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FINAL GRADE: 100%

# **COMMENTS:**

"task 3: calculation errors -1 point (the info gain calculation is incorrect)"

### Task 1:

Each run ~10 times

decision\_tree.py pendigits\_training pendigits\_test optimized 50 = .8393(1 minute)

decision\_tree.py pendigits\_training pendigits\_test 1 50 = .6933 - .7573(5 Seconds)

decision\_tree.py pendigits\_training pendigits\_test 3 50 = .8209 – 8789(1 minute)

# Task 2:

a) 
$$H(0.8, 0.2) = -0.8\log 2(0.8) - 0.2\log 2(0.2) = .7219$$

b) Entropy(Wait) = 
$$-[(20/35) * \log 2(20/35) + (15/35) * \log 2(15/35)] = .4134$$

Entropy(NW) = 
$$-[(60/65) * log2(60/65) + (5/65) * log2(5/65)] = .2058$$

Weight(W) = 
$$35 / 100$$

Weight(NW) = 
$$65 / 100$$

$$Gain = Initial Entropy (Entropy(A)) - [Weight(W) * Entropy(W) + Weight(NW) * Entropy(NW)]$$

- = 0.44354
- c) Gain = -0.3291
- d) Node F, Will Wait
- e) Node E, Will Not Wait

#### Task 3:

$$A1 = x, x, x$$

$$A2 = x,y,y,y$$

$$A3 = x,y,y$$

$$H(E) = 1 \rightarrow \text{ even amount of 'X' and 'Y'}$$

$$H((A1/10),(A2/10),(A3/10)) = H(E) - [(3/10)*(1.0*log2(1) + 0*log2(0)) +$$

$$(4/10)*(-0.25*log2(0.25) -0.75*log2(0.75)) +$$

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(3/10*(-0.33*log2(0.33) - 0.67*(log2(0.67))] = Gain_A
H(E)-(0 +(.32451) + (-.04221))
1 - .2823 = .7177
B1 = x,y,y,y
B2 = x, x, x, y
B3 = x, y
H(E) = 1 \rightarrow \text{ even amount of 'X' and 'Y'}
H((B1/10),(B2/10),(B3/10)) = H(E) - [(4/10)*(-.25*log2(.25) - .75*log2(.75)) +
(4/10)*(-0.75*log2(0.75) -0.25*log2(0.25)) +
(2/10*(-0.5*log2(0.5) - 0.5*log2(0.5)) = Gain B
H(E)-((.32451) + (.32451) + .2)
1 - .84902 = .15098
C1 = x,y,y,y,y
C2 = x, x, x, y
C3 = x
H(E) = 1 \rightarrow \text{ even amount of 'X' and 'Y'}
H((C1/10),(C2/10),(C3/10)) = H(E) - [(5/10)*(-.2*log2(.2) - .8*log2(.80) +
(4/10)*(-0.75\log 2(0.75) -0.25\log 2(0.25)) +
(1/10*(-1.0\log 2(1.0) - 0(\log 2(0)))] = Gain_CH(E) - ((.36096) + (.32451) + (0))
1 - .68547 = = .31453
A is the highest so choose it
Task 4:
a. H(A,B,C,D) max = -(250/1000)\log 2(250/1000) - (250/1000)\log 2(250/1000) -
(250/1000)\log 2(250/1000) - (250/1000)\log 2(250/1000) = 2
b. H(A,B,C,D)_min = -(1000/1000)\log 2(1000/1000) - (0/1000)\log 2(0/1000) -
(0/1000)\log 2(0/1000) - (0/1000)\log 2(0/1000) = 0
Task 5:
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If your classifier correctly predicts if the home team will win 28% of the time, using the contrapositive, it also means that it incorrectly predicts if the "away" team will lose 72% of the time.

By switching the classifier to predict if the "away" team will lose, and then outputting the opposite of the given answer, you essentially have a classifier that can predict if the "home" team will win 72% of the time.