

①

$$\text{Entropy}(S) = -(P_{\oplus}) \log_2 P_{\oplus} - P_{\ominus} \log_2 P_{\ominus}$$

↓  
+ means 1 in target  
- means 0 in target

4+  
2-

$$S: [4+, 2-]$$

$$E = -\frac{4}{6} \times \log_2 \frac{4}{6} - \frac{2}{6} \times \log_2 \frac{2}{6} = -\frac{4}{6} \times (-0.585) - \frac{2}{6} \times (-1.585)$$

$$= 0.389 + 0.528 = 0.917$$

		Target	
		1	0
Weather	cold	3	0
	hot	1	2

$$E(T, X) = \sum_n P(n) E(n)$$

$$E(3, 0) = 0, E(1, 2) = -\frac{1}{3} \log_2 \frac{1}{3} - \frac{2}{3} \log_2 \frac{2}{3}$$

$$= -\frac{1}{3} \times 0.585 - \frac{2}{3} \times (-0.585) = 0.528 + 0.389 = 0.917$$

$$\Rightarrow E(\text{target}, \text{weather}) = P(\text{cold}) E(3, 0) + P(\text{hot}) E(1, 2)$$

$$= \frac{3}{6} \times 0 + \frac{3}{6} \times 0.917 = 0.4585$$

		Target	
		1	0
price	cheap	2	0
	exp.	2	2

$$E(2, 0) = 0$$

$$E(2, 2) = -\frac{2}{4} \log_2 \frac{2}{4} - \frac{2}{4} \log_2 \frac{2}{4} = 1$$

$$E(\text{target}, \text{price}) = P(\text{cheap}) E(2, 0) + P(\text{exp}) E(2, 2) = 0 + \frac{4}{6} = \frac{4}{6} = \frac{2}{3}$$

$$= 0.667$$



target  
1 0  
weight light 2 1  
heavy 2 1

~~$$E(2,1) = -\frac{2}{3} \log_2 \frac{2}{3} - \frac{1}{3} \log_2 \frac{1}{3}$$~~

~~$$= 0.917$$~~

$$E(2,1) = -\frac{2}{3} \log_2 \frac{2}{3} - \frac{1}{3} \log_2 \frac{1}{3} = 0.917$$

~~$$E(\text{target}, \text{weight}) = P(\text{light}) E(2,1) + P(\text{heavy}) E(2,1)$$~~

~~$$= \frac{1}{2} \times 0.917 + \frac{1}{2} \times 0.917 = 0.917$$~~

$$E(\text{target}, \text{weight}) = P(\text{light}) E(2,1) + P(\text{heavy}) E(2,1)$$

$$\frac{1}{2} \times 0.917 + \frac{1}{2} \times 0.917 = 0.917$$

$$\text{Information gain} = H(\text{target}) - E(\text{target}, \boxed{\phantom{0.917}})$$

weather

$$\Rightarrow 0.917 - 0.4585 = 0.4585$$

price

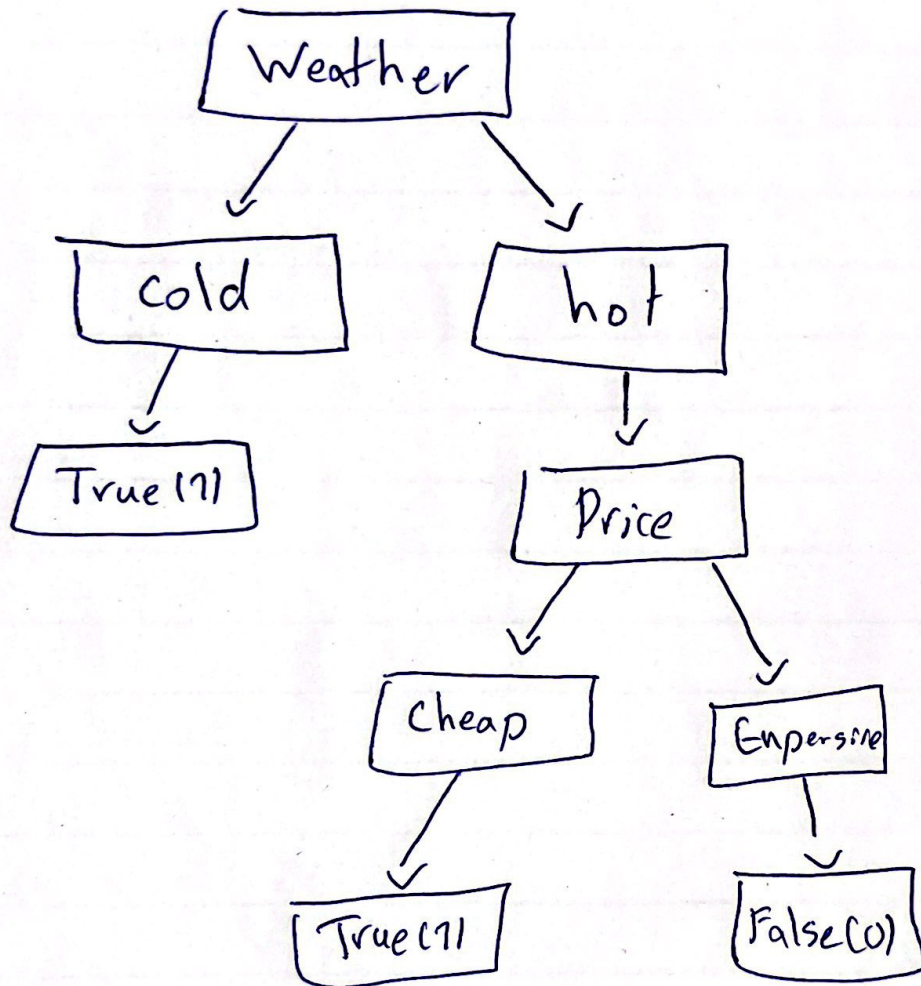
$$\Rightarrow 0.917 - 0.667 = 0.25$$

weight

$$\Rightarrow 0.917 - 0.917 = 0$$



①



۹۷۳۲۴۵۵

سایان ویدیو

③ ML Learning vs MAP Learning;

~~MAP Learning~~

در ML وقتی مشاهدات ما زیاد می شود، اهمیت  $P(h_i)$  کم شده و تقریباً می توانیم آن را  
از بین ببریم و فقط  $P(d|h_i)$  برای ما مهم است در MAP  $P(d|h_i)P(h_i)$  را  
ماکسیم کردن  
را minimize می کنیم.



④  $\forall s, u$  [  $\text{student}(s, u) \wedge \neg \text{cheat}(s) \wedge \text{studied}(s, u) \Rightarrow \text{pass}(s)$  ]

Diagram:  $\text{student}(s, u)$  with  $s$  pointing to a circled 's' and  $u$  pointing to 'university'.  $\text{studied}$  points to a circled 's'.

②  $\text{student}(\text{Ashkan}, \text{Shiraz university})$

③  $\forall s, u$  [  $\text{student}(s, u) \wedge \text{valid}(u) \rightarrow \neg \text{cheat}(s)$  ]

④  $\text{study}(\text{Ashkan})$

⑤  $\text{valid}(\text{Shiraz university})$

convert to CNF.

①  $\Rightarrow \neg \text{student}(s) \vee \text{cheat}(s) \vee \neg \text{student}(s, u) \vee \text{pass}(s)$

③  $\Rightarrow \neg \text{student}(s, u) \vee \neg \text{valid}(u) \vee \neg \text{cheat}(s)$

now we add our rules with  $\neg \text{pass}(\text{Ashkan})$  so we reach a False statement

$(\neg \text{student}(s) \vee \text{cheat}(s) \vee \neg \text{student}(s, u) \vee \text{pass}(s)) \wedge (\neg \text{student}(s, u) \vee \neg \text{valid}(u) \vee \neg \text{cheat}(s))$

$\neg \text{student}(s) \wedge (\neg \text{student}(s, u) \vee \neg \text{cheat}(s)) \vee \neg \text{valid}(u) \vee \neg \text{cheat}(s) \vee \neg \text{pass}(\text{Ashkan})$



$\Rightarrow \sim \text{student}(\text{Ashkan}, \text{Shiraz}) \vee \sim \text{valid}(\text{Shiraz}) \wedge \sim \text{pass}(\text{Ashkan})$

$\Rightarrow \sim \text{student}(\text{Ashkan}, \text{Shiraz}) \wedge \sim \text{pass}(\text{Ashkan})$

Qab rule (سبب) است پس  $\text{pass}(\text{Ashkan})$  درستی

(5)

snow(Sanandaj)

$\forall n [At(Kordestan, n) \rightarrow cold(n)]$

$\forall n [cold(n) \wedge cloudy(n) \rightarrow snow(n)]$

cloudy(Sanandaj)

center of (Sanandaj, Kordestan)

$\forall n, y [center(n, y) \rightarrow At(n, y)]$

~~Q1~~

