

MIES - CT1

18EC10054

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1. a) Total instances possible = 72
 Total concepts possible = 2^{72}

b) Adding two other attribute values (?, ϕ) to each space:
 Total hypotheses = $4 \times 5 \times 4 \times 4 \times 5 = 1600$

All hypotheses containing even a single ϕ are the same

\therefore statistically different = $1 + (3 \times 4 \times 3 \times 3 \times 4) = 1 + 432$
 $= 433$ Ans

c) We apply Find-s on the given data.

$S_0 = \langle \phi, \phi, \phi, \phi, \phi \rangle$ [Most specific hypothesis]

Training on ex-1: Nothing changes (-ve ex), $S_1 = S_0$

Training on ex-2: $S_2 = \langle \text{many, big, no, exp, one} \rangle$

Training on ex-3: $S_3 = S_2$ (-ve ex, nothing changes)

Training on ex-4: $S_4 = \langle \text{many, ?, no, exp, ?} \rangle$

Training on ex-5: $S_5 = \langle \text{many, ?, no, ?, ?} \rangle$

This is the most specific final hypothesis.

x_i	S_i	A_1	A_2	A_3	A_4	T
1	High	High	Good	High	73	Pass
2	High	High	Fair	Low	60	Fail
3	Low	Low	Good	High	64	Pass
4	Low	Low	Good	Low	57	Fail
5	High	High	Fair	High	66	Pass
6	Low	Low	Fair	High	59	Fail

Values (A_1) = High, Low

$$A_1 = [3+, 3-]$$

$$A_{1, \text{High}} = [3+, 1-]$$

$$A_{1, \text{Low}} = [1+, 2-]$$

$$\text{Gain}(S, A_1) = \text{Entropy}(A_1) - \frac{1}{2} \text{Entropy}(A_{1, \text{High}}) - \frac{1}{2} \text{Entropy}(A_{1, \text{Low}})$$

$$= -\frac{1}{2} \log_2\left(\frac{1}{2}\right) - \frac{1}{2} \log_2\left(\frac{1}{2}\right) - \frac{1}{2} \left[-\frac{2}{3} \log_2\left(\frac{2}{3}\right) - \frac{1}{3} \log_2\left(\frac{1}{3}\right) \right. \\ \left. - \frac{2}{3} \log_2\left(\frac{2}{3}\right) - \frac{1}{3} \log_2\left(\frac{1}{3}\right) \right]$$

$$= 1 + \left(\frac{2}{3} \log_2\left(\frac{2}{3}\right) + \frac{1}{3} \log_2\left(\frac{1}{3}\right) \right)$$

$$= 1 - .918$$

$$= \underline{\underline{0.082}}$$

Similarly for A_2 ,

$$A_2 = [3+, 3-]$$

$$A_{2, \text{Good}} = [2+, 1-]$$

$$A_{2, \text{Fair}} = [1+, 2-]$$

Being exactly similar to A_1 , $\text{Gain}(S, A_2) = \underline{\underline{0.082}}$

For $A3$,

$$A3 = [3+, 3-]$$

$$A3_{\text{High}} = [3+, 1-]$$

$$A3_{\text{Low}} = [0+, 2-]$$

$$\begin{aligned} \text{gain}(s, A3) &= 1 + \left(\frac{4}{6} \log \left(\frac{3}{4} \log \left(\frac{3}{4} \right) + \frac{1}{4} \log \left(\frac{1}{4} \right) \right) + \frac{2}{6} (0) \right) \\ &= 1 - 0.541 \end{aligned}$$

$$\text{gain}(s, A3) = \underline{\underline{0.459}}$$

We use the comparator $c = 65$ to convert A_4 from a continuous attribute to a discrete attribute.

$$A_4 = [3+, 3-] \quad [-5, 10] =$$

$$A_4 < 65 = [1+, 3-]$$

$$A_4 > 65 = [2+, 0-]$$

Being exactly similar to in statistics to A_3 , we find

$$\text{Gain}(S, A_4) = \underline{\underline{0.459}}$$

By having the lowest attribute number for the highest entropy gain, we see that A_3 is our first node.

Entropy($A_{3\text{low}}$) = 0, hence no further nodes are needed in that branch

$$\text{Entropy}(A_{3\text{high}}) = 0.811$$

Now we evaluate A_1 , A_2 and A_4 under $A_{3\text{high}}$

$$\begin{aligned} \text{Gain}(A_{3\text{high}}, A_1) &= 0.811 + \frac{1}{2} \left(\frac{1}{2} \log \frac{1}{2} \right) \\ &= 0.561 \end{aligned}$$

$$\begin{aligned} \text{Gain}(A_{3\text{high}}, A_2) &= 0.811 + \frac{1}{2} \left(0 + \frac{1}{2} \log \left(\frac{1}{2} \right) \right) \\ &= 0.561 \end{aligned}$$

$$\begin{aligned} \text{Gain}(A_{3\text{high}}, A_4) &= 0.811 + \frac{1}{2} \left(0 + \frac{1}{2} \log \left(\frac{1}{2} \right) \right) \\ &= 0.561 \end{aligned}$$

By similar rules as before, we pick $A1$ as the second node on $A3_{high}$ branch.

$A3_{high} \rightarrow A1_{high}$ has 0 entropy, so no further nodes needed here

We evaluate $A2$ and $A4$ on $A3_{high} \rightarrow A1_{low}$

$$\text{Gain}(A3_{high} \rightarrow A1_{low}, A2) = 1 - 1 = 0$$

= 0, which is perfect classification for both $A2_{good}$ & $A2_{pass}$.

Final tree

