision values and probabilities:

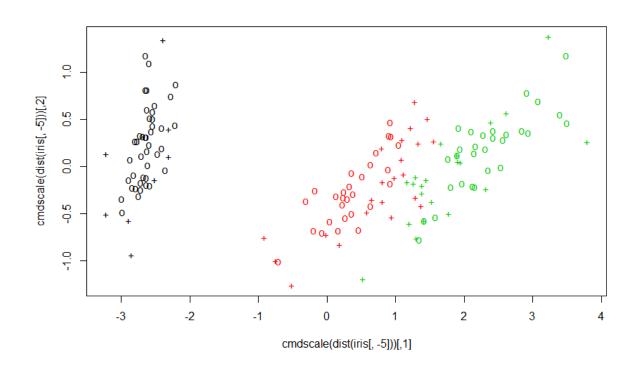
- pred <- predict(model, x, decision.values = TRUE)</pre>
- > attr(pred, "decision.values")[1:4,]



# visualize (classes by color, SV by crosses):

ightharpoonup plot(cmdscale(dist(iris[,-5])), col = as.integer(iris[,5]), pch = c("o","+")[1:150 %in% model\$index + 1])





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- summary(obj)

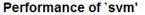
```
> summary(obj)
Parameter tuning of 'svm':
- sampling method: 2-fold cross validation
- best parameters:
 cost
 0.5
- best performance: 0.03333333
- Detailed performance results:
            error dispersion
  cost
1 0.5 0.03333333 0.00942809
2 1.0 0.05333333 0.00000000
3 2.0 0.04000000 0.00000000
4 4.0 0.03333333 0.00942809
  8.0 0.05333333 0.01885618
6 16.0 0.05333333 0.01885618
```

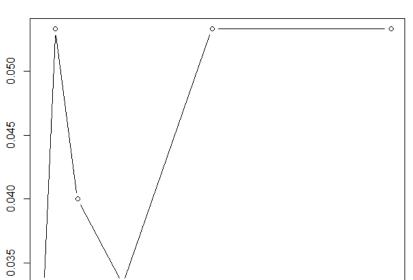
- plot(obj)

Levels:

setosa versicolor virginica

summary(obj\$best.model)





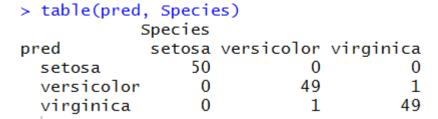
10

15



```
> summary(obj$best.model)
Call:
best.tune(method = svm, train.x = Species \sim ., data = iris, ranges = list(cost = 2\wedge(-1:4)),
    tunecontrol = tune.control(sampling = "cross", cross = 2), kernel = "linear")
Parameters:
   SVM-Type: C-classification
 SVM-Kernel:
              linear
             0.5
       cost:
      gamma: 0.25
Number of Support Vectors: 36
 (2 17 17)
Number of Classes: 3
```

- pred <- predict(obj\$best.model, x)</pre>
- > table(pred, Species)





- summary(obj)

```
> summary(obj)
Parameter tuning of 'svm':
- sampling method: 2-fold cross validation
best parameters:
 gamma cost
 0.25 2

    best performance: 0.05333333

- Detailed performance results:
  gamma cost error dispersion
1 0.25 0.5 0.07333333 0.02828427
2 0.50 0.5 0.08000000 0.03771236
```



- pred <- predict(obj\$best.model, x)</pre>
- > table(pred, Species)
  - > table(pred, Species)

```
Species
pred setosa versicolor virginica
setosa 50 0 0
versicolor 0 48 2
virginica 0 2 48
```

by obj <- tune(svm, Species~., data = iris, kernel = "radial", ranges = list(gamma = 2^(-7:12), cost = 2^(-7:14), tunecontrol = tune.control(sampling = "cross", cross = 2)))

```
> summary(obj)
```

Parameter tuning of 'svm':

- sampling method: 2-fold cross validation
- best parameters:gamma cost0.0625 4
- best performance: 0.01333333
- Detailed performance results:
- gamma cost error dispersion 1 7.8125e-03 7.8125e-03 0.72000000 0.01885618
- 2 1.5625e-02 7.8125e-03 0.72000000 0.01885618



- pred <- predict(obj\$best.model, x)</pre>
- ➤ table(pred, Species)

