

# Applying the AC-3 Algorithm to Graph Coloring: Ensuring Different Colors for Neighboring States

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October 19, 2024

## 1 Introduction

The AC-3 algorithm ensures arc consistency between variables in a Constraint Satisfaction Problem (CSP). In a CSP, each variable has a domain of values, and constraints exist between variables. AC-3 operates by eliminating values from the domains that do not satisfy constraints. A CSP is defined by:

- **Variables:**  $X = \{X_1, X_2, \dots, X_n\}$
- **Domains:**  $D = \{D_1, D_2, \dots, D_n\}$  where each  $D_i$  represents the possible values for variable  $X_i$
- **Constraints:**  $C$ , a set of binary constraints between pairs of variables that restrict their simultaneous values.

The goal of the AC-3 algorithm is to enforce *arc consistency*. A variable  $X_i$  is said to be arc-consistent with a neighboring variable  $X_j$  if for every value  $x \in D_i$ , there is some value  $y \in D_j$  such that  $(x, y)$  satisfies the constraint between  $X_i$  and  $X_j$ .

## 2 Pseudocode

Pseudocode for AC-3

Pseudocode for Remove Inconsistent Values

## 3 Mathematical Definition and Analysis

Given a CSP defined by:

- A set of variables  $X = \{X_1, X_2, \dots, X_n\}$

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**Algorithm 1** AC-3 Algorithm

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0: **Input:** A CSP problem with variables, domains, and constraints.  
0: **Output:** True if CSP is arc-consistent, False if inconsistent.  
0: Initialize the queue with all arcs  $(X_i, X_j)$  in CSP.  
0: **while** queue is not empty **do**  
0:    $(X_i, X_j) \leftarrow$  dequeue from queue  
0:   **if** REMOVE-INCONSISTENT-VALUES( $X_i, X_j$ ) **then**  
0:     **if**  $D_i$  is empty **then**  
0:       **return** False  
0:     **end if**  
0:     **for all**  $X_k$  in NEIGHBORS( $X_i$ ) **do**  
0:       Enqueue  $(X_k, X_i)$   
0:     **end for**  
0:   **end if**  
0: **end while**  
0: **return** True = 0

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**Algorithm 2** Remove Inconsistent Values

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0: **Input:** Variables  $X_i$  and  $X_j$ , domains  $D_i$  and  $D_j$ , and a constraint between them.  
0: **Output:** True if domain of  $X_i$  is revised, False otherwise.  
0: revised  $\leftarrow$  False  
0: **for all**  $x \in D_i$  **do**  
0:   **if** no  $y \in D_j$  allows  $(X_i = x, X_j = y)$  to satisfy the constraint **then**  
0:     Remove  $x$  from  $D_i$   
0:     revised  $\leftarrow$  True  
0:   **end if**  
0: **end for**  
0: **return** revised = 0

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- Domains  $D = \{D_1, D_2, \dots, D_n\}$ , where each  $D_i$  is the domain of variable  $X_i$
- Binary constraints  $C_{ij}$  between pairs of variables  $X_i$  and  $X_j$

We define a CSP as **arc-consistent** if for every pair of variables  $(X_i, X_j)$ , for every value  $x \in D_i$ , there exists a value  $y \in D_j$  such that the constraint  $C_{ij}(x, y)$  is satisfied:

$$\forall x \in D_i, \exists y \in D_j \text{ such that } C_{ij}(x, y) = \text{True}$$

If no such value  $y$  exists for some  $x$ ,  $x$  is removed from  $D_i$ , and we revise the domain as:

$$D'_i = D_i \setminus \{x \mid \forall y \in D_j, C_{ij}(x, y) = \text{False}\}$$

The AC-3 algorithm works by iterating over all arcs and enforcing arc consistency. If a domain becomes empty during this process, the CSP is deemed *inconsistent*.

### 3.1 Time Complexity Analysis

The time complexity of the AC-3 algorithm is  $O(ed^3)$ , where:

- $e$  is the number of arcs (or binary constraints) between the variables.
- $d$  is the size of the largest domain.

#### Simplified Derivation:

- For each arc  $(X_i, X_j)$ , the algorithm checks all values in the domain of  $X_i$  (which has size  $d$ ).
- For each value in  $D_i$ , it compares with all values in  $D_j$  (also of size  $d$ ).
- This means each consistency check takes  $O(d^2)$  time.
- Since each arc might be added back to the queue up to  $d$  times, the worst-case complexity for an arc is  $O(d^3)$ .

Therefore, the total time complexity is:

$$O(e \cdot d^3)$$

where  $e$  is the number of arcs in the CSP.

### 3.2 Space Complexity Analysis

The space complexity of AC-3 is  $O(e + d)$ , where:

- $e$  is the number of arcs (constraints) stored in the queue.
- $d$  is the size of the largest domain, which we must store for each variable.

## 4 Application to Graph Coloring

In the Graph Coloring Problem, variables represent the nodes, and the domains represent possible colors. Constraints exist between adjacent nodes that must not share the same color. AC-3 helps by pruning the domain of possible colors for each node based on its neighbors.

### Example

Consider a node  $X_i$  with a domain of colors:

$$D_i = \{R, G, B\}$$

If one of its neighbors is assigned the color  $R$ , AC-3 will remove  $R$  from the domain of  $X_i$ , leaving:

$$D_i = \{G, B\}$$

This pruning of the search space helps reduce the number of possibilities, thereby making it easier to find a valid coloring for the entire graph. The use of AC-3 in this context significantly improves the efficiency of the search process for a solution.

## 5 Conclusion

The AC-3 algorithm is a powerful technique for enforcing arc consistency in graph coloring problems, such as ensuring that neighboring states in a map have different colors. Its efficiency in terms of both time and space makes it ideal for practical applications where constraints between adjacent regions must be satisfied. By iteratively revising the color domains for each state, AC-3 ensures that neighboring states remain properly distinguished or detects potential conflicts early in the process.

## 6 References

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