Name	

- 1. You are playing a betting game, if you win you get \$1 and if you lose you give \$3. According to the odds you have 80% chance of winning, what is the expected winning amount for you.
- A. \$1.4
- B. \$1
- C. \$0.2
- D. \$0
- 2. True- False: Overfitting is more likely when you have huge amount of data to train?
  - A. TRUE
  - **B. FALSE**
- 3. In which of the following scenario a gain ratio is preferred over Information Gain?
  - A. When a categorical variable has very large number of category
  - B. When a categorical variable has very small number of category
  - C. Number of categories is the not the reason
  - D. None of these
- 4. There are 3 boxes, A, B and C. The probability of choosing box A, B and C are 0.5, 0.3 and 0.2 respectively. Box A, B and C contains 20%, 30% and 50% rotten apples. What is the probability that an apple is drawn from box A given that it is rotten? To solve this kind of problem, what theorem should we use?

Let A = event that an apple is drawn from box A, B = event that an apple is drawn from box B, C = event that an apple is drawn from box C, R = a rotten apple is drawn.

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Given P(A) = 0.5, P(B) = 0.3, P(C) = 0.2.

Also, P(R|A) = 0.2, P(R|B) = 0.3, P(R|C) = 0.5.

Our aim is to find P(A|R) = P(A \text{ intersect } R)/P(R)

= P(A) \times P(R|A) / [P(A) \times P(R|A) + P(B) \times P(R|B) + P(C) \times P(R|C)]

= 0.5 \times 0.2 / (0.5 \times 0.2 + 0.3 \times 0.3 + 0.2 \times 0.5)

= 0.3448.
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Bayesian inference is a branch of statistic inference that arises from this theorem.

5. Give decision tree to represent the boolean function: A V [B  $\Lambda$   $\neg$ C]

