

Project Report Draft - CMPT 417

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I. INTRODUCTION

This project aims to solve the **Pizza** problem given in the *LP/CP Programming Contest 2015*, where some students in the University College Cork want to make a large order of pizza for a party such that they use as many vouchers collected throughout the year as they can. The final objective is to use the vouchers so to obtain all ordered pizzas with the least possible cost.

Because we want to minimize the total possible cost for all ordered pizzas, the problem is then an optimization problem and not only satisfiability, meaning that we start our initial K , where K is our target cost, with the sum of prices of all pizzas with no vouchers used, as this is the maximum possible cost for this problem and, therefore, if we were to choose any K larger than that, K could be improved to at least as good as the total sum of the pizza prices. We repeat the process by looking for satisfiable instances of that K' that are smaller than our previous K and stop whenever we cannot satisfy K' , meaning that K' is the least possible value of K .

In this document, we first specify the problem by defining our vocabulary \mathcal{L} of functions and constant symbols for an \mathcal{L} -structure \mathcal{M} . Then we will show that, to solve this problem, we must find a vocabulary \mathcal{L}' (with a \mathcal{L}' -structure \mathcal{M}') that extends \mathcal{L} such that when we find a satisfiable instance of \mathcal{M} for all constraints in \mathcal{L}' , we find a satisfying instance for the initial problem, since \mathcal{M}' and \mathcal{M} share the same universe M . After specifying our vocabularies and constraint relations, we will show some tests instances of the problem run on **Minizinc** solver that will be then have their performance empirically evaluated. Finally, we will have a short discussion about the process of solving the Pizza problem.

II. SPECIFICATION

The pizza problem is as follows:

- The final goal is to obtain all ordered pizzas for the least possible cost.
- A voucher is a pair of numbers (a, b) where you pay for a pizzas and obtain b pizzas for free as long as each of the b pizzas cost no more than each of the a pizzas.
- A voucher does not need to be completely used and not all vouchers need to be used.

A vocabulary \mathcal{L} then can be defined by the symbols $[price, buy, free, n, m, k]$ where n is the number of pizzas, m is the number of vouchers, k is the cost bound, $price :$

$[n] \rightarrow \mathbb{N}$ is an unary function that maps the price of each one of the n pizzas, $buy : [m] \rightarrow \mathbb{N}$ is an unary function that maps the number of pizzas that must be bought for each one of the m vouchers and $free : [m] \rightarrow \mathbb{N}$ is an unary function that maps the number of pizzas that come for free when a voucher is used, for each one of the m vouchers. The universe M are all the numbers appearing in the structure. The **Minizinc** equivalent form of \mathcal{L} is:

```
int: n;  
array[1..n] of int: price;  
int: m;  
array[1..m] of int: buy;  
array[1..m] of int: free;
```

We need to find an assignment of pizzas and vouchers that minimize the total cost k and for this reason we will create a vocabulary \mathcal{L}' that extends \mathcal{L} , where \mathcal{L} and \mathcal{L}' share the same universe. Let \mathcal{L}' be defined by the symbols in \mathcal{L} and the symbols $[Paid, Used, Justifies, UsedFor]$, where the symbol $Paid : [n] \rightarrow \{0, 1\}$ is an unary symbol representing the set of paid pizzas, $Used : [m] \rightarrow \{0, 1\}$ is an unary symbol representing the set of used vouchers, $Justifies : [m] \rightarrow [n] \rightarrow \{0, 1\}$ is a binary symbol representing the set of vouchers v that will be used by paying for pizzas p and $UsedFor : [m] \rightarrow [n] \rightarrow \{0, 1\}$ is a binary symbol representing the set of free pizzas p that were obtained by using vouchers v . The **Minizinc** equivalent form of \mathcal{L}' is:

```
array[1..n] of var bool: Paid;  
array[1..m] of var bool: Used;  
array[1..m, 1..n] of var bool: Justifies;  
array[1..m, 1..n] of var bool: UsedFor;
```

To find a solution to instance \mathcal{L} we must solve for vocabulary \mathcal{L}' and \mathcal{L}' -structure \mathcal{M} and satisfy the problem constraints in **Minizinc**:

- C_1 . If we paid for a pizza p , then it cannot be in the set of free pizzas
constraint forall (p in $1..n$)
 $(Paid[p] \leftrightarrow \text{not exists}(v \text{ in } 1..m)(UsedFor[v, p]));$
- C_2 . If voucher v is used, then it must get at least one free pizza p with it
constraint forall (v in $1..m$)
 $(Used[v] \leftrightarrow \text{exists}(p \text{ in } 1..n)(UsedFor[v, p]));$
- C_3 . Any used voucher v must be justified by paying for exactly some pizzas p

constraint forall (v in $1..m$)
 ($Used[v] \rightarrow$
 $\text{sum}(p \text{ in } 1..n)(Justifies[v, p]) \geq buy[v]$);

- C_4 . The number of free pizzas cannot be greater than what is possible by using voucher v

constraint forall (v in $1..m$)
 ($\text{sum}(p \text{ in } 1..n)(UsedFor[v, p]) \leq free[v]$);

- C_5 . For every two pizzas p_1, p_2 , if p_1 was a pizza we got for free with voucher v and p_2 is a pizza we paid with voucher v , then the price of p_1 must be less or equal the price of p_2

constraint forall(p_1, p_2 in $1..n$ **where** $p_1 \neq p_2$,
 c in $1..m$)
 ($((UsedFor[c, p_1] \wedge Justifies[c, p_2]) \rightarrow$
 $price[p_1] \leq price[p_2])$);

- C_6 . Two vouchers v_1, v_2 cannot be justified by using the same paid pizza p

constraint forall (v_1, v_2 in $1..m$ **where** $v_1 \neq v_2$,
 p in $1..n$)
 ($Justifies[v_1, p] \rightarrow \text{not}(Justifies[v_2, p])$);

- C_7 . We pay for every pizza p used to justify use of a voucher v

constraint forall (p in $1..n$, v in $1..m$)
 ($Justifies[v, p] \rightarrow Paid[p]$);

- C_8 and C_9 . The pairs in *Justifies* and *UsedFor* can only be consisting of a voucher v and a pizza p

constraint forall (v in $1..m$, p in $1..n$)
 ($Justifies[v, p] \rightarrow (v \text{ in } 1..m \wedge p \text{ in } 1..n)$);
constraint forall (v in $1..m$, p in $1..n$)
 ($Used[v, p] \rightarrow (v \text{ in } 1..m \wedge p \text{ in } 1..n)$);

- C_{10} . The total cost (k , the sum of all paid pizzas) must be less or equal the sum of all pizza prices

int: $total = \text{sum}(price)$;
var int: $k = (\text{sum}(p \text{ in } 1..n)(Paid[p] * price[p]))$;
constraint $k \leq total$;

- And finally, we want to minimize the whole cost K
solve minimize k ;

III. TESTING

We created initially three test instances (the ones given in the *LP/CP Programming Contest 2015*) so to test if our specification was correct as defined on the contest and added seven so to evaluate how efficient the specification solve different instances of the same problem. Each data in the test assign values to the number n of pizzas, the number m of vouchers, the map *price* of pizzas prices and the maps *buy*

and *free* of vouchers, therefore establishing the universe M (all the numbers appearing in the structure). The detailed list of test instances were as below:

- Test 1: given in *LP/CP Programming Contest 2015*

$n = 4$;
 $price = [10, 5, 20, 15]$;
 $m = 2$;
 $buy = [1, 2]$;
 $free = [1, 1]$;

- Test 2: given in *LP/CP Programming Contest 2015*

$n = 4$;
 $price = [10, 15, 20, 15]$;
 $m = 7$;
 $buy = [1, 2, 2, 8, 3, 1, 4]$;
 $free = [1, 1, 2, 9, 1, 0, 1]$;

- Test 3: given in *LP/CP Programming Contest 2015*

$n = 10$;
 $price = [70, 10, 60, 60, 30, 100, 60, 40, 60, 20]$;
 $m = 4$;
 $buy = [1, 2, 1, 1]$;
 $free = [1, 1, 1, 0]$;

- Test 4: Testing instances when all pizzas have same price

$n = 10$;
 $price = [100, 100, 100, 100, 100, 100, 100, 100, 100, 100]$;
 $m = 5$;
 $buy = [3, 1, 1, 2, 3]$;
 $free = [3, 1, 1, 1, 2]$;

- Test 5: Testing instances when all pizzas have prices mixed

$n = 9$;
 $price = [7, 20, 80, 47, 54, 68, 46, 38, 7]$;
 $m = 4$;
 $buy = [3, 5, 1, 4]$;
 $free = [1, 6, 2, 1]$;

- Test 6: Testing when the $n = 1$

$n = 1$;
 $price = [1]$;
 $m = 4$;
 $buy = [3, 4, 5, 10]$;
 $free = [10, 5, 20, 15]$;

- Test 7: Testing with non obvious choice of vouchers

$n = 5$;
 $price = [6, 91, 45, 45, 30]$;
 $m = 10$;
 $buy = [4, 1, 4, 2, 1, 3, 4, 4, 4, 2]$;
 $free = [3, 4, 4, 1, 4, 4, 2, 3, 2, 3]$;

- Test 8: Testing with a simple case of a choice of a voucher
 $n = 2;$
 $price = [7, 8];$
 $m = 1;$
 $buy = [1];$
 $free = [1];$
- Test 9: Testing multiple choices of vouchers
 $n = 5;$
 $price = [100, 99, 25, 10, 1];$
 $m = 5;$
 $buy = [1, 2, 3, 4, 5];$
 $free = [1, 1, 1, 1, 1];$
- Test 10: Testing with pizzas prices that are always double of the price of the previous
 $n = 7;$
 $price = [1, 2, 4, 8, 16, 32, 64];$
 $m = 5;$
 $buy = [3, 4, 5, 3, 4];$
 $free = [1, 1, 1, 2, 2];$

When implementing the constraints on **Minizinc**, we initially used the instances Test 1, Test 2 and Test 3 so to guarantee that the specification was minimally correct before testing the other test instances Test 4 through Test 10. The conclusion is that the specification was working correctly as per the table below:

Instance	Least cost (k)	Output cost (k)	Runtime
Test 1	35	35	149 msec
Test 2	35	35	153 msec
Test 3	340	340	176 msec
Test 4	500	500	863 msec
Test 5	225	225	170 msec
Test 6	1	1	144 msec
Test 7	91	91	171 msec
Test 8	8	8	138 msec
Test 9	135	135	155 msec
Test 10	115	115	162 msec

IV. EMPIRICAL PERFORMANCE

For means of empirical performance of the project, we will try to answer the question #1: "Which is the best way to express a constraint? There is often more than one reasonable way to express a particular constraint on the solutions in a particular language, and sometimes these choices can have a very large effect on running time. You can compare performance with different versions of one constraint, or compare two very different specifications".

We will use two different constraint specifications for the pizza problem and will compare them. While writing the specification on **Minizinc**, there was an impression that when we specify constraints in linear form, where "linear" means

that the problem is not specified nestedly such as **constraint forall**(p in $1..n$, c in $1..m$)(...), they seem to be optimized and perform faster than their nested equivalent form **constraint forall**(p in $1..n$)(**forall** c in $1..m$)(...). We will then evaluate our test instances in each one of the two slightly different specifications, measure and compare their running time.

The tests will be performed using the solver *Gecode* 6.1.0 built-in on **Minizinc** and the evaluation environment is a Macbook Pro with a 2.2GHz quad-core Intel Core i7 processor and 16GB RAM.

The actual representation of the outcome will be in the final version of the project report, as this only a draft.

V. DISCUSSION

Solving problems with *SAT solvers* was initially very challenging because it is a completely different way of programming per se, as the problem is solved completely with a sort of declarative programming paradigm by constructing a specification with constraints and letting the computer do the actual problem solving, which is very different from the approach of solving problems I was accustomed (imperative programming). For this reason, I can say that I learned a new way of solving problems.

I also learned that there are Domain Specific Languages tools specialized in solving *SAT* problems and got intuition of how to make one with techniques such as breaking the problem in vocabularies and structures, then perform grounding in it so to later reduce the *SAT* problem to our problem and verify if the formula is satisfiable and that, by running *SAT* many times, you can optimize the solution for your problem.

Finally, I also learned that there many people in this field trying to improve the algorithms of *SAT solvers* and that depending on the tool you use and the constraints you specify, the running time of the process of problem solving might vary for better or for worse.

VI. DATA