

Kernels SVM

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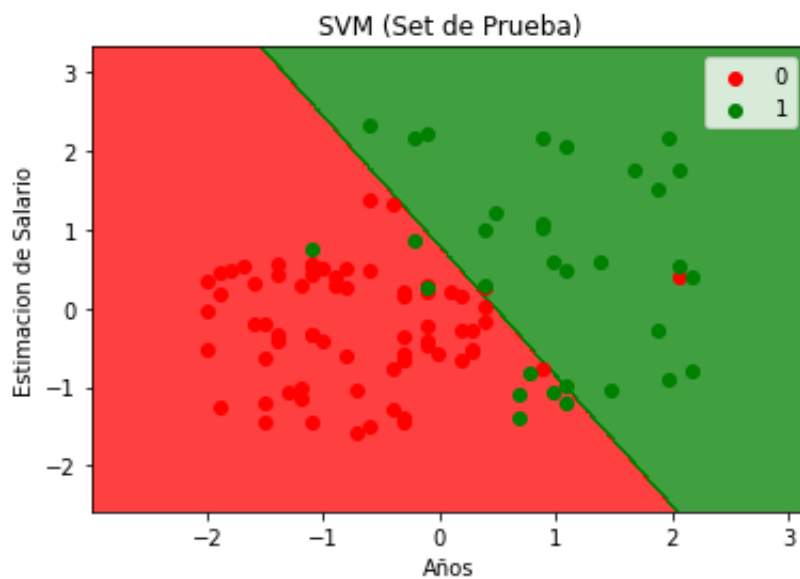
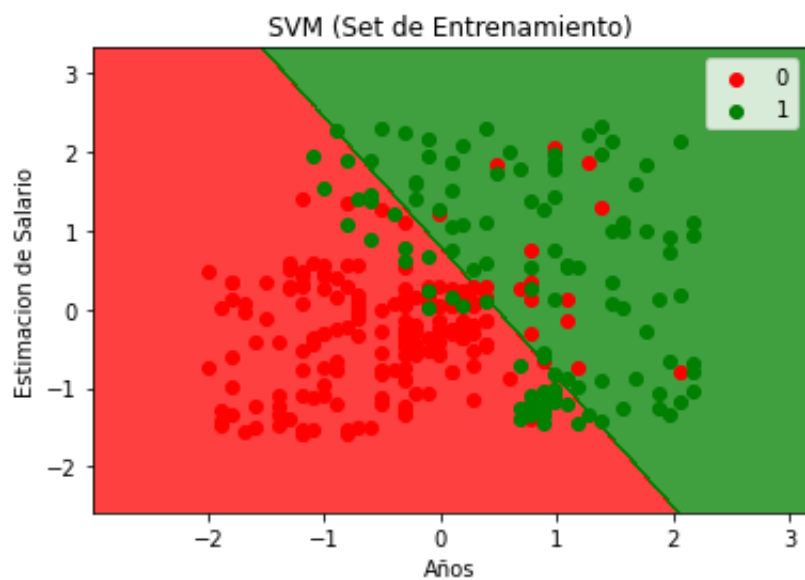
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SVM

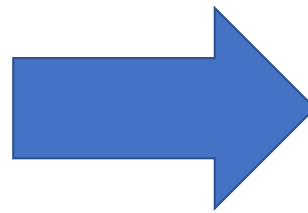
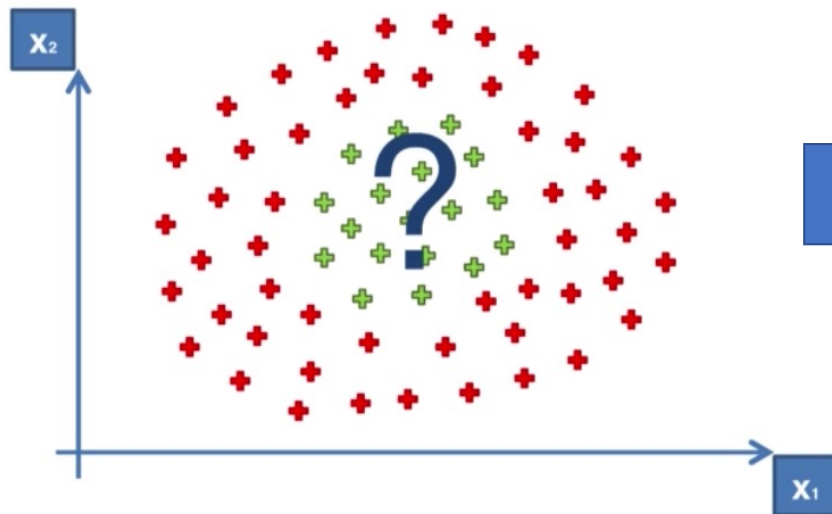
```
# Entrenamiento SVM
from sklearn.svm import SVC
classifier = SVC(kernel = 'linear', random_state = 0)
classifier.fit(X_train, y_train)
```

Resultados



SVM

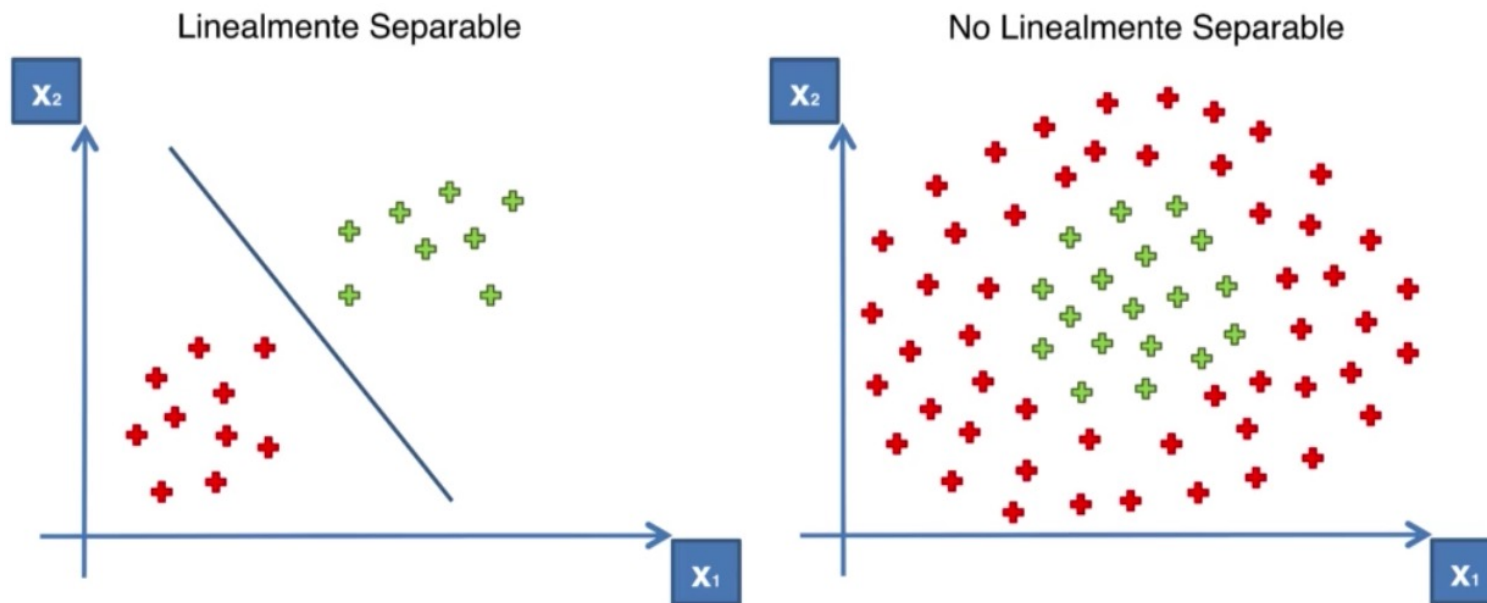
Por que Utilizarlos



Porque esta nube de puntos no es
LINEALMENTE SEPARABLE

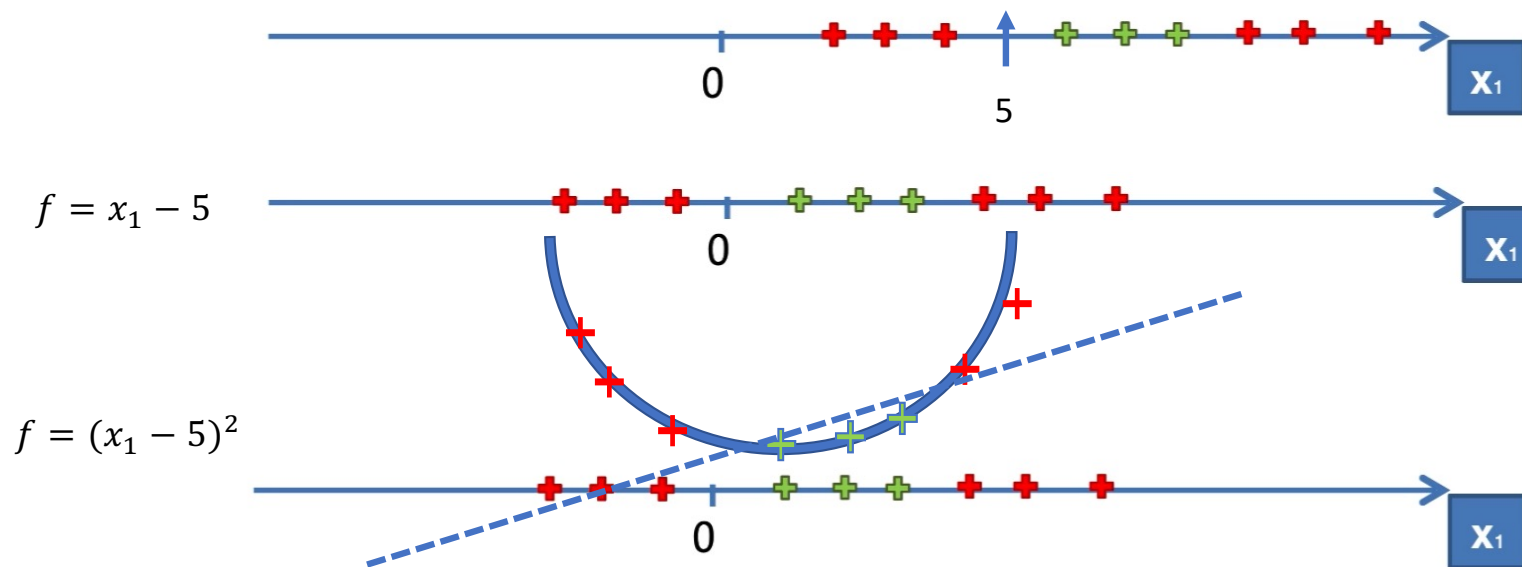
SVM

Por que Utilizarlos

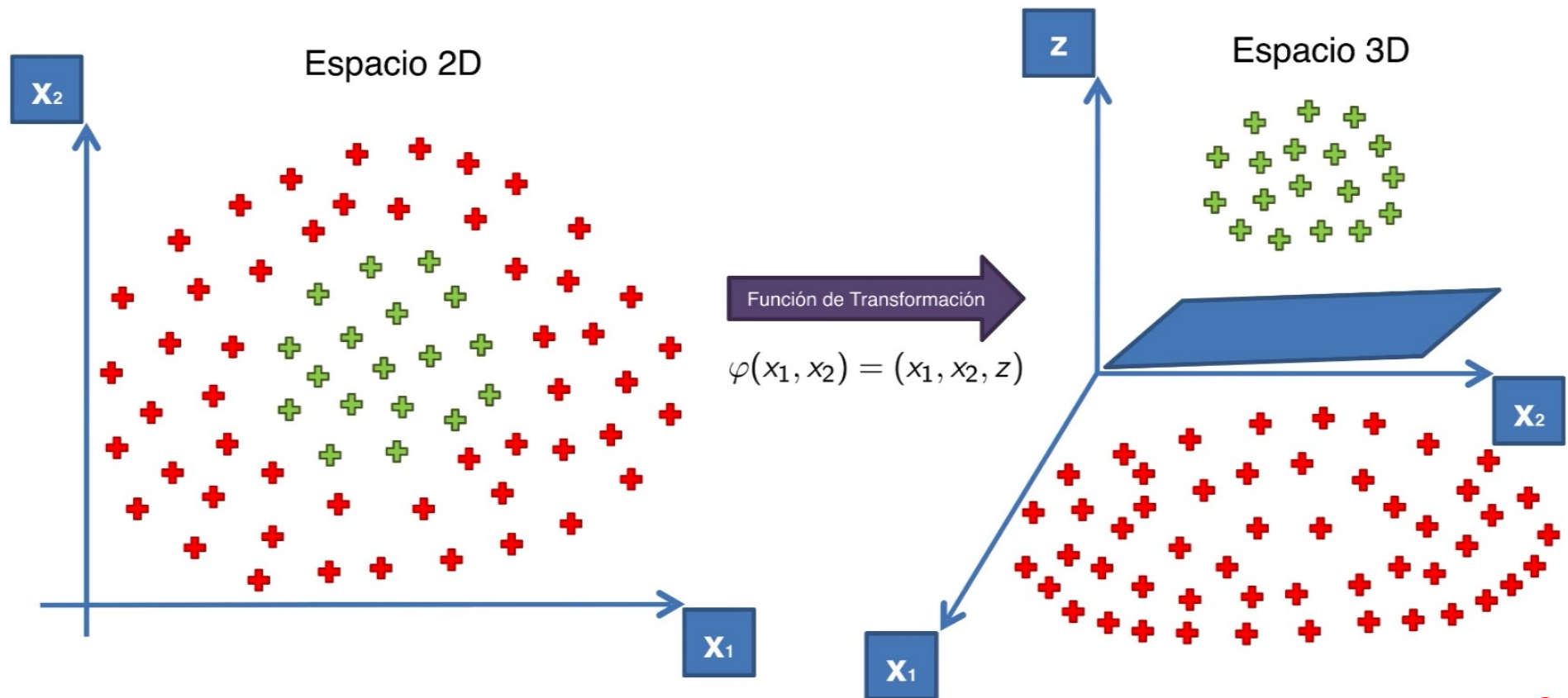


Idea del uso de Kernel

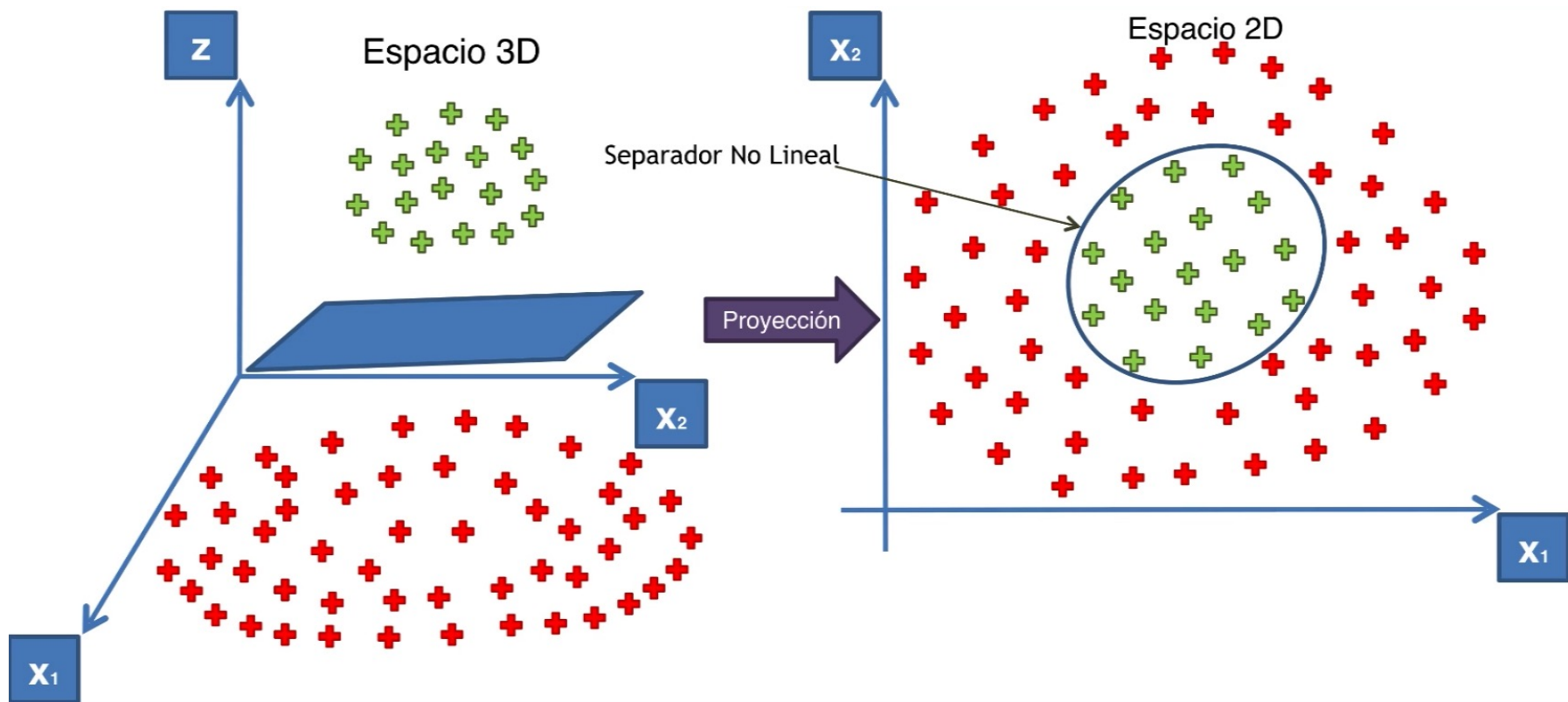
- Aumentando dimensiones



SVM a una Dimensión Superior



SVM a una Dimensión Superior



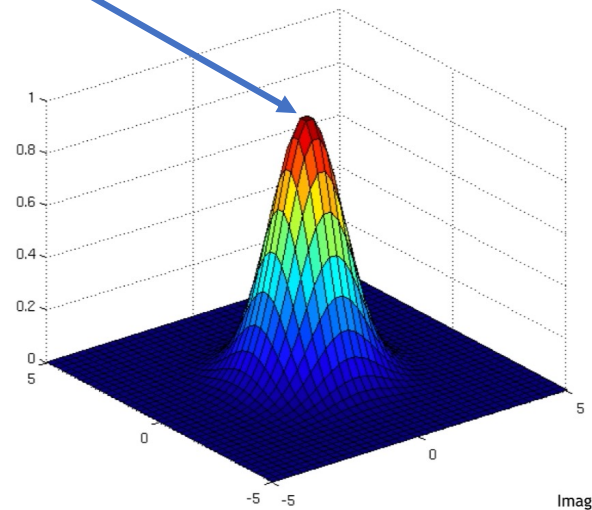
SVM a una Dimensión Superior

Transformar una variable a un
Espacio de Dimension
Superior puede ser muy
costoso computacionalmente

Kernel Gaussiano RBF

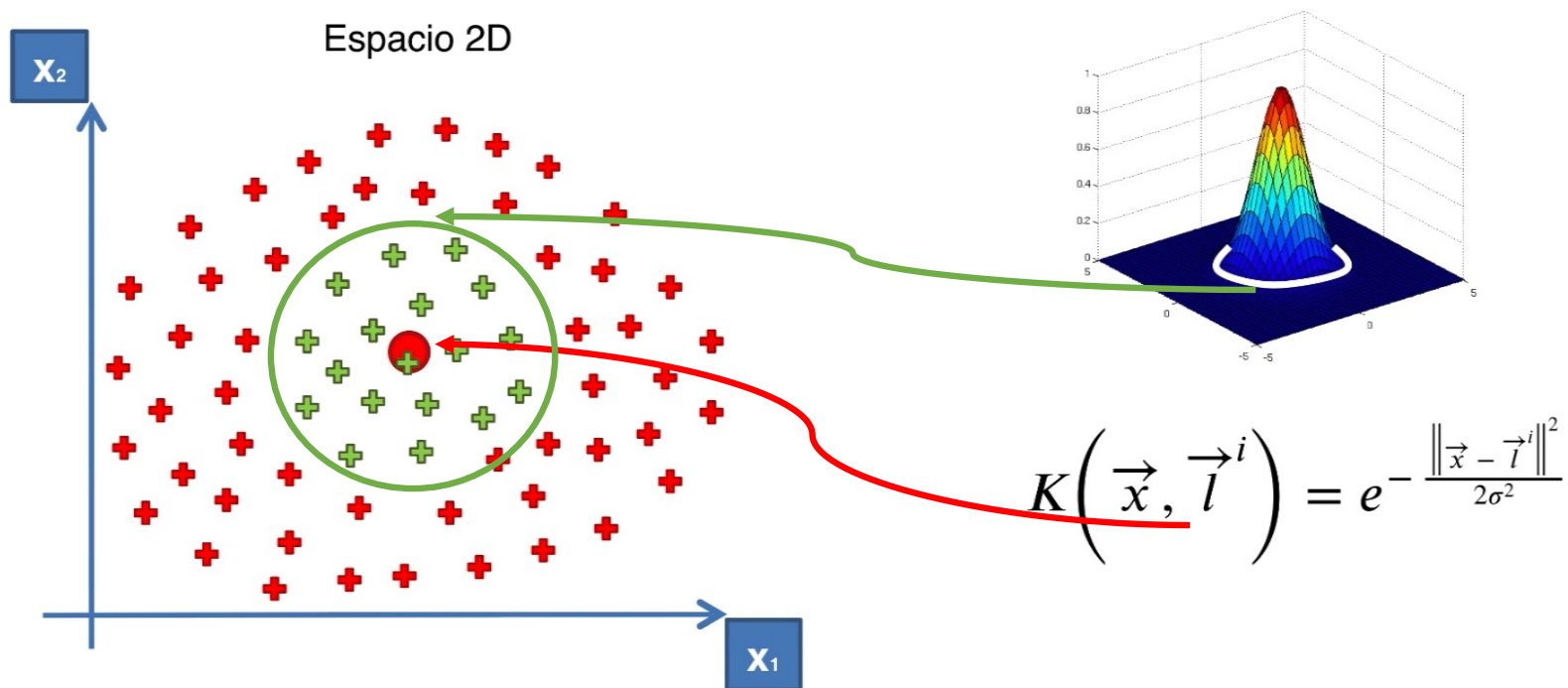
$$K\left(\vec{x}, \vec{l}^i\right) = e^{-\frac{\|\vec{x} - \vec{l}^i\|^2}{2\sigma^2}}$$

Media= Punto Central

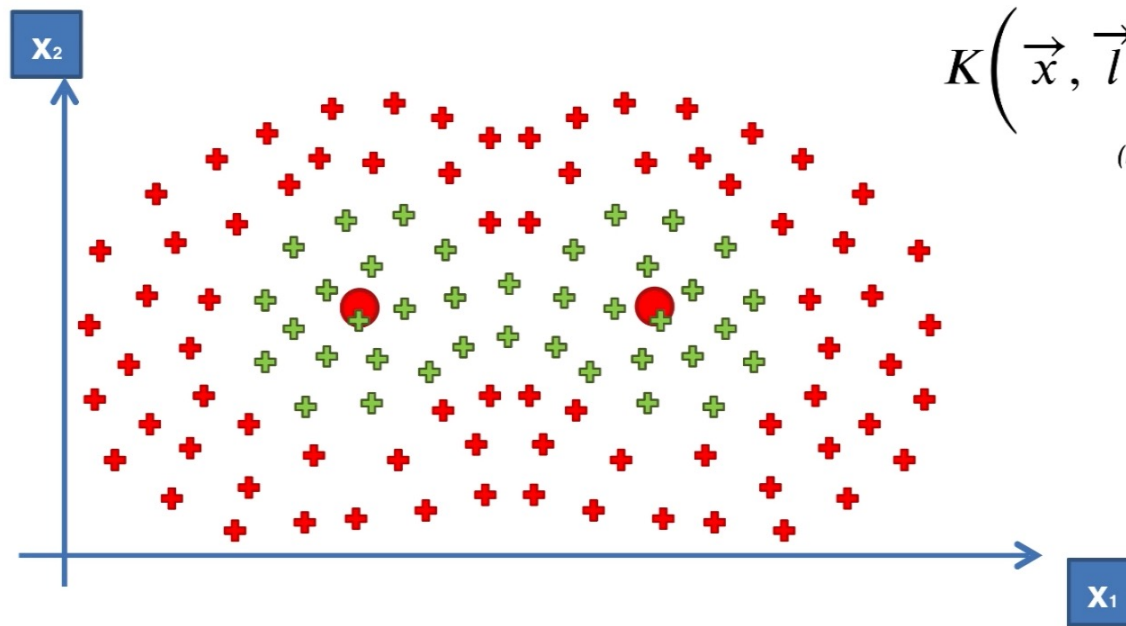


Kernel Gaussiano RBF

Media= Punto Central



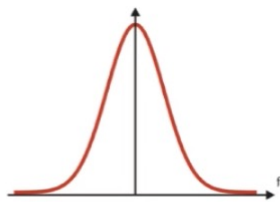
Kernel Gaussiano RBF



$$K\left(\vec{x}, \vec{l}^1\right) + K\left(\vec{x}, \vec{l}^2\right)$$

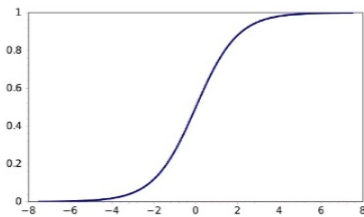
(Fórmula Simplificada)

Típos de kernel



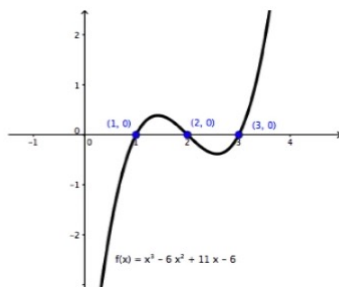
Kernel Gaussiano RBF

$$K\left(\vec{x}, \vec{l}^i\right) = e^{-\frac{\|\vec{x} - \vec{l}^i\|^2}{2\sigma^2}}$$



Kernel Sigmoide

$$K(X, Y) = \tanh(\gamma \cdot X^T Y + r)$$



Kernel Polinómico

$$K(X, Y) = (\gamma \cdot X^T Y + r)^d, \gamma > 0$$

sklearn.svm.SVC

```
class sklearn.svm.SVC(*, C=1.0, kernel='rbf', degree=3, gamma='scale', coef0=0.0, shrinking=True, probability=False,
tol=0.001, cache_size=200, class_weight=None, verbose=False, max_iter=-1, decision_function_shape='ovr', break_ties=False,
random_state=None)
```

[\[source\]](#)

C : float, default=1.0

Regularization parameter. The strength of the regularization is inversely proportional to C. Must be strictly positive. The penalty is a squared l2 penalty.

kernel : {'linear', 'poly', 'rbf', 'sigmoid', 'precomputed'} or callable, default='rbf'

Specifies the kernel type to be used in the algorithm. If none is given, 'rbf' will be used. If a callable is given it is used to pre-compute the kernel matrix from data matrices; that matrix should be an array of shape (n_samples, n_samples).

degree : int, default=3

Degree of the polynomial kernel function ('poly'). Ignored by all other kernels.

gamma : {'scale', 'auto'} or float, default='scale'

Kernel coefficient for 'rbf', 'poly' and 'sigmoid'.

- if `gamma='scale'` (default) is passed then it uses $1 / (n_features * X.var())$ as value of gamma,
- if 'auto', uses $1 / n_features$.

Changed in version 0.22: The default value of `gamma` changed from 'auto' to 'scale'.

coef0 : float, default=0.0

Independent term in kernel function. It is only significant in 'poly' and 'sigmoid'.

shrinking : bool, default=True

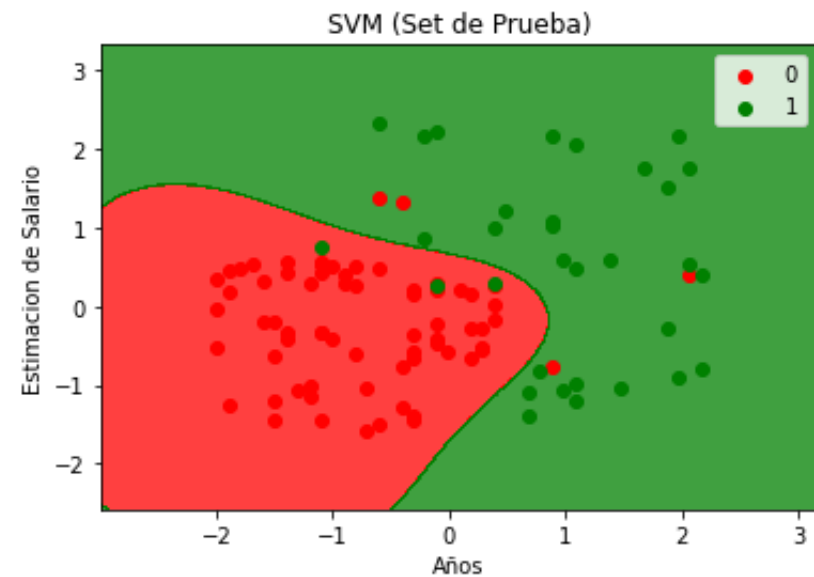
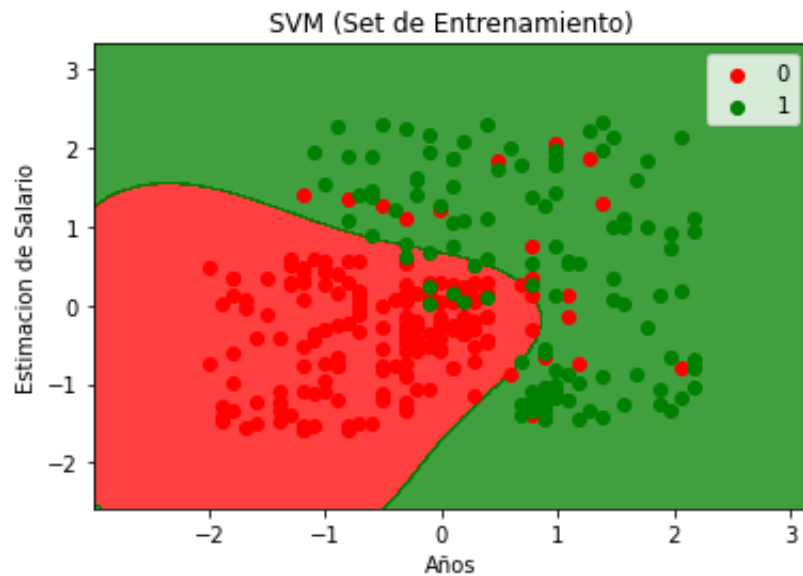
Whether to use the shrinking heuristic. See the [User Guide](#).

sklearn.svm.SVC

```
class sklearn.svm.SVC(*, C=1.0, kernel='rbf', degree=3, gamma='scale', coef0=0.0, shrinking=True, probability=False,
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random_state=None)
```

[\[source\]](#)

```
# Entrenamiento SVM
from sklearn.svm import SVC
classifier = SVC(kernel = 'rbf', random_state = 0)
classifier.fit(X_train, y_train)
```



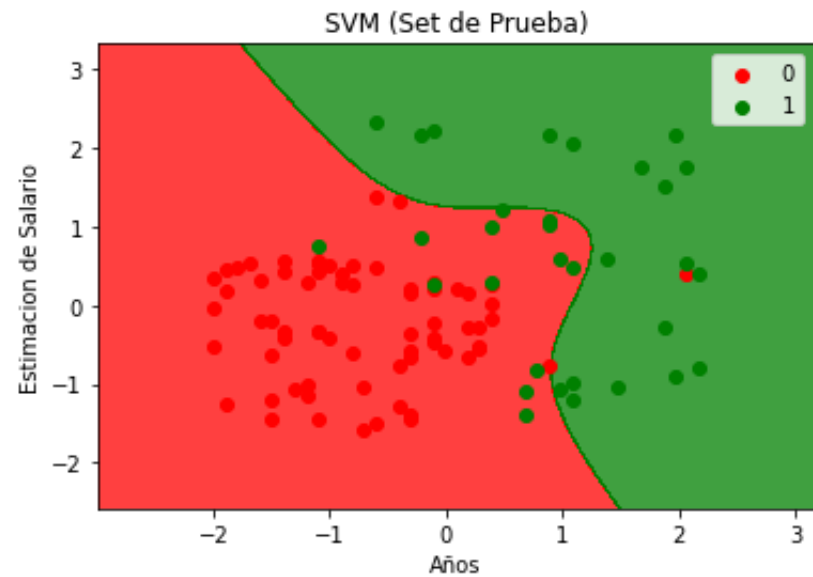
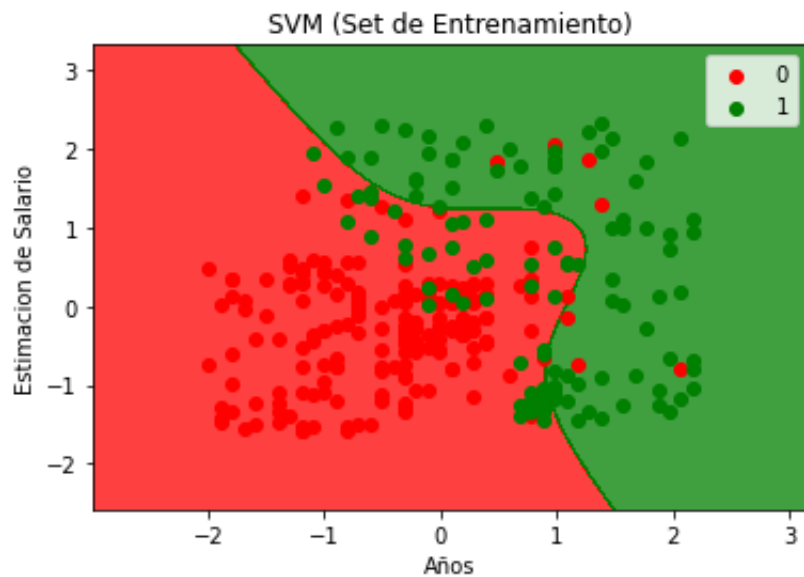
sklearn.svm.SVC

```
class sklearn.svm.SVC(*, C=1.0, kernel='rbf', degree=3, gamma='scale', coef0=0.0, shrinking=True, probability=False,
tol=0.001, cache_size=200, class_weight=None, verbose=False, max_iter=-1, decision_function_shape='ovr', break_ties=False,
random_state=None)
```

[\[source\]](#)

Entrenamiento SVM

```
from sklearn.svm import SVC
classifier = SVC(kernel = 'poly', random_state = 0)
classifier.fit(X_train, y_train)
```

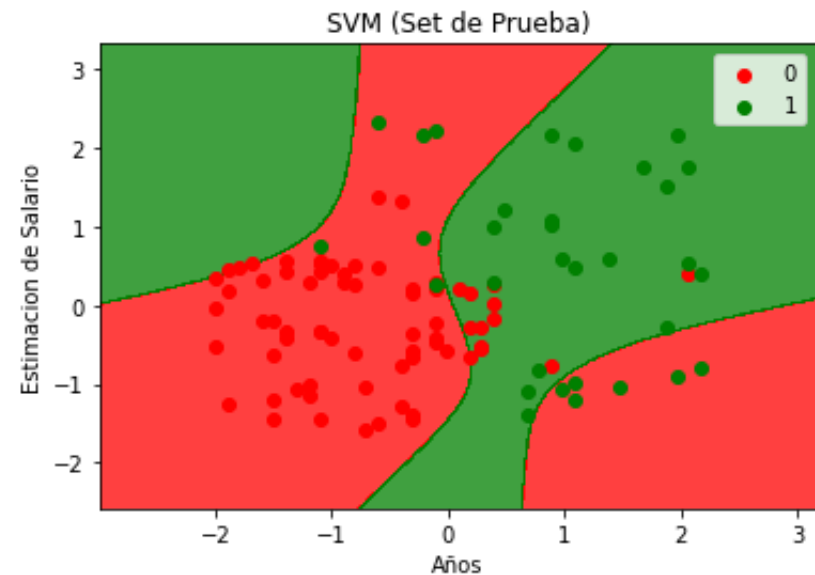
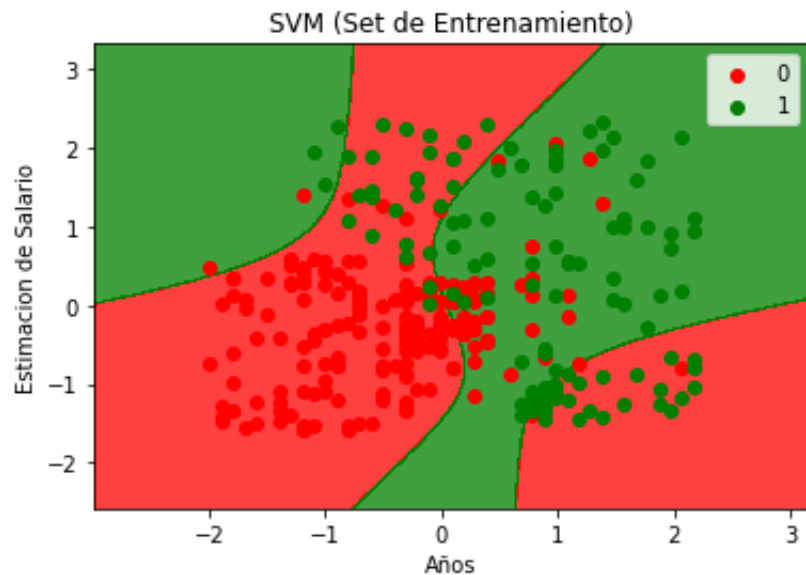


sklearn.svm.SVC

```
class sklearn.svm.SVC(*, C=1.0, kernel='rbf', degree=3, gamma='scale', coef0=0.0, shrinking=True, probability=False,
tol=0.001, cache_size=200, class_weight=None, verbose=False, max_iter=-1, decision_function_shape='ovr', break_ties=False,
random_state=None)
```

[\[source\]](#)

```
# Entrenamiento SVM
from sklearn.svm import SVC
classifier = SVC(kernel = 'sigmoid', random_state = 0)
classifier.fit(X_train, y_train)
```



Continuara.....

