# Latency for periodic multicasting

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## 1 Introduction

### 1.1 Context

Mobile networks are controlled by many radio stations, composed of 2 major parts :

- 1. Base Band Units (BBU): are responsible for computing the signals, and communicate with the core network.
- 2. Remote Radio Head (RRH): are the antennas and have only to communicate with the mobile devices.

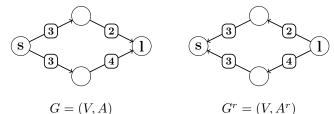
The part of the network between BBU and the core network is called backhaul. The operators are searching for a way to dissociate BBU from RRH, and regroup many BBU in some data centers, that would allow easier updates and maintenance. Nevertheless such a splitting enforces hard throughput and latency constraints on the communication network between BBU and RRH. This part of the network is called fronthaul.

In this article, we will work on latency constraints in fronthaul.

### 1.2 Definitions

We consider a directed graph G = (V, A) with two non intersecting subsets of vertices: a subset L of nodes which are called *leaves* and a subset S of nodes which are called *sources*. The indegree of nodes in S is equal to S, and the outdegree of nodes in S is also equal to S. We denote by S the cardinal of S and by S the cardinal of S. The digraph S models the network, S the possible BBU and S the set of RRH. Each arc S in S has an integer weight S representing the time taken by the signal to go from S to S by using this arc.

We consider G = (V, A) and  $G^r = (V, A^r)$  wherein the set of vertices is the same and  $A^r$  represents the edges of A directed in the other way.



Note that there is a special case in which the graph may have oriented cycles (optical ring topology), otherwise, there is no oriented cycle.

A route r is a sequence of arcs  $a_0, \ldots, a_{n-1}$ , with  $a_i = (u_i, u_{i+1}) \in A$  such that  $u_0 \in S$  and  $u_n \in L$ . The *latency* of a vertex  $u_i$  in r, with  $i \geq 1$ , is defined by

$$\lambda(u_i, r) = \sum_{0 \le k < i} Dl(a_k)$$

We also define  $\lambda(u_0, r) = 0$ . The latency of the route r is defined by  $\lambda(r) = \lambda(u_n, r)$ . In graph theory, a route is a simple path in the graph, and its latency is its weight.

A routing function  $\mathcal{R}$  is an application associating a route  $\mathcal{R}(s, l)$  to each couple  $(s, l) \in S \times L$  in G. Moreover  $\mathcal{R}$  satisfies the *coherent routing* property: the intersection of two routes must be a path.

For simplicity, we assume that we have as many source nodes as we have leaves  $(S = \mathcal{L})$ . A  $\mathcal{R}$ -matching is a bijection  $\rho : S \to L$  which associates to each  $s_i \in \{s_0, ..., s_{S-1}\}$  a  $l_i \in \{l_0, ..., l_{\mathcal{L}-1}\}$ . A  $\mathcal{R}$ -matching defines a set  $\{r_0, ..., r_{\mathcal{L}-1}\}$  of  $\mathcal{L}$  routes in R such that  $\forall i, r_i = R(s_i, l_i)$ .

For each arc  $(u,v) \in A$ , we denote by  $\rho(u,v)$  the subset of routes of  $\rho$  containing (u,v). We define the load of an arc (u,v) as the number of routes that use this arc thus:  $load(u,v) = |\rho(u,v)|$ .

The quintuplet  $N = (G, S, L, R, \rho)$  defines a **matched graph**. We call  $N^r$  the quintuplet  $(G^r, S, L, R, \rho^r)$ , where  $\rho^r$  is the  $\mathcal{R}$ -matching obtained using the same routes, with inverted arcs.

Let P > 0 be an integer called *period*. A P-periodic affectation of N, a matched graph consists in a set  $\mathcal{M} = (m_0, \dots, m_{\mathcal{L}-1})$  of  $\mathcal{L}$  integers that we call offset. Each time window is divided in P slots and the number  $m_i$  represents the slot number used by the route  $r_i$  at its source. We define the time slot used by a route  $r_i$  at any vertex v of the route by

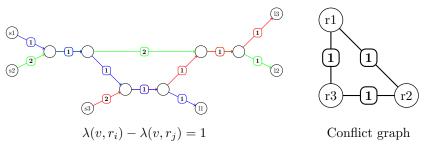
$$t(v, r_i) = m_i + \lambda(u, r_i) \mod P$$
.

A *P*-periodic affectation must have no *collision* between two routes in  $\rho$ , that is  $\forall r_i, r_j \in \rho, i \neq j$ , two routes intersecting in u, and containing the same arc (u, v), we have

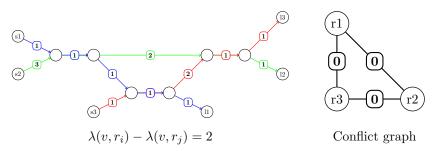
$$t(u, r_i) \neq t(u, r_j)$$
.

We define  $\alpha(u, r_i, r_j) = |\lambda(u, r_i) - \lambda(u, r_j)|$ . This will be useful later, when we'll introduce route conflict graphs.

Notice that the notion of P-periodic affectation is **not monotone** with regard to P. Indeed, we can build a  $\mathcal{R}$ -matching of a graph, with  $\mathcal{L}$  routes  $r_1, \ldots, r_l$  which all intersect two by two and such that if  $r_i$  and  $r_j$  have v as first common vertex we have  $\lambda(v, r_i) - \lambda(v, r_j) = 1$ . Therefore there is a 2-periodic affectation by setting all  $m_i$  to 0.



On the other hand if we set all  $\lambda(v, r_i) - \lambda(v, r_j) = P$ , there is no P-peridodic affectation if P < l.



Here for P=2, there is no P-Periodic affectation.

Therefore if we choose P odd and l=P+1, there is no P-perioddic affectation but modulo 2 all  $\lambda(v,r_i)-\lambda(v,r_j)$  are equal to one thus we have a 2-periodic affectation.

### 1.3 Fronthaul networks

The application we address here by studying the problems defined above is the following. Consider a fronthaul network in which source nodes in S represent computing units in data centers, each one able to do a remote process for base stations modeled by leaves node in L. Consider a  $\mathcal{R}$ -matching  $\rho$  of S in L. Consider a leaf l, its dedicated source node s and R(l,s) the route from l to s in R. We consider a P-periodic affectation of N, and also another P-periodic affectation of  $N^r$ . The periodic process is then the following one:

1. During each period of duration P, l sends a message to its dedicated source s, using their dedicated routes and considering the P-periodic affectation of N.

- 2. When each s receives this message, it makes a computation to answer l. This computation time is fixed.
- 3. After this computation time, s sends a message to l considering the Pperiodic affectation of  $N^r$ .

On source nodes, between the end of computation time and the emission time of the message (given by the P-periodic affectation), there is a waiting time. We define by  $\omega: r \to \mathbb{N}$  the waiting time of a route r in the  $\mathcal{R}$ -matching considered, i.e. the time during which the message is "sleeping", waiting to be sent through the network.

Source and leave nodes periods have to be the same. Indeed, leave nodes have to receive a message at every period, so source nodes have to send a message at every same period. So, we will find two P periodic affectations from the same P in both ways. If the messages have to be buffered, we want to do it in sources nodes. We define by  $\theta$  the computation time required at the source node before sending an answer to it's leave node.

Let us call T(r) the process time on a route r:

$$T(r) = 2\lambda(r) + \omega(r) + \theta$$

Since  $\theta$  is the same on every routes, we can simplify the model by removing this *theta*. Whether we want to consider it, we only have to lengthen all links before source nodes from an integer corresponding to  $\theta$ .

We consider 3 main problems:

### Problem Periodic Routes Assignment (PRA)

**Input:** matched graph N, integer P.

**Question:** does there exist a P-periodic affectation of N?

Optimisation goal: minimizing P.

This problem is our basic problem. We need to find the offset for each routes such that there is no collision between the signals emitted by sources at any nodes in the graph.

### Problem Network Periodic Assignment (NPA)

**Input:** G = (V, A), S, L, a routing  $\mathcal{R}$ , integers P

Question: does there exist a  $\mathcal{R}$ -matching  $\rho$  of S in L such that there exists a

P-periodic affectation  $\mathcal{M}$  of  $\rho$ ? Optimisation goal: minimizing P.

NPA is the same problem than PRA in which is added the liberty to find the R-matching.

NPA and PRA are two general graph problems, but in our network application, some parameters are already fixed:  $\theta$  is set to 2.6ms, the period P is equal to 1ms and T(r) must be less or equal to 3ms.

In our network application, as P and the R-matching are given, we do not need to minimize P because generally P is big enough to carry the dataload. Therefore we want to optimize the time taken by the messages to do the two way trip in order to ensure a good level of quality of service.

A **2-way-trip affectation** of N is a set of doublets  $((x_0, m_0), ..., (x_{\mathcal{L}-1}, m_{\mathcal{L}-1}))$  in which  $\mathcal{M} = (m_0, ..., m_{\mathcal{L}-1})$  is a P-periodic affectation of  $N^r$ , and  $x_i$  is calculated as following:

$$x_i = (m_i + \lambda(r_i) + \theta + \omega_i) mod P$$

where  $\Omega = (\omega_0, ..., \omega_{\mathcal{L}-1})$  is *P*-periodic affectation of N. This means that  $\forall i, j, x_i \notin [x_j, ..., x_j + T]$ , to avoid collisions T is the size of the message.

Note that i design the route and that the second P-periodic affectation  $\Omega$  correspond to the waiting times of routes:  $\omega_i = \omega(r_i)$ .

Our real network problem is the following:

### Problem Periodic Assignment for Low Latency(PALL)

**Input:** matched graphs N, integer P,  $T_{max}$ .

**Question:** does there exist a 2-way-trip affectation of N, such that  $\forall r \in \rho$ ,  $T(r) \leq T_{max}$ .

Optimization goal 1: minimizing max(T(r)).

Minimizing the more expensive in time of all routes allows to ensure that any route will exceed the deadline of 3ms.

**Optimization goal 2:** minimizing  $\sum_{r \in \rho} T(r)$  (equivalent to minimizing  $\sum_{r \in \rho} \omega(r)$ .

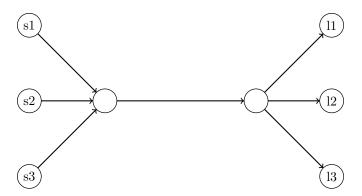
By minimizing the sum of all the routes, we allow a better global Quality of Service trough the network.

# 2 Real Network Topologies

We consider 3 kinds of topologies corresponding to the different kinds of fronthaul networks:

- 1. Topology 1: a basic network topology, composed of some base stations, represented by source nodes S, all connected to the same switch, which will be a vertex, connected himself to another vertex, corresponding to a switch, connected to some leave nodes L representing the Antennas.
- 2. Topology 2: A network containing an optical ring, such that some sources nodes may join it anywhere, not intersecting themselves before the ring, and some set of leave nodes are also leaving at any point of the ring.
- 3. Topology 3 : A "random" network, generated with some realistic properties (number of vertices, degree ...)

# 2.1 Topology 1



Example of topology 1.

In this topology, there is 3 cases:

- 1: When the weight of the links from source nodes are all equal. This is the closest case to the reality, but also the easier to solve.
- 2: When the weight of the leave nodes links are all equal .
- 3: When all the links have different weights.

The middle link can be represented without weight, because it is common to all the routes, so it can be simplified in calculations. Likewise, we do not consider  $\theta$ , the calculation time.

The first case is the easiest to solve. Indeed, it is obvious that if we schedule all messages following each others, we will obtain an optimal  $T_{max}$ : there is no waiting times on any routes. Also, all the message will cross the first switch, without collisions, then they will not be another conflict point between all routes.

This solution is optimal because the messages will not wait on source nodes, so  $\omega(r) = 0$ ,  $\forall r \in \rho$ .  $T_{max}$  is equal to the physical delay of the longer route, that we can not minimize.

Given this observation, one can try to solve a simpler problem for cases 2 and 3: is it possible to find a two way trip with all  $w_i = 0$ ?

It is always possible when the windows P is large enough (for example: send all messages one by one, waiting the previous one to be back), therefore the problem is to find the smallest windows such that it is possible.

The following example shows us that you can find a 2-way-trip affectation such that there is no waiting time, but it needs a greater time window P than a solution with waiting times.

Consider the graph:



Consider messages of 10 slots. If you search for a P-periodic affectation, without using waiting times, there are 2 choices :

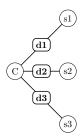
- 11 is sent before 12: the message from 11 will leave A after 11 slots.X is he minimal time slot in which 12 can emit. To avoid collisions with the message from 11 in A,12 must emit after 1 slot:  $X \ge 1$ . In the other way, the message from 11 will use B from time slot 21 to 30. The route from 12 is 12 units long from 12 to B. So, to avoid collisions with the message from 11 in B  $X + 12 \ge 31 \rightarrow x \ge 19$ . Then, by taking X = 19, the time windows is at least 38 slots in A (corresponding to the transit time of the 2 messages and the time lost between those 2 transits).
- 12 is sent before 11: symmetrically the collision the collision occurs in node A if 11 emits before time 19. The time window in B is at least 38 too.

If we allow waiting times: there is an easy scheduling using only a 3 time slots windows: Sending message 1 at time 0, message 2 at time 1, so the messages will not cross A in the same time. to be sent back, message 2 comes back immediately and is using B from slots 13 to 22, then message 1 is using B from slots 23 to 32, waiting 2 time slots at s1.

So the smaller time windows is 20, half smaller than best solution without waiting times.

TODO: temps de traje 22 contre 24 On peut utiliser l'exemple décrit en rajoutant un très long chemin de manière à ce que les waiting times ne changent pas la taille du plus long two way trip.

Let us define a simplification of a matched graph. Ignore weights on arcs starting from leaves, and find a starting order without collisions (then to find the real values, add the corresponding delay to each routes). We obtain a graph in which we have a central node, and different routes going to each sources. Such a graph is called a star. A star is a non-oriented graph in which each edge between the central node C and each source node  $s_i$  has a weight  $a_i$  (corresponding to the weight of the arc in N), where i is the number of the route in  $\rho$ . The r-matching  $\rho$  is now associating a route  $r_i$  to a source node  $s_i$ .



example of a star for a matched graph with 3 routes

We can do this simplification of a matched graph N only if we search an affectation without waiting times: in this case, any solution will be optimal, because the  $T_{max}$  will be twice the delay of the longer route. So, a solution for a star, is a solution for a matched graph, without waiting times. If waiting times are allowed, we are forced to consider the delay of the routes and this model is not available anymore.

A star affectation is a 2-way-trip affectation of the corresponding matchedgraph N in which there is no waiting time,  $\omega(r) = 0 \,\forall r \in \rho$ .

In this simplified model, we have two parameters left: the begin offsets  $o_i$  and the travel time of each routes  $2a_i$ . A route may have no collisions with the others:  $\forall$  i,j  $\in \rho$ ,

$$o_i \notin [b_i, ..., b_i + T]$$
 and  $o_i + 2a_i \notin [o_i + 2a_i, ..., o_i + 2a_i + T]$ .

T is the size of the message.

We define a new problem:

Star affectation (SA)

**Input:** a star s, integer P.

**Question:** does there exist a star affectation of s in a period P.

Optimization: minimizing P.

Solving this problem in P is finding the best solution for cases 2 and 3 of the topology 1.

We suggest the following algorithm to find a star affectation in which  $d_0$  is the shortest route, and  $d_{l-1}$  the longest one, l is the number of routes and T is the size of the messages :

### Algorithm 1 Star affectation from shortest to longest

```
Require: star s
Ensure: star affectation in P \le l * T + 2d_{l-1} - 2d_0
of fset \leftarrow 0
for all routes i in \rho from the shortest to the longest do
o_i \leftarrow offset
of fset \leftarrow offset + T
end for
```

The message using the shortest route is sent before the others, so it come back to the central node before the others. Recursively, the second message is sent before the others, and also will be back before the other messages. Because the second message is sent T slots after the first message ( $o_2 = o_1 + T$ , where T is the size of the message), it come back after the end of the message one:  $o_1 + T + 2a_1 \le o_2 + 2a_2$  with  $a_2 \ge a_1$ .

The period given by this algorithm is  $P = 2\lambda(r_{\mathcal{L}-1}) + \mathcal{L} * T - 2\lambda(r_0)$ , if routes  $\{r_0, ..., r_{\mathcal{L}-1}\}$  are ordered from the shortest to the longest.  $2\lambda(r_0)$  is the first slot in which the first message is back, and  $2\lambda(r_{\mathcal{L}-1})$  is the first slot in which the last message is back.  $\mathcal{L} * T$  is the time to transfer the messages from all routes.

We can also use a greedy algorithm, choosing the offset of each packet in turn so that there is no collision. We must chose an offset  $o_i$  for each route  $r_i$ , so that no two routes have a collision on the central link the two times they use it. We denote by  $[o_i]_P$  the set of times  $\{t \mod P \mid o_i \geq t < o_i + T\}$  which are the time used by the route  $r_i$  on the central link on its first use. There are no collision if for all  $i \neq j$   $[o_i]_P \cap [o_j]_P = \emptyset$ . Moreover no two routes must have a collision on the way back, hence we must have for all  $i \neq j$   $[o_i + a_i]_P \cap [o_j + a_j]_P = \emptyset$ .

One can prove that if P is large enough with regard to l and T, then there is always a two way trip with waiting time 0 with this algorithm. Assume  $P \geq 4lT$ , the algorithm works as follows. We cut the time windows of size P into 2l slots of size 2T and we will asign one to each route. A different slot is assigned to each route, where the offset is chosen inside the slot so that no collision on the first go in the central link is possible. If the slot chosen for the route  $r_i$  is [cT, (c+2)T], then we set  $o_i \in [cT, cT+1]$  so that  $o_i + a_i \mod P = dT$ . Assume that the offset of the first i-1 routes have been chosen in the previous way so that there is no collision. Since we have  $P \geq 4lT$ , we have at least 2l-i possible choices of offset without collision on the first use of the central link. Each of these choices uses a time slot of size T on the way back on the central link. Since there are i time slots used, and  $i \leq l$ , we have strictly more than l possible choices and less than l of them are forbidden, therefore there is a choice of offset with no collision. Since the algorithm works for all  $i \leq l$ , it finds a two way trip with waiting times 0.

TODO: On peut faire mieux que 4lT, avec une meilleure analyse ça doit juste gagner un tout pet it peu, avec un meilleur algo de placement peut-on faire 3lT ou même 2lT? Il doit être possible de montrer qu'on ne peut pas faire mieux que 2lT.

### Algorithm 2 start assignment greedy algorithm

```
Require: star s, window size p
Ensure: star affectation in P_ip
P1[p/2T] \text{ slots in first way period.}
P2[p/T] \text{ slots in back way period.}
for all route i in \rho do
for all slot j in P1 do
if P1[j] is free then
\text{shift} \leftarrow \text{T- } 2a_i \mod p
if P2[shift + 2Tj] is free then
o_i \leftarrow 2Tj + shift
end if
end if
end for
```

In this algorithm, for each route, search for a free time slot in P1, then set  $o_i \in [cT, cT+1]$  so that  $o_i + a_i \mod P = dT$ . If [dT, (d+1)T] is free, and allow the route to use the time slot [cT, cT+1] at the first use of middle link.

The following algorithm is a simplified way to find a star affectation

## Algorithm 3 start assignment greedy algorithm v2

```
Require: star s, window size p
Ensure: star affectation in Pip

P1[p] slots in first way period.

P2[p] slots in back way period.

for all route i in \rho do

for all slot j in P1 do

if [j, j + T]_P is free in P1 then

if [j + 2r_j, j + 2r_j + T]_P is free in P2 then

o_i \leftarrow j

end if

end if

end for
end for
```

In this algorithm, for each route, search for the first free slot [j,j+T] in P1 such that the slot  $[j+2r_j,j+2r_j+T]_P$  in P2 is free too, then give the route the offset j(shifted by the distance from the leave to the first node).

Now, the waiting times are allowed, and try to solve PALL on topologie 1. The following heuristic is suggested :

### Algorithm 4 Longest to shortest with waiting times

```
Require: Matched graph N, window size P, packed size T
Ensure: 2way Trip affectation in P
  offset \leftarrow 0
  for all route i in \rho from the longest to the shortest do
     m_i \leftarrow \text{offset}
     offset \leftarrow offset + T
  end for
  offset \leftarrow 0
  take i such that r_i is the first route to come back in sources node
  w_i \leftarrow \text{offset};
  offset \leftarrow a_i + T
  while there is a route which has no w_i do
     take i such that r_i is the longest route able to come back in source node
     before the last one has totally passed
     w_i \leftarrow \text{offset - } a_i;
     offset \leftarrow offset + T
  end while
```

The running of this algorithm is the following one:

- 1 On leaves node, the messages are scheduled so that they are following each others, from the longest route to the shortest route.
- 2 On sources node, let the message able to come back first pass.
- 3 Once this message has passed, look at all the eligibles message able to be back at the sources node during the transit of the previous one.
- 4 Let the eligible message on the longest route transit in the node first after the previous one, possibly delayed by a little waiting time to avoid crossing.
- 5 delay the others with a waiting time.
- 6 while all the messages are not scheduled, get back to 3

This algorithm gives us solution with the minimum period possible :  $\mathcal{L}$  x T. Indeed, all the messages are scheduled following each others on conflict node : there is no wasted time between the transit of two messages.

For the case 1, this algorithm is similar to the obvious way to schedule the messages mentioned above. It gives optimal solution without increase the calculation time.

For the case 2, this algorithm gives optimal solutions: Indeed, in the manner we order messages, from leaves toward sources, the messages are sent from the one using the longest route to the one using the shortest one. Number the routes in this order from 1 to  $\mathcal{L} - \infty$ .

We have  $\forall i < j$ ,  $r_i < r_j$ . The first message leaves the source switch at time  $2r_1 + T$ , the second message, can use it at the time  $2r_2 + T$ . The second message waited  $2r_1 + T - 2r_2 + T = 2(r_1 + r_2)$  to use the middle link. So, the second message leave the source switch at time  $2r_2 + T + 2(r_1 + r_2) = 2r_1 + T$ . Recursively, all messages will take  $2r_1 + T$  slots for the Two Way Trip. Consequently, for this case, the heuristic is optimal, because the longer route has no waiting time and no TwoWayTrip is longer than the longer route one.

For the case 3, after some simulations on values corresponding to reality  $(P \simeq 19500, T \simeq 2500, l \simeq 2000)$ , good solutions are obtained.

The average of solutions given by this heuristics gives solutions close to the average of  $2 * r_m ax$ .

TODO: interessant de dire que la pire solution correspond souvent a celle trouvé par l'algo? (sur ce genre de "pire" solutions, on ne peux pas faire mieux. TODO: mettre l'exemple ou ca marche mal alors qu'en espacant un peu ca rentre tres bien

# 3 Complexity results

The conflict depth of a  $\mathcal{R}$ -matching is the maximum number of arcs in a route which are shared with other routes. In other words it is the maximal number of potential conflicts along one route. The load of a  $\mathcal{R}$ -matching is the maximal number of routes which share the same edge. It is clear that a P-peridodic affectation must satisfy that P is larger or equal to the load.

We give two alternate proofs that PRA is NP-complete. The first one works for conflict depth 2 and is minimimal in this regards since we later prove that for conflict depth one, it is easy to solve PRA. The second one reduces the problem to graph coloring and implies inapproximability. In all this section the reservation  $\delta$  is assumed to be 0.

TODO: ajouter des références à des problèmes de réseau similaires et de transport en commun qui sont également NP complets.

We define two characteristic graphs associated to a  $\mathcal{R}$ -matching.

**Definition 1** (Edge Conflict Graph). The Edge Conflict Graph (ECG) of a  $\mathcal{R}$ -matching of a digraph G = (V, A) is the graph G' = (V', E') such that a vertex of V' is an arc of A with a load greater than one. There is an edge between two vertices of V' if and only if there is a route which contains the two arcs of A corresponding to the two vertices.

This graph can be oriented so that an edge is oriented as the path in a route it represents and we can associate as weight to this edge, the delay of the corresponding path.

**Definition 2** (Route Conflict Graph). The Route Conflict Graph (RCG) of a  $\mathbb{R}$ -matching of a graph G = (V, E) is the graph G' = (V', E') where the vertices of V' are the routes of the  $\mathbb{R}$ -matching and there is an edges between two vertices corresponding to two routes if and only if they share an edge.

We fix an arbitrary ordering on the elements of  $V' = \{v_1, \ldots, v_n\}$ . We can associate a weight to each edge  $(v_i, v_j)$ . Assume that i < j, and that the routes  $r_i$  and  $r_j$  corresponding to  $v_i$  and  $v_j$  have u as first common vertex. Then the weight of  $(v_i, v_j)$  is  $\lambda(u, r_i) - \lambda(u, r_j)$ .

TODO: Dire comment une solution à PRA s'exprime dans ces graphes avant de faire les preuves de NP-complétude

**Proposition 1.** Problem PRA is NP-complete, for a routing with conflict depth two.

Proof. Let G=(V,E) be a graph and d its maximal degree, we want to determine whether it is edge colorable with d or d+1 colors. We reduce this edge coloring problem to PRA: we define a graph H, for each  $v \in V$  there are two vertices connected by an edge  $(v_1,v_2)$  and none of these edges are incident. Those edges are of weight 1. The schedule  $\rho$  of H is the set of routes  $s_{u,v}, u_1, u_2, v_1, v_2, l_{u,v}$  for each edge (u,v) in G. The weight of the two new edges is d(d+1)-1. The parameter  $\delta$  is set to 0.

Let  $\phi$  be an edge coloring with k colors of G. We can build a k periodic schedule by assigning to the source  $s_{u,v}$  of each route an offset of  $\phi(s_{u,v})$ . Indeed, if two routes  $r_1$  and  $r_2$  share the same edge, say  $(v_1,v_2)$  then they represent two edges  $e_1$  and  $e_2$  of G incident to the vertex v. Therefore  $\lambda(v_1,r_1)=\phi(e_1)$  mod k because the delays of the edges before  $v_1$  sum to d(d+1) or 2(d(d+1)) which are equal to 0 modulo k since k=d or k=d+1. Thus  $\lambda(v_1,r_1)-\lambda(v_1,r_2)$  mod  $k=\phi(e_1)-\phi(e_2)$  mod k and  $\lambda(v_1,r_1)-\lambda(v_1,r_2)$  mod k>0 since  $\phi(e_1)\neq\phi(e_2)$ .

Now consider a k-periodic affectation of  $\rho$ . For each (u, v) in G, we define  $\phi(u, v)$  to be the offset of the route beginning at  $s_{u,v}$ . For the same reasons as in the last paragraph,  $\phi$  is an edge coloring with k colors. Therefore we have reduced edge coloring which is NP-hard [2] to PRA which concludes the proof.

**Theorem 1.** Problem PRA cannot be approximate within a factor  $n^{1-o(1)}$  unless P = NP even when the load is two and n is the number of vertices.

*Proof.* We reduce PRA to graph coloring. Let G be a graph instance of the k-coloring problem. We define H in the following way: for each vertex v in G, there is a route  $r_v$  in H. Two routes  $r_v$  and  $r_u$  share an edge if and only if (u,v) is an edge in G and this edge is only in this two routes. We put weight inbetween shared edges in a route so that there is a delay k between two such edges.

As in the previous proof, a k-coloring of G gives a k-periodic schedule of H and conversly. Therefore if we can approximate the value of PRA within a factor f, we could approximate the minimal number of colors needed to color a graph within a fator f, by doing the previous reduction for all possible k. The proof follows from the hardness of approximability of finding a minimal coloring [4].

In particular, this reduction shows that even with small maximal load, the minimal period can be large.

TODO: By using ideas similar to Vizing theorem, we may prove the following theorem for graph of congestion depth 2. In the proof there is just an analog of the lemma used to prove Vizing theorem, we should try to complete the proof. Also can we say something similar when the congestion depth is larger?

**Proposition 2.** The solution to PRA is either the load or the load plus one. Moreover, a solution of load plus one can be built in polynomial time.

*Proof.* First we define an alerning path and how it characterizes an optimal solution of PRA. Let v be a vertex of degree d in the congestion graph (to be defined). We assume that the coloring of the edges is optimal (to define in the case of a congestion graph). The color  $\alpha$  is not used in it neighborood. For every edge of color  $\beta$  from this vertex we build an  $\alpha - \beta$  path, that is a maximal path  $u_0, u_1, \ldots, u_l$  so that  $(u_0, u_1)$  is of color  $\beta, \beta + \lambda(u_0, u_1)$ , then  $(u_1, u_2)$  is of color  $\alpha + \lambda(u_0, u_1), \alpha + \lambda(u_0, u_2)$  and so on (changer  $\lambda$  car on ne suit pas des routes).

A maximal  $\alpha - \beta$  path must be a cycle or the coloring is not minimal.

As a corollary of the previous proposition, NPA is NP-hard since checking a potential feasible solution for NPA implies to solve PRA. Also, PALL is NP-Hard because solving it implies to slove PRA in both ways with a given P (1ms).

TODO: we may try to find where those problems are in the polynomial hierarchy, in the second level?

# 4 Coloring and P-periodic affectation

To the route conflict graph, we can associate a system of inequations representing the constraints that the *P*-periodic affectation must satisfy.

**Definition 3** (Conflict system). Let G be a weighted RCG, we associate to each vertex  $v_i$  of G the variable  $x_i$ . The conflict system is the set of inequations  $x_i \neq x_j + w(e)$  for i < j and  $e = (v_i, v_j)$  edge of G.

It is simple to see that the conflict system of a  $\mathcal{R}$ -matching has a solution modulo P if and only if the  $\mathcal{R}$ -matching has a P-periodic affectation. This kind of system can be solved by SAT solver or CSP solver and those two methods will be investigated on practical instances. TODO: Donner la formulation pour ces solvers et donner le résultat d'expérience dans une partie indépendante. Les comparer avec un solveur de notre cru avec quelques heuristiques simples.

**Definition 4** (Additive coloring). Le G be a weighted graph, we say that G has an additive coloring with p colors if and only if its associated conflict system has a solution over  $\mathbb{Z}/p\mathbb{Z}$ . The minimal p for which the system has a solution is the additive chromatic number of G denoted by  $\chi_+(G)$ .

Finding an optimal additive coloring is thus the same as solving our problem of finding a *P*-periodic affectation and is at least as hard as finding an optimal coloring. Because of the constraints on the edges, the additive chromatic number of a graph may be much larger than its chromatic number. When the graph is a clique both additive and regular chromatic number may be the size of the graph. However, we will now prove that for bipartite graphs the chromatic number is two while its additive chromatic number is arbitrary large.

We have the following values obtained by exhaustive search.

TODO: généraliser chi+ aux graphes sans poids en faisant un max?

**Fact 1.** The values we list here are the maximal ones we obtain by weighting bipartite graphs.

1. 
$$\chi_{+}(K_{2,2}) = \chi_{+}(K_{3,3}) = 3$$
 with weight 0, 1  
2.  $\chi_{+}(K_{3,4}) = \chi_{+}(K_{3,5}) = 3$   
3.  $\chi_{+}(K_{3,6}) = 4$   
4.  $\chi_{+}(K_{4,4}) = \chi_{+}(K_{4,5}) = \chi_{+}(K_{4,6}) = 4$   
5.  $\chi_{+}(K_{5,5}) = ?$ 

A nice theoretical question would be to find the way to put weights on a bipartite graph so that its additive chromatic number is maximal. My conjecture is that a  $K_{l,l}$  can have an additive chromatic number in O(l). The question is also interesting when we restrict the weights to 0, 1.

**Theorem 2.** There is a weighting of 
$$K_{l^2,\binom{l}{l^2}}$$
 such that  $\chi_+(K_{l^2,\binom{l}{l^2}}) > l$ .

Proof. Let  $V_1, V_2$  be the bipartition of the graph we build and let  $|V_1| = l^2$ . For all  $S \subseteq V_1$  with |S| = l, we denote by  $v_1, \ldots, v_l$  its elements, there is a single element  $v_S$  in  $V_2$  which is connected to exactly the elements of S and such that the weight of  $v_S, v_i$  is i-1. Because of this construction,  $V_2$  is of size  $\binom{l}{l^2}$ . Moreover for any additive coloring of the graph we have constructed, a set of l elements in  $V_1$  cannot have all the same color. But by the extended pigeon principle, since there are  $l^2$  elements in  $V_1$  at least l amongst them must have the same color. This prove that the graph we have built cannot have an additive coloring with l colors.

The theorem can be improved so that the number of colors is logarithmic in the size of the bipartite graph.

**Theorem 3.** There is a weighting of 
$$K_{l^2,\binom{l}{l^2}}$$
 such that  $\chi_+(K_{l^2,\binom{l}{l^2}}) > l$ .

*Proof.* Let  $\mathcal{F}$  be a family of perfect hash functions from  $[l^2]$  to [l]. It means that for any subset S of size l of  $[l^2]$ , there is an hash function  $f \in \mathcal{F}$  such that  $f_S$  is injective. The construction of the bipartite graph is similar to the previous proof. Let  $V_1, V_2$  be the bipartition of the graph we build and let

 $V_1 = \{v_1, \ldots, v_{l^2}\}$ . For each function  $f \in \mathcal{F}$  there is a vertex  $v_f \in V_2$  and the weight of  $(v_i, v_f)$  is f(i). By [3, 1] there is a family of perfect hash functions of size  $2^{O(l)}$  therefore  $V_2$  is of size  $2^{O(l)}$ . Again by using the pigeon principle and the perfect property of the family of functions, we prove that no additive coloring with l colors is possible.

TODO: explain the removal of low degree vertices = kernelization + heuristic on the degree Also implement it in the solver

TODO: Comprendre la valeur sur d'autre familles de graphe, notemment celles qu'on peut rencontrer en pratique, par exemple petite tree width

# 5 Special cases study

After proving the complexity of problem PRA and NPA in the general case, we will now study special cases of problem PRA where its complexity is polynomial. First we define the notion of coherent routing.

As a consequence, for each node u in V, the subgraph of G induced by all the routes from u to all the other nodes in G is a tree. In the following, we only deal with coherent routing functions.

# 5.1 The disjoint paths PRA problem: DP-PRA

In this restriction of PRA, we are given a  $\mathcal{R}$ -matching with the following properties:

- 1. There are no common arcs for routes originating from different sources
- 2. The routing  $\mathcal{R}$  is coherent

**Proposition 3.** Problem DP-PRA can be solved in linear time according to the size of A.

The first property of the DP-PRA problem ensures that an arc cannot belong to two routes in  $\mathcal{R}$  originating from different sources in S. The second property ensures that if two routes originate from the same source x, they share the same arcs from x to a given vertex y and cannot share an arc after. This means that  $\forall (u,v) \in A$ , the arcs (u,v) with the highest load  $l_{max} = max(load(u,v))$  are arcs  $a_0^j$  sharing a common origin  $u_0 \in S$  and a P-periodic affectation for those arcs is a P-periodic affectation for all the arcs in A.

In a P-periodic affectation consists in a time schedule where routes on a same arc must be separated by a delay that is strictly superior to an integer  $\delta \geq 0$ . As a consequence, the minimum size of the period P is equal to  $l_{max} \times (\delta + 1)$ . This means that that on the arc with the highest load, we can schedule the first route at the moment  $m_k = 0$ , the second route at a moment  $m_{k'} = \delta + 1$  and so on for all  $l_{max}$  routes.

## 5.2 The disjoint paths NPA-Problem : DP-NPA

In this subsection we study the NPA problem where in the graph induced by the the routing function  $\mathcal{R}$ :

- There are no common arcs for routes originating from different sources
- The routing  $\mathcal{R}$  is coherent

**Proposition 4.** Problem DP-NPA can be solved in linear time according to the size of A.

The minimum size of P is obtained by minimizing the highest load of an arc  $a_0^j$  for any route  $r^j \in \rho$ . As all source vertices in S are connected to all sources vertices in L by a route in  $\mathcal{R}$ , a simple load balancing allows to obtain a maximum load equal to  $\max_{load} = \lceil \frac{\mathcal{L}}{|S|} \rceil$ . Once the  $\mathcal{R}$ -matching is computed, we face the DP-PNA problem and the minimum value of P is thus equal to  $\max_{load} \times (\delta + 1)$ .

### 5.3 The disjoint paths NPAC-Problem : DP-NPAC

In this problem, there is a delay constraint that must be satisfied by a  $\mathcal{R}$ -matching. This means that we must remove the routes in  $r' \in \mathcal{R}$  where  $\lambda(r') > K$ .

**Proposition 5.** Problem DP-NPAC can be solved in polynomial time according to the size of V.

In a similar way to problem DP-PRA, the arcs with the highest load can only be the arcs  $a_0^j$  sharing a common origin  $u_i \in S$ . In order to minimize the size of the period P, we have to find a  $\mathcal{R}$ -matching such that the maximum number  $k_i$  of routes originating from a same source  $u_i \in S$  is minimal. Having found this minimal value k, we face a problem equivalent to the DP-PRA problem.

In order to find a matching of vertices in S with vertices in L such that the maximum number of vertices in L assigned to a vertex  $s_i \in S$  is inferior or equal to k, we will transform our problem in a flow problem.

### 5.3.1 Construction of a flow graph

Let us consider an instance  $I = (G = (V, A), S, L, \mathcal{R}, \mathcal{P}, \delta, \mathcal{K} \text{ and } M)$  of NPA where the routes in  $\mathcal{R}$  are only those that respect the delay constraint K. We first construct a complete bipartite graph G' = (V', A') where V' is made of two sets of vertices: vertices  $V'_1$  corresponding to S and  $V'_2$  corresponding to L. A' is made of all possible arcs from vertices  $v'_1 \in V'_1$  to vertices  $v'_2 \in V'_2$ , with a capacity 1. We then add to V' a source node S' and a sink node T'. Finally we add to A' all the arcs from S' to each vertex  $v'_1 \in V'_1$ , with a capacity K and all the arcs from each vertex  $K'_1 \in K'_2$  to  $K'_1 \in K'_2$ . We have thus obtained a flow graph  $K'_1 \in K'_2$  whose size is polynomial in

#### includegraphics[scale=0.3]DP-NPA.pdf

Figure 1: Reduction of DPA-NPA into a flow problem

regards to the size of G. We will now compute the maximum flow in G' in order to determine if its size is at least  $\mathcal{L}$ , in order to be able to connect all the leaves.

We can compute the size of a maximum flow in G' in a polynomial time using a generic flow algorithm as Ford-Fulkerson (it will terminate as arcs capacity are rational numbers). In order to minimize the objective k, we can begin with a value k=1 and use a dichotomic approach to find the minimal value of k for which a maximal flow of size  $\mathcal{L}$  exists in G'. The maximum value of k is M. If k=M and the maximum flow value in  $G'<\mathcal{L}$ , then there is no valid  $\mathcal{R}$ -matching of S in L.

The complexity of minimizing k is thus  $0(m^2n \times log_2n)$  where m = |A| and n = |V|. We obtain in the end a  $\mathcal{R}$ -matching minimizing the maximal number of routes originating from a single source  $u_s \in S$  and we are faced with the DP-PRA problem for this instance.

# 5.4 Intersecting paths problems for PRA

Next we study more general cases of PRA where routes originating from different sources can intersect. Let us consider an instance  $I = (G, S, L, \mathcal{R}, \rho, \mathcal{P}, \delta)$  of PRA. We first define the collision induced graph of a  $\mathcal{R}$ -matching in G = (V, A).

### 5.4.1 The 1-arc collision PRA problem

In this restriction of PRA, we are given a  $\mathcal{R}$ -matching with the following properties:

- 1. There is a bottleneck: a single arc  $a_b = (u_b, u_b')$  for which  $load(a_b) \ge 1$  if the routes in  $\rho(a_b)$  come from different sources in S
- 2. The routing  $\mathcal{R}$  is coherent

**Proposition 6.** Problem 1-arc collision PRA can be solved in linear time according to the size of A.

In order to have a P-periodic affectation, the period P must be big enough for all arcs in  $\rho(a_b)$ , thus the minimal size of P is  $P = load(a_b) \times (\delta + 1)$ . Moreover, this is also the minimum solution for PRA. If we consider a scheduling  $m_i \in \mathcal{M}'$  of each route  $r_i \in \rho(a_b)$  such that there is a P-periodic affectation of this routes on the arc  $a_b$  with a size of  $P = load(a_b) \times (\delta + 1)$ , the schedule  $m_j$  in  $\mathcal{M}$  corresponding to  $r_i$  will be  $m_j = m_i - \lambda(u_b) mod P$ . As the routing is coherent, no two routes originating from a same source can use different routes to attain  $a_b$  thus the P-periodic affectation is valid for any arc  $(u,v) \in A$  if it is valid on  $a_b$ .

### 5.5 The Data-center model

In this subsection we assume that we have a few datacenters which creates a few edges of high load in the routing graph. We further assume that to routes can share at most two edges with others routes, the first one with multiple routes at the exit of the datacenter and thes econd one with a single other route.

This model is a special case of conflict depth two and generalizes the disjoint path case. Its route conflict graph is very simple and better heuristics should work. Can we prove this case hard or easy? TODO: Investigate this case and others arising from true data. For trees the problem is simple, what notion measure the proximity of a DAG to tree? We could try the tree width ot the graph seen as not oriented, but it does not seem really adapted.

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