1 Greedy Activity Selection (3 pt.)

*Activity Selection* is a classic algorithms problem. It works as follows: you are given a schedule with *n* activities, each of which has a start and finish time. You must select a subset of the greatest possible size, subject to the constraint that none of the activities within overlap.

Below is an example schedule.

<table>
<thead>
<tr>
<th>Start Times</th>
<th>Finish Times</th>
</tr>
</thead>
<tbody>
<tr>
<td><em>a</em>&lt;sub&gt;1&lt;/sub&gt;</td>
<td>1 3 3 3 5 8 5.5</td>
</tr>
<tr>
<td><em>a</em>&lt;sub&gt;2&lt;/sub&gt;</td>
<td>2 4 4 4 7 12 10</td>
</tr>
<tr>
<td><em>a</em>&lt;sub&gt;3&lt;/sub&gt;</td>
<td><em>a</em>&lt;sub&gt;4&lt;/sub&gt;</td>
</tr>
<tr>
<td><em>a</em>&lt;sub&gt;6&lt;/sub&gt;</td>
<td><em>a</em>&lt;sub&gt;7&lt;/sub&gt;</td>
</tr>
<tr>
<td><em>a</em>&lt;sub&gt;8&lt;/sub&gt;</td>
<td></td>
</tr>
</tbody>
</table>

Two valid solutions to this schedule are \{*a*<sub>1</sub>, *a*<sub>2</sub>, *a*<sub>6</sub>, *a*<sub>7</sub>\}, and \{*a*<sub>1</sub>, *a*<sub>3</sub>, *a*<sub>6</sub>, *a*<sub>7</sub>\}. Two invalid solutions are \{*a*<sub>1</sub>, *a*<sub>2</sub>, *a*<sub>6</sub>\} (We only include three activities when we could include four) and \{*a*<sub>1</sub>, *a*<sub>2</sub>, *a*<sub>6</sub>, *a*<sub>8</sub>\} (two of the activities overlap).

Consider the following greedy algorithm for activity selection. The idea is that at each step, we greedily add a valid activity with the fewest conflicts with other valid activities. (An activity is *valid* if it doesn’t conflict with an already selected activity).
def greedyActivitySelection(Activities):
    result = {}
    # initialize overlap counts array OV
    OV = initOV(Activities)
    while size(OV) > 0:
        a1 = argmin(OV) # activity with the least overlaps
        result += a1
        Activities -= a1
        Activities -= {activities that conflict with a1}
        # reinitialize OV using smaller schedule
        OV = initOV(Activities)
    return result

The number of conflicts to begin with (represented in the array OV) are:

<table>
<thead>
<tr>
<th>a1</th>
<th>a2</th>
<th>a3</th>
<th>a4</th>
<th>a5</th>
<th>a6</th>
<th>a7</th>
<th>a8</th>
</tr>
</thead>
<tbody>
<tr>
<td>0</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>3</td>
<td>1</td>
<td>1</td>
<td>2</td>
</tr>
</tbody>
</table>

The algorithm (breaking ties arbitrarily) could choose a1, then a6, then a7, then a2.

Is this algorithm correct?

[We are expecting: Either a short English explanation for why this algorithm always succeeds, or a counterexample to show that it doesn’t.]

# 2 Knapsacks with Nondiscrete Items

In this exercise we’ll practice designing and analyzing greedy algorithms. We’ll look at a continuous variant of the knapsack problem that we saw in class. You have a knapsack with a capacity of $Q$ ounces and there are $n$ items; the difference between this exercise and the version that we saw in class is that you can take a fractional amount of each item. For example, perhaps one item is 3.6 ounces of brightly colored sand; you can choose to take 2.5235 ounces of sand for your knapsack if that’s how much you want.

Each item $i$ has a value per ounce $v_i > 0$ (measured in units of dollars per ounce) and a quantity $q_i > 0$ (measured in ounces). There are $q_i$ ounces of item $i$ available to you, and for any real number $x \in [0, q_i]$, the total value that you derive from $x$ ounces of item $i$ is $x \cdot v_i$.

Your goal is to choose an amount $x_i \geq 0$ to take for each item $i$ in order to maximize the value $\sum_i x_i v_i$ that you receive while satisfying:

1. you don’t overfill the knapsack (that is, $\sum_i x_i \leq Q$), and
2. you don’t take more of an item than is available (that is, $0 \leq x_i \leq q_i$ for all $i$).

Assume that $\sum_i q_i \geq Q$, so there always is some way to fill the knapsack.
2.0 (0 pt.)
Suppose that you already have partially filled your knapsack, and there is some amount of each item left. What item should you take next, and how much?

[We are expecting: Nothing, this part is worth zero points, but it’s a good thing to think about before you go on to the next part.]

2.1 (3 pt.)
Design a greedy algorithm which takes as input \( Q \) along with the tuples \((i, v_i, q_i)\) for \( i = 0, \ldots , n - 1 \), and outputs tuples \((i, x_i)\) so that (1) and (2) hold and \( \sum_i x_i v_i \) is as large as possible. Your algorithm should take time \( O(n \log n) \).

Note: If you have a list of tuples \((a_i, b_i, c_i)\), it is perfectly acceptable to say something like “Sort the list by \( c_i \)” in your pseudocode.

[We are expecting:

• Pseudocode AND an English explanation of what it is doing.

• A justification of the running time.

]

2.2 (3 pt.)
Fill in the inductive step below to prove that your algorithm is correct.

• Inductive hypothesis: After making the \( t \)'th greedy choice, there is an optimal solution that extends the solution that the algorithm has constructed so far.

• Base case: Any optimal solution extends the empty solution, so the inductive hypothesis holds for \( t = 0 \).

• Inductive step: \( (\text{you fill in}) \)

• Conclusion: At the end of the algorithm, the algorithm returns a set \( S^* \) of tuples \((i, x_i)\) so that \( \sum_j x_j = Q \). Thus, there is no solution extending \( S^* \) other than \( S^* \) itself. Thus, the inductive hypothesis implies that \( S^* \) is optimal.

[We are expecting: A proof of the inductive step: assuming the inductive hypothesis holds for \( t - 1 \), prove that it holds for \( t \).]

2.3 Ethics (2 pt.)
Suppose you work as an admissions officer at a prestigious university and you thought perhaps using an algorithm to help rank applicants would help streamline the admissions process.
However, you know that greedily admitting applicants with the highest SAT/ACT score would not necessarily result in the best class of students. Hence, you decide to take other criteria into account, such as community involvement, leadership and distinction in extracurricular activities, and personal qualities and character.

Using the concept of incommensurability, explain why an algorithm might have a difficult time deciding how to rank two applicants.

- Two things are incommensurable when we lack a common measure of value. Incommensurability makes it difficult to establish ranking relationships, such as “more than” or “less than,” “better than” or “worse than.”

[We are expecting: Two to four sentences explaining how two applicants can have values that are incommensurable.]

### Problems

The following questions are problems. You may talk with your fellow CS 161-ers about the problems. However:

- Try the problems on your own before collaborating.
- Write up your answers yourself, in your own words. You should never share your typed-up solutions with your collaborators.
- If you collaborated, list the names of the students you collaborated with at the beginning of each problem.

### 3 Devon and the Redwood Tree

Devon the Duck is making her nest in a Redwood tree. She sometimes drops her eggs, so she will place her nest in the highest branch at which she can drop an egg without it breaking. If an egg will break when dropped from branch $i$, it will also break when dropped from any branch $j$ as long as $j \geq i$.

Chester the Chicken has kindly agreed to loan Devon some of his excess eggs so that Devon can run an experiment to determine which branch to place her nest on. Chicken eggs break when dropped from a branch if and only if Duck eggs do. Once it breaks, an egg can no longer be used to run experiments. Devon must be entirely certain of which branch she will place her nest on by the time she runs out of chicken eggs.

Count the minimum number of drops that Devon needs to make in the worst case, given that
Chester has given her $k$ chicken eggs.

[A Plucky the Penguin moment: if all of the branches on the tree would break Devon’s eggs, then she nests on branch 0, which indicates that she nests on the ground. If there is one branch and it won’t break her eggs, then Devon nests on branch 1. You won’t need to worry about this too much in the following problems, just know that one branch on the tree implies two possible places to nest.]

For $n \geq 0$ and $k \geq 1$, let $D[n, k]$ be the optimal worst-case number of drops that Devon needs to determine the correct branch out of $n$ branches using $k$ eggs. That is, $D[n, k]$ is the number of drops that the best algorithm would use in the worst-case.

### 3.1 (1 pt.)

For any $1 \leq j \leq k$, what is $D[0, j]$? What is $D[1, j]$?

[We are expecting: Your answer. No justification required.]

### 3.2 (1 pt.)

For any $1 \leq m \leq n$, what is $D[m, 1]$?

[We are expecting: Your answer, with a brief (1 sentence) justification.]

### 3.3 (2 pt.)

Suppose the best algorithm drops the first egg from branch $x \in \{1, \ldots, n\}$. Write a formula for the optimal worst-case number of drops remaining in terms of $D[x - 1, k - 1]$ and $D[n - x, k]$.

[We are expecting: Your formula and an informal explanation of why this formula is correct.]

### 3.4 (2 pt.)

Write a formula for $D[n, k]$ in terms of values $D[m, j]$ for $j \leq k$ and $m < n$.

Hint: Use part [3.3]

[We are expecting: Your formula and an informal explanation of why this formula is correct.]

### 3.5 Dynamic Programming Algorithm (5 pt.)

Design a dynamic programming algorithm which will compute $D[n, k]$ in time $O(n^2k)$.

[We are expecting: Pseudocode AND a brief English description of how it works, as well as an informal justification of the running time. You do not need to justify that it is correct. ]
4 Making Change

Lucky the lackadaisical lemur works at McDonald’s, and often he has to make change. If a customer paid $5 for an order that costs $4.48, he would have to provide 52¢ in change. Assuming Lucky has unlimited access to pennies, nickels, dimes, and quarters, he could pay that 52¢ with two quarters and two pennies (4 total coins), or five dimes and two pennies (7 total coins), or 52 pennies (52 total coins), or a number of other combinations. Lucky’s manager hates having extra change lying around, so Lucky’s goal is to use as many total number of coins as possible.

Consider the more general problem: Lucky has \( k \) coins which are worth distinct values \( v_1, v_2, \ldots, v_k \), such that \( 0 < v_1 < v_2 < \ldots < v_k \). He has to make change amounting exactly to \( x \). More precisely, he has to provide \([c_1, c_2, c_3, \ldots, c_m]\) with each \( c_j \in \{v_1, v_2, \ldots, v_k\}\) so that \( \sum_1^m c_i = x \). His goal is to maximize \( m \), or in other words, to use as many coins as possible.

4.1 Greedy Approach

Consider the following greedy algorithm for making change that Lucky came up with.

```python
def makeChange(Coins, x):
    if x == 0:
        return []  # empty list
    if x < 0:
        return -1
    recursiveSol = makeChange(Coins, x-Coins[0])
    if recursiveSol == -1:
        return -1
    else:
        return [Coins[0]] + recursiveSol
```

4.1.1 (2 pt.) If it returns a valid way to make change, does this algorithm return one that is optimal (i.e., that uses the maximum total number of coins)?

[We are expecting: Either a short English explanation for why this algorithm always succeeds, or a counterexample to show that it doesn’t.]

4.1.2 (1 pt.) Does this algorithm always return a valid way of making change if one exists?

[We are expecting: Either a short English explanation for why this algorithm always succeeds, or a counterexample to show that it doesn’t.]
4.2 Dynamic Programming Approach (5 pt.)

Design a DP algorithm which computes the maximum number of coins Lucky can use to make change. Your algorithm should run in time $O(kx)$, where $x$ is the amount you must make change for and $k$ is the number of types of coins available to use.

[We are expecting: Pseudocode and a short English description of your algorithm. No runtime justification is necessary.]

5 Bear Island

One day, you get stranded on a beach on an island with $n$ bears. You will be here a while, so you want to make friends and introduce yourself to all the bears. However, you don’t want to leave the beach since the rest of the island looks dangerous. Each bear $i$ will only be on the beach during one time interval $[a_i, b_i]$, inclusive. You can only introduce yourself to bear $i$ during this time interval.

Your plan is to stand in the center of the beach and use the megaphone you brought with you to say “Hello” at certain times $t_1, \ldots, t_m$. Any bear on the beach at one of the times $t_j \in \{t_1, \ldots, t_m\}$ will hear your introduction. It’s okay if a bear hears your introduction more than once, but you want to make sure every bear hears your introduction at least once.

5.1 Greedy Algorithm (6 pt.)

The battery in your megaphone is low, so you want to minimize $m$, the number of times you use your megaphone. Devise a greedy algorithm that takes as input the list of intervals $[a_i, b_i]$ and outputs a list of times $t_1, \ldots, t_m$ such that $m$ is as small as possible but every bear hears your introduction. Your algorithm should run in time $O(n \log n)$ where $n$ is the number of bears on the island. You can assume every bear visits the beach at some point.

[We are expecting: Pseudocode and an English description of your algorithm, as well as a short justification of the runtime]

5.2 Proof of Correctness (4 pt.)

Give a formal proof that your algorithm is correct.

[We are expecting: A proof by induction]