

SENG 474 Assignment 2

$$1 \text{ a) Success rate} = (\text{red})(\frac{1}{3}) + (\text{blue})(\frac{1}{3}) + (\text{yellow})(\frac{1}{3}) = \frac{1}{3}$$

$$\text{Error rate} = 1 - \text{success rate} = 1 - \frac{1}{3} = \boxed{\frac{2}{3}}$$

$$1 \text{ b) Success rate} = (\text{red})(\frac{1}{3}) + (\text{blue})(\frac{1}{3}) + (\text{yellow})(\frac{1}{3}) = \frac{1}{3}$$

$$\text{Error rate} = 1 - \text{success rate} = 1 - \frac{1}{3} = \boxed{\frac{2}{3}}$$

$$1 \text{ c) Success rate} = (\text{red})(\frac{1}{2}) + (\text{blue})(\frac{1}{4}) + (\text{yellow})(\frac{1}{4}) = \frac{1}{2}$$

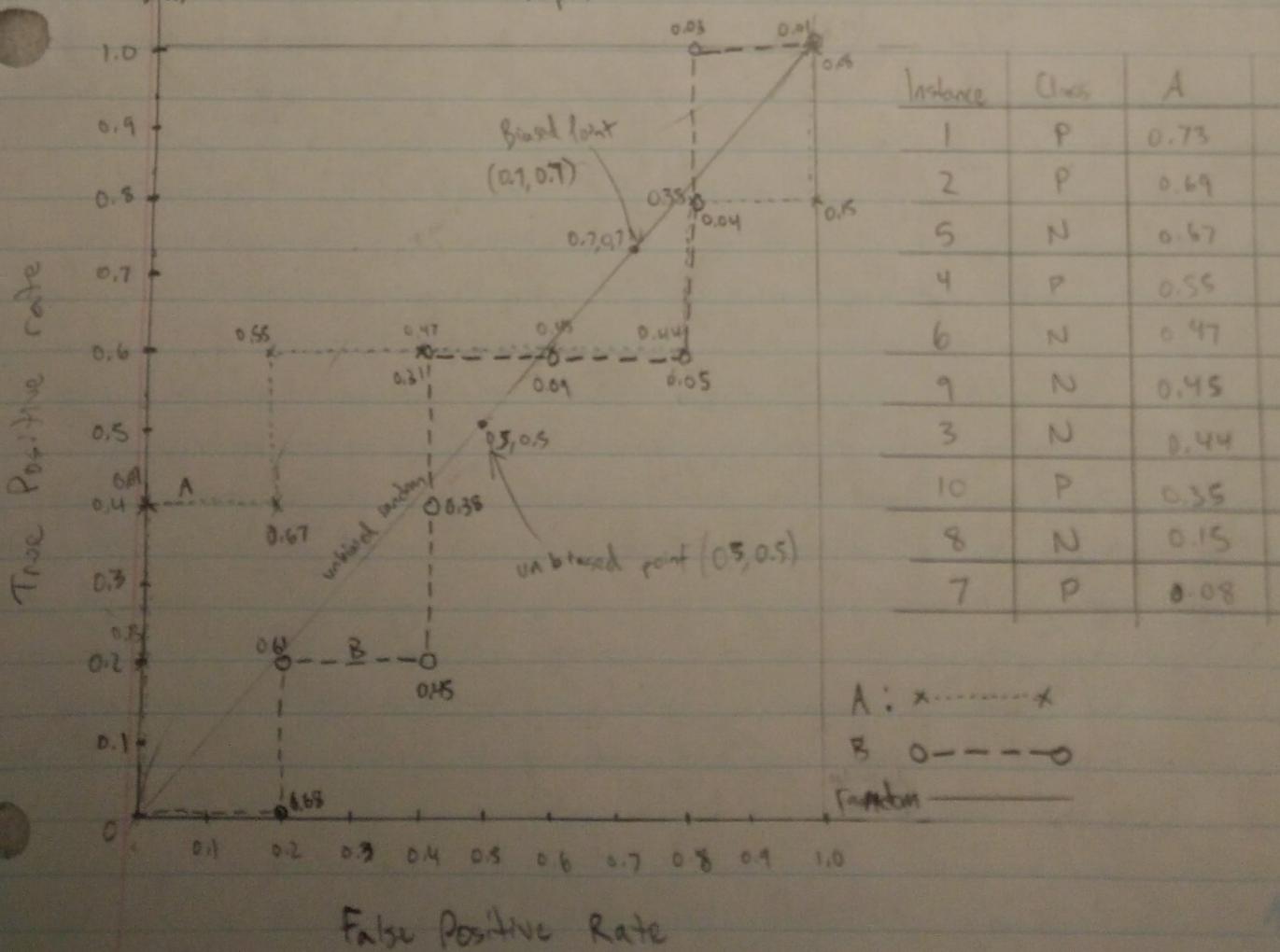
$$\text{Error rate} = 1 - \text{success rate} = 1 - \frac{1}{2} = \boxed{\frac{1}{2}}$$

$$1 \text{ d) Success rate} = (\text{red})(\frac{1}{2}) + (\text{blue})(\frac{1}{4}) + (\text{yellow})(\frac{1}{4}) = 0.425$$

$$\text{Error rate} = 1 - \text{success rate} = 1 - 0.425 = \boxed{0.575}$$

2. a)

ROC Graph



Instance	Class	B
3	N	0.68
1	P	0.61
5	N	0.45
7	P	0.38
4	P	0.31
6	N	0.09
8	N	0.05
10	P	0.04
2	P	0.03
9	N	0.01

b)  $t = 0.5$

Precision:  $\frac{3}{3+1} = 75\%$ . Precise

Recall:  $\frac{3}{3+2} = 60\%$ . Recall

F-measure:  $2 \times (P \times R) / (P + R) = 2 \frac{(0.75)(0.6)}{(0.75 + 0.6)} = \frac{2}{3} = 66.6\%$ .

c)  $t = 0.5$

Precision:  $\frac{1}{1+1} = 50\%$ . Precise

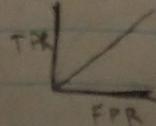
Recall:  $\frac{1}{1+4} = 20\%$ . Recall

F-measure:  $2 \times (P \times R) / (P + R) = 2 \frac{(0.5)(0.2)}{(0.5 + 0.2)} = \frac{2}{7} = 28.6\%$ .

Since A has a greater F-measure (F1) than B, A is a better classifier. This is expected based on the ROC from part a) since classifier A clearly has a better precision and recall for  $t = 0.5$ .

d) A performs better than the unbiased random classifier between  $t = 0.45$  and  $t = 1$ . B performs better than the unbiased random classifier between  $t = 0.01$  and  $t = 0.04$  as well as between  $t = 0.09$  and  $t = 0.38$ . Unbiased random is a straight line from bottom left to top right.

e) This is plotted approximately as follows.



See graph for details.

$$\begin{aligned}
 3.a) P(D|\theta) &= \prod_{i=1}^N p(x_i|\theta) \\
 &= \prod_{i=1}^N \theta^{x_i} (1-\theta)^{1-x_i} \\
 &= \theta^{\text{#T}} (1-\theta)^{\text{#F}}
 \end{aligned}$$

#T is #True = 7 (from D)  
 #F is #False = 3

solve for  $\max \theta$

$$\frac{d}{d\theta} P(D|\theta) = \frac{d}{d\theta} \theta^{\text{#T}} (1-\theta)^{\text{#F}}$$

$\Rightarrow$  set derivative to 0

$$0 = \text{#T} \theta^{\text{#T}-1} (1-\theta)^{\text{#F}} - \text{#F} \theta^{\text{#F}-1} (1-\theta)^{\text{#T}}$$

$$\text{#T} \theta^{\text{#T}-1} (1-\theta)^{\text{#F}} = \text{#F} \theta^{\text{#F}-1} (1-\theta)^{\text{#T}}$$

$$7\theta^6 (1-\theta)^3 = 3\theta^7 (1-\theta)^2$$

$$7(1-\theta) = 3\theta$$

$$7 - 7\theta = 3\theta$$

$$7 = 10\theta$$

$$P(X=T) = \theta = \frac{7}{10}$$

$$3. b) \text{ Beta} \sim \Theta^{\beta_1-1} (1-\Theta)^{\beta_2-1}$$

$$\beta_1 = 4, P(X=T) = \frac{1}{2}, \alpha_T = 7, \alpha_F = 3$$

$$P(X=T) = \frac{\beta_1 - 1}{(\beta_1 - 1) + (\beta_2 - 1)} = \frac{1}{2}$$

$$\frac{4-1}{(4-1) + (\beta_2 - 1)} = \frac{1}{2}$$

$$\frac{3}{\beta_2 + 2} = \frac{1}{2}$$

$$b = \beta_2 + 2$$

$$\beta_2 = 4$$

$$\begin{aligned} P(\Theta|D) &\propto P(D|\Theta)P(\Theta) \\ &= \Theta^{\alpha_T} (1-\Theta)^{\alpha_F} \Theta^{\beta_1-1} (1-\Theta)^{\beta_2-1} \\ &= \Theta^{10} (1-\Theta)^6 \end{aligned}$$

$$0 = \frac{\partial}{\partial \Theta} P(\Theta|D) = \frac{\partial}{\partial \Theta} \Theta^{10} (1-\Theta)^6$$

$$\frac{d}{d\Theta} \log P(\Theta|D) = \frac{d}{d\Theta} 10 \log \Theta + 6 \log(1-\Theta)$$

$$0 = \frac{10}{\Theta} - \frac{6}{1-\Theta}$$

$$\frac{10}{\Theta} = \frac{6}{1-\Theta}$$

$$\frac{1-\Theta}{\Theta} = \frac{6}{10}$$

$$\frac{1-\Theta}{\Theta} = \frac{6}{10}$$

$$\frac{1}{\Theta} = \frac{16}{10}$$

$$\Theta = \frac{10}{16} = \frac{5}{8}$$

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$$4.a) E(X) = \frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})^2$$

$$b) \frac{\partial}{\partial w_0} E(X)$$

$$= \frac{\partial}{\partial w_0} \frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})^2$$

$$= \frac{1}{2N} \sum_{i=1}^N \frac{\partial}{\partial w_0} (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})^2$$

$$= \frac{1}{2N} \sum_{i=1}^N 2(y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}) (-1)$$

$$= -\frac{1}{N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})$$

so the update for  $w_0$  is

$$w = w - K \left( -\frac{1}{N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}) \right)$$

$$w \leftarrow w + \frac{1}{N} K$$

$$4. c) \frac{\partial}{\partial w_1} E(X)$$

$$= \frac{\partial}{\partial w_1} \frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})^2$$

$$= \frac{1}{2N} \sum_{i=1}^N \frac{\partial}{\partial w_1} (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})^2$$

$$= \frac{1}{2N} \sum_{i=1}^N 2(y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}) (-x_{i,1})$$

$$= -\frac{1}{N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})(x_{i,1})$$

so the update for  $w_1$  is

$$w = w - K \left[ -\frac{1}{N} \sum_{i=1}^N (x_{i,1})(y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}) \right]$$

$$d) \frac{\partial}{\partial w_2} E(X)$$

$$= \frac{\partial}{\partial w_2} \frac{1}{2N} \sum_{i=1}^N (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})^2$$

$$= \frac{1}{2N} \sum_{i=1}^N \frac{\partial}{\partial w_2} (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})^2$$

$$= \frac{1}{2N} \sum_{i=1}^N 2(y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}) (-x_{i,2})$$

$$= -\frac{1}{N} \sum_{i=1}^N (x_{i,2})^2 (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2})$$

so the update for  $w_2$  is

$$w = w - K \left[ -\frac{1}{N} \sum_{i=1}^N (x_{i,2})^2 (y_i - w_0 - w_1 x_{i,1} - w_2 x_{i,2}) \right]$$