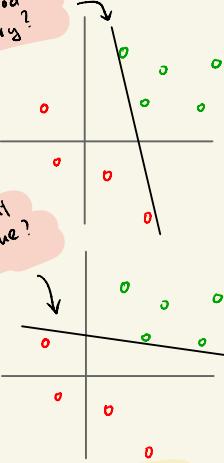
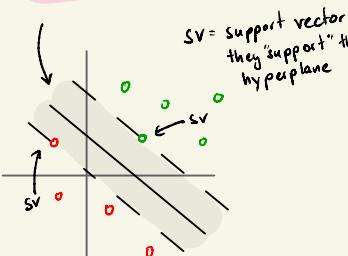


This a good Boundary?

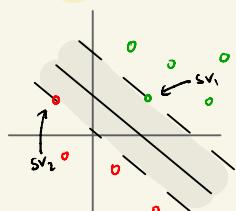


How about this one?

Want:  
fit a street between  
the classes that is  
as wide as possible.



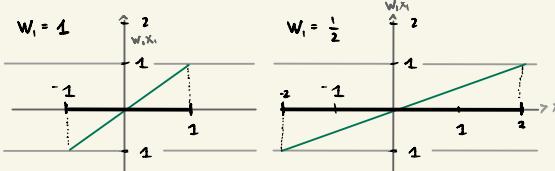
Question: if we move some  
points a tiny bit, for  
which does the Decision Boundary  
change? & which not?



Choose  $w$   
such that  $w^T x - b = 1$   
goes through  $SV_1$ ,  
and  $w^T x - b = -1$   
goes through  $SV_2$ .

**Maximum Margin:**  
maximize the distance from  $SV_1, SV_2$  to  
the hyperplane (or minimize  $\|w\|$ ) under  
conditions  $y_i(w^T x_i - b) \geq 1$

Intuitive picture:  
smaller  $w \Rightarrow$  wider margin



Let  $sv = (x_1)$   
then  $sv = t \frac{w}{\|w\|}$   
because  $w^T x = 1$   
 $t \cdot w^T w = 1 \Rightarrow t = \frac{1}{\|w\|^2}$   
 $x_1 = 0$

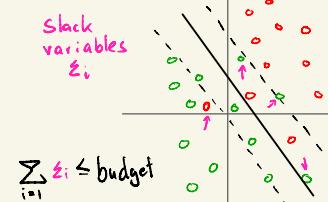
$$\begin{aligned} \text{Margin} &= (SV_1 - SV_2)^T \frac{w}{\|w\|} \\ &= SV_1^T \frac{w}{\|w\|} - SV_2^T \frac{w}{\|w\|} \\ &= \frac{b+1}{\|w\|} - \frac{b-1}{\|w\|} \\ &= \frac{2}{\|w\|} \end{aligned}$$

How?

**Problem!**  
impossible if classes are not  
linearly separable

**IDEA!**

Cut classifier some slack  
and give it a budget for  
margin violations and misclassifications



$$\sum_i \epsilon_i \leq \text{budget}$$

Look familiar? →  
Before (i.e. Ridge)

minimize  $A + \lambda B$  bigger  $\lambda$  more weight  
on  $B$   
now  
minimize  $CA + B$  smaller  $C$  more weight  
on  $B$   
 $\Rightarrow C = \frac{1}{\lambda}$

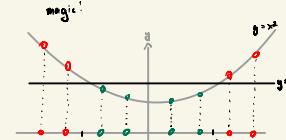
Reminder: Dot Product &  
Projections



Error: 1 : 2  
2 : 1  
3 : 0  
4 : 0  
5 : 1  
6 : 2

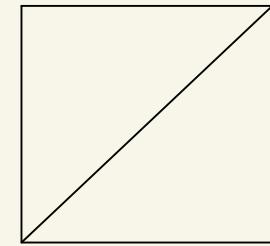
TLDR:

not separable by line Kernel Trick

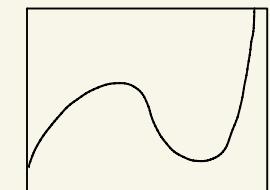


Shapes of Decision Boundaries  
from different Kernels

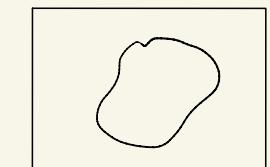
Rules of Thumb!



Straight lines ⇒ linear

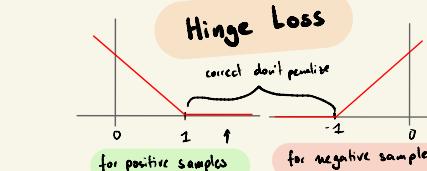


Curves = polynomial



islands = RBF  
radial basis f.

How?



**Hinge Loss**

minimize:  
 $\frac{1}{n} C \sum_i (\max(0, 1 - y_i(w^T x_i - b)) + \|w\|^2)$

**Bonus!**

also less sensitive to outliers!

