# Solution for Home Assignment 2 (Theoretical part) Danis Alukaev BS19-02

### 3. Connected lines least squares.

Given n points on a 2D plane  $(x_1, y_1), (x_2, y_2), ..., (x_n, y_n)$  with  $x_1 < x_2 < ... < x_n$  that lie roughly on a sequence of several line segments. The idea is to approximate this data set not with a single line, but with several connected lines. So, our goal is to find the minimum cost (defined below) for given n points in the plane using dynamic programming.

Let MIN(j)= minimum cost for points  $(x_1, y_1), (x_2, y_2), ..., (x_j, y_j),$  e(i, j) = minimum sum of squared errors for points  $(x_i, y_i), (x_{i+1}, y_{i+1}), ..., (x_j, y_j),$ Sum of squared errors  $SSE = \sum_{i=1}^{n} (y_i - (ax_i + b))^2$ , that is minimized when

$$a = \frac{n\sum_i (x_iy_i) - (\sum_i x_i)(\sum_i y_i)}{n\sum_i x_i^2 - (\sum_i x_i)^2} \text{ and } b = \frac{\sum_i y_i - a\sum_i x_i}{n}$$

Now, we can define trade-off (cost) function for a particular solution of connected lines least squares problem: Trade-off = (the sum of SSE in each segment) + (the number of lines in our approximation).

### 1.Define what are the subproblems.

The minimal cost of the particular point  $(x_j, y_j)$  can be either e(i, j) for the solution that keep the line started in  $(x_i, y_i)$  to approximate point  $(x_j, y_j)$  where  $1 \le i < j \le n$ , or e(i, j - 1) + 1 for solution that requires introduction of a new line. We notice that for some problem MIN(j) the segment (with the right endpoint  $(x_j, y_j)$  and the optimal partition) starts at the some earlier point  $(x_i, y_i)$ . Therefore, we can solve the problem MIN(j) if we know the solution for the subproblem MIN(i - 1) for  $1 \le i \le j \le n$ .

## 2.Recognize and solve the base cases.

Since the calculation of MIN(j) requires the value of MIN(i-1) and  $i \ge 1$  (number of points), our base case will be MIN(0). Let's consider such a case: if we have no points, then we do not introduce any lines, and minimal cost of the solution will be equal to 0. Therefore, MIN(0) = 0.

## 3. Write down the recurrence that relates subproblems.

There are two possible scenarios for a particular point  $(x_j, y_j)$ , where  $1 \le j \le n$ :

First scenario is to keep the line started in some point  $(x_i, y_i)$  where  $1 \le i < j$ , and use this line to approximate current point  $(x_j, y_j)$ . The cost of such a solution will be the e(i, j).

Second scenario is to introduce new line started in  $(x_j, y_j)$  and the cost of such a solution will be the e(i, j - 1) + 1.

So, for a particular MIN(j) we try to find the balance of MIN(i - 1) and the e(i, j), where  $1 \le i \le j \le n$ . It means, that we consider the partition defined in the previous iteration and decide is it optimal to introduce new line, or just keep the line continuous from some point  $(x_i, y_i)$  to approximate the current point  $(x_i, y_i)$ .

Hence, the recurrence relation for the connected lines least squares problem can be defined as:

$$MIN(j) = min(e(i, j) + MIN(i - 1) + 1)$$
, for  $1 \le i \le j$  and  $j \ne 0$   $MIN(j) = 0$ , for  $j = 0$ .

## **4.Write the pseudocode (always using DP) that returns MIN(n).** The pseudocode shown in the table Algorithm 1 (page 3).

Firstly, we declare two arrays to store computations of the minimum cost and minimum SSE for corresponding points. The bottom-up approach allows us to avoid the overlapping of subproblems, i.e. it optimizes the process of e(i,j) and MIN(i) computation. Also, we define base case for connected lines least squares problem that is MIN(0) = 0. Then, we calculate and store minimum SSE for all combinations of points  $(x_i, y_i), (x_{i+1}, y_{i+1}), ..., (x_j, y_j)$  with  $1 \le i \le j \le n$ . Next step is the computing and storing the minimum cost for all sequences of points  $(x_1, y_1), ..., (x_t, y_t)$  with  $t \in [1..n]$ . To do that we use the derived in previous task recurrent formula  $MIN(j) = \min(e(i, j) + MIN(i-1) + 1)$ , for  $1 \le i \le j \le n$  and  $j \ne 0$ . Finally, the result will be stored in MIN[n], and we can return this value.

### 5. State and justify the complexity of your pseudocode.

To treat all pairs (i,j) with  $1 \le i \le j \le n$  we need the quadratic time. Every computation of e(i,j) takes linear time, because we process all points  $(x_i, y_i), ..., (x_j, y_j)$  to compute their squared errors. Therefore, the computation of all e(i,j) for  $1 \le i \le j \le n$  takes  $O(n^3)$  time.

Then, we process given n points, and on each iteration j we try to find such an i that  $1 \le i \le j$  holds and  $[e(i,j)+\mathrm{MIN}(i-1)+1]$  minimized. It can be performed in  $O(n^2)$  time. Hence, the time complexity of proposed algorithm is  $O(n^3) + O(n^2) = O(n^3 + n^2) = O(n^3)$ .

## Algorithm 1 Connected lines least squares.

```
1: // declare arrays
 2: new array Min[n+1] // to store computations of the minimum cost for
   corresponding points
 3: new array E[n+1][n+1] // to store computations of the minimum SSE
   for corresponding points
 4: // set the base case
 5: Min[0] \leftarrow 0
 6: for j = 1 to n do
       for i = 1 to j do
 7:
           // compute and store the minimum SSE for points
 8:
 9:
           //(x_i, y_i), (x_{i+1}, y_{i+1}), ..., (x_j, y_j)
           E[i][j] \leftarrow e(i,j)
10:
11: for j = 1 to n do
12:
       // introduce new variable that will store the MIN(j)
       \min \leftarrow \texttt{MAX\_VALUE}
13:
       for i = 1 to j do
14:
15:
           // compute the cost for corresponding points
16:
           temp \leftarrow Min[i-1] + E[i][j] + 1
           // check whether the obtained value is the minimum
17:
           // among all previously calculated values
18:
           if temp < min then
19:
20:
               // set the preliminary minimum cost
               \min \leftarrow \text{temp}
21:
       // store the result of MIN(j)
22:
       Min[j] \leftarrow min
23:
24: return Min[n]
```