# 区間の集合上での動的計画法

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区間の集合上で定義される問題で動的計画法を用いれば解くことのできるような問題を考え、動的計画法を効率的に実行する手法を述べる。応用として、グラフの最適系列分割や円弧グラフの最大重みクリークを  $O(m \log n)$  の手間で求めることができるようになる。また、区間グラフの最小重み支配点集合も  $O(n \log n)$  で求められる。

## Dynamic Programming on Intervals

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We consider problems on intervals which can be solved by dynamic programming. Specifically, we give an efficient implementation of dynamic programming on intervals. As applications, an optimal sequential partition of a graph G = (V, E) and a maximum-weight clique of a weighted circular-arc graph G = (V, E) can be obtained in  $O(m \log n)$  time, where n = |V| and m = |E|. We also present an  $O(n \log n)$  time algorithm for finding a minimum weight dominating set of an interval graph G = (V, E), provided G is given by its intersection model of n intervals.

## 1 Introduction

Dynamic programming is one of the most popular techniques for designing efficient algorithms and has been applied to many problems. For example, an optimal sequential partition of a graph and a maximum weight clique of a circular-arc graph can be obtained by dynamic programming on intervals [4, 6]. In this paper we propose an efficient implementation of dynamic programming on intervals.

For an interval z, we denote by x(z) and y(z) the left and right endpoints of z, respectively. We assume x(z) and y(z) are integers from 1 to p. Then the problem we consider is: for a set Z of intervals, a set  $SZ = \{z_1, z_2, \ldots, z_p\}$  with  $y(z_i) = i$  and a real-valued cost function c(z) for  $z \in Z$ , compute

$$E[i] = \min\{D[j] + C(j,i)\},\tag{1}$$

for each  $1 \le i \le p$ . Here we assume that the minimum is taken over all the integers j < i contained in  $z_i \in SZ$ . We also assume that D[i] can be computed from E[i] in constant time and that C(j,i) is a constant multiple of the total cost of intervals  $z \in Z$  that contain both i and j, i.e.,

$$C(j,i) = \alpha \sum_{j,i \in z \in Z} c(z).$$

Furthermore, we assume E[1] = 0. This problem can easily be solved by the following Algorithm DP based on dynamic programming:

For i := 1 to p, compute E[i] by (1) and D[i].

It is clear that Algorithm DP correctly computes all E[i] (and D[i]) since all D[j] (j < i) are already computed when computing E[i], and if we let q = |Z|, it requires  $O(q + p^2)$  time. In this paper we shall present an efficient implementation of Algorithm DP and thus obtain an  $O(q \log p)$  time algorithm for computing all E[i].

As applications, we can obtain an optimal sequential partition of a graph G = (V, E) and a maximum weight clique of a weighted circular-arc graph G = (V, E) in  $O(m \log n)$  time, where n = |V| and m = |E|. Similarly, we can obtain other efficient algorithms on interval graphs and circular-arc graphs. For example, a minimum-weight dominating set of a weighted interval graph can be obtained in  $O(n \log n)$  time and a minimum-weight dominating set of a weighted circular-arc graph can be obtained in  $O(n^2 \log n)$  time, provided that their intersection models are given. These algorithms improve the complexity of existing algorithms [3, 5, 6].

## 2 Efficient Implementation

In this section, we present an efficient implementation of Algorithm DP. By this implementation, all E[i] in (1) can be computed in  $O(q \log p)$  time and O(q) space.

For the efficient implementation, we consider the following problem. We are given a set of numbers  $X = \{1, 2, ..., p\}$   $(p \ge 2)$  and a set DZ of intervals each of which has its both endpoinds in X. Also each interval  $z \in DZ$  has a cost cost(z). For a number x, csum(x, DZ) is defined as  $csum(x, DZ) = \sum_{x \in z \in DZ} cost(z)$  if  $DZ \ne \emptyset$  and  $csum(x, DZ) = \infty$  if  $DZ = \emptyset$ . We consider the following three operations.

$$findmin(z, DZ)$$
: return  $\min_{x \in z} \{csum(x, DZ)\},\$ 

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insert(z, DZ): replace DZ with DZ \cup \{z\}, delete(z, DZ): replace DZ with DZ - \{z\}.
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A crucial observation is:

Observation 1. Each of the operations findmin, insert and delete can be done in  $O(\log p)$  time with a data structure of O(p) space constructed by an O(p) time preprocessing.

Although this observation can be easily shown by using the segment tree in computational geometry [2],[7], we will prove briefly later in Section 2.2 for completeness. Now we assume Observation 1 and show that each E[i] and D[i] can be computed in  $O(\log p)$  time.

## 2.1 Implementation of Algorithm DP

In this subsection we give an implementation of Algorithm DP. It is based on a kind of (line) sweep. In the procedure, M is supposed to be a sufficiently large positive constant.

#### Procedure IMPLDP;

```
begin {comment computing E[i] and D[i];}
   X := \{1, 2, \dots, p\}; DZ := \emptyset;
   z_0 := [1, p]; cost(z_0) := M; insert(z_0, DZ);
   for i := 1 to p do begin
      if i = 1 then E[i] := 0 else E[i] := findmin(z_i, DZ);
4
      compute D[i] from E[i];
5
6
      y_i := [i, i]; cost(y_i) := D[i] - M; insert(y_i, DZ);
7
      if i is the right endpoint of an interval in Z then
8
          for each z \in Z with y(z) = i do delete(z, DZ);
      if i is the left endpoint of an interval in Z then
9
          for each z \in Z with x(z) = i do
10
              begin cost(z) := \alpha c(z); insert(z, DZ) end
11
   end
end;
```

The following observations are useful to show the correctness of Procedure IMPLDP.

- (A) If we choose M such that  $M > 2 \sum_{z \in Z} |\alpha_c(z)|$ , then just before the *i*-th iteration of Line 3, csum(x, DZ) > M/2 for all points  $x \in [1, p]$  except integer points x < i.
- (B) Just before the *i*-th iteration, csum(x, DZ) = D[x] + C(x, i) for an integer point x < i.

We will prove Observations (A) and (B) by induction on i.

Initially (just after Line 2), csum(x, DZ) = M > M/2 for all  $x \in [1, p]$  and thus, just before the first iteration (i = 1), (A) and (B) hold.

We assume that (A) and (B) hold just before the *i*-th iteration and consider the *i*-th iteration. Note that csum(x,DZ) < M/2 holds only if virtual interval  $y_x$  has been inserted into DZ in Line 6. Thus, during the *i*-th iteration, only csum(i,DZ) can newly become less than or equal to M/2 (this will happen when D[i] is less than or equal to M/2). Thus, (A) holds immediately after *i*-th iteration (i.e., just before the i+1-th iteration). Next we

consider (B). Define  $Z_R(x)$  to be the set of intervals in Z having x as the right endpoint, i.e.,

$$Z_R(x) = \{ z \in Z \mid y(z) = x \}.$$

Similarly, define  $Z_L(x)$  and Z(x,y) for x < y as follows.

$$Z_L(x) = \{ z \in Z \mid x(z) = x \},$$

$$Z(x,y) = \{ z \in Z \mid x,y \in z \}.$$

Note that  $C(x,y) = \alpha \sum_{z \in Z(x,y)} c(z)$ . During the *i*-th iteration, all intervals in  $Z_R(i)$  are deleted from DZ and all intervals in  $Z_L(i)$  are inserted to DZ. Furthermore, D[x] is not changed for an integer x < i. Thus,  $Z(x,i+1) = Z(x,i) - Z_R(i)$  and just before the i+1-th iteration, we have

$$csum(x, DZ) = D[x] + C(x, i + 1),$$

for an integer x < i. Similarly, for a sufficiently small  $\varepsilon > 0$ ,

$$Z(i, i+1) = (Z(i-\varepsilon, i) - Z_R(i)) \cup Z_L(i).$$

Thus,

$$csum(i, DZ) = D[i] + C(i, i + 1)$$

after *i*-th iteration, since  $y_i$  is inserted. This implies that (B) holds just before the i + 1-th iteration. Thus, we complete our proof of (A) and (B).

These observations imply that each E[i] (and D[i]) is correctly computed by  $E[i] := findmin(z_i, DZ)$  (or E[i] := 0) in Line 4 during the *i*-th iteration when  $i = y(z_i)$  and  $findmin(z_i, DZ) < M/2$ , since csum(x, DZ) = D[x] + C(x, i) for an integer point x < i and

$$findmin(z_i, DZ) = \min_{x \in z_i} \{csum(x, DZ)\}$$
$$= \min_{x \leq i, x \in z_i, x \text{:integer}} \{D[x] + C(x, i)\},$$

by Observations (A) and (B) (note that csum(i, DZ) > M/2 just before the *i*-th iteration). If  $findmin(z_i, DZ) > M/2$ , then the correct value of E[i] is  $\infty$  and in the procedure E[i] (and D[i]) is set a value > M/2. Thus we can obtain the following theorem by Observation 1 since the number of operations findmin, insert and delete is O(q).

**Theorem 1.** All E[i] can be correctly computed in  $O(q \log p)$  time.

## 2.2 Data Structure Supporting findmin, insert and delete

In this subsection, we shall describe a data structure supporting the operations findmin, insert and delete defined before. Since delete(z, DZ) can be replaced by insert(z', DZ) using z = z' and cost(z') = -cost(z), we consider only findmin and insert.

We consider a balanced binary search tree T(X) for a set  $X = \{1, 2, ..., p\}$ . That is, T(X) satisfies the following (i)-(iii).

(i) Each node of T(X) has either no children or two children, its left child  $\ell(v)$  and its right child r(v) (a node with no children is a leaf and a node with children is an inner node).

(ii) Each node v of T(X) contains a key  $key(v) \in X$  and keys are arranged in symmetric order: each  $i \in X$  is stored in exactly one leaf of T(X) as a key, and each inner node v has key(v) satisfying

$$\max_{u\in L(v)}\{key(u)\}=key(v)<\min_{u\in R(v)}\{key(u)\},$$

where L(v) denotes the set of descendants of the left child  $\ell(v)$  of v and R(v) denotes the set of descendants of the right child r(v) of v.

(iii)  $depth(v) = O(\log p)$  for each node v of T(X).

Note that this balanced binary search tree T(X) can be constructed in O(p) time and O(p) space. For simplicity, we write  $x_i$  to denote the leaf containing i as a key. We consider two values cmin(v) and cin(v) for each node v of T(X). We maintain cmin(v) and cin(v) so that they satisfy

$$cmin(v) = \min\{cmin(\ell(v)) + cin(\ell(v)), cmin(r(v)) + cin(r(v))\}$$
 (2)

for each inner node v of T(X). Let P(u, v) be the simple path of T(X) from node u to node v. We also use P(u, v) to denote the set of nodes on the path P(u, v). Initially (when  $DZ = \emptyset$ ), cmin(v) = 0 and cin(v) = 0 for each node v of T(X). Equation (2) together with  $cmin(x_i) = 0$  for each leaf  $x_i$  will imply  $cmin(v) = \min\{\sum_{u \in P(v, x_i) - \{v\}} cin(u)\}$  for each inner node v of T(X), where the minimum is taken over all the leaves  $x_i$  that are descendants of v.

We first consider insert(z, DZ). Let z = [i, j]  $(i \le j)$ . Then insert(z, DZ) can be done as follows. Let  $r_{ij}$  be the nearest common ancestor of  $x_i$  and  $x_j$ . Consider the paths  $P(r_{ij}, x_i)$  and  $P(r_{ij}, x_j)$  from  $r_{ij}$  to  $x_i$  and  $x_j$ . Let  $v_i$  be the deepest node on  $P(r_{ij}, x_i)$  such that  $v_i$  is the right child of a node on  $P(r_{ij}, x_i)$  if such  $v_i$  exists, otherwise we set  $v_i := r_{ij}$ . Similarly, let  $v_j$  be the deepest node on  $P(r_{ij}, x_j)$  such that  $v_j$  is the left child of a node on  $P(r_{ij}, x_j)$  if such  $v_j$  exists, otherwise we set  $v_j := r_{ij}$ . Define  $R(r_{ij}, v_i)$  and  $L(r_{ij}, v_j)$  as follows. If  $v_i = v_j = r_{ij}$  then  $R(r_{ij}, v_i) = \{\ell(r_{ij})\}$ . If  $v_i \neq r_{ij} = v_j$  then  $R(r_{ij}, v_i) = \{\ell(r_{ij})\}$ . If  $v_i \neq r_{ij}$  then

$$R(r_{ij}, v_i) = \{ v = r(p(v)) \mid p(v) \in P(\ell(r_{ij}), p(v_i)), v \text{ is not in } P(\ell(r_{ij}), p(v_i)) \}$$

that is,  $v \in R(r_{ij}, v_i)$  if and only if v is the right child of a node on the path  $P(\ell(r_{ij}), p(v_i))$  but v itself is not on the path  $P(\ell(r_{ij}), p(v_i))$ . Thus,  $v_i \in R(r_{ij}, v_i)$  if  $v_i \neq r_{ij}$ . If  $v_j \neq r_{ij}$  then

$$L(r_{ij}, v_j) = \{v = \ell(p(v)) \mid p(v) \in P(r(r_{ij}), p(v_j)), v \text{ is not in } P(r(r_{ij}), p(v_j))\}$$

that is,  $v \in L(r_{ij}, v_j)$  if and only if v is the left child of a node on the path  $P(r(r_{ij}), p(v_j))$  but v itself is not on the path  $P(r(r_{ij}), p(v_j))$ . Thus,  $v_j \in L(r_{ij}, v_j)$  if  $v_j \neq r_{ij}$ . Note that, for each  $k \in X$ ,  $k \in z = [i, j]$  if and only if there is exactly one node  $v \in R(r_{ij}, v_i) \cup L(r_{ij}, v_j)$  such that  $x_k$  is a descendant of v. We first set cin(v) := cin(v) + cost(z) for each  $v \in R(r_{ij}, v_i) \cup L(r_{ij}, v_j)$ . Next, we have to modify cmin(u) for all  $u \in P(r_{ij}, p(v_i)) \cup P(r_{ij}, p(v_j)) \cup P(r, r_{ij})$ , since cin(v) for all  $v \in R(r_{ij}, v_i) \cup L(r_{ij}, v_j)$  are now changed  $(P(r, r_{ij})$  is the path from the root r to  $r_{ij}$ ). This can be done by traversing nodes along the paths  $P(r_{ij}, p(v_i))$ ,  $P(r_{ij}, p(v_j))$  and  $P(r, r_{ij})$  in decreasing order of their depth and looking their children. Thus, insert(z, DZ) can be done in  $O(\log p)$  time, since the depth(v) of each node v of T(X) is  $O(\log p)$ .

Next we consider findmin(z, DZ). Let z = [i, j]  $(i \le j)$ . Let  $r_{ij}$ ,  $P(r_{ij}, x_i)$ ,  $P(r_{ij}, x_j)$ ,  $v_i$ ,  $P(r_{ij}, v_i)$ ,  $P(r_{ij}, v_j)$  and  $P(r, r_{ij})$  be the same as above. If we let

$$cs(v) := cmin(v) + \sum_{u \in P(r,v)} cin(u)$$

for each  $v \in R(r_{ij}, v_i) \cup L(r_{ij}, v_j)$ , then  $\min_{v \in R(r_{ij}, v_i) \cup L(r_{ij}, v_j)} \{cs(v)\}$  is the desired value findmin(z, DZ). This value can be computed in  $O(\log p)$  time by traversing nodes along the paths  $P(r_{ij}, p(v_i))$ ,  $P(r_{ij}, p(v_j))$  and  $P(r, r_{ij})$  in decreasing order of their depth and looking their children. Thus, findmin(z, DZ) can also be done in  $O(\log p)$  time. Note that T(X) with cmin(v) and cin(v) for each node v of T(X) can be represented in O(p) space and is almost the same as the segment tree in computational geometry [2], [7].

## 3 Applications and Related Problems

In this section, we present some applications of an efficient implementation of Algorithm DP. The first application is to optimal sequential partitions of graphs proposed by Kernighan [6].

### 3.1 Optimal Sequential Partitions of Graphs

We are given a graph G = (V, E), a cost function  $c: E \to R^+$ , a weight function  $w: V \to R^+$  and a positive number K. Here we assume the vertices of the graph are labeled from 1 to n. For a subset  $B = \{b_1, b_2, \ldots, b_s\}$  of V such that  $b_1 = 1$  and  $b_i < b_j$  for i < j, we obtain a family of graphs  $G(B) = \{G_1, G_2, \ldots, G_s\}$ , where each  $G_i$  is the subgraph of G induced by the set  $\{b_i, b_i + 1, \ldots, b_{i+1} - 1\}$  (for convenience, we assume  $b_{s+1} - 1 = n$ ). We will call P = G(B) a sequential partition (or partition, for short) of the graph G = (V, E) defined by G. The cost of partition G(B), denoted by G is defined to be the total cost of the edges of G joining different graphs in G(B). A partition G(B) is called addmisible if all the graphs in G(B) have the weights G (the weight of G) is called optimal if G and G is called optimal if G is cost G for all addmisible partitions G of G.

We would like to find an optimal partition of G. Kernighan [6] first proposed an algorithm for finding an optimal partition and later Kaji and Ohuchi [5] corrected the complexity analysis of his algorithm. They showed that the complexity of the Kernighan's algorithm is  $O(n^2)$ .

Here we will show that the Kernighan's algorithm can be implemented to run in  $O(m \log n)$  time, where m = |E|. We first review the Kernighan's algorithm. For a subset  $B_i$  of  $\{1, 2, \ldots, i\}$  containing 1 and i, the corresponding partition  $G(B_i)$  of G will be called a partial addmissible partition if all graphs  $G_j$  in  $G(B_i)$  except the graph containing the vertex i have the weights  $\leq K$ . Let T(i) be the minimum cost of a partial addmissible partition  $G(B_i)$  of G for a subset  $B_i \subset \{1, 2, \ldots, i\}$  containing 1 and i. Then Kernighan observed the following.

$$T(i) = \min\{T(j) - A(j-1,i) + A(i-1,i)\},\tag{3}$$

where minimum is taken over all j's such that  $w(j) + w(j+1) + \cdots + w(i-1) \leq K$  and A(x,y) (x < y) is the total cost of the edges which contain both x and y (we consider an edge e = (i,j) to be an interval [i,j] on the real line).

Thus, if we let E[i] := T(i) - A(i-1,i), D[j] := T(j) and C(j-1,i) := -A(j-1,i) by setting  $\alpha := -1$ , then we can rewrite (3) as follows.

$$E[i] = \min\{D[j] + C(j-1,i)\}.$$

This is almost the same as (1) and can be solved by a similar method described before. Here we reduce (3) to the same form as in (1). We consider the following intervals and their costs.

$$Z(E) = \{z(e) = [2x + 1, 2y] \mid e = [x, y] \in E\},\$$

$$Z(V) = \{z(i) = [2x(i), 2i] \mid i \in V, x(i) = \min\{j \mid w(j) + w(j+1) + \dots + w(i-1) \le K\}\},$$
 and

$$c(z(e)) = c(e)$$
 for  $z(e) \in Z(E)$ .

Let Z=Z(E) and  $SZ=Z(V)\cup\{z=[2i-1,2i-1]\mid i=1,2,\ldots,n\}$ . We use the same notation as in Section 3 (q=m and p=2n). Note that if we set  $\alpha:=-1$  then C(2j,2i)=-A(j-1,i). Thus if we let E[2i]:=T(i)-A(i-1,i) and D[2j]:=T[j], then we have the problem of computing E[2i] and D[2i] defined by (1)  $(E[2i-1] \text{ and } D[2i-1] \text{ will become } \infty)$ . Note that all A(i-1,i)=-C(2i,2i) can be computed in O(m) time in advance and we store them in a table. Thus we can compute D[2i] from E[2i] in costant time. Since there are m+n intervals and 2n endpoints, all E[2i] and D[2i] can be computed in  $O(m\log n)$  time by the method in Section 2.

## 3.2 Maximum Weight Clique of a Circular-Arc Graph

There are several variations of the problem described in Section 1. Here we consider one variation which appears in an algorithm for finding a maximum weight clique of a circular-arc graph [1, 4] and dominates the complexity of the algorithm. We are given a sequence  $Z = (z_1, z_2, \ldots, z_q)$  of intervals and a subsequence  $SZ = (z_{\sigma(1)}, z_{\sigma(2)}, \ldots, z_{\sigma(r)})$  of Z and a real-valued cost function c(z) for  $z \in Z - SZ$ , and compute

$$E[i] = \max\{D[j] + C(j,i)\},\tag{4}$$

for each  $1 \leq i \leq r$ . Here we assume that the maximum is taken over all the integers j < i such that  $\ell(z_{\sigma(j)})$  is contained in  $z_{\sigma(i)} \in SZ$ . We also assume that D[i] can be computed from E[i] in constant time and that C(j,i) is a constant multiple of the total cost of intervals  $z_k \in Z - SZ$  such that z appears after  $z_{\sigma(j)}$  and before  $z_{\sigma(i)}$  in the sequence Z (i.e.,  $\sigma(j) < k < \sigma(i)$ ) and contains both  $\ell(z_{\sigma(i)})$  and  $\ell(z_{\sigma(j)})$ . Furthermore, we assume E[1] = 0.

This problem can also be solved by an similar method as one described before. Thus, a maximum weight clique of a circular-arc graph G = (V, E) can be obtained in  $O(m \log n)$  time and O(n) space provided that G is given by its intersection model of arcs of a circle, where m = |E| and n = |V|. See [1] for details.

## 3.3 Minimum Weight Dominating Set

A dominating set of a graph G=(V,E) is a subset U of V such that, for every vertex  $v \in V$ ,  $v \in U$  or there is an edge  $e=(u,v) \in E$  connecting a vertex  $u \in U$  and v. A minimum weight dominating set is a dominating set of minimum weight, where each vertex has a positive weight. To find a minimum dominating set of an interval graph, we first

consider a shortest path problem on the interval graph. Here the length of a path is the sum of the weights of vertices on the path. We assume an interval graph G = (V, E) is given by its intersection model of intervals  $Z = \{z_1, z_2, \ldots, z_n\}$  on the line. We also assume  $x(z_i) \leq x(z_j)$  for i < j. We now want to compute the distances of shortest paths from  $z_1$  to other vertices  $z_i$ . Let D[i] be the distance of a shortest path from  $z_1$  to  $z_i$ . Then  $D[1] = w(z_1)$  and, for  $i = 2, 3, \ldots, n$ ,

$$D[i] = \min_{z_i \cap z_i \neq \emptyset} \{D[j] + w(z_i)\}.$$

Thus, if we let  $E[i] := D[i] - w(z_i)$  then we have

$$E[i] = \min_{z_j \cap z_i \neq \emptyset} \{D[j]\}.$$

This can be solved by a similar method described before. In fact, a simpler method can be applied and all E[i] can be computed in  $O(n \log n)$  time.

To find a minimum weight dominating set of an interval graph G=(V,E), we consider new intervals obtained from Z. Specifically, we consider an interval  $z_i'$  with  $w(z_i')=w(z_i)$  corresponding to  $z_i$  as follows:  $x(z_i')=x(z_i)$  and  $y(z_i')$  is the leftmost right endpoint y(z) of the intervals  $z\in Z$  such that  $y(z_i)< x(z)$ . Let  $z_0$  be a virtual interval with weight 0 such that  $x(z_0)=x(z_1)$  and  $y_0=\min\{y(z)\mid z\in Z\}$ . Then the distance of a shortest path from  $z_0$  to  $z_n'$  is the weight of a minimum dominating set of the graph G=(V,E). Thus, a minimum weight dominating set of an interval graph can be obtained in  $O(n\log n)$  time. Since a minimum weight dominating set of a circular-arc graph G=(V,E) can be obtained by solving n times the minimum weight dominating set problems of interval graphs, it can be obtained in  $O(n^2\log n)$  time. This improves the previous complexity [3].

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