CIS 472/572, Spring 2021 Homework 1 (Written): Decision Trees DUE DATE: Submit via Canvas by Thursday, April 15th at 11:55pm.

Your answers should be typewritten, except for figures which may be hand-drawn. Some questions in this exercise refer to Mitchell's Machine Learning, chapter 3. You can access it using the following link:

http://www.cs.princeton.edu/courses/archive/spr07/cos424/papers/mitchell-dectrees.pdf

Question 1: (Mitchell, Exercise 3.1, page 77). Give decision trees to represent the following boolean functions:

- (a) $A \wedge \neg B$
- (b) $A \vee [B \wedge C]$
- (c) $A \times OR B$
- (d) $[A \wedge B] \vee [C \wedge D]$

Question 2: Consider the samples in the Play-tennis dataset from Table 3.2 in Mitchell's textbook (linked above). If you calculate the information gain for all of the attributes of this set, you will observe that the attribute "Outlook" has the largest information gain, which is equal to 0.246. Therefore, the attribute "Outlook" is the best heuristic choice for the root node.

- 1. List the labels of the new tree branches below the root node.
- 2. Which partition of the data will be assigned to each branch by ID3? Please list the sample IDs that will be assigned to each branch.
- 3. Calculate the information gain for the remaining attributes in each branch, and determine which attribute will be chosen as the root of the sub-tree in each branch.

Question 3: Suppose a bank makes loan decisions using two decision trees, one that uses attributes related to credit history and one that uses other demographic attributes. Each decision tree separately classifies a loan applicant as "High Risk" or "Low Risk." The bank only offers a loan when both decision trees predict "Low Risk."

1. Describe an algorithm for converting this pair of decision trees into a single decision tree that makes the same predictions (that is, it predicts non-risky only when both of the original decision trees would have predicted non-risky).

2. Let n_1 and n_2 be the number of leaves in the first and second decision trees, respectively. Provide an upper bound on n, the number of leaves in the single equivalent decision tree, expressed as a function of n_1 and n_2 .