



Weighted Acyclic Di-Graph Partitioning by Balanced Disjoint Paths

H. Murat AFSAR – Olivier BRIANT

Murat.Afsar@g-scop.inpg.fr

Olivier.Briant@g-scop.inpg.fr

G-SCOP Laboratory
Grenoble Institute of Technology



Presentation Plan

Introduction

Problem Definition

Solution Method

Numerical Results

Conclusions and Perspectives



Partitioning Problems - Disjoint Paths

- ▶ Covering all the nodes
- ▶ Each node appears on just one path

Polynomially solvable - Flow



Problem Context

- ▶ Telecommunication networks
Noninterfering messages
- ▶ Transportation
Aircraft Rotation Problems



Problem Definition

- ▶ A directed acyclic graph $G(V, A)$,
- ▶ Find K disjoint and weight balanced paths p_1, p_2, \dots, p_K such that
Each node V is covered by exactly one path.



Difficulty

► Objective : $\min \sum_{k=1}^K c_{p_k}$

Where

► $c_{p_k} = g\left(\sum_{v \in p_k} w_v\right) = \frac{1}{M} \left(M - \sum_{v \in p_k} w_v\right)^2$

► And M is the ideal path weight

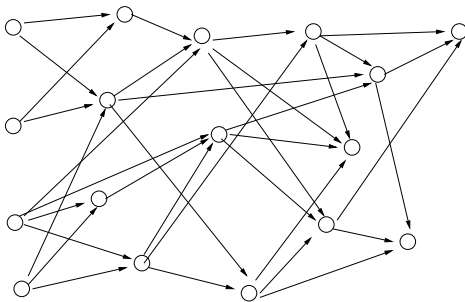
$$M = \frac{1}{K} \sum_{v \in V} w_v$$



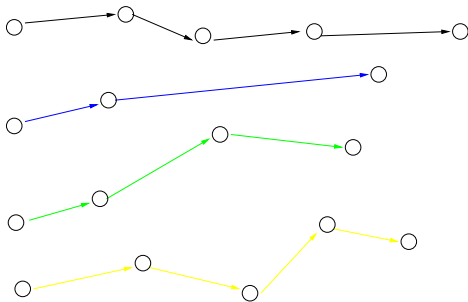
Similar Problems

- ▶ Finding 2 node disjoint paths with bounded weight on a *DAG* NP-complete
- ▶ Vertex covering by minimum weight paths, on trees NP-Hard (even in unweighted case)
- ▶ Partitioning problem NP-hard

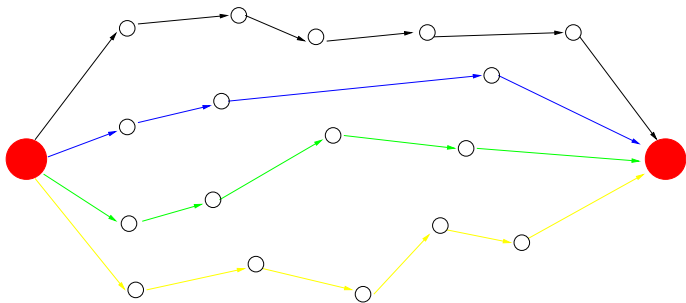
Partitoning Graph by Disjoint Paths



Partitoning Graph by Disjoint Paths



Partitoning Graph by Disjoint Paths



Problem Formulation

$$\min \sum_{v \in V \setminus \{s, t\}} \frac{(My_{(s,v)} - x_{(s,v)})^2}{M}$$

subject to

$$\sum_{(v,v') \in \delta^-(v')} y_{v,v'} - \sum_{(v',v) \in \delta^+(v)} y_{v',v} = 0 \quad \forall v' \in V \setminus \{s, t\}$$

$$\sum_{(v,v') \in \delta^-(v')} x_{v,v'} - \sum_{(v',v) \in \delta^+(v')} x_{v',v} = w_{v'} \quad \forall v' \in V \setminus \{s, t\}$$

$$\sum_{v' \in \delta^+(v)} y_{v,v'} = 1 \quad \forall v \in V \setminus \{s, t\}$$

$$\sum_{(s,v') \in \delta^+(s)} y_{s,v'} = K$$

$$x_{v,v'} \leq W_v y_{v,v'} \quad \forall (v, v') \in A$$

$$x_{v,v'} \in \mathbb{R} \quad \forall (v, v') \in A \quad y_{v,v'} \in \{0, 1\} \quad \forall (v, v') \in A$$



Decomposition Dantzig-Wolfe

- ▶ Master Problem : Combining paths to partition the graph
- ▶ Slave Problem : Finding *good* paths



Master Problem

$$\min \sum_{p \in P} c_p \lambda_p$$

subject to

$$\sum_{p \in P} \lambda_p = K$$

$$\sum_{p \in P | v \in p} \lambda_p = 1 \quad \forall v \in V$$

$$\lambda_p \in \{0, 1\} \quad \forall p \in P$$



Reduced Cost of a Path

$$c_p = g\left(\sum_{v|v \in p} w_v\right)$$

$$\bar{c}_p(u) = g\left(\sum_{v|v \in p} w_v\right) - \sum_{v|v \in p} u_v - u_0$$

Where u_v and u_0 are the dual variables associated to the node v and to the first constraint, respectively



Slave Problem Solution Methods

- ▶ Max Flow - Max Cost to cover all the nodes
- ▶ 2OPT on flow solution to minimize the minimum reduced cost :
Finding a feasible solution with promising paths.

- ▶ Optimal dynamic programming :

A multi label setting dynamic model to find $\max \sum_{v \in S \rightarrow v'} u_v$ for

each node v' for each total weight $\sum_{v \in S \rightarrow v'} w_v = W_{S \rightarrow v'}$



Max Flow - Max Cost

- ▶ Flow 1 on graph $G(V, A)$: $cost(a) = \sum_{p|a \in p} \lambda_p \quad \forall a \in A$
- ▶ Flow 2 on graph $G'(V, A')$: With only a limited number of exiting arcs from all node v such that

$$\sum_{p|a \in p} \lambda_p \geq 0.5$$

- ▶ For the paths p, p' in the solution S , for every couple of nodes $v \in p$ and $v' \in p'$ compute

$$\bar{c}_p^{new} = g(W_{s \rightarrow v} + W_{v' \rightarrow t}) - (U_{s \rightarrow v} + U_{v' \rightarrow t})$$

$$\bar{c}_{p'}^{new} = g(W_{s \rightarrow v'} + W_{v \rightarrow t}) - (U_{s \rightarrow v'} + U_{v \rightarrow t})$$

Where $U_{s \rightarrow v}$ is the sum of dual values of the nodes on the path between s and v

- ▶ If $\min\{\bar{c}_p^{new}, \bar{c}_{p'}^{new}\} < \min\{\bar{c}_p, \bar{c}_{p'}\}$ then accept the swap



Dynamic programming

$$f(v, W) = u_v + \max_{v' \in \delta^-(v)} f(v', W - w_v)$$

And naturally

$$f(s, 0) = 0$$

Where W is the total weight of the path between s and v



Dynamic programming

$$f(v, W) = u_v + \max_{v' \in \delta^-(v)} f(v', W - w_v)$$

And naturally

$$f(s, 0) = 0$$

Where W is the total weight of the path between s and v

$$\min_{\substack{p \in P(v) \\ \sum_{v \in p} w_v = W}} \bar{c}_p(u) = g(W) - f(v, W)$$

Where $P(v)$ is the set of paths ending at the node v



Lagrangian Lower Bound

$$\tilde{c}_p(u) = g\left(\sum_{v|v \in p} w_v\right) - \sum_{v|v \in p} u_v$$



Lagrangian Lower Bound

$$\tilde{c}_p(u) = g\left(\sum_{v|v \in p} w_v\right) - \sum_{v|v \in p} u_v$$

$$\theta_1^*(u) = K \times \min_{p \in P} \tilde{c}_p(u) + \sum_{v \in V} u_v$$

$$\theta_2^*(u) = \sum_{i \in \mathcal{K}} \min_{p \in P(v_i)} \tilde{c}_p(u) + \sum_{v \in V} u_v$$

Where $P(v_i)$ is the set of paths ending with the node v_i
 So each path ends at a different node

Obviously $\theta_2^*(u) \geq \theta_1^*(u)$



Instance Characteristics

Table: General DAG with 700 nodes and $K = 80$

A_0.05	5%	$w_{min} = 40$	$w_{max} = 120$
A_0.10	10%	$w_{min} = 40$	$w_{max} = 120$
A_0.15	15%	$w_{min} = 40$	$w_{max} = 120$
A_0.20	20%	$w_{min} = 40$	$w_{max} = 120$
A_0.25	25%	$w_{min} = 40$	$w_{max} = 120$
B_0.05	5%	$w_{min} = 1$	$w_{max} = 1$
B_0.10	10%	$w_{min} = 1$	$w_{max} = 1$
B_0.15	15%	$w_{min} = 1$	$w_{max} = 1$
B_0.20	20%	$w_{min} = 1$	$w_{max} = 1$
B_0.25	25%	$w_{min} = 1$	$w_{max} = 1$

Instance Characteristics

Table: Flight-like nodes with 700 nodes, $w_{min} = 40$, $w_{max} = 120$, $K = 80$

F_40_80	$wait_{min} = 40$	$wait_{max} = 80$	nb arcs : 14 405
F_40_100	$wait_{min} = 40$	$wait_{max} = 100$	nb arcs : 20 468
F_40_120	$wait_{min} = 40$	$wait_{max} = 120$	nb arcs : 26 887
F_40_150	$wait_{min} = 40$	$wait_{max} = 150$	nb arcs : 36 377
F_40_240	$wait_{min} = 40$	$wait_{max} = 240$	nb arcs : 60 791



RMP Linear Relaxation vs Lagrangean Relaxation

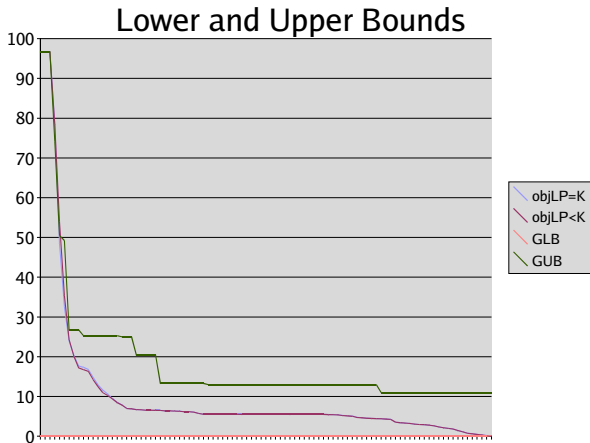
Restricted Master Linear Relaxation 1 :

$$\sum_{p \in P} \lambda_p = K$$

Restricted Master Linear Relaxation 2 :

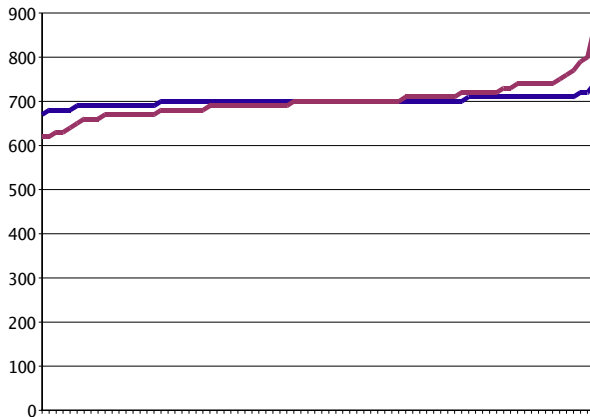
$$\sum_{p \in P} \lambda_p \leq K$$

G-SCOP RMP Linear Relaxation vs Lagrangean Relaxation



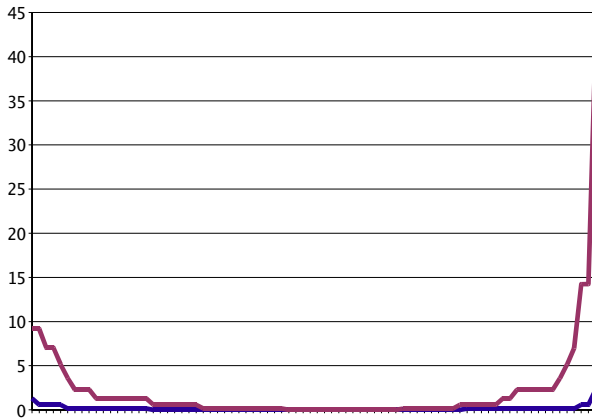
G·SCOP RMP Linear Relaxation vs Lagrangean Relaxation

Initial and Final Solutions



G·SCOP RMP Linear Relaxation vs Lagrangean Relaxation

Initial and Final Solution Costs





Numerical Results

Instance	iter	Z_{LP1}^*	Z_{LP2}^*	GLB	GUB	Columns
A_0.05	95	0,141	0,070	0,141	10,996	9649
A_0.10	18	0,141	0,002	0,141	7,282	4913
A_0.15	11	0,141	0,002	0,141	3,283	3216
A_0.20	7	0,141	0,096	0,141	0,712	1918
A_0.25	6	0,141	0,106	0,141	0,427	1467
B_0.05	34	1,714	0,555	1,714	1,714	4648
B_0.10	6	1,714	0,940	1,714	1,714	1484
B_0.15	6	1,714	0,892	1,714	1,714	1586
B_0.20	6	1,714	0,937	1,714	1,714	1552
B_0.25	5	1,714	1,714	1,714	1,714	1217
F_40_80	6	2,945	2,945	2,945	2,945	1639
F_40_100	6	2,624	1,665	2,624	2,624	1839
F_40_120	5	2,689	2,689	2,689	2,689	1195
F_40_150	5	2,120	2,120	2,120	2,120	1064
F_40_240	5	1,946	1,946	1,946	1,946	1102



Total CPU Time (in sec.) per Method

Instance	Total	Master	Slave	Flow	2OPT
A_0.05	473	265	56	8	133
A_0.10	153	27	19	9	85
A_0.15	176	24	13	4	109
A_0.20	161	13	10	5	96
A_0.25	161	8	9	3	91
B_0.05	63	15	1	2	45
B_0.10	53	5	0	2	46
B_0.15	152	5	0	3	143
B_0.20	184	5	3	4	170
B_0.25	172	2	0	5	162
F_40_80	210	9	0	2	199
F_40_100	253	8	1	1	242
F_40_120	262	4	2	1	255
F_40_150	169	2	2	4	160
F_40_240	290	2	2	3	282



Conclusions

- ▶ Gap between lower and upper bounds when the graph is sparse
- ▶ Optimality proof of heuristic approach when the graph is dense
- ▶ 2-OPT and Master problem are dominating the resolution time



GSCOP Perspectives

- ▶ More tests to come
- ▶ Improving the lower bound
- ▶ Branch & Price for the sparse graphs to show the optimality



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