

Machine Learning -- Assignment 2

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Question 1

Test hypothesis h has errors $r = 240$ on sample S of $n = 800$.

$$\begin{aligned} error_s(h) &= \frac{r}{n} \\ &= \frac{240}{800} \end{aligned}$$

$$error_s(h) = 0.3$$

$$\begin{aligned} \text{The standard deviation for } error_s(h) &= \sqrt{\frac{r*(1-r)}{n}} \\ &= \sqrt{\frac{0.3*0.7}{800}} \end{aligned}$$

$$\text{Standard deviation for } error_s(h) = 0.0162$$

Example in class had $n = 40$ and $r = 12$

$$error_s(h) = \frac{12}{40} = 0.3$$

$$\text{Thus, standard deviation for } error_s(h) = \sqrt{\frac{0.3*0.7}{40}} = 0.0725$$

As no other information is given, most probable value of $error_D(h)$ is $error_s(h)$. this also assumes standard unbiased estimator for $error_D(h)$ and data sample S is independent of discrete-valued hypothesis h .

Question 2

Sample S contains $n = 100$.

Incorrect classification, $r = 100 - \text{correct classification}$.

$$r = 100 - 79.$$

$$\text{Error, } r = 21.$$

95% confidence Implies, $N\% = 95, Z_{0.95} = 1.96$

$$\begin{aligned} \text{Standard deviation} &= \sqrt{\frac{error_s(h)*(1-error_s(h))}{n}} \\ &= \sqrt{\frac{(0.21)*(1-0.21)}{100}} = 0.0407 \end{aligned}$$

$$error_D(h) = error_s(h) \mp Z_N \cdot \sqrt{\frac{error_s(h)*(1-error_s(h))}{n}}$$

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$$= 0.21 \mp (1.96) * (0.0407)$$

$$\text{error}_D(h) = 0.21 \mp 0.0798$$

$$\text{Upper bound} = 0.21 + 0.0798 = 0.29$$

$$\text{Lower bound} = 0.21 - 0.0798 = 0.1302$$

Question 3

Range of interval, $L = 0.3$ and $U = 0.6$

$$\text{Therefore, midpoint, } M = (L + U)/2 = 0.45 = P$$

$$\text{And } (1 - p) = 1 - 0.45 = 0.55$$

$$\text{Lower bound, } L = p - z \sqrt{\frac{p(1-p)}{n}} \quad \text{upper bound, } U = p + z \sqrt{\frac{p(1-p)}{n}}$$

$$\text{Confidence interval width is } U - L = 2z \sqrt{\frac{p(1-p)}{n}}$$

$$95\% \text{ of two level confidence, we use } Z_{0.975} = 1.96$$

$$\text{Minimum number of examples } n \geq \frac{4 * z^2 * p * (1-p)}{(U-L)^2}$$

$$n \geq \frac{4 * (1.96)^2 * 0.45 * 0.55}{(0.1)^2}$$

$$n \geq 380.3$$

Thus minimum number of examples needed to collect should be **greater than or equal to 380**.

Question 4

$$r = 10, n = 75$$

90% 2-sided true error:

$$\text{error}_D(h) = \text{error}_S(h) \mp Z_N \cdot \sqrt{\frac{\text{error}_S(h) * (1 - \text{error}_S(h))}{n}}$$

$$= \frac{10}{75} \mp (1.64) * \sqrt{\frac{(0.133) * (0.867)}{75}}$$

$$= 0.133 \mp (1.64) * 0.0392$$

$$\text{error}_D(h) = 0.133 \mp 0.06429$$

95% 1-sided true error:

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$$\begin{aligned} error_D(h) &= error_S(h) + Z_{.90} \cdot \sqrt{\frac{error_S(h) \cdot (1 - error_S(h))}{n}} \quad \dots \text{consider only Upper bound} \\ &= 0.133 + (1.64) \cdot 0.0392 \\ \mathbf{error_D(h)} &= 0.133 + 0.064 = 0.197 \end{aligned}$$

80% 1-sided true error:

$$\begin{aligned} error_D(h) &= error_S(h) + Z_{.80} \cdot \sqrt{\frac{error_S(h) \cdot (1 - error_S(h))}{n}} \quad \dots \text{consider only Upper bound} \\ &= 0.133 + (1.28) \cdot 0.0392 \\ \mathbf{error_D(h)} &= \mathbf{0.133 + 0.0501 = 0.1831} \end{aligned}$$

Question 5

$V_c(H) = 3$ V_c Dimension for linear separator in 2-dimensional.

$$\delta = (100 - 90)\% = 0.1 \epsilon = 0.05$$

For Upper bound,

$$\begin{aligned} m &\geq \frac{1}{\epsilon} \left(4 \log_2 \left(\frac{2}{\delta} \right) + 8 V_c(h) \log_2 \left(\frac{13}{\epsilon} \right) \right) \\ m &\geq \frac{1}{0.05} \left(4 \log_2 \left(\frac{2}{0.1} \right) + 8 V_c(h) \log_2 \left(\frac{13}{0.05} \right) \right) \\ m &\geq \frac{1}{0.05} (4 * 4.321 + 8 * 3 * 8.022) \\ m &\geq 20 * (17.28 + 192.53) \\ m &\geq 4196.33 \end{aligned}$$

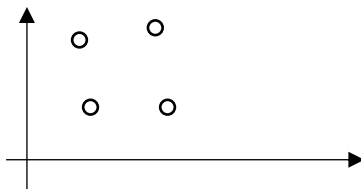
This bound does not seem to be realistic.

Because a hyperplane in 2 dimension is a line which has to be defined by 2 points. So if 90% confidence with at most 5% error will break the line. So this might be the reason to sound unrealistic.

Question 6

(a) For rectangle,

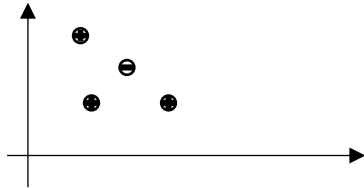
Consider point in XY - plane and $a < x < b$ and $c < y < d$



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Consider that 3 points are positive and a point is negative in the following case. Here it is not possible to shatter the points in $XY - plane$.



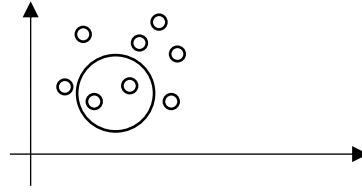
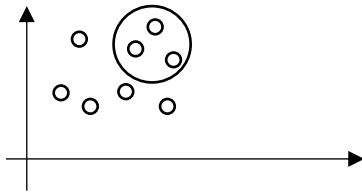
Hence, the $V_c(h) = 4$

(b) For circle,

Points in circle are positive and outside is negative.

$H_c = \text{circle in } XY - \text{plane}$

V_c dimension for circle in $XY - plane$ is at least 3, as 3 points make up non-degenerated triangle.



It is possible to shatter 1,2 or 3 positive points so,

$$V_c = 3$$

Question 7

(a)

Region bounded by point (0,0) and (n,n) in the interval (0,100) implies that $n = 101$.

$$\delta \text{ with probability } 95\% = 100 - 95 = 0.05$$

$$\delta = 0.05$$

$$|h| = \left(\frac{n(n+1)}{2}\right)^2$$

$$\epsilon = 0.15$$

Size of hypothesis m is calculated as:

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$$m \geq \frac{1}{\epsilon} \left[\ln |H| + \ln \left(\frac{1}{\delta} \right) \right]$$

$$m \geq \frac{1}{0.15} \left[\ln \left(\frac{n(n+1)}{2} \right)^2 + \ln \left(\frac{1}{0.05} \right) \right]$$

$$m \geq \frac{1}{0.15} \left[\ln \left(\frac{101(101+1)}{2} \right)^2 + \ln(20) \right]$$

$$m \geq \frac{1}{0.15} [\ln(5151)^2 + \ln(20)]$$

$$\mathbf{m \geq 133.93}$$

(b)

Region bounded by point (0,100) a read values, $n = \infty$

$$V_c(h) \leq 4 \text{ if } a < x < b : z = 1 \text{ else } z = 0$$

$$a < x < b : z = 0 \text{ else } z = 1$$

$$\text{if } c < y < d : z = 1 \text{ else } z = 0$$

$$c < y < d : z = 0 \text{ else } z = 1$$

$$\mathbf{\delta = 0.05}$$

$$\epsilon = 0.15$$

$$m \geq \frac{1}{\epsilon} (4 \log_2 \left(\frac{2}{\delta} \right) + 8 V_c(h) \log_2 \left(\frac{13}{\epsilon} \right))$$

$$m \geq \frac{1}{0.15} (4 \log_2 \left(\frac{2}{0.05} \right) + 8 * 4 * \log_2 \left(\frac{13}{0.15} \right))$$

$$m \geq \frac{1}{0.15} (4 * 5.321 + 8 * 4 * 6.437)$$

$$m \geq \frac{1}{0.15} (227.26)$$

$$\mathbf{m \geq 1515.12}$$

Question 8

(a)

The tree has depth 2 and 4 leaves in total.

So, syntactically distinct trees are $2^4 n(n-1)$

$$f(n) = 2^4 n(n-1)$$

$$H_{rd2} = f(x)$$

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(b)

$$|h_{rd2}| = 2^4 n(n-1)$$

Upper bound for number of examples, m with $error = \epsilon$ & $confidence = \delta$

$$m \geq \frac{1}{\epsilon} \left[\ln |h_{rd2}| + \ln \left(\frac{1}{\delta} \right) \right]$$

$$m \geq \frac{1}{\epsilon} \left[\ln 2^4 n(n-1) + \ln \left(\frac{1}{\delta} \right) \right]$$

Question 9

(a)

$$n = 100. N\% = 95\%$$

$error_D(h)$ should be calculated to find the error.

$$error_S(h) = \frac{r}{n} = \frac{0}{100} \quad \dots \text{given}$$

As $error_S(h) = 0$, standard deviation = 0 and true error is also 0.

This implies that it is difficult to calculate the true error with 95% probability as $error_D(h) = 0$

$$error_D(h) = error_S(h) \mp Z_N \cdot \sqrt{\frac{error_S(h) \cdot (1 - error_S(h))}{n}}$$

$$Z_N = 1.96$$

$$error_D(h) = error_S(h) \mp 1.96 \cdot \sqrt{\frac{error_S(h) \cdot (1 - error_S(h))}{n}}$$

$$error_D(h) = 0$$

$$p[error_D(h) > error_D(h) + \epsilon] \leq e^{-2m\epsilon^2}$$

$$p[error_D(h) > error_D(h) + \epsilon] \geq 95\%$$

Hence it is difficult to calculate the true error since $error_S(h) = 0$ it is difficult to find the interval.

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(b)

$$n = 100 \text{ \& } r = 30$$

$$error_s(h) = \frac{r}{n} = 0.3$$

$$N\% = 90\% \text{ implies } Z_{.90} = 1.96$$

$$error_D(h) = error_s(h) \mp Z_N \cdot \sqrt{\frac{error_s(h) * (1 - error_s(h))}{n}}$$

$$error_D(h) = 0.3 \mp 1.96 \cdot \sqrt{\frac{0.3 * 0.7}{100}}$$

$$error_D(h) = 0.3 \mp 0.0898$$

Upper bound = 0.3898 & Lower bound = 0.2101

The interval in which this true error will fall is 0.2101 to 0.3898