



MASTER IN COMPUTER SCIENCE

Course: Constraint Programming

# Mini-solver in OCaml based on binary Constraint Satisfaction Problem

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# 1 Introduction

The goal of this project is to realize a binary constraint solver based on the arc consistency filtering algorithms AC-3, AC-4, AC-6, AC-2001. As we have seen in the Constraint Programming course provided by Mr. Régin, a constraint satisfaction problem is a problem made of a set of variables  $V = \{V_1, ..., V_n\}$  such that each variable  $V_i$  is defined over a domain  $D_i$ . A domain is a set of values that can be assigned to the corresponding variable.

A constraint is a relations between a set of values taken from some of the given domains such that each domain of each constraint appear at most one time.

Let n be the number of variables of the problem, the constraint problem can be represented as an undirected n-partite hyper-graph G = (V, E) where each partition of the graph is made of the values of each domain. V is made of all the values of each variable and E represent the constraints of the graph.

**Example 1.** If we have the variables  $V_1 = \{1, 2, 3\}$  and  $V_2 = \{1, 2\}$  and  $V_3 = \{0, 1, 2\}$  and the constraint  $C_1 \triangleq |v_1 - v_2| = v_3$ , we build the hyper-graph G made of the vertices  $\{1_{V_1}, 2_{V_1}, 3_{V_1}, 1_{V_2}, 2_{V_2}, 0_{V_3}, 1_{V_3}, 2_{V_3}\}$  and the hyper-edges are made such that the constraint  $C_1$  is respected, for instance we can build the multi-edge  $e = \{1_{V_1}, 1_{V_2}, 0_{V_3}\}$  since the absolute value of the difference between the value 1 from  $V_1$  and the value 1 from  $V_2$  gives 0 in  $V_3$ .

A solution of a constraint satisfaction problem is a simple-path passing exactly one time through each partition of the hyper-graph, this is an equivalent for the classic definition which says that a solution is a the choice of a value for each variable such that every constraint are satisfied.

A value  $v_i \in D_i$  of the variable  $V_i$  is supported in the hyper-graph if for each constraint  $c_1$  involving the domains  $D(C) = \{d1, \ldots, d_n\}$  we have  $d_i \in D(C)$  and for each domain  $d_j \in D(C)$  there exists a value  $v_j \in D_j$  having a relation with  $v_i$ . A not-supported value can be removed from its domains since it cannot be part of a solution of the problem.

We can find a solution of a constraint problem by choosing an arbitrary value  $v_i$  from a domain  $D_i$  and removing all the other values in  $D_i$ . We look for all the domains  $D_j$  linked to  $D_i$  and remove all the values in  $D_j$  that are no more supported; these values are called *delta domain*. We repeat this operation until no modification can be performed. This operation is called *propagation*. If after the propagation there exists an empty domain it means that there does not exist a solution containing  $v_1$ . We repeat the procedure with backtracking the state of the problem before  $v_i$  was chose and we choose a new value in  $D_i$  different from  $v_i$ . If on the other hand, after propagation we have no empty domains, we take a value  $v_j$  from another domain  $D_j$  and repeat the procedure. If we are able to select a value for each value which not produce an empty domain, it means we have found a solution.

# 2 Binary constraints and arc consistency

An interesting property of constraint satisfaction problems is that they can always been rewritten in an equivalent problem having only binary constraint. A binary constraint is a constraint linking only two variables and thanks to this strategy the corresponding graph will have no more hyper-edges.

**Example 2.** If we retake the problem depicted in Example 1, we can change its model by adding an auxiliary variable  $V_{aux}$  representing the "index" of each multi-relation of the original problem. For example, if we take  $\{1_{V_1}, 1_{V_2}, 0_{V_3}\}$ , we can say that  $1_{V_{aux}}$  is the index of this tuple of values. The constraint  $C_1$  is split in 3 sub-constraints:  $C_1^1$  representing the link between  $V_{aux}$  and  $V_1$ ,  $C_1^2$  representing the link between  $V_{aux}$  and  $V_2$ ,  $V_1^3$  representing the link between  $V_{aux}$  and  $V_3$ . Note that constraint  $C_1^3$  is made in order to respect the original constraint  $C_1$ . A more detailed example will be provided in Section 5.1.2

In the state of the art we can find a lot of algorithms aiming to filter the values of a domains returning the set of not-supported values in a binary constraint satisfaction problem after deletion of a value  $v_i$  in a domain  $D_i$ .

In the following paragraph we will sketch the main ideas behind the algorithms AC-3, AC-4, AC-6, AC-2001.

**AC-3** In the AC-3 algorithm, after the deletion of a value  $v_i$  from a domain  $D_i$ , AC-3 will iterate over each domain  $D_j$  with a relation with  $D_i$  and for each value of  $v_j \in D_j$ , if there does not exist a value in  $D_i$  supporting  $v_j$ ,  $v_j$  will be returned.

**AC-4** This algorithm has an internal data structure in order to improve the search of not-supported variables. Each value of each domain is associated to the list of the values supporting it. When we remove a value  $v_i \in D_i$ , we can directly know which variables  $v_j$  depend on  $v_i$  and if  $v_j$  has no other variable in  $D_i$  supporting it,  $v_j$  is returned.

**AC-6** In AC-6, the internal data structure is similar to the one of AC-6, but instead of associating each value  $v_i$  to all the value  $v_j$  supporting  $v_i$ , we only store the first value in each domain supporting  $v_i$ . In this way, we reduce the amount of data to store in memory and when a value  $v_i \in D_i$  is removed, we look for the values  $v_j$  supporting it and if the support of  $v_j$  in  $D_i$  is different from  $v_i$  nothing is done. Otherwise, we look for a new support in  $D_i$  starting from the value  $v_i$ . Note that in this algorithm it is important to give an order to the values in the domain.

**AC-2001** In order to use the minimum amount of space, AC-2001 stores, for each value  $v_i \in D_i$ , the first element  $v_j$  in the related domain supporting it. When  $v_i$  is removed, we only look for the values depending on  $v_j$  and for them we look for a new support starting from  $v_i$  in  $D_i$ , if such value doesn't exist it means that  $v_j$  can be removed from  $D_j$ .

# 3 My Implementation

I have developed my solver in OCaml (v. 4.13.1) a functional programming language using pointer and the *Base* library since I have noticed better speed performances compared of the standard OCaml modules. In the following subsections I will provide a brief explanation of some of the most important modules I have implemented.

# 3.1 Doubly linked lists

A doubly linked list (dll) is a list whose elements have a pointer to their corresponding following and preceding element. One can note that the predecessor (resp. successor) of the first (resp. the last) element of a doubly linked list are represented by a fictive object: in my case I have used the *None* option type. Doubly linked lists are particularly useful since the insertion and the deletion of an element of a dll can be done in constant time: this is particularly useful to backtrack a list to a previous state.

```
type 'e node = {
  value : 'e;
  id : int;
  dll_father : 'e t;
  mutable prev : 'e node option;
  mutable next : 'e node option;
  mutable is_in : bool;
}
and 'e sentinel = { mutable first : 'e node; mutable last : 'e node }
and 'e t = { id_dom : int; name : string; mutable content : 'e sentinel option }
```

We can see that the type node has a prev and a next field which are of type optional. The doubly linked list, itself, is represented by the type t (following the OCaml convention) and it contains a sentinel pointing on the first and the last element of the dll.

To represent in a unique way nodes and dll, I have added the field id (resp.  $id\_dom$ ) in order to find them quickly when looking inside Hash-Tables.

The id of those records are generated through the generator:

```
let gen =
  let x = ref 0 in
  fun () -> incr x; !x
```

Inside the module *DoublyLinkedList*, I have added all of those utility functions allowing to modify the content of a *dll*. In particular, I can create, remove, insert, append or prepend a node inside a *dll* modifying correspondingly the *prev* and the *next* fields.

Moreover, I took inspiration from the List module in OCaml and I have added some higher-order functions in order to check if an element belongs to a dll, if a property is verified for every element in the dll (a kind of foreach in Java streams), etc.

#### 3.2 Graph, domains and constraints representation

The graph, the domains and the constraints are all implemented in the *Graph* module since they contains the information about the problem.

```
open Base
module DLL = DoublyLinkedList

type 'a domain = 'a DLL.t

type 'a value = 'a DLL.node

type 'a relation = 'a value -> 'a value -> bool

type 'a table_type = (int * int) Hash_set.t

type 'a graph = {
  tbl : 'a table_type;
  relation : 'a relation;
  constraint_binding : ('a domain, 'a domain DLL.t) Hashtbl.t;
  domains : 'a domain Hash_set.t;
}
```

The type graph is a record containing a *Hash\_set* of pairs of integers, that are the *id* of two values supporting each other; a *relation* taking two node and returning if they are linked in the constraint graph (it can be seen as the edges of the graph). Finally the *constraint\_binding* is a *Hash-Table* associating to the *id* of each domain the set of domain linked through a constraint.

We can add constraints between values through the auxiliary function:

This function take in parameter a graph, the name of a variable  $v_1$  followed by the name of its domain  $d_1$  and a second variable  $v_2$  with the name of its corresponding domain  $d_2$ . In this function,  $d_1$  and  $d_2$  are inserted to the list of domains and the constraint between the node  $v_1$  and  $v_2$  is added.

#### 3.3 The solver

The solver is the engine behind the resolution of a CP problem. The solver select the values of the domains and each time a selection is performed, the AC algorithm is asked to give back the delta domains.

My solver in OCaml is a *functor* taking in parameter a module of type *Arc\_consistency*.

The solver has two public functions:

```
module type Solver = sig
  module DLL = DoublyLinkedList

val initialization : ?verbose:bool -> string Constraint.graph -> unit

val find_solution :
  ?debug:bool ->
  ?count_only:bool ->
```

```
?only_valid:bool ->
?verbose:bool ->
?one_sol:bool ->
unit ->
unit
```

These function aim to initiate the problem inside the solver taking a graph g in entry, and to find one or all the solutions obtainable from g. All the optional argument of the  $find\_solution$  method are detailed in Section 4.

Inside the functor we can find all the auxiliary attributes and functions allowing to solve the given problem.

```
type 'a stack_type :
    (string AC.stack_operation * string Graph.value) option Stack.t
val backtrack_mem : 'a stack_type
val stack_op : 'a stack_type
val remove_by_node : ?verbose:bool -> string Graph.value -> unit
val propagation_remove_by_node : ?verbose:bool -> string Graph.value -> unit
val propagation_select_by_node : ?verbose:bool -> string Graph.value -> unit
val back_track : unit -> unit
```

**stack\_op:** is the stack containing all the operation made inside the Arc-Consistency algorithm.

**backtrack\_mem:** is the stack containing all the pointers to a previous state in the exploration tree in order to backtrack.

**remove\_by\_node:** when we remove a value  $v_i$  from a domain  $D_i$ , we call the Arc-Consistency algorithm passed to the solver *functor* and we add to the stack of undo operation the delta domain the set of values to remove for propagation in a second moment. Note that if after remove  $v_i$  from  $D_i$ , we throw the  $Empty\_domain$  exception which will be caught in order to backtrack and find other solutions.

**propagation\_remove\_by\_node:** this is an recursive function which propagates the deletion of a value  $v_i$  which keep to remove all the value inside the delta domain until it is not empty.

**propagation\_select\_by\_node:** is selected is a function that calls the  $propagation\_remove\_by\_node$  for all the values inside  $D_i$  that are different from the current value  $v_i$ . At each selection of a value  $v_i$ , we add to the  $backtrack\_mem$  a pointer to the actual state of the solver in order to get it back during the backtrack step.

#### 3.4 The Arc Consistency Algorithms

The Arc Consistency algorithms are modules respecting the signature:

```
module type Arc_consistency = sig
  exception Not_in_support of string

module DLL = DoublyLinkedList

type 'a data_struct
  type 'a stack_operation
  type 'a remove_in_domain = string Graph.value list

val name : string
  val print_data_struct : string data_struct -> unit
  val initialization : ?print:bool -> string Graph.graph -> string data_struct

val revise :
    string Graph.value ->
```

```
string data_struct ->
    string stack_operation * string remove_in_domain

val back_track : string stack_operation -> unit
end
```

In fact, an AC algorithm must have an initialization function allowing to clean the graph and instantiate the internal data structure, a revise function to remove a value  $v_i$  from its domain and return the delta domains and the stack operation to be done during the backtrack to restore the internal state of the internal data structure.

#### 3.4.1 AC-3

This algorithm has no data structure, therefore, its only usefull implementation is the revise function witch is the mere application of the AC-3 definition:

We iterate over every node in the domains with a constraint binding with the domain of the value  $v_1 \in D_1$  passed in argument of the function. If there is a value  $v_j \in D_j$  with no support in  $D_i$  then  $v_j$  is appended to the list of delta domains.

#### 3.4.2 AC-4

The internal structure of AC-4 is complex and I have tried to make it as efficient as possible.

```
type 'a double_connection = {
  node : 'a Graph.value;
  mutable assoc : 'a double_connection DLL.node option;
}

type 'a cell_type = ('a Graph.domain, 'a double_connection DLL.t) Hashtbl.t
type 'a data_struct = ('a Graph.value, 'a cell_type) Hashtbl.t
type 'a stack_operation = 'a double_connection DLL.node list
```

The data structure is a  $Map \ \mathcal{M}_1$  associating to each value  $v_i$  of each domain a second map  $\mathcal{M}_{\in}$ .  $\mathcal{M}_2$  is a map containing all the domains  $D_j$  having a relation with  $v_i$ , to each domain  $D_j$  we associate a list of  $double\_connection$ . A  $double\_connection$  is a record containing the value of  $v_j \in D_j$  supporting  $v_i$  and a pointer to the reciprocal  $double\_connection$  going from  $v_j$  to  $v_i$ . This pointer is useful to speed the deletion of the support  $v_j$  from the support of  $v_i$  if  $v_j$  is deleted. The revise function returns the  $double\_connection$  nodes removed from the internal data structure of AC-4 and the list of delta domains.

## 3.4.3 AC-6

The AC-6 algorithm is the one that caused me a lot of problem of implementation, since it works with both S-Lists and the Last value.

```
type 'a cell = {
   s_list : ('a Graph.value * 'a cell) DLL.t;
   last : ('a Graph.domain, 'a Graph.value DLL.t) Hashtbl.t;
}
type 'a int_struct = ('a Graph.value, 'a cell) Hashtbl.t
type 'a data_struct = 'a Graph.graph * 'a int_struct
type 'a stack_operation =
   ('a Graph.value * 'a cell) DLL.node list * 'a Graph.value DLL.node list
```

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The data structure of AC-6 is made by the graph on one side and a Hashtbl which associates to each value a record of type 'a cell. A cell contains:

- the *s\_list* that is a *dll* associating to each value of the graph a pointer to the *cell* in the internal data structure. This pointer allow to fast find which last values should be modified when a value is deleted.
- the last value that is a *Hashtbl* associating a domain to its corresponding last value.

#### 3.4.4 AC-2001

```
type 'a remove_in_domain = string Graph.value list
type 'a last = (Graph.ValueDomain.t, 'a Graph.value DLL.t) Hashtbl.t
type 'a data_struct = { last : 'a last; graph : 'a Graph.graph }
type 'a stack_operation = 'a Graph.value DLL.node list
```

The AC-2001 data structure is particularly easy to implement:  $data\_struct$  contains the original graph and the field allowing to improve the support search after the deletion of a value  $v_i \in D_i$ . Inside the last field, we find an Hashtbl associating to each couple  $v_i, D_j$  (where  $D_j$  is a domain with a constraint with  $D_i$ ) the last support of the node  $v_i$  in  $D_j$ . When a node  $v_i$  is remove in  $D_i$ , we lopp for all the domains  $D_i$  with a constraint with  $D_i$  and for all value  $v_j \in D_j$  if the last support of  $v_j$  is  $v_i$  then we look for a new value  $v_i' \in D_i$  starting from  $v_i$  supporting  $v_j$ . If this support does not exists then  $v_j$  will be returned with the delta domains, otherwise  $v_i'$  will replace the last of  $v_j$ .

# 4 Run the project

In order to simplify the interaction with the solver, the main program accepts a list of optional parameters to set the input of the problem, the verbose print mode, the arc consistency mode *etc*.

Here a list of the parameters and their behavior can be printed in the stout passing the-help option. Here the full list of commands:

```
Set the filtering algo among 3, 4, 6, 2001 - default : 3
-ac
-ν
             Set the verbose mode
-f
             Set the input file
-first
             Stop after the first valid solution
-queens [N]
             Set the size of the queen solver (min N: 5)
-all-int [N] Set the size of the allIntervalSeries solver (min N: 3)
-only-sol
             Only print the number of fails and solutions
-only-valid
             Print only the valid solutions
             Debug mode
             Print the input graph
-print-inp
-help
             Display this list of options
--help
             Display this list of options
```

Note that the all-int problem is run if you pass both the all-int and the *queens* parameters. The -f parameter takes a file following the grammar proposed in Section 4.1.

#### 4.1 The parser

A simple parser have been added to the main in order to easily enter file for texting the solver. A valid file to be parsed respect the following grammar:

```
P := variables* "\n--\n" constraints*
variables := name ":" (domains " ")* "\n"
constraint := variable " " value " " variable " " value
name := [a-zA-Z]*
domains := [a-zA-Z]*
```

An example of file input can be seen in Appendix A.1

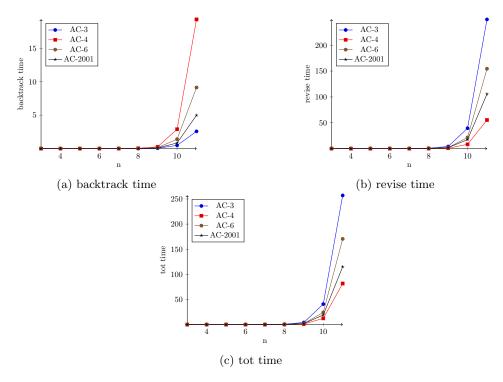


Figure 1: Time taken by AllIntervalSeries

# 4.2 Example of commands

• dune exec -- main -queens 3 -ac 3 -only-sol: run the 3-queens problem with the AC-3 arc consistency filter and print only the solutions on the console. Result:

The number of fails is 0
The number of solutions is 6
Total Time: 0.000068
Time of backtracks: 0.000003
Time of revise: 0.000013

- dune exec -- main -f ./graphs/input\_4.txt -ac 2001 -v produces the result proposed in Appendix  $A.2^1$
- the bash executables ./allInt.sh and ./queens.sh run respectively the All Interval Series and the N-Queens problem for n going from 0 to respectively 12 and 11. Not that both executable accept two integers in parameter changing the default bound of computation. For example, ./allInt.sh 3 7 will launch the All Interval Series for n from 3 to 7.

# 5 Benchmark

## 5.1 AllIntervalSeries problem

#### 5.1.1 Stats

#### 5.1.2 Generation of the problem

<sup>&</sup>lt;sup>1</sup>The file ./graphs/input\_4.txt is the one depicted in Appendix A.1

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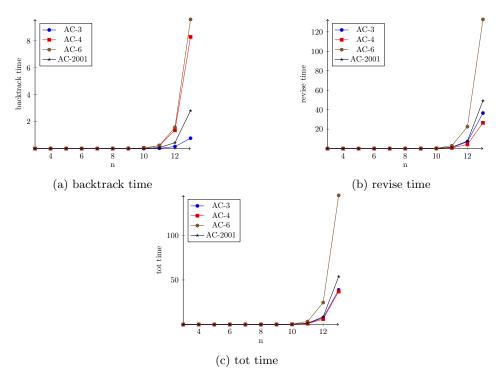


Figure 2: Time taken by Queens

# 5.2 Queens problem

# **5.2.1** Stats

# 5.2.2 Generation of the problem

# 6 Conclusion

# A An output example

#### A.1 Example file to parse

```
d1: a b c
d2: 2 3 4 5
d3: e f g h
--
d1 b d2 2;
d1 c d2 2;
d1 c d2 3;
d2 2 d3 f;
d2 3 d3 e;
d2 3 d3 g;
d2 4 d3 e;
d2 5 d3 e;
```

# A.2 Example of output

```
Initialization : removing a from d1
Initialization: removing 4 from d2
Initialization: removing 5 from d2
Initialization: removing h from d3
The data structure is:
node : (d2,2), last : (d1,b)
node : (d2,2), last : (d3,f)
node : (d3,f), last : (d2,2)
node : (d1,c), last : (d2,2)
node : (d1,b), last : (d2,2)
node : (d3,e), last : (d2,3)
node : (d3,g), last : (d2,3)
node : (d2,3), last : (d1,c)
node : (d2,3), last : (d3,e)
The domains are
-- Start Domains --
d1 : b;c;
d2: 2;3;
d3 : e;f;g;
--- End Domains ---
--> Selecting b from d1
* Removing c from d1
List of values having no more support = [(d2,3)]
* Removing 3 from d2
List of values having no more support = [(d3,g), (d3,e)]
* Removing g from d3
List of values having no more support = []
* Removing e from d3
List of values having no more support = []
--> Selecting 2 from d2
--> Selecting f from d3
A solution : [(d3,f), (d2,2), (d1,b)] !!
--> Selecting c from d1
* Removing b from d1
List of values having no more support = []
--> Selecting 2 from d2
* Removing 3 from d2
List of values having no more support = [(d3,g), (d3,e)]
* Removing g from d3
```

List of values having no more support = []

```
* Removing e from d3
List of values having no more support = []
--> Selecting f from d3
A solution : [(d3,f), (d2,2), (d1,c)] !!
--> Selecting 3 from d2
* Removing 2 from d2
List of values having no more support = []
--> Selecting e from d3
* Removing f from d3
List of values having no more support = []
* Removing g from d3
List of values having no more support = []
A solution : [(d3,e), (d2,3), (d1,c)] !!
--> Selecting f from d3
* Removing e from d3
List of values having no more support = []
* Removing g from d3
List of values having no more support = []
A solution : [(d3,f), (d2,3), (d1,c)] !!
--> Selecting g from d3
* Removing e from d3
List of values having no more support = []
* Removing f from d3
List of values having no more support = []
A solution : [(d3,g), (d2,3), (d1,c)] !!
The number of fails is 0
The number of solutions is 5
Total Time: 0.000270
Time of backtracks: 0.000002
Time of revise: 0.000015
```