Exercise 2.8:

Part a:

A: \overline{g} represent the average of the function $g(g_1, g_2, g_3 \dots g_n)$, so \overline{g} is linear combination of

 g_1 , g_2 , g_3 ... g_n . Since H is closed under linear combination, and g_1 , g_2 , g_3 ... g_n belongs to H, so \overline{g} has to be in the H.

Part b:

A:
$$H = \{g_1, g_2\}$$

$$g_{_{1}} = 0$$

$$g_{2} = 1$$

 \overline{g} in this example is not in H.

Part c:

 \overline{g} can be a binary function but it won't always be a binary function. As part b as example, \overline{g} is not a binary function, but if there's another g_3 in H and g_3 = -1. Now \overline{g} is a binary function.

Problem 2.14

Part a:

So we know
$$d_{vc}(H) \leq \sum_{1}^{K} d_{vc}(H_{1})$$

Each Hi has a VC dimension of dvc

$$\sum_{i=1}^{K} d_{vc}(H_i) = K * d_{vc}$$

$$d_{vc}(H) \le K * d_{vc}$$

Since K is a positive number, if we add K to the right side the less or equal sign will become less sign.

So:
$$d_{vc}(H) < K * d_{vc} + K$$

$$d_{yc}(H) < K * (d_{yc} + 1)$$

Part b:

We know:
$$m_{_H}(l) <= l^{^{d}_{_{vc}}} + 1$$

Since K and $l^{d_{vc}}$ are both positive number

$$l^{d_{vc}} + 1 \le K(l^{d_{vc}} + 1)$$

$$m_{_{H}}(l) <= K l^{d_{vc}} + K$$

$$Kl^{d_{vc}} >= K$$
, so $Kl^{d_{vc}} + K < = 2Kl^{d_{vc}}$

From the question we know $2^{1} > 2Kl^{d_{vc}}$

So
$$m_{_H}(l)$$
 < 2^I

Since
$$m_{_H}(l) \le 2^{\Lambda}d_{_{VC}}(H)$$

So $d_{_{VC}}(H) \le 1$

Part c:

From part a we know $d_{vc}(H) < K * (d_{vc} + 1)$, so only need to prove $d_{vc}(H) <= 7(d_{vc} + K)log_2 d_{vc} K$. We set I from part b to $7(d_{vc} + K)log_2 d_{vc} K$.

Plug into $2^{l} > 2Kl^{d_{vc}}$

We have $2^{\Lambda}(7(d_{vc}+K)log_2d_{vc}K) > 2K(7(d_{vc}+K)log_2d_{vc}K)^{d_{vc}}$

Take log from each side

$$\begin{aligned} &7d_{vc}\log_{2}d_{vc}\mathsf{K} + 7\mathsf{K}log_{2}d_{vc}\mathsf{K} > 1 + \log_{2}\mathsf{K} + 7\log_{2}d_{vc} + \log_{2}d_{vc}(d_{vc} + \mathsf{K}) + \ \log_{2}\log_{2}d_{vc}\mathsf{K} \\ &\mathrm{Since}\ d_{vc}\log_{2}d_{vc}\mathsf{K} > 1 \\ &d_{vc}\log_{2}d_{vc}\mathsf{K} > \log_{2}\mathsf{K} \\ &d_{vc}\log_{2}d_{vc}\mathsf{K} > \log_{2}\log_{2}d_{vc}\mathsf{K} \\ &d_{vc}\log_{2}d_{vc}\mathsf{K} > 7\log_{2}d_{vc} \\ &\mathsf{K} > 7\log_{2}d_{vc} \end{aligned}$$

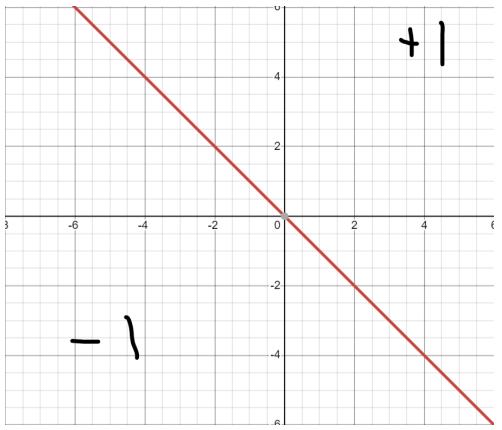
So left side must be greater than right side, so 2^I > 2KI^d_{vc} when I = $7(d_{vc}+K)log_2d_{vc}K$, so we can use conclusion from part b $d_{vc}(H)$ <= I, at I = $7(d_{vc}+K)log_2d_{vc}K$, so $d_{vc}(H)$ <= $7(d_{vc}+K)log_2d_{vc}K$

So we proved $d_{vc}(H) < K * (d_{vc} + 1)$ from part 1, and $d_{vc}(H) <= 7(d_{vc} + K)log_2 d_{vc} K$ from part b.

Problem 2.15

Part a:

$$h(x_1, x_2) = sign(x_1 + x_2)$$



Part h

From the hint, assume we generated the first point x_a and generating the second point by increase x_1 of x_a , and decreasing x_2 of x_a , and continue doing this after generating N points. Any two points in those N points won't have any relation. Thus, m(H) = 2^N, $d_{vc} = \infty$

Problem 2.24

Part a:

Give
$$\mathcal{D} = \{(x_1, x_1^2), (x_2, x_2^2)\}$$
. And $g(x) = ax + b$ $x_1^2 = ax_1 + b$ $x_2^2 = ax_2 + b$ There we know $a = x_1 + x_2$ $b = -x_1x_2$ $g(x) = (x_1 + x_2)x - x_1x_2$ $\overline{g}(x) = \frac{1}{2} * \frac{1}{2} \int_{-1}^{1} \int_{-1}^{1} (x_1 + x_2)x - x_1x_2 dx_1 dx_2 = 0$

Part b:

$$E_{\text{out}}(g^{(\mathcal{D})}) = \mathbb{E}_{\mathbf{x}} \left[(g^{(\mathcal{D})}(\mathbf{x}) - f(\mathbf{x}))^2 \right]$$

$$\mathsf{bias}(\mathbf{x}) = (\bar{g}(\mathbf{x}) - f(\mathbf{x}))^2,$$

$$\operatorname{var}(\mathbf{x}) = \mathbb{E}_{\mathcal{D}}[(g^{(\mathcal{D})}(\mathbf{x}) - \bar{g}(\mathbf{x}))^2],$$

Iterate N time, each time generate two random points in range [-1,1] each time. Get gout in iteration, and compute the average after iteration. calculate g(x) by getting average of all g(x), compute bias by using g(x) and given g(x), compute var by compare g(x) to each g(x) Part d:

$$E_{\text{out}}(g^{(\mathcal{D})}) = \mathbb{E}_{\mathbf{x}} \left[(g^{(\mathcal{D})}(\mathbf{x}) - f(\mathbf{x}))^2 \right]$$

$$g(x) = (x1+x2)x - x1x2$$

 $f(x) = x^2$
Eout = E[((x1+x2)x - x1x2 -x^2)^2]

$$= \frac{1}{2} \cdot \frac{1}{2} \cdot \frac{1}{2} \cdot \int_{1}^{1} \int_{1}^{1} ((x1+x2)x - x1x2 - x^2)^2 dx 1 dx 2 dx$$

= 0.533

$$\mathsf{bias}(\mathbf{x}) = (\bar{g}(\mathbf{x}) - f(\mathbf{x}))^2,$$

$$\overline{g}(x) - f(x) = (0-x^2)$$

Bias =
$$\frac{1}{2} \int_{1}^{1} x^{2}$$

=0.2

$$\operatorname{var}(\mathbf{x}) = \mathbb{E}_{\mathcal{D}}[(g^{(\mathcal{D})}(\mathbf{x}) - \bar{g}(\mathbf{x}))^2],$$

Var =
$$\frac{1}{2} \frac{1}{2} \frac{1}{2} \int_{-1-1-1}^{1} \int_{-1}^{1} (x1+x2)x - x1x2)^2 dx1dx2 dx$$

Var = 0.333