

costobayesiano

August 14, 2018

```
In [1]: import numpy as np
```

```
In [2]: import pandas as pd
```

```
In [3]: import matplotlib.pyplot as plt
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```
In [4]: from sklearn.model_selection import cross_val_predict, StratifiedKFold, GridSearchCV, tra
```

```
In [5]: from sklearn.metrics import roc_auc_score
```

```
In [6]: from sklearn import tree
```

```
In [7]: from sklearn import datasets #Importamos el conjunto de datos
```

```
In [8]: from sklearn.model_selection import train_test_split
```

```
In [9]: np.random.seed(0)
```

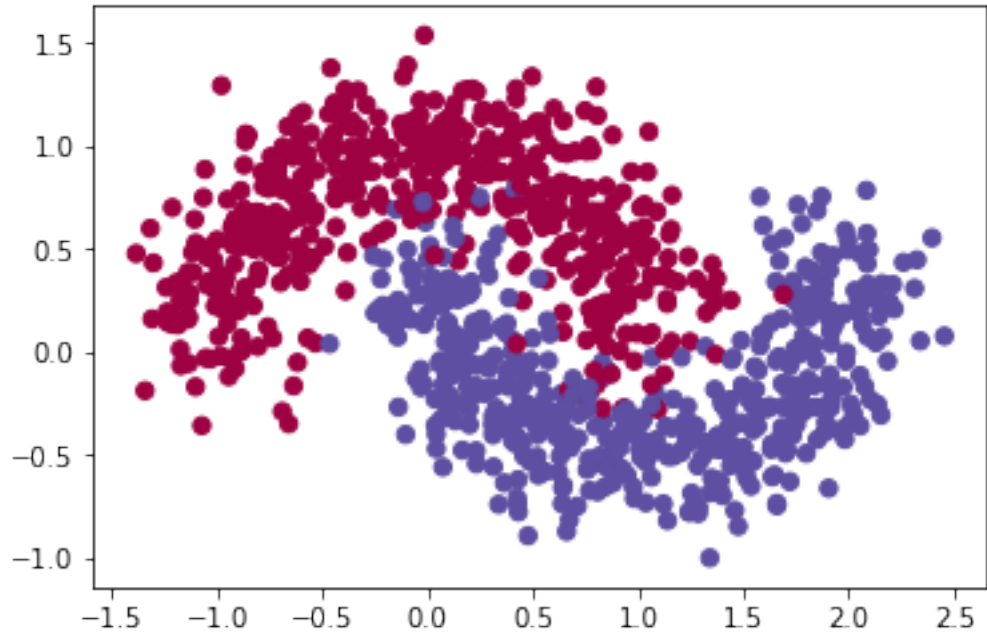
```
In [10]: X, y = datasets.make_moons(1000, noise=0.20)
```

```
In [11]: #Dividimos nuestros datos en "conjunto de entrenamiento y de prueba
```

```
In [12]: X_train, X_test, y_train, y_test = train_test_split(X, y)
```

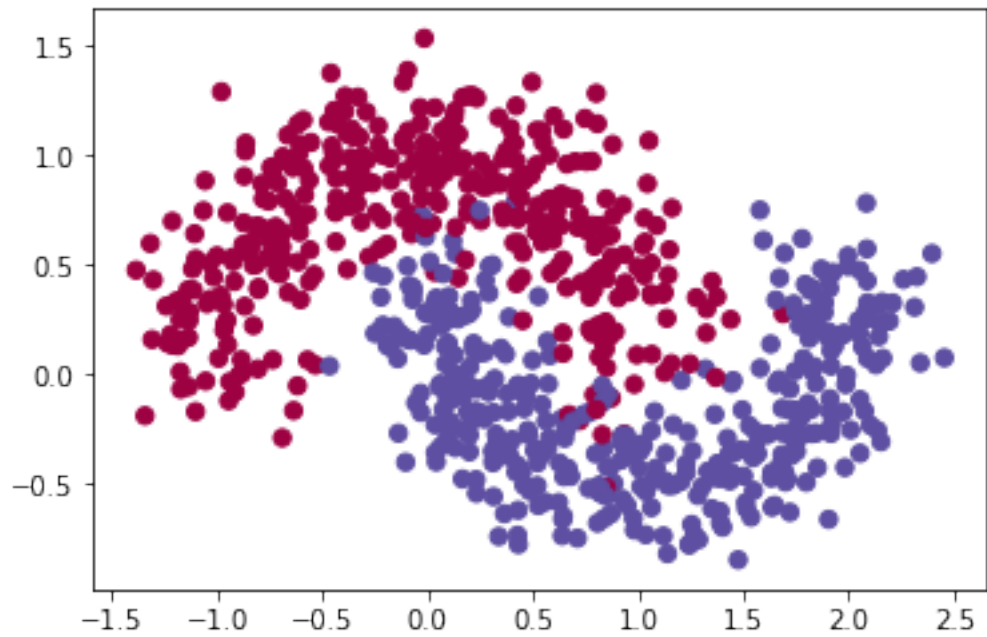
```
In [13]: plt.scatter(X[:,0], X[:,1], s=40, c=y, cmap=plt.cm.Spectral)
```

```
Out[13]: <matplotlib.collections.PathCollection at 0x1a0940c0d0>
```



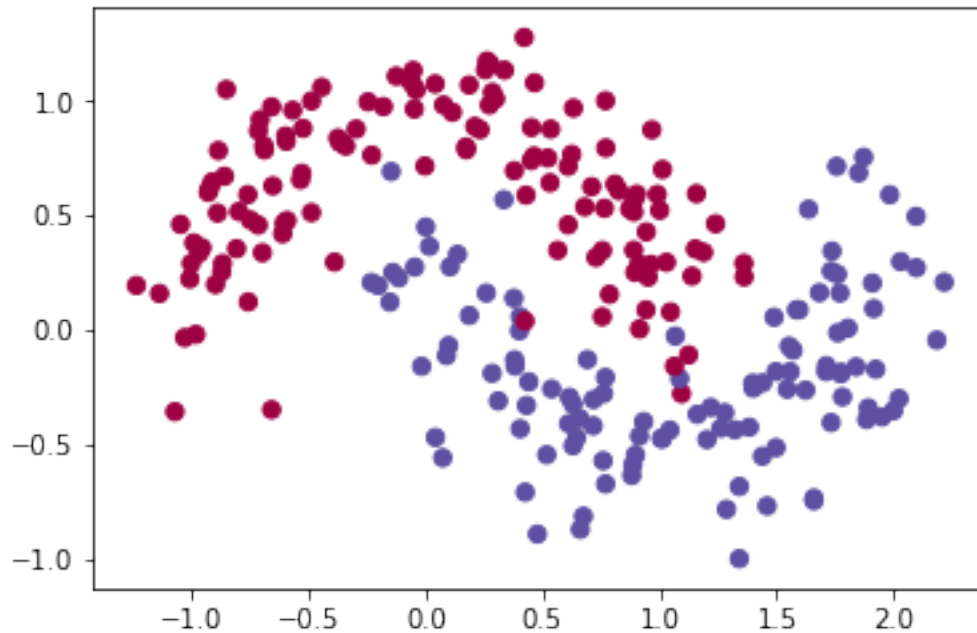
```
In [14]: plt.scatter(X_train[:,0], X_train[:,1], s=40, c=y_train, cmap=plt.cm.Spectral)
```

```
Out[14]: <matplotlib.collections.PathCollection at 0x1a0952b250>
```



```
In [15]: plt.scatter(X_test[:,0], X_test[:,1], s=40, c=y_test, cmap=plt.cm.Spectral)
```

```
Out[15]: <matplotlib.collections.PathCollection at 0x1a096108d0>
```



```
In [118]: #entreno un arbol como clasificador y como estimador de probabilidades
```

```
clf=tree.DecisionTreeClassifier(criterion='entropy',  
                               min_samples_split=20,  
                               min_samples_leaf=5,  
                               max_depth = 10)
```

```
clf.fit(X_train,y_train)
```

```
# estimo probabilidades para cada muestra de test
```

```
proba=clf.predict_proba(X_test)
```

```
cost_01=10000
```

```
cost_00=0
```

```
cost_10=1
```

```
cost_11=0
```

```
#Calculo riesgos para cada decision
```

```
Riesgo1=proba[:,1]*cost_11+proba[:,0]*cost_10
```

```
Riesgo0=proba[:,1]*cost_01+proba[:,0]*cost_00
```

```
y_pred_test_bmr=(Riesgo0-Riesgo1)
```

```
#si el riesgo de asignar 0 es mayor que el de 1 decido 1
```

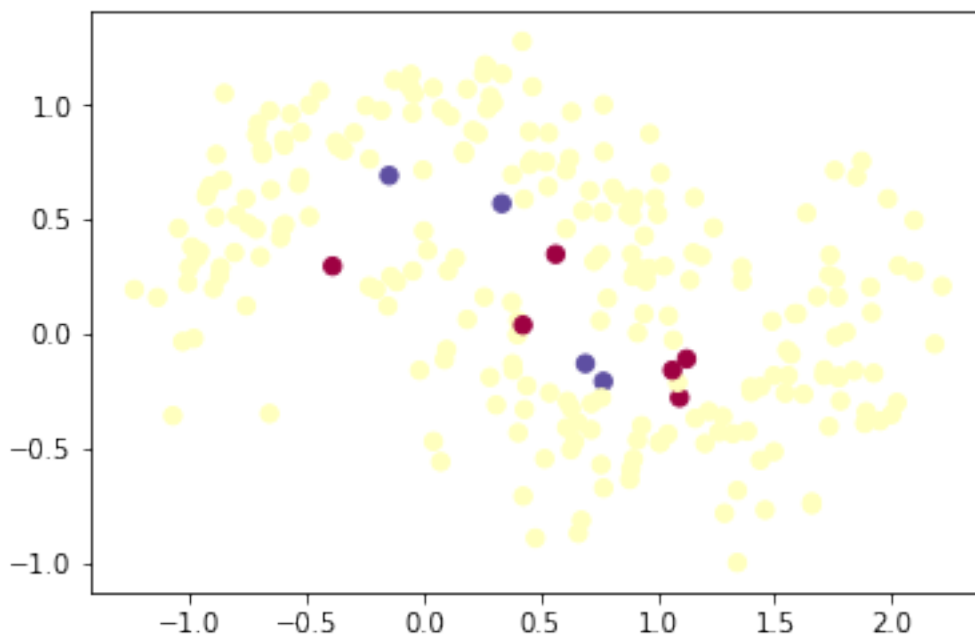
```
y_pred_test_bmr1=(np.sign(y_pred_test_bmr)+1)/2
```

```
In [119]: diff_test=y_test-clf.predict(X_test)
```

```
In [120]: # resultados solo con el arbol
```

```
In [121]: plt.scatter(X_test[:,0], X_test[:,1], s=40, c=diff_test, cmap=plt.cm.Spectral)
```

```
Out[121]: <matplotlib.collections.PathCollection at 0x1a0adb5650>
```



```
In [122]: #resultado al incluir los costos de los errores
```

```
In [123]: diff_test=y_test-y_pred_test_bmr1
```

```
In [124]: plt.scatter(X_test[:,0], X_test[:,1], s=40, c=diff_test, cmap=plt.cm.Spectral)
```

```
Out[124]: <matplotlib.collections.PathCollection at 0x1a0ae25f50>
```

