0,1.

1) According to theorem 2.4. if $m_{\mu}(N) \angle 2^{N}$ then $\exists k$ for which $m_{\mu}(N) \angle 2^{k}$, so $m_{\mu}(N) \leq \sum_{i=0}^{k-1} \binom{N}{i}$

which is a polynomial; therefore, there is no such hypothesis set.

2) a) error bar= \frac{1}{2N} ln \frac{2ul}{3}

 $E_{in} = \sqrt{\frac{1-\ln 2.1000}{2.400}\ln \frac{2.1000}{0.05}} = 0.115$

 $\epsilon_{test} = \sqrt{\frac{1}{2.200} \ln{\frac{2.1}{0.05}}} = 0.096$

Ein $7 \in \text{test.} \Rightarrow 50$ the estimate provided by test sample is better.

b) The more examples we use for testing, the fewer examples we use for training which means, we may not get a good hypothesis from training set. Suppose N=1 for training set 4 the rest for (K) testing = $E_{out} \leq E_{in} + \sqrt{\frac{1}{2K} ln \frac{2}{4\pi}}$

Which is bounded tighter. However, this accounts For Overfitting. Therefore having less training set & more testing set is not always a good idea.

* 1 -1 +1 * X X +0001 X X X -1 +1 -1 - XXXX COOXXXX it b a negative positive -1 +1 -1 simpler way! -X+O-1X-+12 6-1 From which we can see that only +1,-1,+1 dichetomy is added, so we have iddied only 1 dichotomy for N=3 points $\Rightarrow \frac{N^2}{2} + \frac{N}{2} + 1 + (N-2) = \frac{N^2}{2} + \frac{3N}{2} - 1$ To find duc: $2^{N} = \frac{N^{2}}{2} + \frac{3N}{2} - 1$ N_{max} = 3. Check: $2^{3} \stackrel{?}{=} \frac{9}{2} + \frac{9}{2} - 1 = \frac{16}{2}$

So, dvc = 3

4) a) First we construct (D+1) X (D+1) matrix. d non-singular Non-singular matrix, means det X ≠ 0 & X is a square matrix. Suppose $X = \begin{bmatrix} 1 & x_0^1 & \dots & x_n^n \\ 1 & x_n^1 & \dots & x_n^n \end{bmatrix}$ where x_k are $x_k^1 & \dots & x_n^n \\ \vdots & x_n^1 & \dots & \vdots \\ 1 & n^1 & \dots & n^n \end{bmatrix}$ different. 1 x' - xo Consider $y = (y_0, ..., y_0)^T \in \{-1, +1\}^{D+1}$ $\{ \text{ let } C = (c_0, ..., c_D)^T = X, y \Rightarrow XC = y$ $\Rightarrow h_c(x) = sigh\left(\frac{2}{i=0}c_ix_k^i\right) = y_k \quad k = 0,..., D$ => mu (D+1) = 2 0+1 & dvc > D+1. b) Since X has D+1 vectors, any D+2 vectors of length D+1 have to be linearly dependent. $\Rightarrow (x_m, x_m', \dots, x_m) = \sum_{k \neq m} a_k (x_k, x_k', \dots, x_k')$ (linear combination) for each ax #0 Suppose $y_m = -1 + y_k = sign(a_k) = sign(\sum_{i=0}^{\infty} c_i x_k^i)$ $\Rightarrow \sum_{i=0}^{\infty} c_i a_k x_k^i = 70$ (because signs are the same) let's consider $(x_{m}^{0}, x_{m}^{1}, x_{m}^{0}) = \sum_{k \neq m}^{\infty} \sum_{i=0}^{\infty} c_{i} a_{k} x_{k}^{i}$ $\Rightarrow \sum_{k \neq m}^{\infty} \sum_{i=0}^{\infty} c_{i} a_{k} x_{k}^{i} = \sum_{i=0}^{\infty} c_{i} x_{m}^{i} = 0$

$$\Rightarrow f_m = +1$$

$$f_m = -1 \quad \text{ff}_{m=+1} \quad \text{is a contradiction.}$$

$$50 \quad D+2 \text{ points cannot be shattered by H.}$$

$$\Rightarrow d_{vc} \leq D+1$$

$$\Rightarrow d_{vc} = D+1$$

5) suppose
$$x_1 = 10^1$$
, $x_2 = 10^2$ - $x_N = 10^N$
 $4 \text{ let } y = (y_1, ..., y_N)^T \in \{-1, +1\}^N$

Consider
$$d = \frac{1}{10K}$$
 if $y = -1$ f $d = \frac{2}{10K}$ if $y = +$

$$\Rightarrow h_{\chi}(x_{\chi}) = (-1)^{L \chi \cdot 10^{K} \int} = y_{\chi} \quad \text{for } K=1,...,N$$

$$\Rightarrow H(x_1,...,x_N) = \{-1,+1\}^N \Leftrightarrow m_{\mathcal{H}}(N) = 2^N$$

X is invertible matrix, because it's in Vardermonde matrix form & it has $(d+1) \times (d+1)$ dimension. So, for $y = (y_1, y_2, \dots, y_{d+1})^T$ we have a solution $(x_1, x_2, \dots, x_{d+1})^T$ In other words, d+1 points car be shattered, therefore dy 7, d+1

b) We know that any (d+2) vectors in (d+1) dimension are linearly dependent, so we have a linear combination of Z_i $\frac{d+1}{2} = \sum_{i=1}^{d+1} k_i z_i$ $W^{T}_{2d+2} = \sum_{i=1}^{d+1} w^{T}_{k_i} z_i$ $Sign(w^{T}_{2d+2}) = sign(\sum_{i=1}^{d+1} k_i w^{T}_{2i})$ $A (sign(z_1), sign(z_2) - sign(z_{d+1}), -sign(z_{d+2}))$ This case cannot be shattered, so $d_{ie} \leq d+1$

=> dre=d+1.

1) let
$$g(X) = (h(x) - f(x))^{2}$$
, then
$$g(X) = 0 \text{ or } g(X) = 1 \text{ , since}$$

$$\frac{h(x)}{g(x)} = \frac{f(x)}{g(x)} = \frac{f(x)}{g(x)$$

then
$$E(g(X)) = \sum_{n=1}^{N} g(X) P(X = x_n) = \sum_{n=1}^{N} g(X) \cdot \frac{1}{N} = \frac{1}{N} \sum_{n=1}^{N} (0 | 1) = P(h(\bar{x}) \neq f(x))$$

From Ch. 1.

2) From Ch. 1.

$$E(g(X)) = \sum_{j=1}^{N} g(X) P(X = x_h) = \frac{1}{N} \sum_{j=1}^{N} (2^{2} | 0^{2}) = \frac{1}{N} \sum_{j=1}^{N} (4|0)$$

$$\frac{h}{-1} = \frac{1}{4} \frac{1}{E(g(x))}$$

3)
$$m_{H}(N) \leq \frac{d^{N}}{d^{N}} \binom{N}{d}$$

if $d_{N_{0}} = 1 \Rightarrow \sum_{i=0}^{2} \binom{N}{i} = \binom{N}{N} + \binom{N}{N} = 1 + \frac{N!}{(N-1)!} = 1 + N$

if $d_{N_{0}} = 2 \Rightarrow \sum_{i=0}^{2} \binom{N}{i} = \binom{N}{N} + \binom{N}{N} + \binom{N}{N} = 1 + \frac{N!}{(N-1)!} + \frac{N!}{(N-2)!} = 1 + N + \frac{N(N-1)}{2} = 1 + N + \frac{N(N-1)}{2} = 1 + N + \frac{N(N-1)!}{2} = 1 + N +$