



SVM review

And some feedback from the class

Is this a valid kernel?

Assume $\mathbf{x} = [x_1, x_2]$

- 1) $k(\mathbf{x}, \mathbf{x}') = x_1 + x_2 + x_1' + x_2'$
- 2) $k(\mathbf{x}, \mathbf{x}') = x_1^2 + x_2^2 + x_1'^2 + x_2'^2$
- 3) $k(\mathbf{x}, \mathbf{x}') = \exp((x_1 - x_1')^2 + (x_2 - x_2')^2)$



Is this a valid kernel?

Assume $\mathbf{x} = [x_1, x_2]$

- 1) $k(\mathbf{x}, \mathbf{x}') = x_1 + x_2 + x_1' + x_2'$
No – this is not positive semi-definite
- 2) $k(\mathbf{x}, \mathbf{x}') = x_1^2 + x_2^2 + x_1'^2 + x_2'^2$
probably – requires a more careful look
- 3) $k(\mathbf{x}, \mathbf{x}') = \exp((x_1 - x_1')^2 + (x_2 - x_2')^2)$
yes – this is the Gaussian kernel



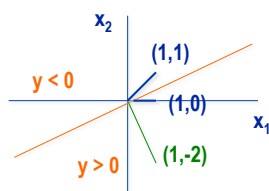
Hyperplanes

- ◆ Given the hyperplane defined by the line
 - $y = x_1 - 2x_2$
 - $y = (1, -2)^T \mathbf{x} = \mathbf{w}^T \mathbf{x}$
- ◆ Is this point correctly predicted?
 - $y = 1, \mathbf{x} = (1, 0)$?
 - $y = 1, \mathbf{x} = (1, 1)$?
- ◆ How should the weight be adjusted to reduce the loss function?
 - .



Hyperplanes

- ◆ Given the hyperplane defined by the line
 - $y = x_1 - 2x_2$
 - $y = (1, -2)^T \mathbf{x}$



So (1,0) having $y=1$ is correct and (1,1) having $y=1$ is not correct



Hyperplanes

- ◆ True or False? When solving for a hyperplane specified by $\mathbf{w}^T \mathbf{x} + b = 0$ one can always set the margin to 1;

$$\mathbf{w}^T \mathbf{x}_1 + b = -1 \text{ and } \mathbf{w}^T \mathbf{x}_2 + b = 1$$

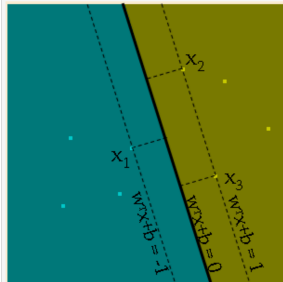


Hyperplanes

- ◆ **True or False?** When solving for a hyperplane specified by $\mathbf{w}^T \mathbf{x} + b = 0$ one can always set the margin to 1;

$$\mathbf{w}^T \mathbf{x}_1 + b = -1 \text{ and } \mathbf{w}^T \mathbf{x}_2 + b = 1$$

true



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Projections

- ◆ The projection of a point $\mathbf{x}$ onto a line $\mathbf{w}$ is $\mathbf{x}^T \mathbf{w} / \|\mathbf{w}\|_2$
- ◆ The distance of a point $\mathbf{x}$ to a hyperplane defined by the line $\mathbf{w}$ is $\mathbf{x}^T \mathbf{w} / \|\mathbf{w}\|_2$

Why?

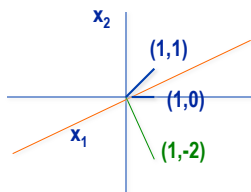


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Hyperplanes

- ◆ The projection of a point $\mathbf{x}$ on a plane defined by a line $\mathbf{w}$ is $\mathbf{x}^T \mathbf{w} / \|\mathbf{w}\|_2$
- ◆ The distance of $\mathbf{x}$ from the hyperplane defined by $(1, -2)$ is what
 - for $\mathbf{x} = (-1, 2)$
 - for $\mathbf{x} = (1, 0)$
 - for $\mathbf{x} = (1, 1)$



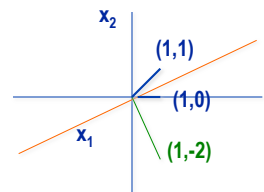
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Hyperplanes

- ◆ The projection of a point $\mathbf{x}$ on a plane defined by a line $\mathbf{w}$ is $\mathbf{x}^T \mathbf{w} / \|\mathbf{w}\|_2$
- ◆ The distance of $\mathbf{x}$ from the hyperplane defined by $(1, -2)$ is what
 - for $\mathbf{x} = (-1, 2)$
 - for $\mathbf{x} = (1, 0)$
 - for $\mathbf{x} = (1, 1)$

The same as before – you project onto the line



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Hyperplanes

- ◆ **True or False?** When solving for a hyperplane specified by $\mathbf{w}^T \mathbf{x} + b = 0$ one can always set the margin to 1;

$$\mathbf{w}^T \mathbf{x}_1 + b = -1 \text{ and } \mathbf{w}^T \mathbf{x}_2 + b = 1$$

- ◆ **True or False?** This then implies that the margin is the distance of the support vectors from the separating hyperplane

$$\frac{\mathbf{w}^T (\mathbf{x}_2 - \mathbf{x}_1)}{2\|\mathbf{w}\|_2} = \frac{1}{\|\mathbf{w}\|_2}$$



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(non)separable SVMs

- ◆ **True or False?** In a real problem, you should check to see if the SVM is separable and then include slack variables if it is not separable.

- ◆ **True or False?** Adding slack variables is equivalent to requiring that all of the α_i are less than a constant in the dual

$$\mathbf{w}^T \phi(\mathbf{x}) + b = \sum_i \alpha_i y_i k(\mathbf{x}_i, \mathbf{x}) + b$$



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(non)separable SVMs

- ◆ **True or False?** In a real problem, you should check to see if the SVM is separable and then include slack variables if it is not separable.
 - False: you can just run the slack variable problem in either case
- ◆ **True or False?** Adding slack variables is equivalent to requiring that all of the α_i are less than a constant in the dual

$$\mathbf{w}^\top \phi(\mathbf{x}) + b = \sum_i \alpha_i y_i k(\mathbf{x}_i, \mathbf{x}) + b$$

true



Why do SVMs work well?

- ◆ Why are SVMs fast?
- ◆ Why are SVMs often more accurate than logistic regression?



Why do SVMs work well?

- ◆ Why are SVMs fast?
 - They work in the dual, with relatively few points
 - The kernel trick
- ◆ Why are SVMs often more accurate than logistic regression?
 - SVMs use kernels
 - but logistic regression can also do that
 - Logistic regression uses all the data points, assuming a probabilistic model, while SVMs ignore the points that are clearly correct.



Class feedback

- ◆ What you liked
 - My enthusiasm, focus on intuition
 - The HW
 - The subject
 - Clickers



What could be improved?

- ◆ Make homework clearer
 - fewer (maybe longer) HWs
 - More short questions to help clarify notation
 - latex is a pain
- ◆ More math
- ◆ Less math
- ◆ More (applied) projects
- ◆ More theory
- ◆ Less work



What could be improved?

- ◆ More detailed explanation of clicker questions
 - post answers
- ◆ Some (extra) session where we do more of a "hands-down" approach to problems.
- ◆ Change exam questions to be more like homework



Class feedback

◆ Speed

- 36% too fast
- 61% good
- 3% too slow

◆ Average hours/week

- 17

◆ Homeworks

- All very close to “4 – probably keep”

