

2.16

(a) To solve the polynomial $\sum_{i=0}^D C_i X^i$

we need $D+1$ points. The resulting

polynomial will pass through all of

these points. So we will choose $D+1$ points

and assign the y value with the sign we

need (+1 / -1). After solving the system

we will get the coefficients that will produce

the intended signs on the $D+1$ points. so H

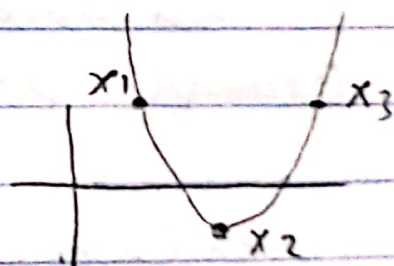
shatter $D+1$ points.

• Example: for $D=2$

Points: $(x_1, +1)$

$(x_2, -1) \Rightarrow$ solve and get C_0, C_1, C_2 .

$(x_3, +1)$



we also choose any positive number other than (1)
or any negative number other than (-1).

(b) Consider the same approach as in the

previous step, we can only control

the sign of $D+1$ points, we trying to

check $D+2$, we will notice that there

is extra point which its sign can not be

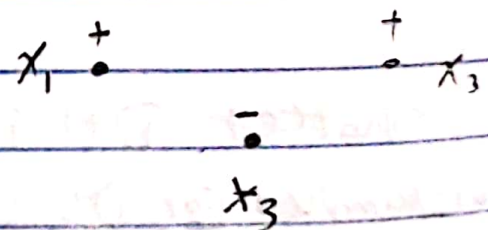
controlled. The case will cause that two

neighboring points will have the same

sign which will prevent us from generating

all dichotomies

Examples let $D=1$



This case cannot be solved by
a straight line (no other orientation we let it be
solved)
So H don't shatter $D+2$ point

*From (a), (b) the VC dimension of H is
exactly $(D+1)$.

2.24

$$(a) E_{in}(g) = \sum_{i=1}^2 (f(x_i) - h(x_i))^2 \\ = \sum_{i=1}^2 (x_i^2 - (ax_i + b))^2$$

set the derivatives with respect to a and b to zero

$$\frac{\partial E_{in}(g)}{\partial a} = -2 \sum_{i=1}^2 x_i (x_i^2 - ax_i - b) = 0 \rightarrow (1)$$

$$\frac{\partial E_{in}(g)}{\partial b} = -2 \sum_{i=1}^2 (x_i^2 - ax_i - b) = 0 \rightarrow (2)$$

$$x_1 * (1) - x_2 * (2)$$

$$\left. \begin{aligned} x_1^2 - ax_1 - b &= 0 \\ x_2^2 - ax_2 - b &= 0 \end{aligned} \right\} \text{solve}$$

$$a = x_1 + x_2 \\ b = -x_1 x_2$$

$$\stackrel{p}{=} g(x) = \overbrace{(x_1 + x_2)}^a x - \overbrace{x_1 x_2}^b$$

$$\bar{g}(x) = E_D [g^D(x)]$$

$$= E_D [(x_1 + x_2)x - x_1 x_2]$$

$$= E_D [x_1 x] + E_D [x_2 x] - E_D [x_1 x_2]$$

$$= \underline{E_D [x_1] x + E_D [x_2] x - E_D [x_1] E_D [x_2]}$$

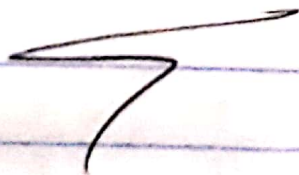
↓
due to independence
between x_1 and x_2

As the data follow a uniform distribution between -1, 1 : $E(x_1) = 0$ & $E_D(x_2) = 0$

$$E_D^D(x) = 0$$

$$\therefore \bar{g}(x) = 0 \times x + 0 \times x - 0 \times 0$$

$$= 0$$



(d)

$$\text{Variance} = E_x [E_D [(g^D(x) - \bar{g}(x))^2]]$$

$$= E_x [E_D ((X_1 + X_2)X - X_1 X_2 - 0)^2]$$

$$= E_x [E_D ((X_1 + X_2)X - X_1 X_2)^2]$$

$$= E_x [E_D ((X_1 + X_2)^2 X^2 + X_1^2 X_2^2 - 2X_1 X_2 (X_1 + X_2)X)]$$

$$= E_x [X^2 E_D (X_1^2 + X_2^2 + 2X_1 X_2) + E_D (X_1^2 X_2^2)$$

$$- 2X E_D (X_1^2 X_2 + X_1 X_2^2)]$$

As the data follow a uniform distribution between $-1, 1$:

$$E(X) = 0$$

$$E(X^2) = \frac{1}{3}$$

$$E(X^3) = 0$$

$$E(X^4) = \frac{1}{5}$$

$$= \text{Var} = E_x \left[X^2 \left(\frac{1}{3} + \frac{1}{3} \right) + \left(\frac{1}{3} - \frac{1}{3} \right) \right]$$

other term with X_1 & X_2 yields zero

$$= E_x \left(\frac{2}{3} X^2 + \frac{1}{9} \right)$$

$$= \frac{2}{3} \times \frac{1}{3} + \frac{1}{9} = \frac{1}{3}$$

$$\text{Bias} = E_x \left[(\hat{g}(x) - f(x))^2 \right]$$

$$= E_x \left[(0 - x^2)^2 \right]$$

$$= E_x \left[X^4 \right] \rightarrow \text{from prev page}$$

$$= \frac{1}{5}$$

$$E_{\text{out}} = \text{Bias} + \text{Variance}$$

$$= \frac{1}{3} + \frac{1}{5} = \frac{8}{15}$$

$$E_x[X] = \frac{1}{2} \int_{-1}^1 x dx = \frac{1}{2} \times \frac{1}{2} [1^2 - (-1)^2] = 0$$

$$E_x[X^2] = \frac{1}{2} \int_{-1}^1 x^2 dx = \frac{1}{2} \times \frac{1}{3} [1^3 - (-1)^3] = \frac{1}{3}$$

$$E_x[X^3] = \frac{1}{2} \int_{-1}^1 x^3 dx = \frac{1}{2} \times \frac{1}{4} [1^4 - (-1)^4] = 0$$

$$E_x[X^4] = \frac{1}{2} \int_{-1}^1 x^4 dx = \frac{1}{2} \times \frac{1}{5} [1^5 - (-1)^5] = \frac{1}{5}$$

3.12

$$H^2 = X(X^T X)^{-1} X^T X(X^T X)^{-1} X^T$$

$$\text{Let } X^T X = A$$

$$H^2 = X A^{-1} A^{-1} X^T$$

$$= X A^{-1} X^T$$

$$= X(X^T X)^{-1} X^T$$

\hat{y} is the projection of y onto the column space of X .