

## Example

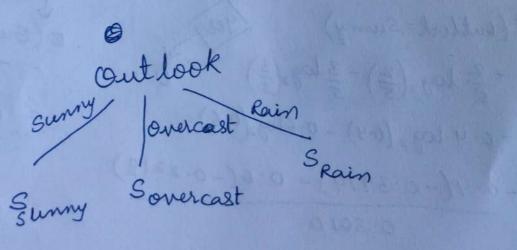
Day	Outlook	Temperature	Humidity	Wind	PlayTennis
D1	Sunny	Hot	High	Weak	No
D2	Sunny	Hot	High	Strong	No
D3	Overcast	Hot	High	Weak	Yes
D4	Rain	Mild	High	Weak	Yes
$D_5$	Rain	Cool	Normal	Weak	Yes
D6	Rain	Cool	Normal	Strong	No
D7	Overcast	Cool	Normal	Strong	Yes
D8	Sunny	Mild	High	Weak	No
D9	Sunny	Cool	Normal	Weak	Yes
D10	Rain	Mild	Normal	Weak	Yes
D11	Sunny	Mild	Normal	Strong	Yes
D12	Overcast	Mild	High	Strong	Yes
D13	Overcast	Hot	Normal	Weak	Yes
D14	Rain	Mild	High	Strong	No

```
Decision Tree Model Building
    Entropy (s) = - P + log 2 P + - P - log 2 P -
    Entropy ([94, 5-]) = -9/14 log_2(9/14) - 5 log_2(5/14)
        = -0.6428 \log (0.6428) - 0.3571 \log (0.3571) \log (0.3571)
       = -0.6428 x (-0.1919) - 0.3571 x(-0.4472)
                          6.3010
          0.1283+0.1597
               \frac{23370.1597}{0.3010} = \frac{0.283}{0.3010} = 0.9401
 Gain (s, Outlook) = Entropy (s) - & |So| Entropy (So)
                                         3 Sunny,
                                         Overcast.
                                         Rainz
                   [9+,5-]
                Outlook
         Sunny overcast Rein
                                     ¿D4, D5, D6.
   ₹D1, D2, D8, D9, ₹ D3, D7,
                                       D10, D143
                    D12, D13 ?
                                     [3+,2-]
     [2+,3-] [4+,0-]
                                           E (Outlook = Rain)
 E (Outlook = Sunny)
                          (yes)
                                             = -\frac{3}{5}\log_2(\frac{3}{5}) - \frac{2}{5}\log_2(\frac{2}{5})
 = -\frac{2}{5} \log_2(\frac{2}{5}) - \frac{3 \log_2(\frac{3}{5})}{5}
= -0.4 log 2 (0.4) -0.6 log 2 (0.6)
                                           = 0.9707
= -0.4(-0.3979)-0.6(-0.2218)
              0.3010
= 0.9707
```

Gain (s, outlook) =

$$Entropy(s) - |S_{sumy}| = (outlook \ge Sumny)$$
 $- |S_{overcast}| = (outlook = Overcast)$ 
 $- |S_{Rain}| = (outlook = Rain)$ 
 $= 0.94 - \frac{5}{14}(0.9707) - \frac{4}{14}(0) - \frac{5}{14}(0.9707)$ 
 $= 0.94 - 0.6933$ 
 $= 0.2468 \Rightarrow Gain(S, outlook) = 0.2468$ 
 $= 0.2468 \Rightarrow Gain(S, wind) = 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 
 $= 0.084$ 

As "outlook" provides the best prediction, we pick "outlook" attribute to split the root node into three branches viz., "sunny, overcast, rain



sunny

Temperature	Humidity	wind	playtennis
Hot	High High High Normal Normal	Weak	No
Hot		strong	No
Mild		Weak	No
cool		Weak	Yes
Mild		Strong	Yes

Sovercast

	Humidity \	wind	Play Tenn	15
Temperature	Harracas		Yes	7
Hot	High	Weak		/
Cool	Normal	strong	Yes	>
Mild	High	strong		
	Normal	Weak	1 Yes	)
Hot				

9 Rain

	Humidity	wind	Playtennis
Temperature		Weak	Yes
Mild	High	Weak	Yes .
cool	Normal	strong	No
cool	Normal	weak	Yes
Mild	Itigh	strong	No

$$E(S_{unny}) = -\frac{3}{5} log_2(\frac{3}{5}) - \frac{2}{5} log_2(\frac{2}{5}) = 0.971$$

Gain (Sunny, Temperature)

$$= E(S_{unny}) - \int_{-2}^{2} E(0,2) + 2 E(1,1) + 1 E(1,0)$$

= 
$$E(S_{sunny}) - \left[\frac{2}{5}E(0,2) + \frac{2}{5}E(1,1) + \frac{1}{5}E(1,0)\right]$$
  
=  $0.971 - 0.4 = 0.571$ 

Grain (Sunny, Temperature) = 0.571

Illy Gain (Sunny, Humidity) = 0.971

Grain (Sunny, Windy) = 0.020

