Vrund Patel

1 Theory

1.

(a)

$$\begin{split} H(Y) &= H(P(y=0), P(y=1)) = -P(y=0) * \log_2 P(y=0) + -P(y=1) * \log_2 P(y=1) \\ &= \left(-\frac{9}{21}\right) * \log_2 \frac{9}{21} + \left(-\frac{12}{21}\right) * \log_2 \frac{12}{21} \\ &= .98522 \end{split}$$

(b)
$$IG(A) = H(\frac{9}{21}, \frac{12}{21}) - \mathbb{E}(H(A))$$

Feature $x_1 = \{0, 1\} = \{F, T\}$

$$IG(A) = H\left(\frac{12}{21}, \frac{9}{21}\right) - \left(\frac{p_0 + n_0}{p + n}H\left(\frac{p_0}{p_0 + n_0}, \frac{n_0}{p_0 + n_0}\right) + \frac{p_1 + n_1}{p + n}H\left(\frac{p_1}{p_1 + n_1}, \frac{n_1}{p_1 + n_1}\right)\right)$$

$$IG(A) = H\left(\frac{12}{21}, \frac{9}{21}\right) - \left(\frac{5 + 8}{12 + 9}H\left(\frac{5}{5 + 8}, \frac{8}{5 + 8}\right) + \frac{7 + 1}{12 + 9}H\left(\frac{7}{7 + 1}, \frac{1}{7 + 1}\right)\right)$$

$$IG(A) = 0.98522 - \left(\frac{13}{21}H\left(\frac{5}{5 + 8}, \frac{8}{5 + 8}\right) + \frac{8}{21}H\left(\frac{7}{7 + 1}, \frac{1}{7 + 1}\right)\right)$$

$$IG(A) = 0.9852 - \left(\frac{13}{21}\left(-\frac{5}{13}*\log_2\left(\frac{5}{13}\right) + \left(-\frac{8}{13}*\log_2\left(\frac{8}{13}\right)\right)\right) + \frac{8}{21}H\left(\frac{7}{7 + 1}, \frac{1}{7 + 1}\right)\right)$$

$$IG(A) = 0.9852 - \left(0.5951 + \frac{8}{21}\left(-\frac{7}{8}*\log_2\left(\frac{7}{8}\right) + \left(-\frac{1}{8}*\log_2\left(\frac{1}{8}\right)\right)\right)\right)$$

$$IG(A) = 0.183$$

Feature
$$x_2 = \{0, 1\} = \{F, T\}$$

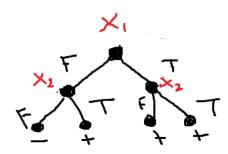
$$IG(A) = H\left(\frac{12}{21}, \frac{9}{21}\right) - \left(\frac{p_0 + n_0}{p + n}H\left(\frac{p_0}{p_0 + n_0}, \frac{n_0}{p_0 + n_0}\right) + \frac{p_1 + n_1}{p + n}H\left(\frac{p_1}{p_1 + n_1}, \frac{n_1}{p_1 + n_1}\right)\right)$$

$$IG(A) = 0.9852 - \left(\frac{5 + 6}{12 + 9}H\left(\frac{5}{5 + 6}, \frac{6}{5 + 6}\right) + \frac{7 + 3}{12 + 9}H\left(\frac{7}{7 + 3}, \frac{3}{7 + 3}\right)\right)$$

$$IG(A) = 0.9852 - \left(\frac{11}{21}\left(-\frac{5}{11} * \log_2\left(\frac{5}{11}\right) + \left(-\frac{6}{11} * \log_2\left(\frac{6}{11}\right)\right)\right) + \frac{10}{21}\left(-\frac{7}{10} * \log_2\left(\frac{7}{10}\right) + \left(-\frac{3}{10} * \log_2\left(\frac{3}{10}\right)\right)\right)\right)$$

$$IG(A) = 0.0449$$

(c)



2.

(a)
$$P(A = Yes) = \frac{3}{5} = 0.6$$

$$P(A = No) = \frac{2}{5} = 0.4$$

(b)

$$\begin{aligned} \text{Chars mean} &= \frac{^{216+69+302+60+393}}{5} = \frac{^{1040}}{5} = 208 \\ \text{Chars standard deviation} &= \sqrt{\frac{^{(216-208)^2+(69-208)^2+(302-208)^2+(60-208)^2+(393-208)^2}}{5-1}} = \\ &= \sqrt{\frac{^{84350}}{4}} = 145.21 \end{aligned}$$

Word Length mean =
$$\frac{5.68+4.78+2.31+3.16+4.2}{5} = 4.03$$
 Word Length standard deviation =
$$\sqrt{\frac{(5.68-4.03)^2+(4.78-4.03)^2+(2.31-4.03)^2+(3.16-4.03)^2+(4.2-4.03)^2}{5-1}}$$
 = 1.33

# of Chars Standardized	Average Word Length	Give an A
	Standardized	

$$\frac{216 - 208}{145.21} = 0.055$$

$$\frac{69 - 208}{145.21} = -0.957$$

$$\frac{302 - 208}{145.21} = 0.647$$

$$\frac{60 - 208}{145.21} = -1.019$$

$$\frac{393 - 208}{145.21} = 1.274$$

$$\frac{4.78 - 4.03}{1.33} = 0.564$$

$$\frac{2.31 - 4.03}{1.33} = -1.293$$

$$\frac{3.16 - 4.03}{1.33} = -0.654$$

$$\frac{4.2 - 4.03}{1.33} = 0.128$$
No

C = Characters Standardized =
$$\frac{242-208}{145.21}$$
 = .234
L = Word Length standardized = $\frac{4.56-4.03}{1.33}$ = .398

$$P(A = yes) = 0.6$$

$$P(A = yes | C = .234, L = .398)$$

$$P(C = .234) = 0.34$$

$$P(L = .398) = 0.34$$

$$P(A = Yes | C = .234, L = .398) = \frac{P(A)P(C, L|A)}{P(C,L)}$$

$$P(A = Yes|C = 0.2341, L = 0.4028) = \frac{0.6 * P(C|A)P(L|A)}{P(C)P(L)}$$

$$P(A = Yes|C = 0.2341, L = 0.4028) = \frac{0.6 * P(C|A)P(L|A)}{0.34 * 0.34}$$

2 Logistic Regression Spam Classification

- (a) Precision **0.8671454219030521**
- (b) Recall **0.8370883882149047**
- (c) F-measure **0.8518518518519**

(d) Accuracy **0.8904109589041096**

3 Naive Bayes Classifier

- (a) Precision **0.6465116279069767**
- (b) Recall **0.9636048526863085**
- (c) F-measure **0.7738343771746694**
- (d) Accuracy **0.7879973907371167**

4 Decision Trees

I was not able to compute the stats as I had not completed finished implementing the DTL algorithm.