Given  $\{Y_1, \ldots, Y_n\}$  and M, design an algorithm to construct a non-decreasing set  $\{X_1, \ldots, X_n\}$  to minimize

$$cost(X) = \sum_{i=1}^{n} |X_i - Y_i|^2.$$

1. A subproblem is finding a non-decreasing sequence  $\{X_1,\ldots,X_n\}$  such that

$$0 \le X_1 \le \dots \le X_n \le m$$
,

where  $0 \le m \le M$ , and M is the original problem parameter. Problems are ordered by increasing m. Denote the solution sequence corresponding to m by  $X^{(m)}$ .

- 2. The base case is where m = 0. Then the sequence  $X^{(0)}$  is  $\{0, \ldots, 0\}$  (i.e. all 0's), and the cost of the base case is  $cost(X^{(0)}) = \sum_{i=1}^{n} Y_i^2$ .
- 3. First we'll consider what the sequence *is*, before considering the cost of the squence. Symbolically,

$$X^{(j)} = \{X_1^{(j-1)}, \dots, X_k^{(j-1)}, \underbrace{j, \dots, j}_{n-k}\}.$$

So the non-decreasing sequence minimizing cost with upper limit j is equal to the first k terms of the sequence with upper limit j-1, followed by n-k elements of value j, where k is chosen by inspection to minimize the cost of  $X^{(j)}$ . Note that k may be n, so it is possible that  $X^{(j)}$  does not contain a term with value j, and k may be 0, so that  $X^{(j)}$  does not contain any of the elements from  $X^{(j-1)}$ .

Thus the recurrence relation for the cost of the sequence is:

$$\mathrm{cost}(X^{(j)}) = \min(\mathrm{cost}(\{X_1^{(j-1)}, \dots, X_k^{(j-1)}, \underbrace{j, \dots, j}_{n-k}\}) | k \in \{0, \dots, n\})).$$

4. Now we prove that the recurrence relation is correct.

*Proof.* Note that when m = 0, there is only one feasible sequence of elements, that with all 0's. The cost of this sequence is therefore the solution to the subproblem for m = 0.

So we know that for some i, we can correctly solve the subproblem for m = i, so we know the sequence  $X^{(i)}$ . We claim that we can therefore solve the subproblem for m = i + 1. Since we know the sequence  $X^{(i)}$ , we can construct feasible solutions to the subproblem for m = i + 1 by constructing each of the sequences

$$\{X_1^{(i)}, \dots, X_k^{(i)}, \underbrace{i+1, \dots, i+1}_{n-k}\}\)$$
 for  $k \in \{0, \dots, n\}$ .

We know that the optimal sequence for m = i + 1 must be one of these, since the only way we could possibly improve the sequence from the solution for m = i is to increase the value of some of the elements in the sequence. Then we simply set  $X^{(i+1)}$  to the sequence with minimum cost. It follows that

$$\mathrm{cost}(X^{(i+1)}) = \min(\mathrm{cost}(\{X_1^{(i)}, \dots, X_k^{(i)}, \underbrace{i+1, \dots, i+1}_{n-k}\}) | k \in \{0, \dots, n\})).$$

Noting that this is the same relationship as our recurrence relation, we conclude that our recurrence relation is correct.  $\Box$ 

5. Now we give an algorithm for finding a solution to the subproblem.

Suppose we have correctly constructed  $X^{(i-1)}$ , let this be denoted by an array x of length n, and let best be positive infinity. We're trying to construct  $X^{(i)}$ . Let k be n, let min\_index be n. Save x in an array keep, so we remember the correct solution for m = i - 1.

Change the kth element of x to i, and compute the cost of the new x. If this cost is less than best, then let best be this cost, and let  $min_index$  be k. Decrement k, and repeat until k is 0.

Finally reset x so that it is the first  $min_index$  elements of keep, followed by  $n-min_index$  elements of value i.

6. Finally we give an algorithm for the original problem.

Given sequence y of length n and M, first let x be an array of zeroes of length n. Let best be positive infinity.

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for (m=0; m <= M; m++) {
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Copy x into keep. Let i be n, let m\_best be positive infinity, let min\_index be n.

Change the ith element of x to m. Compute the cost of the new x. If this cost is less that m\_best, then let m\_best be the cost and let min\_index be i. Decrement i and repeat until i is 0.

Reset x so that it is the first min\_index elements of keep, followed by  $n-\min_i$  elements of value m.

Compute the cost of x again, and if it's less than best, then set best equal to the cost.

}

Return best.

Thus the solution to the original problem is the cost of  $X^{(M)}$ , i.e. the cost of the solution sequence to the subproblem corresponding to M.

7. The running time of the algorithm is O(nM). Essentially, we're solving M subproblems that correspond to constructing sequences with upper limit equal to each non-negative integer less than or equal to M; thus the outer loop in our algorithm runs M times. At each step in the loop, we search for the min\_index, which takes O(n) operations, since we have to go through the entire x array of length n.