1. Consider the recurrence relation $T(0) = T(1) = 2$ and for $n > 1$
\[ T(n) = \sum_{i=1}^{n-1} T(i)T(i-1) \]
We consider the problem of computing $T(n)$ from $n$.

(a) Show that if you implement this recursion directly in say the C programming language, that the program would use exponentially, in $n$, many arithmetic operations.

(b) Explain how, by not recomputing the same $T(i)$ value twice, one can obtain an algorithm for this problem that only uses $O(n^2)$ arithmetic operations.

(c) Give an algorithm for this problem that only uses $O(n)$ arithmetic operations.

2. Give a polynomial time algorithm that takes three strings, $A$, $B$ and $C$, as input, and returns the longest sequence $S$ that is a subsequence of $A$, $B$, and $C$.

3. Give an efficient algorithm for finding the shortest common super-sequence of two strings $A$ and $B$. $C$ is a super-sequence of $A$ if and only if $A$ is a subsequence of $C$.

4. Consider the algorithm that you developed for the previous problem.

   (a) Show the table that your algorithm constructs for the inputs $A = zxyyzz$, and $B = zzyxzy$

   (b) Explain how to find the length of the shortest common super-sequence in your table.

   (c) Explain how to compute the actual shortest common super-sequence from your table by tracing back from the table entry that gives the length of the shortest common super-sequence.

5. The input to this problem is a pair of strings $A = a_1 \ldots a_m$ and $B = b_1 \ldots b_n$. The goal is to convert $A$ into $B$ as cheaply as possible. The rules are as follows. For a cost of 3 you can delete any letter. For a cost of 4 you can insert a letter in any position. For a cost of 5 you can replace any letter by any other letter. For example, you can convert $A = abcabc$ to $B = abacab$ via the following sequence: $abcabc$ at a cost of 5 can be converted to $abaabc$, which at cost of 3 can be converted to $abac$, which at cost of 3 can be converted to $abacab$. Thus the total cost for this conversion would be 19. This is almost surely not the cheapest possible conversion. Give a polynomial time dynamic programming algorithm.

   NOTE: The Unix diff command essentially solves this problem.

6. Find the optimal binary search tree for keys $K_1 < K_2 < K_3 < K_4 < K_5$ where the access probabilities/weights are .5,.05,.1,.2,.25 respectively using the algorithm discussed in class and in the notes. Construct one table showing the optimal expected access time for all subtrees considered in the algorithm, and another showing the roots of the optimal subtrees computed in the other table. Show how to use the table of roots to recompute the tree.
7. Give an efficient algorithm for the following problem. The input is an \( n \) sided convex polygon. Assume that the polygon is specified by the Cartesian coordinates of its vertices. The output should be the triangulation of the polygon into \( n - 2 \) triangles that minimizes the sum of the perimeters of the triangles. Note that this is equivalent to minimizing the length of the cuts required to create the triangulation.

HINT: Fix a side, \( e \), of the polygon. What are the possible triangles (in the final triangulation) that can have \( e \) as an edge? How can you determine which triangle has \( e \) for an edge in the final triangulation? How did we answer this question when picking the correct root for the optimal BST in our recursive solution? If \( e \)'s triangle were fixed, how would you solve for the rest of the polygon?

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8. The input to this problem is a sequence \( S \) of integers (not necessarily positive). The problem is to find the consecutive subsequence of \( S \) with maximum sum. “Consecutive” means that you are not allowed to skip numbers. For example if the input was

\[
12, -14, 1, 23, -6, 22, -34, 13
\]

the output would be \( 1, 23, -6, 22 \). Give a linear time dynamic programming algorithm for this problem.

In your write-up, first explain why a naive recursive solution is not possible. That is, figure out why knowing the nth number, and the maximum consecutive sum of the first n-1 numbers, is not sufficient information to compute the maximum consecutive sum of the first n numbers. These examples, which show you need to strengthen the inductive hypothesis, should give you a big hint how to strengthen the inductive hypothesis to compute two different different consecutive subsequences, the maximum consecutive sum subsequence, and one other one.

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9. The input to this problem is a tree \( T \) with integer weights on the edges. The weights may be negative, zero, or positive. Give a linear time algorithm to find the shortest simple path in \( T \). The length of a path is the sum of the weights of the edges in the path. A path is simple if no vertex is repeated. Note that the endpoints of the path are unconstrained.

HINT: This is very similar to the problem of finding the largest independent set in a tree.

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10. The input for this problem consists of \( n \) keys \( K_1, \ldots, K_n \), with \( K_1 < K_2 < \ldots, K_n \), and associated probabilities \( p_1, \ldots, p_n \). The problem is to find the AVL tree that minimizes \( \sum_{i=1}^{n} d(K_i) \cdot p_i \), the expected depth of a key. An AVL tree is a binary search tree with the property that every node has balance factor \(-1, 0, \) or \( 1 \). The balance factor or a node is the height of its right subtree minus the height of its left subtree. Give a polynomial time algorithm for this problem.

HINT: You will have to strengthen the inductive hypothesis. Obviously the height of a subtree is relevant.

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11. The input consists of \( n \) intervals over the real line. The output should be a collection, \( C \), of non-overlapping intervals such the sum of the lengths of the intervals in \( C \) is maximized. Give a polynomial time algorithm for this problem.

HINT: Strengthen the induction hypothesis. Consider the intervals by increasing order of their left endpoints. Just like in the Longest Increasing Subsequence (LIS) problem, there are two pieces of information that are relevant about a partial solution. One is obviously the length of the intervals, what is the other? Think about what the answer was in the LIS problem.

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12. Consider the code for the Knapsack program from class.
(a) Explain how one can actually find the highest valued subset of objects, subject to the weight constraint, from the Value table computed by this code. So you need to explain how to backtrack from the bottom of the table back to the top of the table to actually construct the subset of objects in the knapsack.

(b) Explain how to solve the Knapsack problem using only $O(L)$ memory/space and $O(nL)$ time. You need only find the value and weight of the optimal solution, not the actual collection of objects.

13. Our goal is now to consider the Knapsack problem, and develop a method for computing the actual items to be taken in $O(L)$ space and $O(nL)$ time.

(a) Consider the following problem. The input is the same as for the knapsack problem, a collection of $n$ items $I_1, \ldots, I_n$ with weights $w_1, \ldots, w_n$, and values $v_1, \ldots, v_n$, and a weight limit $L$. The output is in two parts. First you want to compute the maximum value of a subset $S$ of the $n$ items that has weight at most $L$, as well as the weight of this subset. Let us call this value and weight $v_a$ and $w_a$. Secondly for this subset $S$ you want to compute the weight and value of the items in $\{I_1, \ldots, I_{n/2}\}$ that are in $S$. Let use call this value and weight $v_b$ and $w_b$. So your output will be two weights and two values. Give an algorithm for this problem that uses space $O(L)$ and time $O(nL)$.

(b) Explain how to use the algorithm from the previous subproblem to get a divide and conquer algorithm for finding the items in the Knapsack problem a and uses space $O(L)$ and time $O(nL)$.

HINT: First call the algorithm for the previous subproblem. What recursive call do you need to make to find the items in the final answer from the items in $\{I_1, \ldots, I_{n/2}\}$? What recursive call do you need to make to find the items in the final answer from the items in $\{I_{n/2+1}, \ldots, I_n\}$?

HINT HINT: The solution to the recurrence relation $T(k) = T(a) + T(b) + k$ is $O(k)$ if $a + b = k/2$.

14. Give an algorithm for the following problem whose running time is polynomial in $n + L$.

Input: positive integers $v_1, \ldots, v_n$, with $L = \sum_{i=1}^{n} v_i$.

Output: A solution (if one exists) to $\sum_{i=1}^{n} (-1)^{x_i} v_i = 0$ where each $x_i$ is either 0 or 1.

HINT: Very similar to subset sum algorithm.

15. Give an algorithm for the following problem whose running time is polynomial in $n + \log L$:

Input: positive integers $v_1, \ldots, v_n$ and $L$.

Output: A solution (if one exists) to $(\sum_{i=1}^{n} x_i v_i) \mod n = L \mod n$ where each $x_i$ is either 0 or 1. Here $x \mod y$ means the remainder when $x$ is divided by $y$.

HINT: This is like the subset sum problem but you will have to be more severe in your pruning since you can’t afford to have $L$ partial solutions anymore.

16. Give an algorithm for the following problem whose running time is polynomial in $n + W$:

Input: positive integers $w_1, \ldots, w_n, v_1, \ldots, v_n$, and $W$.

Output: The maximum possible value of $\sum_{i=1}^{n} x_i v_i$ subject to $\sum_{i=1}^{n} x_i w_i \leq W$ and each $x_i$ is a nonnegative integer.

HINT: The tree of feasible solutions will be different than the one in the knapsack/subset sum problem since you may include an item in the final sum more than once. Each node in the tree of feasible solutions may now have up to $W$ children.
17. Give an algorithm for the following problem whose running time is polynomial in \( n + L \), where \( L = \max(\sum_{i=1}^{n} c_i^3, \prod_{i=1}^{n} v_i) \).

Input: positive integers \( v_1, \ldots, v_n \)

Output: A subset \( S \) of the integers such that \( \sum_{v_i \in S} c_i^3 = \prod_{v_i \in S} v_i \).

18. The input to this problem is a set of \( n \) gems. Each gem has a value in dollars and is either a ruby or an emerald. Let the sum of the values of the gems be \( L \). The problem is to determine if it is possible to partition the gems into two parts \( P \) and \( Q \) such that each part has the same value, the number of rubies in \( P \) is equal to the number of rubies in \( Q \), and the number of emeralds in \( P \) is equal to the number of emeralds in \( Q \). Note that a partition means that every gem must be in exactly one of \( P \) or \( Q \). You algorithm should run in time polynomial in \( n + L \).

HINT: Start as in the subset sum example. Your pruning rule will have to be less severe. That is, first ask yourself why you may not be able to prune two potential solutions that have the same aggregate value.

19. The input to this problem consists of an ordered list of \( n \) words. The length of the \( i \)th word is \( w_i \), that is the \( i \)th word takes up \( w_i \) spaces. (For simplicity assume that there are no spaces between words.) The goal is to break this ordered list of words into lines, this is called a layout. Note that you can not reorder the words. The length of a line is the sum of the lengths of the words on that line. The ideal line length is \( L \). No line may be longer than \( L \), although it may be shorter. The penalty for having a line of length \( K \) is \( L - K \). The total penalty is the maximum of the line penalties. The problem is to find a layout that minimizes the total penalty. Give a polynomial time algorithm for this problem.

HINT: Consider whether how many layouts of the first \( m \) words, which have \( k \) letters on the last line, you need to remember.

20. The input to this problem is two sequences \( T = t_1, \ldots, t_n \) and \( P = p_1, \ldots, p_k \) such that \( k \leq n \), and a positive integer cost \( c_i \) associated with each \( t_i \). The problem is to find a subsequence of \( T \) that matches \( P \) with maximum aggregate cost. That is, find the sequence \( i_1 < \ldots < i_k \) such that for all \( j, 1 \leq j \leq k \), we have \( t_{i_j} = p_j \) and \( \sum_{j=1}^{k} c_{i_j} \) is maximized.

So for example, if \( n = 5 \), \( T = XYXYXY \), \( k = 2 \), \( P = XY \), \( c_1 = c_2 = 2 \), \( c_3 = 7 \), \( c_4 = 1 \) and \( c_5 = 1 \), then the optimal solution is to pick the second \( X \) in \( T \) and the second \( Y \) in \( T \) for a cost of \( 7 + 1 = 8 \).

(a) Give a recursive algorithm to solve this problem. Then explain how to turn this recursive algorithm into a dynamic program.

(b) Give a dynamic programming algorithm based on enumerating subsequences of \( T \) and using the pruning method.

(c) Give a dynamic programming algorithm based on enumerating subsequences of \( P \) and using the pruning method.

21. The input to this problem is a sequence of \( n \) points \( p_1, \ldots, p_n \) in the Euclidean plane. You are to find the shortest routes for two taxis to service these requests in order. Let us be more specific. The two taxis start at the origin. If a taxi visits a point \( p_i \) before \( p_j \) then it must be the case that \( i < j \). (Stop and think about what this last sentence means.) Each point must be visited by at least one of the two taxis. The cost of a routing is just the total distance traveled by the first taxi plus the total distance traveled by the second taxi. Design an efficient algorithm to find the minimum cost routing.

HINT: Use dynamic programming. Consider exhaustively enumerating the possible tours one point at a time. So after the \( i \)th stage you would consider all ways to visit the points \( p_1, \ldots, p_i \). Then find a pruning rule that will reduce the number of tours we need to remember down to a polynomial number.
22. The input to this problem is \( n \) points \( x_1, \ldots, x_n \) on a line. A good path \( P \) has the property that one endpoint of \( P \) is the origin and every \( x_i \) is covered by \( P \). Note that \( P \) need not be simple, that is, it can backtrack over territory that it has already covered. Assume a vehicle that starts at the origin at time 0 and can move along the line at unit speed. The response time \( r_i \) for each \( x_i \) is the time until the vehicle first reaches \( x_i \). The problem is to find the good path that minimizes \( \sum_{i=1}^{n} r_i/n \), the average response time. For example, if the points are \( x_1 = 1 \), \( x_2 = 8 \) and \( x_3 = -2 \) and the path visited the points in the order \( x_1, x_3, x_2 \), the average response time for this path would be \( 1/3 + (1 + 3)/3 + (1 + 3 + 10)/3 \). Give a polynomial time algorithm for this problem.

23. Consider the problem where the input is a collection of \( n \) train trips within Germany. For the \( i \)th trip \( T_i \), you are given the date \( d_i \) of that trip, and the non-discounted fare \( f_i \) for that trip. The German railway system sells a Bahncard for \( B \) Marks that entitles you to a 50% fare reduction on all train travel within Germany within \( L \) days of purchase. The problem is to determine when to buy a Bahncard to minimize the total cost of your travel.

For example if the input was \( d_1 = \) January 11, 1997, \( f_1 = 20 \) Marks, \( d_2 = \) February 11, 1998, \( f_2 = 200 \) Marks, \( d_3 = \) January 11, 1999, \( f_3 = 200 \) Marks, \( d_4 = \) March 13, 1999, \( f_4 = 100 \) Marks, \( d_5 = \) February 11, 2002, \( f_5 = 200 \) Marks, and \( d_6 = \) January 11, 2003, \( f_6 = 600 \) Marks, \( B = 240 \), and \( L = 365 \) then you might buy a Bahncard on February 11, 1998, and February 11, 2002, resulting in a total cost of 1200 Marks.

Give a polynomial time algorithm for this problem. The running time of your algorithm should be independent of \( B \) and \( L \).

24. Give a polynomial time algorithm for the following problem. You are responsible for a web server that serves exactly one html file. The problem input consists of a sequence \( R = R_1, \ldots, R_n \) of non-negative integers, and a positive integer \( k \). The number \( R_i \) represents the number of requests for your html file that are made at time \( i \) and \( k \) represents the maximum number of times the html file may be broadcast. The html page may only be broadcast at integer times and if the server broadcasts the html file at time \( t \), then any requests received strictly before \( t \) are satisfied. If the earliest broadcast, that is strictly later than time \( i \), is at time \( t > i \), then the wait time of request \( R_i \) is \( t - i \). The goal is to pick the \( k \) times to broadcast the file in order to minimize the sum over all requests’ wait times.

As an example, assume that the input was \( R = 3, 4, 0, 5, 2, 7 \) (so \( n = 6 \)) and \( k = 3 \). Then one possible solution (there is no claim that this is the optimal solution) would be to broadcast at times 2, 4, and 7 (note that it is obvious that in every optimal schedule that there is a broadcast at time \( n + 1 \) if \( R_n \neq 0 \)). The 3 requests at time 1 would then have to wait 1 time unit. The 4 requests at time 2 would then have to wait 2 time units. The 5 requests at time 4 would then have to wait 3 time units. The 2 requests at time 5 would then have to wait 2 time units. The 7 requests at time 6 would then have to wait 1 time units. Thus the total waiting time for this solution would be

\[
3 \times 1 + 4 \times 2 + 5 \times 3 + 2 \times 2 + 7 \times 1
\]

25. Assume that you are given a collection \( B_1, \ldots, B_n \) of boxes. You are told that the weight in kilograms of each box is an integer between 1 and some constant \( L \), inclusive. However, you do not know the specific weight of any box, and you do not know the specific value of \( L \). You are also given a pan balance. A pan balance functions in the following manner. You can give the pan balance any two disjoint sub-collections, say \( S_1 \) and \( S_2 \), of the boxes. Let \( |S_1| \) and \( |S_2| \) be the cumulative weight of the boxes in \( S_1 \) and \( S_2 \), respectively. The pan balance then determines whether \( |S_1| < |S_2| \), \( |S_1| = |S_2| \), or \( |S_1| > |S_2| \). You have nothing else at your disposal other than these \( n \) boxes and the pan balance. The problem is to determine if one can partition the boxes into two disjoint sub-collections of equal
weight. Give an algorithm for this problem that makes at most $O(n^2 L)$ uses of the pan balance. For partial credit, find an algorithm where the number of uses is polynomial in $n$ and $L$.

26. Give an algorithm that takes a positive integer $n$ as input, and computes the number of possible orderings of $n$ objects under the relations $<$ and $=$. For example, if $n = 3$ the 13 possible orderings are as follows:

- $a = b = c$,
- $a = b < c$,
- $a < b = c$,
- $a < b < c$,
- $a < c < b$,
- $a < c < b$,
- $b < a = c$,
- $b < a < c$,
- $b < c < a$,
- $b = c < a$,
- $c < a = b$,
- $c < a < b$, and
- $c < b < a$.

Your algorithm should run in time polynomial in $n$.

27. Give a polynomial time algorithm for the following problem. Consider a generalization of problem 24.

Instead of a single HTML page, your webserver now has $m$ pages. Thus, the input now consists of an $n$ by $m$ array, $R$, of non-negative integers, and an integer $k$. The value $R_{t,p}$ gives the number of requests for page $p$ at time $t$ and $k$ is the maximum broadcasts allowed. At each integer time, the server can broadcast as many different pages as it likes. If the server broadcasts page $p$ at time $t$, then all requests for page $p$ strictly before time $t$ are satisfied. The wait time of a request for page $p$ at time $i$ is defined as in problem 24. The goal is to pick the $k$ times and pages to broadcast in order to minimize the sum over all pages of all requests’ wait times.

HINT: Although the problem definition doesn’t explicitly say this, you should make sure you understand why it must be the case that broadcasting $q > 1$ different pages at time $t$ lowers the total broadcasts allowed by $q$ and not by 1, otherwise this problem is trivial given you’ve solved problem 24.