Internship Report

Solal Rapaport

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1 Introduction

The goal is to build formulas that will allow robots, spread on a ring, to gather. We have k robots and we will use view vectors to build those formulas. The formulas will be an interpretation of the pseudo-code given in the research report [1].

The formulas we are building, will be used with formulas given in an other research report [2], and then will be tested in the acceleration algorithm using an interpolant [2]

2 Configuration with single multiplicity

The strategy ϕ_{SM} is true if the given configuration has a single multiplicity and that the robot calling the strategy should move toward the robot at distance d_0 :

$$\phi_{SM}(d_0, \dots, d_{k-1}) := (\bigvee_{i=0}^{k-1} (d_i = 0 \bigwedge_{j=0}^{k-1} _{j \neq i} (d_j > 0 \lor (d_j = 0 \land d_{j-1} = 0)))) \land (d_{k-1} \neq 0) \land ((d_1 = 0 \land d_{k-2} = 0 \land d_0 \le d_{k-1}) \lor (d_1 = 0 \land d_{k-2} \neq 0))$$

In order to test our strategy we need a function that will initialize our first configuration and make it one with a single multiplicity without being already a winning one. Here is its implementation in python:

```
def InitSM(p, s, t, taille_anneau):
    Initialize the given configuration
    The configuration will have no multiplicity
    and won't be a winning one
    tmpOr = []
    tmpAnd = []
    for i in range(len(p)):
        tmpOr.append(p[i] != p[(i+1)\%len(p)])
    tmpAnd.append(Or(tmpOr))
    # we make sure there is no winning configuration
    # as a result of this function
    tmpOr = []
    for i in range(len(s)):
        tmpAnd.append(p[i] >= 0)
        tmpAnd.append(p[i] < taille_anneau)
        tmpAnd.append(s[i] = -1)
        tmpAnd.append(t[i] == 0)
    for i in range(len(p)):
        tmpAndbis = []
        tmpAndbis.append(p[i] == p[(i+1)\%len(p)])
        tmpOrbis = []
        for j in range(len(p)):
```

```
if ((j != i)  and (j != (i+1)):
            tmpOrbis.append(p[j] = p[i])
            tmpAndter = []
            for h in range(len(p)):
                if h != j:
                     tmpAndter.append(p[h] != p[j])
            tmpOrbis.append(And(tmpAndter))
    tmpAndbis.append(Or(tmpOrbis))
    tmpOr.append(And(tmpAndbis))
tmpAnd.append(Or(tmpOr))
return And(tmpAnd)
```

We want to test this function, we choose a configuration with a ring of size 5 and 3 robots. We use the parameters below as an input:

```
# Taille de l'anneau
taille_anneau = 5
nb\_robots = 3
                         # Nombre de robot sur l'anneau
p = [ Int('p\%s', \% i) for i in range(nb_robots)
s = [Int('s\%s', \%i)] for i in range(nb_robots)
t = [ Int('t\%s', \% i) for i in range(nb_robots) ]
tabInit = InitSM(p, s, t, taille_anneau)
print("tabInit_:\n", tabInit)
solv1 = Solver()
solv1.add(tabInit)
\mathbf{print} ("solv1 \square: \square", solv1.check())
if(solv1.check() = sat):
    print("model_:\n", solv1.model())
The code above produces the following output:
tabInit:
 And (Or(p0 != p1, p1 != p2, p2 != p0),
    p0 >= 0,
    p0 < 5,
    s0 = -1,
    t0 = 0,
    p1 >= 0,
    p1 < 5,
    s1 = -1,
    t1 = 0.
    p2 >= 0,
    p2 < 5,
    s2 = -1,
    t2 = 0,
    Or(And(p0 = p1, Or(p2 = p0, And(p0 != p2, p1 != p2))),
       And(p1 = p2, Or(p0 = p1, And(p1 != p0, p2 != p0))),
       And(p2 = p0,
           Or(p0 = p2,
              And (p1 != p0, p2 != p0),
               p1 == p2,
               And (p0 != p1, p2 != p1)))))
solv1 : sat
model:
 [p1 = 1,
 p0 = 0,
 p2 = 0,
```

```
\begin{array}{lll} t2 &=& 0\,,\\ s2 &=& -1\,,\\ t1 &=& 0\,,\\ s1 &=& -1\,,\\ t0 &=& 0\,,\\ s0 &=& -1\,] \end{array}
```

What the model shows us is a configuration where robots 0 and 2 are in position 0 and robot 1 is in position 1.

Now we implement in python the $\phi_S M$ strategy:

```
def phiSM(distances):
    tabAnd = []
    tabOr = []
    for i in range(len(distances)):
        tabAndBis = []
        tabAndBis.append(distances[i] == 0)
        for j in range(len(distances)):
            if j != i:
                 tabOrBis.append(distances[j] > 0)
                 tabOrBis.append(And(distances[j] = 0,
                     \operatorname{distances}[j-1] = 0)
                 tabAndBis.append(Or(tabOrBis))
        tabOr.append(And(tabAndBis))
    tabAnd.append(Or(tabOr))
    tabAnd.append(distances[-1] != 0)
    tabOr = []
    tabOr.append(And(distances[1] == 0, distances[-2] == 0,
        distances[0] \le distances[-1])
    tabOr.append(And(distances[1] == 0, distances[-2] != 0))
    tabAnd.append(Or(tabOr))
    return And(tabAnd)
```

We test this function in the same configuration than given before with the following input :

```
# Taille de l'anneau
taille_anneau = 5
nb\_robots = 3
                        # Nombre de robot sur l'anneau
p = [Int('p\%s', \%i)] for i in range(nb\_robots)
s = [ Int(',s%s', % i) for i in range(nb_robots)
t = [ Int('t\%s', \% i)  for i  in range(nb_robots)
d0 = [Int('d\%s' \% i) for i in range(nb_robots)]
       Int('d\%s' \% i) for i in range(nb_robots)
d2 = [Int('d\%s', \% i)] for i in range(nb_robots)
tabInit = InitSM(p, s, t, taille_anneau)
tabConfig1 = ConfigView(taille_anneau, nb_robots, 0, p, d0)
tabConfig2 = ConfigView(taille_anneau, nb_robots, 1, p, d1)
tabConfig3 = ConfigView(taille_anneau, nb_robots, 2, p, d2)
tabPhiSM1 = phiSM(d0)
tabPhiSM2 = phiSM(d1)
tabPhiSM3 = phiSM(d2)
solv1 = Solver()
solv1.add(tabInit)
solv1.add(tabConfig1)
solv1.add(tabPhiSM1)
```

```
print("solv1_:_", solv1.check())
if(solv1.check() = sat):
        \mathbf{print} ("model:\n", solv1.model())
solv2 = Solver()
solv2.add(tabInit)
solv2.add(tabConfig2)
solv2.add(tabPhiSM2)
\mathbf{print} ("solv2 : : ", solv2 . check ())
if(solv2.check() = sat):
        \mathbf{print} ("model:\n", solv2.model())
solv3 = Solver()
solv3.add(tabInit)
solv3.add(tabConfig3)
solv3.add(tabPhiSM3)
print("solv3_:_", solv3.check())
if(solv3.check() = sat):
        print("model_:\n", solv3.model())
  It produces the following output:
solv1 : sat
model:
 [d0 = 1,
d2 = 4,
p1 = 1,
p0 = 0,
p2 = 1,
d1 = 0,
t2 = 0,
s2 = -1,
t1 = 0,
s1 = -1,
t0 = 0,
s0 = -1
solv2: sat
model:
 [d0 = 1,
d2 = 4,
p1 = 0,
p0 = 1,
p2 = 1,
d1 = 0,
t2 = 0,
s2 = -1,
t1 = 0,
s1 = -1,
t0 = 0,
s0 = -1
solv3 : sat
model:
 [d0 = 1,
d2 = 4,
p1 = 0,
p0 = 0,
p2 = 4,
```

```
\begin{array}{lll} d1 &=& 0\,,\\ t2 &=& 0\,,\\ s2 &=& -1\,,\\ t1 &=& 0\,,\\ s1 &=& -1\,,\\ t0 &=& 0\,,\\ s0 &=& -1\,] \end{array}
```

What it shows us is that every robot can move in this kind of configuration. The solver find a way to place the robot calling the strategy outside of the multiplicity and then allow the strategy to be true, meaning, the robot moves toward the multiplicity.

3 Gathering rigid configurations

Let d_{ij} be the value j of the view vector of the robot i, and ds_{ij} the value j of the symmetrical view of the robot i. The robot is calling the strategy ϕ_R , here are all the logic formulas used in order to build ϕ_R :

AllView is true if $d_{00}, \ldots, d_{k-1k-1}$ are all the views you can obtain from a single view vector $dist_0, \ldots, dist_{k-1}$:

$$AllView(dist_0, \dots, dist_{k-1}, d_{00}, \dots, d_{k-1k-1}) := (\bigwedge_{i=0}^{k-1} (\bigwedge_{j=0}^{k-1} (d_{ij} = dist_{(j+i) \mod k})))$$

IsRigid is true if the given configuration is a rigid configuration. Meaning, all the views are distinct, and so there is no multiplicity, the configuration isn't symmetric or periodic.

$$IsRigid(d_{00}, \dots, d_{k-1k-1}, ds_{00}, \dots, ds_{k-1k-1}) := \bigwedge_{i=0}^{k-1} (\bigwedge_{j=0}^{k-1} d_{ij} \neq 0) \land \\ \bigwedge_{i=0}^{k-1} (\bigwedge_{l=0}^{k-1} l_{\neq i} ((\bigvee_{j=0}^{k-1} d_{ij} \neq d_{lj}) \land (\bigvee_{j=0}^{k-1} d_{ij} \neq ds_{lj}) \\ \land (\bigvee_{j=0}^{k-1} ds_{ij} \neq d_{lj}) \land (\bigvee_{j=0}^{k-1} ds_{ij} \neq ds_{lj})))$$

AllCode is true if (α'_r, β'_r) is the set of two natural numbers of the robot r such as α'_r and β'_r are codes of r's views, with $\alpha'_r < \beta'_r$.

$$AllCode(d_{00}, \dots, d_{k-1k-1}, ds_{00}, \dots, ds_{k-1k-1}, \alpha_0, \dots, \alpha_{k-1}, \beta_0, \dots, \beta_{k-1}, \alpha'_0, \dots, \alpha'_{k-1}, \beta'_0, \dots, \beta'_{k-1}) := \\ \bigwedge_{i=0}^{k-1} (\alpha'_i < \beta'_i \land (\alpha'_i = \alpha_i \lor \alpha'_i = \beta_i) \land (\beta'_i = \alpha_i \lor \beta'_i = \beta_i)) \land \\ ((\alpha_0 < \alpha_1 < \dots < \alpha_{k-1} < \beta_0 < \dots < \beta_{k-1}) \land \\ (\bigvee_{p=0}^{k-1} (\bigwedge_{q=0}^{p-1} (d_{0q} = d_{1q}) \land d_{0p} > d_{1p})) \land \dots \land \\ (\bigvee_{p=0}^{k-1} (\bigwedge_{q=0}^{p-1} (ds_{(k-2)q} = ds_{(k-1)q}) \land ds_{(k-2)p} > ds_{(k-1)p})) \lor \\ ((\alpha_0 < \alpha_2 < \alpha_1 < \dots < \alpha_{k-1} < \beta_0 < \dots < \beta_{k-1}) \land \dots) \lor \dots)$$

CodeMaker is true if the configuration is rigid and if $(a_0, \ldots, a_{k-1}, as_0, \ldots, as_{k-1})$ are each code of each view passed as a parameter :

$$CodeMaker(d_{00}, \dots, d_{k-1k-1}, ds_{00}, \dots, ds_{k-1k-1}, a_0, \dots, a_{k-1}) := IsRigid(d_{00}, \dots, d_{k-1k-1}, ds_{00}, \dots, ds_{k-1k-1}) \land \exists \alpha_0, \dots, \alpha_{k-1}, \beta_0, \dots, \beta_{k-1}, \alpha'_0, \dots, \alpha'_{k-1}, \beta'_0, \dots, \beta'_{k-1}, AllCode(d_{00}, \dots, d_{k-1k-1}, ds_{00}, \dots, ds_{k-1k-1}, \alpha_0, \dots, \alpha_{k-1}, \beta_0, \dots, \beta_{k-1}, \alpha'_0, \dots, \alpha'_{k-1}, \beta'_0, \dots, \beta'_{k-1})$$

$$(\bigwedge_{i=0}^{k-1} (\bigwedge_{j=0, j\neq i}^{k-1} ((a_i > a_j \land \alpha'_j > \alpha'_i) \lor (a_i < a_j \land \alpha'_j < \alpha'_i)))) \land \bigwedge_{i=0}^{k-1} (\bigwedge_{j=0, j\neq i}^{k-1} a_i \neq a_j)$$

FindMax is true if Max is the highest value of the view vector passed as a parameter

$$FindMax(dist_0, \dots, dist_{k-1}, Max) := (\bigwedge_{i=0}^{k-1} (Max \ge dist_i) \wedge (\bigvee_{i=0}^{k-1} (Max = dist_i)))$$

5

FindM is true if M is the index of the robot (index in the view vector) which has the largest code of view and a neighboring robot at distance Max:

$$FindM(d_{00}, \dots, d_{k-1k-1}, a_0, \dots, a_{k-1}, Max, dM_0, \dots, dM_{k-1}) := \bigvee_{m=0}^{k-1} ((\bigwedge_{i=0}^{k-1} ((a_m \ge a_i \land (d_{i0} = Max \lor d_{ik-1} = Max)) \lor (d_{i0} < Max \land d_{ik-1} < Max))) \land M = m)$$

FindN is true if N is the index of the robot (index in the view vector) with the largest code of view and M as a neighboring robot at distance Max:

$$FindN(d_{00}, \dots, d_{k-1k-1}, a_0, \dots, a_{k-1}, Max, M, N) := (d_{M0} = Max \wedge d_{Mk-1} = Max \wedge ((N = ((M+1) \mod k) \wedge a_{(M+1) \mod k} > a_{(M-1) \mod k}) \vee (N = ((M-1) \mod k) \wedge a_{(M-1) \mod k} > a_{(M+1) \mod k}))) \vee (d_{M0} = Max \wedge d_{Mk-1} \neq Max \wedge N = ((M+1) \mod k)) \vee (d_{M0} \neq Max \wedge d_{Mk-1} = Max \wedge N = ((M-1) \mod k))$$

Since those formulas can't be implemented in python because it is impossible to work around a variable index, we choose to build a new formula, FindMN that will be true if both vectors dM and dN are the view vector of, respectively, M and N.

```
FindMN(d_{00},\ldots,d_{k-1k-1},a_{0},\ldots,a_{k-1},Max,M,N,\\ dM_{0},\ldots,dM_{k-1},dN_{0},\ldots,dN_{k-1}):=\\ \bigvee_{m=0}^{k-1}(\ (\bigwedge_{i=0}^{k-1}((a_{m}\geq a_{i}\wedge(d_{i0}=Max\vee d_{ik-1}=Max))\\ \lor(d_{i0}< Max\wedge d_{ik-1}< Max)))\wedge M=m\wedge\\ (\ (d_{m0}=Max\wedge d_{mk-1}=Max\wedge\\ ((N=M+1\ \operatorname{mod}\ k\wedge a_{(m+1)\ \operatorname{mod}\ k}>a_{(m-1)\ \operatorname{mod}\ k})\vee\\ (N=M-1\ \operatorname{mod}\ k\wedge a_{(m-1)\ \operatorname{mod}\ k}>a_{(m+1)\ \operatorname{mod}\ k})))\vee\\ (d_{m0}=Max\wedge d_{mk-1}\neq Max\wedge N=M+1\ \operatorname{mod}\ k)\vee\\ (d_{m0}\neq Max\wedge d_{mk-1}=Max\wedge N=M-1\ \operatorname{mod}\ k)\wedge\\ ((N=M-1\ \operatorname{mod}\ k\wedge (\bigwedge_{l=0}^{k-1}(dN_{l}=d_{(m-1\ \operatorname{mod}\ k)l}\wedge dM_{l}=d_{m((k-1)-l)}))))\vee\\ (N=M+1\ \operatorname{mod}\ k\wedge (\bigwedge_{l=0}^{k-1}(dN_{l}=d_{(m+1\ \operatorname{mod}\ k)l}\wedge dM_{l}=d_{m((k-1)-l)}))))))
```

 ϕ_R is true if the robot can move toward a multiplicity. There must be only one multiplicity and the configuration must be rigid:

```
 \phi_R(dist_0,\dots,dist_{k-1}) := \\ \exists d_{00},\dots,d_{k-1k-1}, \ AllView(dist_0,\dots,dist_{k-1},d_{00},\dots,d_{k-1k-1}) \land \\ \exists ds_{00},\dots,ds_{k-1k-1}, \bigwedge_{i=0}^{k-1} (ViewSym(d_{i0},\dots,d_{ik-1},ds_{i0},\dots,ds_{ik-1})) \land \\ \exists Max,a_0,\dots,a_{k-1},dM_0,\dots,dM_{k-1},dN_0,\dots,dN_{k-1}, \\ CodeMaker(d_{00},\dots,d_{k-1k-1},ds_{00},\dots,ds_{k-1k-1},a_0,\dots,a_{k-1}) \land \\ FindMax(dist_0,\dots,dist_{k-1},Max) \land \\ FindMN(d_{00},\dots,d_{k-1k-1},a_0,\dots,a_{k-1},Max,dM_0,\dots,dM_{k-1},dN_0,\dots,dN_{k-1}) \land \\ \exists dM2_0,\dots,dM2_{k-1},dN2_0,\dots,dN2_{k-1}, \\ ((\bigwedge_{i=0}^{k-1}(dM2_i=dM_{i+1}\ \ \text{mod}\ k)) \lor (\bigwedge_{i=0}^{k-1}(dM2_i=dM_{i-1}\ \ \text{mod}\ k))) \land \\ (\bigvee_{i=0}^{k-1}(dM2_i\neq dN_i)) \land \\ ((\bigwedge_{i=0}^{k-1}(dN2_i=dN_{i+1}\ \ \text{mod}\ k)) \lor (\bigwedge_{i=0}^{k-1}(dN2_i=dN_{i-1}\ \ \text{mod}\ k))) \land \\ \bigvee_{i=0}^{k-1}(distM_i=(\sum_{l=0}^{i}dM_l) \land distN_i=(\sum_{l=0}^{i}dN_l)) \land \\ (\bigvee_{i=0}^{k-1}(distM_i=(\sum_{l=0}^{i}dM_l) \land distN_i=(\sum_{l=0}^{i}dN_l)) \land \\ (\bigvee_{i=0}^{k-1}(distM_i=distN_i) \bigwedge_{j=0}^{k-1}(distM_j=dist_j)) \lor \\ (distM_i>distN_i \bigwedge_{q=0}^{i}(distM_q=distN_q) \bigwedge_{j=0}^{k-1}(distN_j=dist_j)) ) )
```

4 Gathering an odd number of robots

We are now building a strategy, ϕ_{ON} , that will gather an odd number of robots on a non-periodic configuration. It is the strategy with the lowest priority, it means that the configuration won't be rigid and won't have any multiplicity.

First we build the formula, IsOddNonPeriodic, that will return true if the number of robots is odd and if the configuration is non-periodic:

$$IsOddNonPeriodic(dist_0, \dots, dist_{k-1}) := \\ ((k+1) \mod 2 = 0) \land \\ \exists p \in [0; \frac{k-1}{2}] \\ \exists d'_0, \dots, d'_{p-1}, \bigwedge_{i=0}^{k-1} (d'_{i \mod p} = dist_i) \land dist_{k-1} \neq d'_{p-1}$$

Now, we build build the ϕ_{OD} strategy, it returns true if the configuration is non-rigid, non-periodic, has no multiplicity and has an odd number of robots. The robot just move in order to create a multiplicity or a rigid configuration that will lead to a rigid configuration or one with a multiplicity.

$$\phi_{ON}(dist_0, \dots, dist_{k-1}) := \\ \exists d_{00}, \dots, d_{k-1k-1}, \ AllView(dist_0, \dots, dist_{k-1}, d_{00}, \dots, d_{k-1k-1}) \land \\ \exists ds_{00}, \dots, ds_{k-1k-1}, \bigwedge_{i=0}^{k-1} (ViewSym(d_{i0}, \dots, d_{ik-1}, ds_{i0}, \dots, ds_{ik-1})) \land \\ \neg IsRigid(d_{00}, \dots, d_{k-1k-1}, ds_{00}, \dots, ds_{k-1k-1}) \land \\ IsOddNonPeriodic(dist_0, \dots, dist_{k-1}) \land \\ (\bigwedge_{i=0}^{k-1} dist_i \neq 0)$$

5 Tests

In order to test the algorithm [2] we will use the python code we show you at the beginning : InitSM and phiSM.

We will use the SAT-solver to test different configurations. We will change the number of robots and the size of the ring from a test to an other.

5.1 Test InitSM

First, we test the function InitSM alone: can we have an initial configuration with a single multiplicity with those parameters?

nb-robot \size-ring	2	3	4	5	6
2	Unsat	Unsat	Unsat	Unsat	Unsat
3	Sat	Sat	Sat	Sat	Sat
4	Sat	Sat	Sat	Sat	Sat
5	Sat	Sat	Sat	Sat	Sat
6	Sat	Sat	Sat	Sat	Sat

The results make sense: we can't create a multiplicity with 2 robots which is not a winning configuration. Else, even on a ring size of 2 we can have a multiplicity on one spot and only one robot on the other spot.

5.2 Test ϕ_{SM}

Now we test ϕ_{SM} through the algorithm [2], we also use the function InitSM that makes sure we have a single multiplicity at the beginning.

nb-robot \size-ring	2	3	4	5	6
3	Timeout	Timeout	Timeout		
4	Timeout	•••	•••		
5					
6					

Same test but with the function *Init* instead.

nb-robot \size-ring	2	3	4	5	6
3	Timeout	Loose	Loose	Loose	
4					
5					
6			•••	•••	

```
For nb_{robot}=3 and size_{ring}=2 we face this problem: Traceback (most recent call last): 

File "algov5.py", line 56, in <module> 

Ip = tree_interpolant(And(Interpolant(And(tmpAndInterpolant)), And(tmpAndContext))) 

File "/usr/lib/python3.8/site-packages/z3/z3.py", line 8297, in tree_interpolant 

res = Z3_compute_interpolant(ctx.ref(),f.as_ast(),p.params,ptr,mptr) 

File "/usr/lib/python3.8/site-packages/z3/z3core.py", line 4074, in Z3_compute_interpolant 

_elems.Check(a0) 

File "/usr/lib/python3.8/site-packages/z3/z3core.py", line 1336, in Check raise self.Exception(self.get_error_message(ctx, err)) 

z3.z3types.Z3Exception: b'theory not supported by interpolation or bad proof'
```

```
1 foreach synchronous winning strategy f do
 2
       k = 1;
       while true \ do
 3
 4
           I(c) = Init(c);
           continue = true;
 \mathbf{5}
           while continue do
 6
              I' = Interpolant(I(c) \land Post(c, c1), Post(c1, c2) \land \cdots \land
 7
                Post(c_{k-1}, c_k) \wedge BouclePerdante(c_k));
              if I' = SAT then
 8
                  if I = Init then
 9
                                                               /* Loosing Strategy */
                      exit;
10
                  else
11
                      k = k + 1;
                      continue = false;
13
                  end
14
              else
15
                  if I' \implies I then
16
                   exit;
                                                               /* Winning Strategy */
17
                  else
                   I = I \vee I';
19
                  end
20
              end
21
           end
22
       end
\mathbf{23}
24 end
```

References

- [1] Ralf Klasing, Euripides Markou, and Andrzej Pelc. Gathering asynchronous oblivious mobile robots in a ring. Tech. rep. RR-1422-07. UMR 5800 Université Bordeaux 1, 351, cours de la Libération, 33405 Talence CEDEX, France: Laboratoire Bordelais de Recherche en Informatique, Jan. 2007.
- [2] Nathalie Sznajder and Souheib Baarir. Algorithme d'accélération par interpolants. (French) [Acceleration Algorithm using an interpolant]. Tech. rep. Laboratoire Informatique de Paris 6 (LIP6), Feb. 2022.