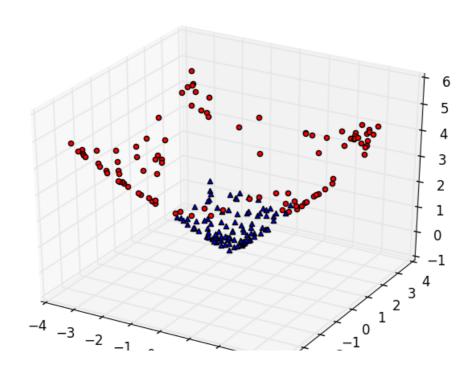
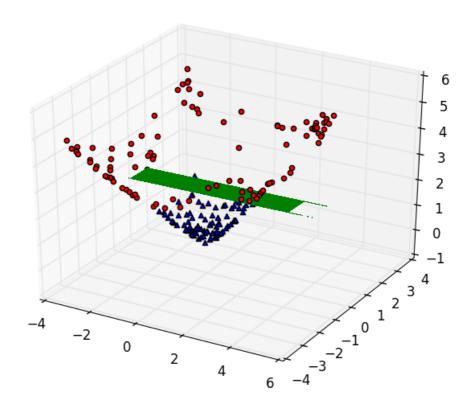
Kernel Trick for separating linearly inseparable data using Perceptron Let x,y be 2d data Kernels used are:

 $1)(X,Y,Z)=(x,y,x^2+y^2)$

Accuracy is 1.000 and the plots are: 1)Data after changed and 2)The final hyperplane





English letter classification using Support Vector Machines

The kernels used were:

1)Linear 2)Polynomial and 3)RBF

And the hyperparameters tested were :

C, Gamma, degree and coef0

C is the parameter for the soft margin cost function.

Gamma is the free parameter of the Gaussian radial basis function.

1)RBF: RBG implicitly maps every point to an infinite dimensional space and the parameters used here are C, Gamma, degree and coef0.

Variation with C:

A low C makes the decision surface smooth, while a high C aims at classifying all training examples correctly by giving the model freedom to select more samples as support vectors. So we have to choose C which is not too high or too low.

Results:

for c= 1 --accuracy: 0.9563, precision: 0.9563, recall: 0.9563, f1: 0.9563 for c= 10 --accuracy: 0.9734, precision: 0.9734, recall: 0.9734, f1:0.9734 for c= 100 --accuracy: 0.9697, precision: 0.9697, recall: 0.9697, f1:0.9697

for c= 1000 --accuracy: 0.971 , precision: 0.971 , recall: 0.971 , f1: 0.971

Variation with Gamma:If gamma is too large, the radius of the area of influence of the support vectors only includes the support vector itself and no amount of regularization with C will be able to prevent overfitting. When gamma is very small, the model is too constrained and cannot capture the complexity or "shape" of the data.

```
So we have to choose Gamma which is not too high or too low. Results:
```

for Gamma= 0.001 --accuracy: 0.8256, precision: 0.8256, recall: 0.8256,

f1: 0.8256

for Gamma= 0.01 --accuracy: 0.9204, precision: 0.9204, recall: 0.9204,

f1: 0.9204

for Gamma= 0.1 --accuracy: 0.9731, precision: 0.9731, recall: 0.9731,

f1: 0.9731

for Gamma= 1 --accuracy: 0.9415, precision: 0.9415, recall: 0.9415, f1:

0.9415

for Gamma= 10 --accuracy: 0.288, precision: 0.288, recall: 0.288, f1: 0.288

. 2)Polynomial:

from the formula we can say that the kernel depends on all the above parameters. Polynomial: (gamma*u'*v + coef0)^degree (using libsvm's nomenclature)

Variation with C:

Results:

for C= 1 --accuracy: 0.9292, precision: 0.9292, recall: 0.9292, f1: 0.9292

for C= 10 --accuracy: 0.952, precision: 0.952, recall: 0.952, f1: 0.952

for C= 100 --accuracy: 0.948, precision: 0.948, recall: 0.948, f1: 0.948

for C= 1000 --accuracy: 0.942, precision: 0.942, recall: 0.942, f1: 0.942

So intermediate c values gives best results. Variation with Gamma:

Results:

for Gamma= 0.01 --accuracy: 0.6059, precision: 0.6059, recall: 0.6059,

f1: 0.6059

for Gamma= 0.1 --accuracy: 0.9494, precision: 0.9494, recall: 0.9494,

f1: 0.9494

for Gamma= 1 --accuracy: 0.9418, precision: 0.9418, recall: 0.9418, f1:

0.9418

for Gamma= 10 --accuracy: 0.9453, precision: 0.9453, recall: 0.9453, f1: 0.9453 similarly intermediate gamma values gives best results.

Variation with Degree:

Variation with degree is data specific, in our data it decreases with degree. Results:

```
Degree 3 --accuracy: 0.9526, precision: 0.9526, recall: 0.9526, f1:
0.9526
for Degree= 5 --accuracy: 0.9062, precision: 0.9062, recall: 0.9062, f1:
0.9062
for Degree= 7 --accuracy: 0.8516, precision: 0.8516, recall: 0.8516, f1:
0.8516
for Degree= 10 --accuracy: 0.7764, precision: 0.7764, recall: 0.7764, f1:
0.7764
Variation with coef0: For this dataset, intermediate coef0 values gives best
results.
Results:
for Coef0= -1 --accuracy: 0.2187, precision: 0.2187, recall: 0.2187, f1:
0.2187
for Coef0= 0.01 --accuracy: 0.9532, precision: 0.9532, recall: 0.9532, f1:
0.9532
for Coef0= 0 --accuracy: 0.9525, precision: 0.9525, recall: 0.9525, f1:
0.9525
for Coef0= 0.1 --accuracy: 0.9599, precision: 0.9599, recall: 0.9599, f1:
0.9599
for Coef0= 1 --accuracy: 0.9621, precision: 0.9621, recall: 0.9621, f1:
0.9621
for Coef0= 5 --accuracy: 0.9565, precision: 0.9565, recall: 0.9565, f1:
0.9565
for Coef0= 10 --accuracy: 0.9537, precision: 0.9537, recall: 0.9537, f1:
0.9537
```

3)Linear: For linear kernel the parameter is C as linear: . It tries to seperate data linearly. Variation with C.

Results:

```
for C= 1 --accuracy: 0.8573 , precision: 0.8573 , recall: 0.8573 , f1: 0.8573 for C= 10 --accuracy: 0.8531 , precision: 0.8531 , recall: 0.8531 , f1: 0.8531 for C= 100 --accuracy: 0.8503 , precision: 0.8503 , recall: 0.8503 , f1: 0.8503
```

So here accuracy decreases with C.