Plot SVM margins using MATLAB and libsvm

I am using symlib to classify linearly two dimensional non-separable data. I am able to train the sym and obtain w and b using symlib. Using this information I can plot the decision boundary, along with the support vectors, but I am not sure about how to plot the margins, using the information that symlib gives me.

Below is my code:

```
model = svmtrain(Y,X, '-s 0 -t 0 -c 100');
w = model.SVs' * model.sv coef;
b = -model.rho;
if (model.Label(1) == -1)
    w = -w; b = -b;
end
y_hat = sign(w'*X' + b);
sv = full(model.SVs);
% plot support vectors
plot(sv(:,1),sv(:,2),'ko', 'MarkerSize', 10);
% plot decision boundary
plot_x = linspace(min(X(:,1)), max(X(:,1)), 30);
plot y = (-1/w(2))*(w(1)*plot x + b);
plot(plot_x, plot_y, 'k-', 'LineWidth', 1)
matlab
        machine-learning
                          svm libsvm
```

asked Feb 17 '15 at 6:52 user115188

The margin is the distance between the decision boundary and the support vectors. How would you like to plot it? – hbaderts Feb 17 '15 at 7:52

1 Answer

It depends on what you mean by "the margins". It also depends on what SVM version you are talking about (separable on non-separable), but since you mentioned libsvm I'll assume you mean the more general, non-separable version.

The term "margin" can refer to the Euclidean distance from the separating hyperplane to the hyperplane defined by wx+b=1 (or wx+b=-1). This distance is given by 1/norm(w).

"Margin" can also refer to the margin of a specific sample \times , which is the Euclidean distance of \times from the separating hyperplane. It is given by

```
(wx+b)/norm(w)
```

note that this is a signed distance, that is it is negative/positive, depending on which side of the hyperplane the point \times resides. You can draw it as a line from the point, perpendicular to the hyperplane.

Another interesting value is the slack variable xi, which is the "algebraic" distance (not Euclidean) of a support vector from the "hard" margin defined by wx+b=+1 (or -1). It is positive only for support vectors, and if a point is not a support vector, its xi equals 0. More compactly:

```
xi = max(0, 1 - y*(w'*x+b))
```

where y is the label.

edited Feb 17 '15 at 19:22

answered Feb 17 '15 at 8:14



I think I'm just not understanding the svm theory correctly. Given your equations, how do i know which support vector to use to obtain the margin? Would it be the correctly classified support vector that is farthest from the decision boundary? — user115188 Feb 17 '15 at 16:30

I'll edit my answer since it's not short enough for a comment – Itamar Katz Feb 17 '15 at 19:07