

Name: Karan Sikka (@ksikka)

15-381 F13

Due: 11/5/13

$$1) H(V) = - \left(\frac{\# \text{ pos}}{\text{total}} \right) \cdot \log \left(\frac{\# \text{ pos}}{\text{total}} \right) - \dots$$

so, 15 positives, 11 not, 26 total \Rightarrow

$$H(V) = - \left(\frac{15}{26} \right) \log \left(\frac{15}{26} \right) - \left(\frac{11}{26} \right) \log \left(\frac{11}{26} \right) = 0.983$$

$$2) IG(P) = H(V) - H(V|P)$$

$H(V)$ computed above, what is $H(V|P)$?

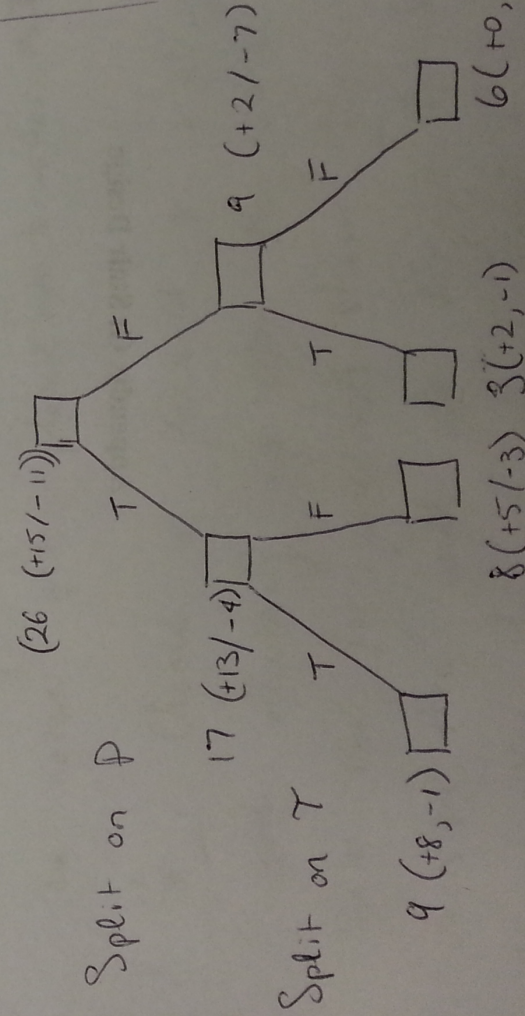
$$H(V|P) = \frac{17}{26} \left[-\frac{13}{7} \log_2 \left(\frac{13}{7} \right) - \frac{4}{17} \log_2 \left(\frac{4}{17} \right) \right] + \frac{9}{26} \left[-\frac{2}{9} \log_2 \left(\frac{2}{9} \right) - \frac{7}{9} \log_2 \left(\frac{7}{9} \right) \right]$$

$$\Rightarrow H(V|P) = 0.779 \Rightarrow IG(P) = 0.204$$

By similar computation,

$$H(V|T) = 0.806 \Rightarrow IG(T) = 0.177$$

3) We split in order of largest info gain.



Key:

A (+8, -6)

$\Rightarrow A = \#$ of data points

B = $\#$ of positive points

C = $\#$ of negative points

- 4) Collecting unimportant features would likely lead to overfitting, since the algorithm would think this data is more significant than it actually is.
- Making useless a continuous attribute would also provide the algorithm with unnecessary information to branch on, leading to overfitting.

- 5) Stopping growth of the tree is a good way to reduce overfitting.
- (Less likely to fit noise). This is because noise will have less information gain, and they will not get fit. ~~rather~~ the noise gets placed into broader data.

- 6) Yes. Choose a threshold x and y which effectively partition the data into the four quadrants shown. The decision tree could use discrete thresholds to partition the continuous space.
- 4 quadrants requires two levels. ✓