

Ch 9.1: Maximum Margin Classifier

Lecture 26 - CMSE 381

Michigan State University

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Dept of Computational Mathematics, Science & Engineering

Mon, Mar 30, 2026

Announcements

Last time:

- Ch 8: Random Forests

This lecture:

- Maximal Margin Classifier
- No jupyter notebook for this class

Announcements:

21	W	3/18	Polynomial & Step Functions	7.1-7.2		
22	F	3/20	Step Functions; Basis functions; Start Splines	7.2-7.4		
23	M	3/23	Regression Splines	7.4		
24	W	3/25	Decision Trees	8.1		Q7
25	F	3/27	Random Forests	8.2.1, 8.2.2	HW #5 Due Sun 3/29	
26	M	3/30	Maximal Margin Classifier	9.1		
27	W	4/1	SVC	9.2		Q8
28	F	4/3	SVM	9.3, 9.4		
29	M	4/6	Single Layer NN	10.1		
30	W	4/8	Multi Layer NN	10.2		Q9
31	F	4/10	CNN	10.3		
32	M	4/13	Unsupervised learning / clustering	12.1, 12.4	HW #6 Due Sun 4/12	
33	W	4/15	Virtual: Project Office Hours			Q10
	F	4/17	Review			
	M	4/20	Midterm #3			
	W	4/22				
	F	4/24				Project Due

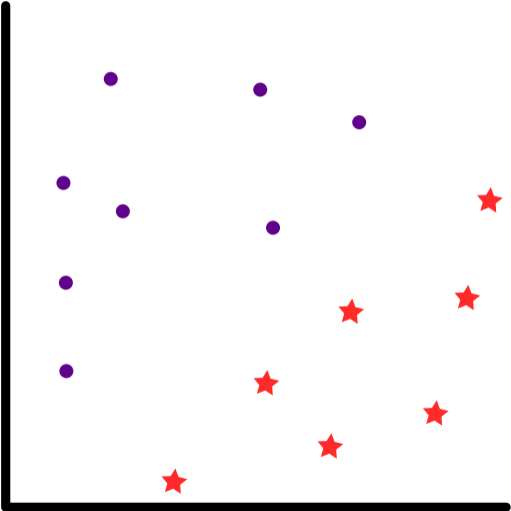
What will you learn today?

- What is hyperplane?
 - ▶ How do you mathematically describe this hyperplane?
 - ▶ How do you mathematically describe the two sides of the hyperplane?
 - ▶ Given the equation of the hyperplane and the coordinates for a point, you should be able to tell which side of the plane the point is on.
- What qualify a hyperplane as a separating hyperplane?
 - ▶ You should be able to describe this mathematically using an inequality.
 - ▶ You should also be able to determine whether a hyperplane is a separating hyperplane given a graph, or given the equation of the plane and the coordinates and class of a few points.
- How to use a separating hyperplane as a classifier?
- What makes a hyperplane a maximal margin hyperplane?
 - ▶ What are its margin and support vectors?
 - ▶ Given a graph, you should be able to clearly label the margin and support vectors. You should also be able to infer the size of the margin from reading the graph.
 - ▶ You should also be able to describe the optimization problem mathematically.
 - ▶ Given the equation of the maximal margin hyperplane and the coordinates of support vectors, you should be able to calculate the size of the margin by hand.

Section 1

Maximal Margin Classifier

The goal

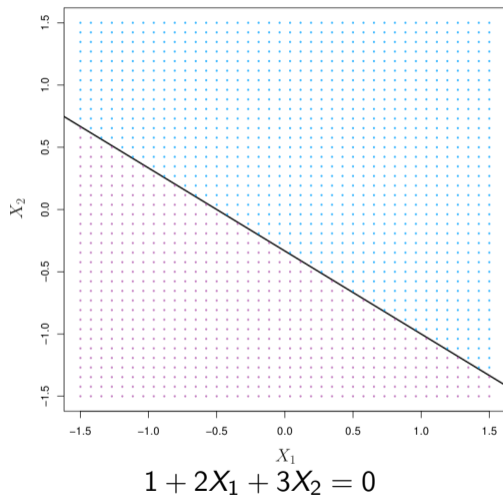


What is a hyperplane?

Mathematical definition of a hyperplane

$$\beta_0 + \beta_1 X_1 + \beta_2 X_2 + \cdots + \beta_p X_p = 0$$

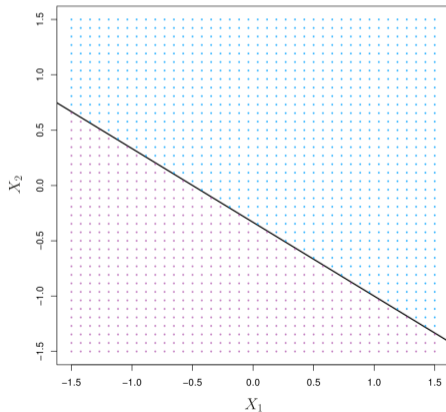
Hyperplane for $p = 2$



There are two sides to every hyperplane

$$\beta_0 + \beta_1 X_1 + \beta_2 X_2 + \cdots + \beta_p X_p < 0$$

$$\beta_0 + \beta_1 X_1 + \beta_2 X_2 + \cdots + \beta_p X_p > 0$$



Classification Setup

Data matrix:

$$X = \begin{pmatrix} - & x_1^T & - \\ - & x_2^T & - \\ & \vdots & \\ - & x_n^T & - \end{pmatrix}_{n \times p}$$

$$x_1 = \begin{pmatrix} x_{11} \\ \vdots \\ x_{1p} \end{pmatrix}, \dots, x_n = \begin{pmatrix} x_{n1} \\ \vdots \\ x_{np} \end{pmatrix}$$

Observations in one of two classes,
 $y_i \in \{-1, 1\}$

$$Y = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{pmatrix}$$

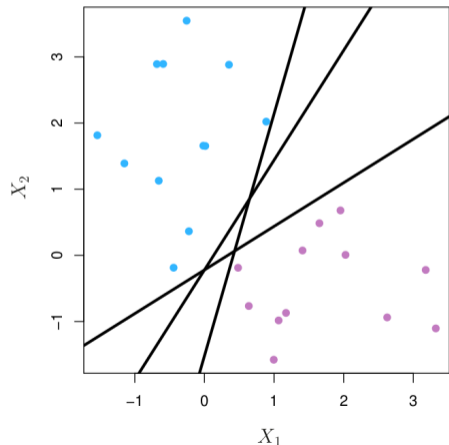
Separate out a test observation

$$x^* = (x_1^* \cdots x_p^*)^T$$

Separating Hyperplane

$$\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \cdots + \beta_p x_{ip} > 0 \text{ if } y_i = 1$$

$$\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \cdots + \beta_p x_{ip} < 0 \text{ if } y_i = -1$$



Another way to say it

$$\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \cdots + \beta_p x_{ip} > 0 \text{ if } y_i = 1$$

$$\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \cdots + \beta_p x_{ip} < 0 \text{ if } y_i = -1$$

For all i :

$$y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \cdots + \beta_p x_{ip}) > 0$$

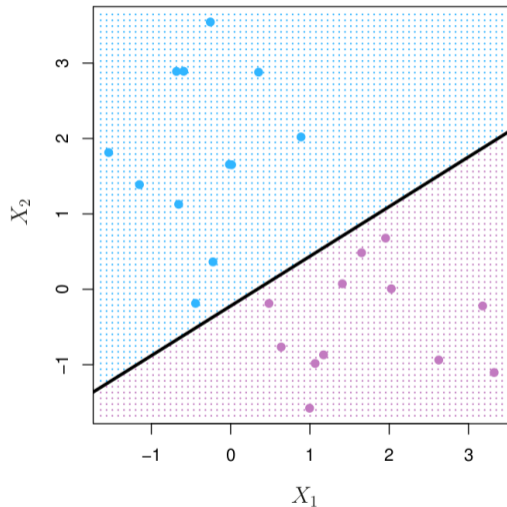
Separating hyperplane becomes a classifier

If you have a separating hyperplane:

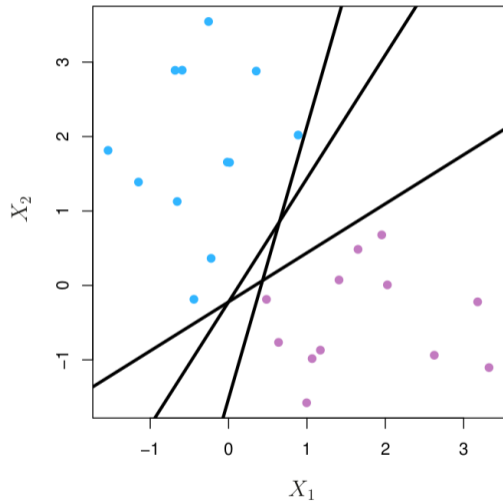
- Check

$$f(x^*) = \beta_0 + \beta_1 x_1^* + \beta_2 x_2^* + \dots + \beta_p x_p^*$$

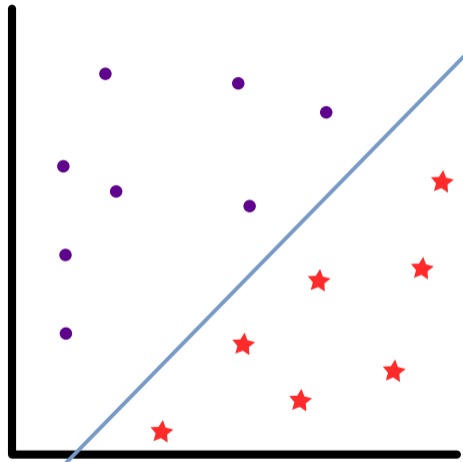
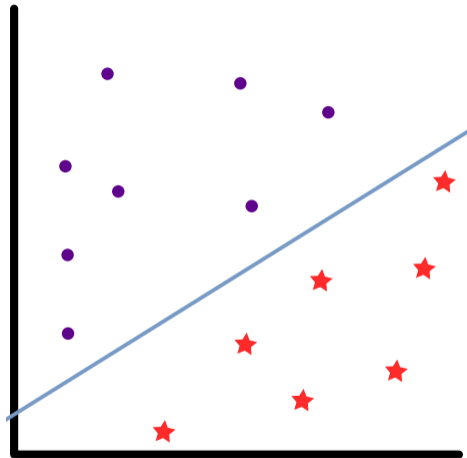
- If positive, assign $\hat{y} = 1$
- If negative, assign $\hat{y} = -1$



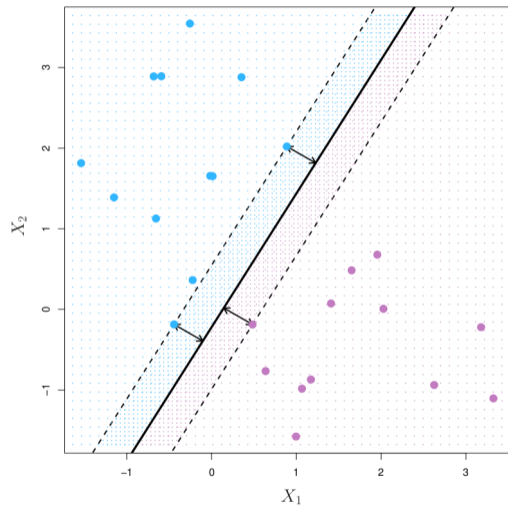
How do we pick?



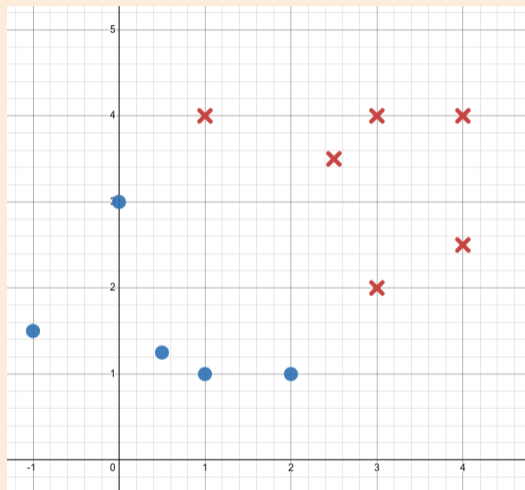
Distance from an observation to a hyperplane



Maximal margin classifier



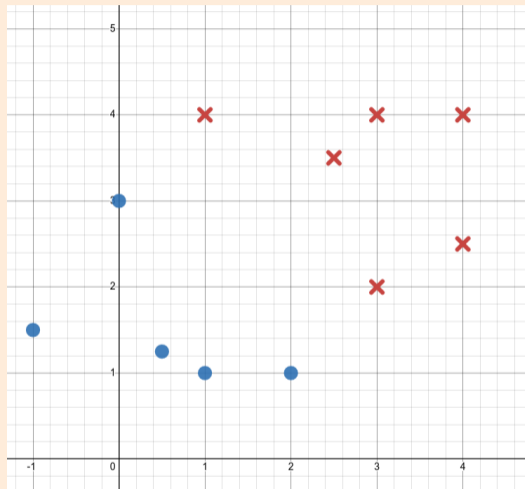
Example



- Sketch the maximal margin hyperplane.
- What is the equation of this line in the form $\beta_0 + \beta_1 X_1 + \beta_2 X_2 = 0$?
- Circle the support vectors. What is their distance from the line?

<https://www.desmos.com/calculator/tklbommiwz>

Extra work space



Respond to [PollEv question](#).

<https://www.desmos.com/calculator/tklbommiwz>

Mathematical Formulation

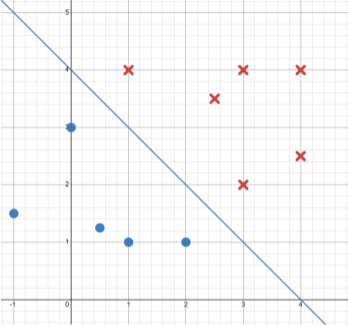
$$\text{maximize } M$$

$\beta_0, \beta_1, \dots, \beta_p, M$

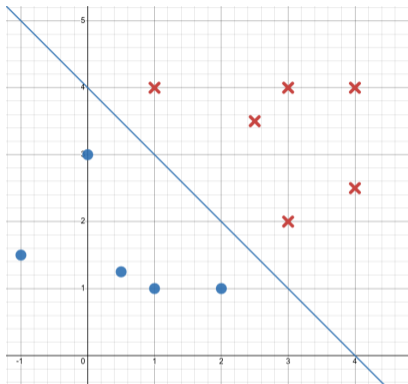
$$\text{subject to } \sum_{j=1}^p \beta_j^2 = 1,$$

$$y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2} + \dots + \beta_p x_{ip}) \geq M \quad \forall i = 1, \dots, n$$

First constraint



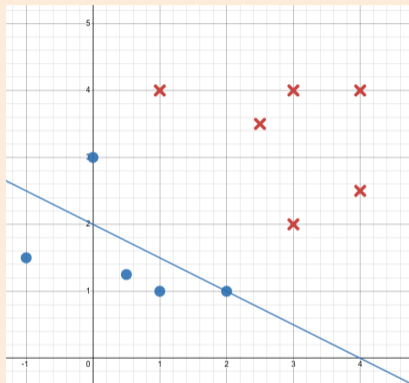
Second constraint



- Blue circles: $y_i = -1$
- Red Xs: $y_i = 1$
- $-2\sqrt{2} + \frac{\sqrt{2}}{2}X_1 + \frac{\sqrt{2}}{2}X_2 = 0$

$$y_i(\beta_0 + \beta_1x_{i1} + \beta_2x_{i2}) \geq M$$

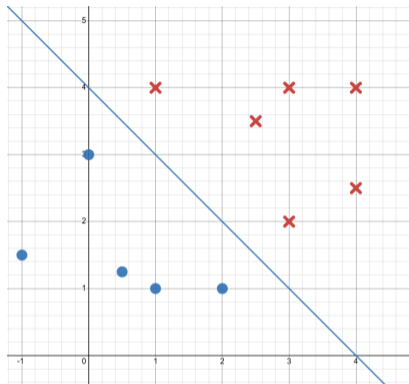
An example with a bad choice of hyperplane



What is $y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2})$ for the point $x_i = (0, 3)$?

- Blue circles: $y_i = -1$
- Red Xs: $y_i = 1$
- $-\frac{4}{\sqrt{5}} + \frac{1}{\sqrt{5}}X_1 + \frac{2}{\sqrt{5}}X_2 = 0$

Second constraint extra space



$$y_i(\beta_0 + \beta_1 x_{i1} + \beta_2 x_{i2}) \geq M$$

- Blue circles: $y_i = -1$
- Red Xs: $y_i = 1$
- $-2\sqrt{2} + \frac{\sqrt{2}}{2}X_1 + \frac{\sqrt{2}}{2}X_1 = 0$

Mathematical Formulation

$$\text{maximize } M$$

$\beta_0, \beta_1, \dots, \beta_p, M$

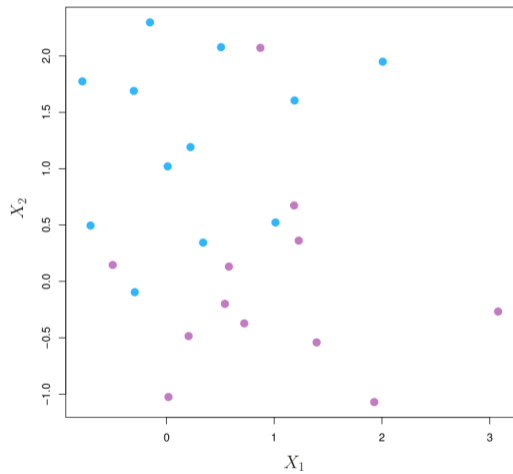
$$\text{subject to } \sum_{j=1}^p \beta_j^2 = 1,$$

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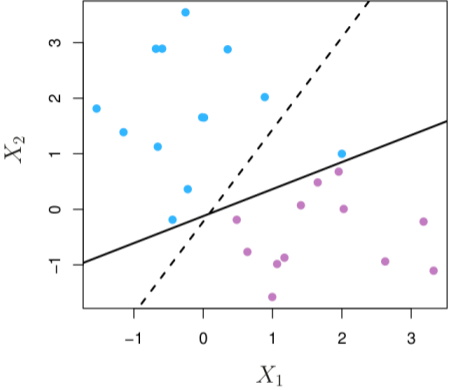
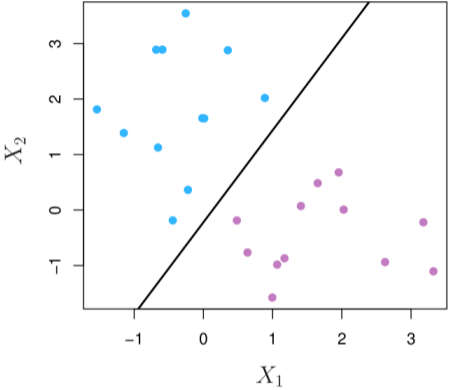
Section 2

Issues with Maximal Margin Classifier

But what if....



Sensitivity to new points



Next time

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