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based on Hurnighty - 3 brouches! his Home work 2.

1, * Before Splitting: 14 samples -> 7 sunny, 7 rainy • Probability of sunny: 7/14 = 1/2• Probability of rainy: 7/14 = 1/2

> Entropy before Splitting: $H(X) = -\sum_{k=1}^{K} p(X=x_k) \log_2 p(X=x_k)$

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

(C) Split based on Wind (2 banbranches: Yes and No)

· Windy (Yes): 6 samples > 1 sunny, 5 rainy

· Probability of sunny: 1/6

· Probability of rainy : 5/6

= Entropy of windy: $H(X) = -\left(\frac{1}{6}\log_2\left(\frac{1}{6}\right) + \frac{5}{6}\log_2\left(\frac{5}{6}\right)\right) = 0.65$ bit · Not Windy (No): 8 samples -> 6 sunny, 2 rainy

· Probability of sunny: 6/8 = 3/4

· Probability of rainy: 2/8 = 1/4

> Entropy of Not Windy: H(X) = - (3 log2(3) + 4 log2(4)) = 0.81 bit

 \rightarrow Weighted Average: $E(H(x)) = \left(\frac{6}{14} \times 0.65 + \frac{8}{14} \times 0.81\right) = 0.74$ bit

@ Split based on Humidity (3 branches: high, mild, low)

· high: 5 samples > 1 sunny, 4 rainy => P(sunny)=115; P(rainy)=415 > # Entropy of high: T(X)=-(1/5 log_2(1/5) + (4/5) log_2(4/5)) = 0.72 bit

· mild: 5 samples -> 3 sunny, trainy => P(sunny) = 315; p(rainy) = 2/5 => Entropy of mild: #(x)= (3) + 2 log (2) = 0.97 bit

-> Weighted A verage: E(H(x)) =

• nuild: 4 samples
$$\rightarrow 2$$
 sunny, 2 rainy $\rightarrow p$ (sunny)= $2/4 = 1/2$; $p(rainy) = 2/4 = 1/2$
 \rightarrow Entropy of nuild: $H(x) = -\left(\frac{1}{2}\log_2(\frac{1}{2}) + \frac{1}{2}\log_2(\frac{1}{2})\right) = 1$ bit

• low: 5 samples
$$\Rightarrow$$
 0 sunny, 5 rainty \Rightarrow p(sunny)=0; p(rainy)=1.
 \Rightarrow Entropy of low: $H(x) = -(0 \times \log_2(0) + 1 \times \log_2(1)) = 0$ bif

- Entropy of low:
$$H(x) = -(0 \times \log_2(0) + 1 \times \log_2(1)) = 0$$
 bit

@ Comparisons between features;

	Temp	Humidity	Wind
Entropy before split:	$= \left(\overline{1}_{8-0} \times \frac{\mathbb{A}}{5} + \right)$	E(H(X))= 1 1 = 0.62	- Weighteel Microge
Entropy after split:	0.29	= \mu0.84 mu2 1	e respondent har beend tuge
Information Grain	1-0.29=0.71	0.16	lin la vacina e

is Temp (it gains the most => The best feature to put on the top of the tree information be based or