

The Distance Constraint on Sequence Variables*

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Abstract

Insertion sequence variables have recently been introduced as a computational domain for modeling routing and sequencing problems in constraint programming. This paper investigates filtering for the (minimum) distance constraint over insertion sequence variables. This global constraint links a sequence to a distance variable based on a given distance matrix. So far, only a simple filtering algorithm has been proposed, which considers partial paths but ignores mandatory nodes. Our contribution introduces stronger lower bounds by integrating mandatory nodes. These bounds further enable the derivation of additional filtering rules for node insertions.

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1 Introduction

Routing and scheduling problems are central in constraint programming applications such as vehicle routing [3, 4], scheduling [5], and the Traveling Salesman Problem (TSP)[6]. Sequence variables [10] provide a flexible framework to model such problems by incrementally constructing solutions through node insertions into a partial sequence while enforcing ordering constraints.

Let V be a finite set of nodes. A sequence $\vec{s}$ is an ordered list without repetition, starting at α and ending at ω and $\vec{s}$ a subsequence. A sequence variable domain is defined by a partial sequence $\vec{s}$, a set of required nodes R , a set of excluded nodes X , and placement constraints. Domain update consists in inserting nodes from $I = V \setminus (\vec{s} \cup X)$ into $\vec{s}$. An insertion (v_i, v_j, v_k) inserts v_j between consecutive nodes v_i and v_k .

Figure 1 shows a sequence domain before and after an insertion.

Constraint propagation over sequence variables is a key component of this framework [1], in particular for cost-based constraints such as the distance constraint. This global constraint links a sequence $\vec{s}$ to $Dist$, its total travel cost: $Dist = \sum_{v_i \xrightarrow{\vec{s}} v_j} d_{i,j}$.

Propagating the distance constraint is challenging due to the combinatorial nature of insertion decisions. Existing approaches rely on relaxed propagation [2], considering insertions independently. More precisely, propagation only reasons on lower bounds computed on the current partial sequence:

* This paper is an extended abstract of a paper accepted at the main conference [8].



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41 $L_O(\vec{s}) = \sum_{v_i \xrightarrow{\vec{s}} v_j} d_{i,j}$, while ignoring the later cost of insertion of required insertable nodes
 42 that must be inserted later. For each insertion (v_i, v_j, v_k) , the associated detour cost is: $\Delta_{i,j,k} =$
 43 $d_{i,j} + d_{j,k} - d_{i,k}$. Insertions exceeding the upper bound are removed: $L_O(\vec{s}) + \Delta_{i,j,k} > \lceil Dist \rceil$.

44 In this paper, we strengthen this reasoning by introducing admissible lower bounds that explicitly
 45 account for required insertable nodes. These bounds are derived from relaxations of the insertion
 46 process and enable additional filtering rules for infeasible insertions.



■ **Figure 1** Sequence domain representation. The partial sequence $\vec{s}$ is represented with solid bold lines, while the remaining edges from the graph are shown with dashed gray lines. The right part shows the domain from the left part after inserting the required node 3 into $\vec{s}$.

2 Contribution

48 Our work introduces three relaxations of the insertion process, leading to stronger lower bounds and
 49 associated filtering rules.

50 The first relaxation considers that every required node must have at least one predecessor and one
 51 successor in the final sequence. Ignoring connectivity constraints, we assign each required node inde-
 52 pendently to its cheapest predecessor (or successor). The resulting bound sums the minimum incoming
 53 (or outgoing) arc costs: $L_{MIO}(\vec{s}) = \max \left(\sum_{v_j \in R} \min_{v_i \in R} d_{i,j}, \sum_{v_j \in R} \min_{v_k \in R} d_{j,k} \right)$.

54 This bound can also be exploited for filtering. When evaluating an insertion (v_i, v_j, v_k) , the
 55 minimum predecessor or successor contribution previously associated with v_j is removed and replaced
 56 by the actual detour induced by the insertion.

57 The second relaxation reasons directly on insertions. Each required insertable node is forced to con-
 58 tribute independently through its minimum insertion detour, while interactions between insertions are
 59 ignored. This gives the following lower bound: $L_{MD}(\vec{s}) = L_O(\vec{s}) + \sum_{v_j \in R \cap I} \min_{v_i, v_k \in R} \Delta_{i,j,k}$.

60 Similarly to the previous relaxation, filtering is achieved by replacing the minimum detour
 61 associated with a node by the actual detour induced by the considered insertion.

62 The third relaxation approximates the global predecessor and successor structure of the sequence
 63 without enforcing connectivity. Each node is constrained to have exactly one predecessor and one
 64 successor, while disconnected components are allowed. The resulting lower bound is obtained
 65 through a minimum cost maximum flow (MinCostMaxFlow), similarly to approaches used for the
 66 cardinality constraint with costs [7, 9].

3 Conclusion

68 We introduced three admissible lower bounds for the distance constraint on sequence variables,
 69 derived from relaxations of the insertion process. Experiments [8] show that the detour-based bound
 70 is particularly effective in reducing the search space and improving performance, highlighting the
 71 benefit of reasoning on future insertions. These results demonstrate the potential of insertion-based
 72 relaxations to strengthen propagation in sequence variables.

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