

1 Smartest Students (25 + 3pts)

- (1) If the only restriction was the total weight W , the problem would be exactly like the 0/1 Knapsack problem we talked about in class. Since there is no restriction on how many students she can take, the only choice is to pick the i^{th} student, so the subproblem would just be $f[i, k]$ with k representing the remaining weight and i representing the current choice of student. Then, $f[i, k]$ would represent the optimal smartness for total weight k using the first i students. Without the restriction, the total students she can take is allowed to be more than T , so the only consideration to be taken is whether to pick student i or not.

- (2) (a): $1 \leq i \leq N$

We know that we need to identify the boundary of i , and we know we can pick from a range of N students. Therefore, i can be any value from 1 to N , and if it is 0, we have reached a base case.

(b): T

We know we need to bound the values of j , and since we know that $0 \leq k \leq W$, we know that the weight dimension of the table is complete. The last remaining dimension would be the amount of students chosen, so the maximum value that j could be is T .

- (3)

$$f[i, j, k] = \max \begin{cases} f[i-1, j, k] \\ f[i-1, j+1, k-w_i] + s_i \end{cases}$$
$$f[0, j, k] = 0$$
$$f[i, j, k] = 0 \quad j > T$$
$$f[i, j, k] = 0 \quad k < w_i$$

The equation for $f[i, j, k]$ is given by deciding whether or not choosing student i is part of the optimal solution. Therefore, we calculate the subproblem of not choosing student i , while looking at the next student and keeping the values of j and k the same. Then, we calculate the subproblem of choosing student i which leads us to increment the value of j since we are adding one student to the chosen T students and we must also subtract the the weight of student i from the value of k since the remaining weight will decrease with student i taking up space on the plane. Since we are choosing to add the student onto the plane, we must also add student i 's smartness to the result of the recursive call. After calculating the two situations of picking student i or not, we choose the larger value to save in the 3-D table at $f[i, j, k]$.

- (4)
- ```
int maxSmart (vector<int>& weights, vector<int>& smartness, vector<vector<vector<int>>>& f,
 int i, int j, int k, int T) {
 if f[i][j][k] is not -1 return f[i][j][k]
 return and save 0 to f[i][j][k] if i = 0, j > T, or k < items.at(i).first

 ans = maxSmart(weights, smartness, f, i-1, j, k, T)
 temp = maxSmart(weights, smartness, f, i-1, j+1, k - weights.at(i), T) + scores.at(i)

 if temp > ans then ans = temp
 f[i][j][k] = ans
 return ans
```

}

Begin with the call,

```
int ans = maxSmart (weights, smartness, f, N, 0, W, T);
```

in the main function. The first two parameters of the function are vectors whose  $i^{th}$  elements represent the  $i^{th}$  student's weight and smartness, respectively. The values in the 3-D table  $f$  are initialized to  $-1$  to represent that the value of subproblem  $f[i, j, k]$  has not been calculated. Then we pass in the initial values for  $i$ ,  $j$ , and  $k$  as  $N$ ,  $0$ , and  $W$ , respectively. Then, the subproblem  $f[N, 0, W]$  represents the optimal smartness when choosing from all  $N$  students, with  $T = 0$ , or all  $T$ , seats available, and all  $W$  pounds available on the plane. In the function,  $\text{maxSmart}()$ , if the value of the subproblem  $f[i, j, k]$  has been calculated, it is returned. Otherwise, if a base case is reached,  $0$  is saved at  $f[i][j][k]$  and returned. Otherwise, the first integer  $ans$  is calculated to represent the optimal smartness of not choosing student  $i$ . The next integer  $temp$  is calculated to represent the total smartness if student  $i$  is chosen. The larger value is saved to the  $ans$  variable and saved at  $f[i][j][k]$ . Then it is returned as the maximum smartness achievable from the first  $i$  students with  $T - j$  seats available and  $k$  pounds remaining on the plane.

- (5) The asymptotic complexity of my algorithm is  $\Theta(n^2)$ . From the algorithm in part (4), we can see that for calculating the total smartness for each of the first  $i$  students for  $1 \leq i \leq n$ , where  $n$  is the number of students to be chosen from, there are  $n$  calls to the function. Therefore, for the algorithm to calculate the best total smartness for  $n$  students, the complexity would be  $\Theta(n^2)$ . Also, there are two calls to the  $\text{maxSmart}()$  function on item  $i - 1$  within the call to  $\text{maxSmart}()$  on item  $i$ . Then, the recursion tree must have a height of  $n$  since there must be  $n$  calls to the function to reach the base case. We also know that with 2 calls to the function, there will be  $2n$  leaves of the recursion tree. Then, the complexity of the algorithm would be the height of the tree times the number of leaves. Then the complexity would be  $\Theta(n \cdot 2n)$  which is simplified to  $\Theta(n^2)$ . This confirms that our algorithm runs in  $\Theta(n^2)$  time.
- (6) (Bonus)

## 2 Morse Code (25pts)

|       |      |       |      |       |      |       |      |       |      |       |      |
|-------|------|-------|------|-------|------|-------|------|-------|------|-------|------|
| A (3) | .-   | B (2) | -... | C (1) | -. . | D (2) | -..  | E (3) | .    | F (3) | ..-  |
| G (2) | --.  | H (3) | .... | I (3) | ..   | J (1) | .--- | K (2) | -.-  | L (1) | .-.. |
| M (2) | --   | N (2) | -. . | O (1) | ---  | P (2) | .--. | Q (2) | --.- | R (2) | .-.  |
| S (1) | ...  | T (2) | -    | U (1) | ..-  | V (1) | ...  | W (1) | .-   | X (2) | -. . |
| Y (2) | -.-- | Z (1) | --.. |       |      |       |      |       |      |       |      |

- (1) Morse code for *UCR* is ..- -.-.-. from the given table. Yes, there are many interpretations of the Morse code for *UCR* given that there is no separator. For example, the code could represent *EACR*, *EPDN*, or *EETTETEETE*. The first two given representations of the code for *UCR* have a low stroke count of 9.

- (2)
- ```

int minStrokes(vector<pair<char, int>>& chars, vector<vector<int, int>>& vals, int
    currChar, int j) {

    if currChar is 0 return 0
    if vals[i][j] has been calculated, return vals[i][j]
    if j - currChar > 3 return 0
    int currStrokes = 0;
    for (int i = 0; i < chars.size(); ++i) {
        if (check(X, currChar, j, chars.at(i).first)) {
            currStrokes = chars.at(i).second
            break
        }
    }
    int ans = minStrokes(chars, vals, currChar-1, j)
    if currStrokes is still 0 {
        vals[i][j] = ans
        return ans
    }

    int temp = minStrokes(chars,vals, currChar-1, currChar-1) + currStrokes
    if (temp < ans || (ans == 0)) ans = temp
    vals[i][j] set to ans
    return ans
}

```

It is assumed that the function `minStrokes()` is initially called in the main function with the call,

```
int ans = minStrokes(chars, vals, X.size(), X.size());
```

The subproblem $s[i, j]$ represents the minimum amount of strokes from a decoded message from the first j symbols while choosing to use the decoded representation of the substring $X[i, j]$ or not. Then the recurrence for the algorithm would look as:

$$s[i, j] = \min \begin{cases} s[i-1, j] \\ s[i-1, i-1] + s_{ij} \end{cases}$$

In the recurrence, the first subproblem moves the i pointer to the symbol in front of the current symbol at i in X , representing choosing not to use the decoded character of $X[i, j]$ and instead look at a longer string to decode. Then, in the subproblem, the substring $X[i, j]$ is one character longer than the

substring $X[i, j]$ of the original problem showing that instead of decoding the original substring, we added a new value to the front of the substring. This will represent a different decoded character than what the original substring represented which means a different stroke number could be found. In the second subproblem, we are choosing to use the decoded character mapped to by the substring $X[i, j]$, so then we start the subproblem at the first symbol before the substring $X[i, j]$, or $i - 1$. We must also add s_{ij} , stroke value of decoded $X[i, j]$, to the result of this subproblem since we are choosing to use the decoded value of $X[i, j]$. After getting a value for the subproblem $s[i - 1, i - 1]$ (minimum number of strokes for the first $i - 1$ symbols), we add the value to s_{ij} which completes the second subproblem. Basically, we are trying to see if we should use the decoded character mapped to by substring $X[i, j]$ with stroke count s_{ij} or if we can achieve a lower count of strokes by not using the decoded character of $X[i, j]$ and instead use the decoded character of $X[i - 1, j]$.

The original call to the function passes in two vectors, chars and vals, respectively. The vals vector is a table that represents the minimum strokes achieved from the first j symbols while choosing the last symbol represented by the substring $X[i, j]$. Before the initial call to the function, all places in vals are initialized to -1 to represent that a value has not been calculated. The vector chars is a vector of pairs with the first value of the pair being a char to represent the decoded character and the second value being the number of strokes for that character. Then we pass in the size of X as both the *currChar* and j parameters, and when *currChar* is 0, we have reached a boundary case to return 0. Here in the first call to the function, the size of substring $X[i, j]$ is just 1, so we are looking at the choice to decode the last symbol or to decrement the i value to decode the last two symbols. The base cases ensure that the distance between *currChar* and j is always less than or equal to 4 places since 4 is the greatest length of any encoded value. After the base cases, we also check to see if the current substring $X[i, j]$ actually represents a decoded value. If not, we can only look at the first subproblem of the recurrence since the second subproblem requires that the substring $X[i, j]$ is an encoded value.

- (3) The time complexity of my algorithm would be $\Theta(n^2)$. Let n be the size of the input string X . Then, while calculating the minimum stroke count for each substring of the first i symbols for $1 \leq i \leq n$, there are n calls to the function. Therefore, for each n number of symbols, there are n calls to the function, so the function has complexity $\Theta(n^2)$.
- (4) The minimum amount of strokes we can get from the sequence - -.- - . with characters:

A (3): .-
 C (1): -.-.
 E (3): .
 G (2): --.
 K (2): -.-
 T (2): -

is 7 strokes. We can achieve this by encoding the string *GEG*.

- 3 Knapsack Algorithms can be Simple I (Bonus, 3pts + 1 candy)
- 4 Knapsack Algorithms can be Simple II (Bonus, 3pts + 1 candy)