



Classification: VC Dimension

Machine Learning: Jordan Boyd-Graber University of Colorado Boulder

.

VC Dimension

To show VC dimension of a set of points

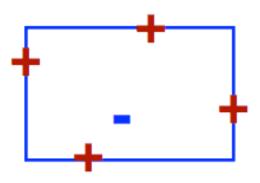
- Show that a set of d can be shattered
- Show that **no** set of d+1 can be shattered

Axis Aligned Rectangles

Axis Aligned Rectangles



Axis Aligned Rectangles



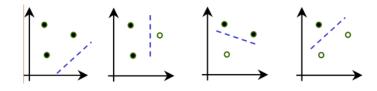




Figure 3.2 Unrealizable dichotomies for four points using hyperplanes in R². (a) All four points lie on the convex hull. (b) Three points lie on the convex hull while the remaining point is interior.



Figure 3.2 Unrealizable dichotomies for four points using hyperplanes in R². (a) All four points lie on the convex hull. (b) Three points lie on the convex hull while the remaining point is interior.

In general, the VC dimension of d-dimensional hyperplanes is d+1

Show that the VC dimension of a finite hypothesis set H is at most $\lg |H|$.

• To shatter a set, it means that every point can take on a different binary label h(x)

- To shatter a set, it means that every point can take on a different binary label h(x)
- If a set has d points, there are 2^d ways to do that

- To shatter a set, it means that every point can take on a different binary label h(x)
- If a set has d points, there are 2^d ways to do that
- Each configuration requires a different hypothesis

- To shatter a set, it means that every point can take on a different binary label h(x)
- If a set has d points, there are 2^d ways to do that
- Each configuration requires a different hypothesis
- Solving for the number of hypotheses gives $\lg |H|$

Next time

- · Getting more practical
- SVMs
- Excellent theoretical properties