

Advances in Forestry Control &
Automation Systems in Europe

INTEGRATION OF OR AND MPC TECHNIQUES TO THE BIOMASS SUPPLY CHAIN FOR ENERGY PRODUCTION

Tatiana M. Pinho^{1,2}, Bruno Oliveira²,
Dmitry Podpokaev³, Jussi Rasinmäki³,
Alexandra Marques², J. Boaventura-Cunha¹

1 UTAD, Portugal

2 InescTec, Portugal

3 Simosol, Finland

alexandra.s.marques@inesctec.pt

SSAFR 2015 | UPPSALA, SWEDEN | 19-21 AUGUST 2015

This project has received funding from the European Union Seventh
Framework Programme (FP7/2007-2013) under grant agreement n° [604286].



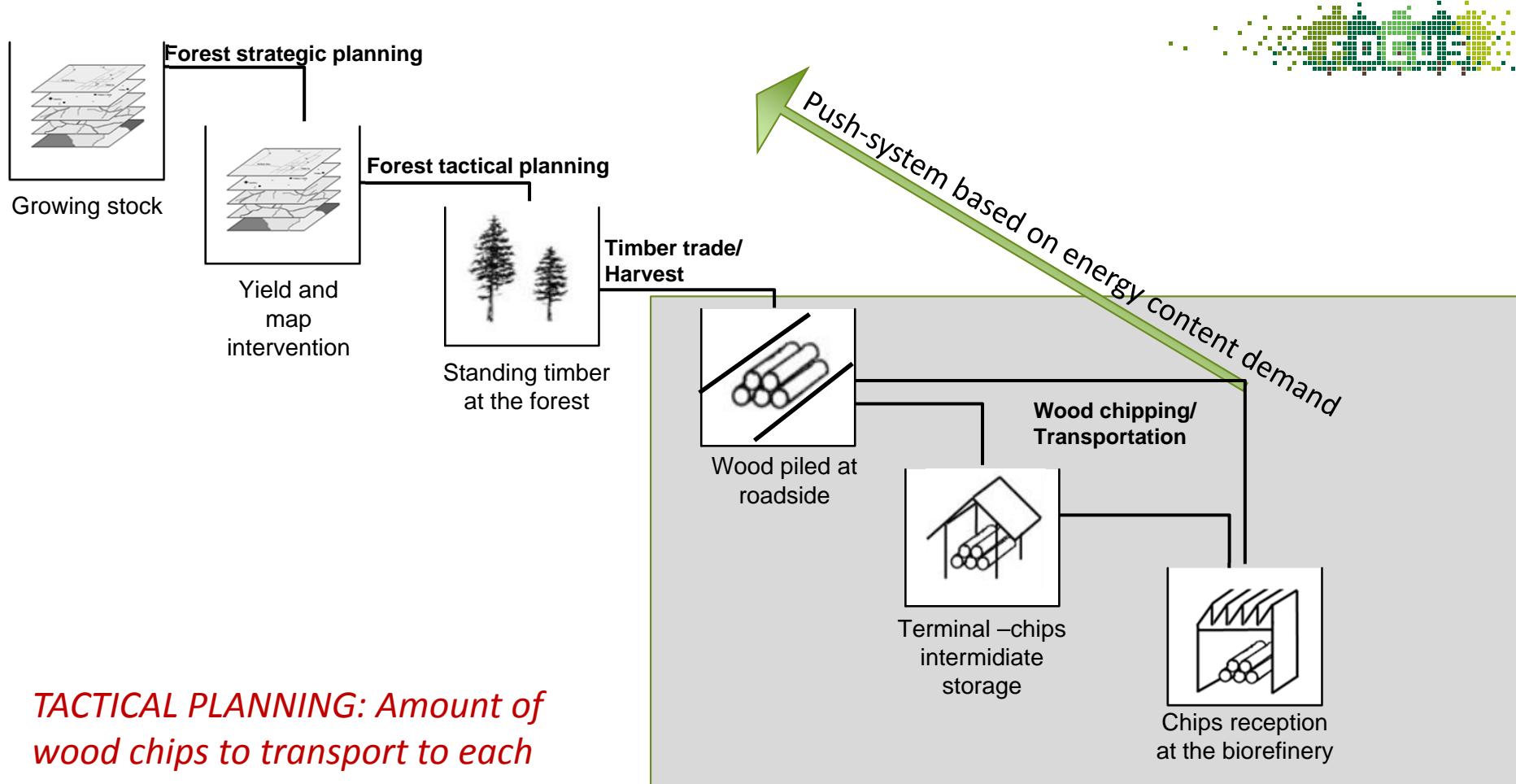
Agenda



Advances in Forestry Control &
Automation Systems in Europe

1. Problem definition
2. Solution method
 1. MIP model
 2. Matheuristic
 3. “Control” level
 4. Feedback loop
3. Preliminary computational results
4. Concluding remarks
5. References

Wood chips delivery planning problem



TACTICAL PLANNING: Amount of wood chips to transport to each location considering its moisture content (flow variables)?

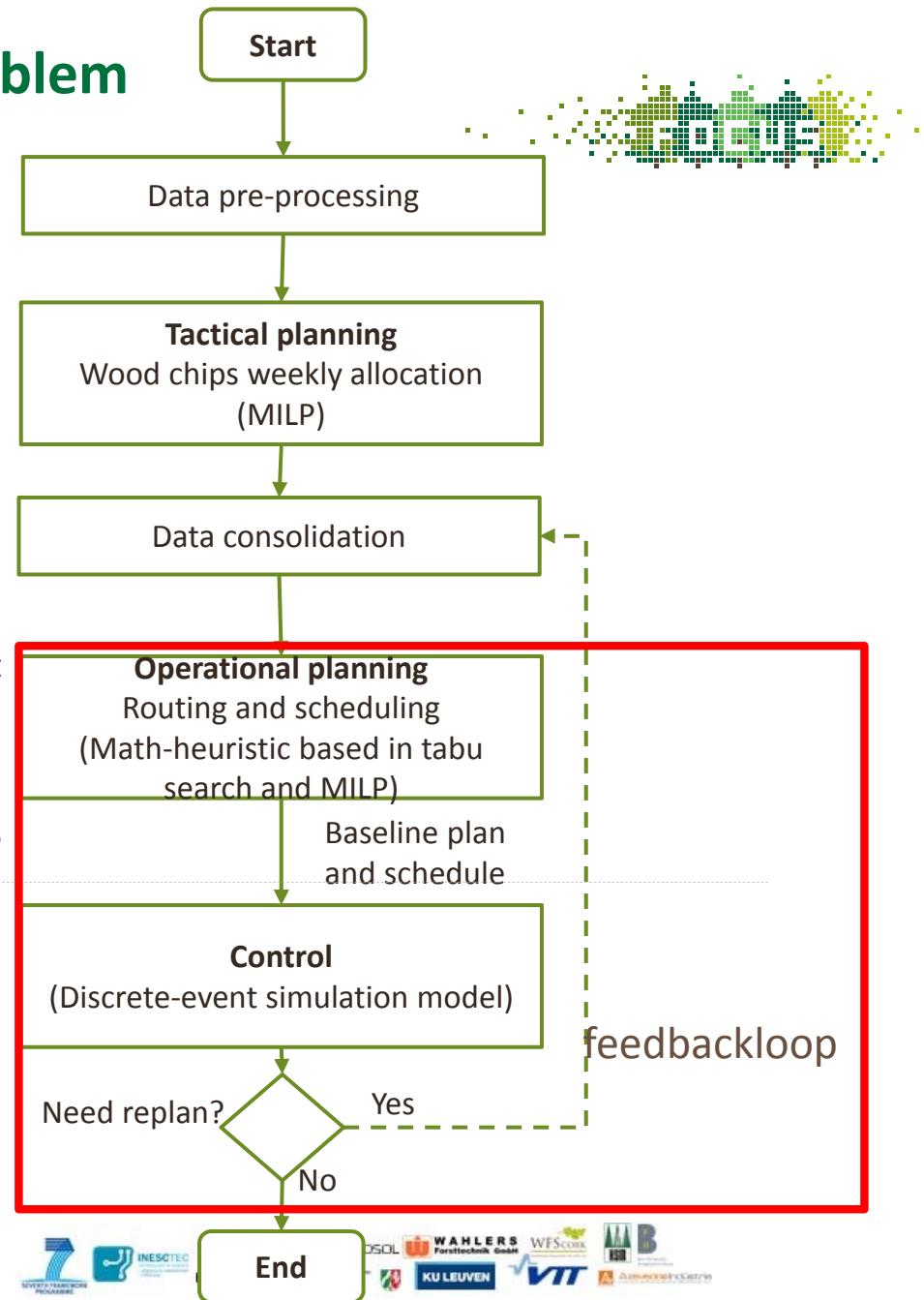
OPERATIONAL PLANNING: Daily routing and scheduling of trucks and chippers

Modelling Approach for the Wood chips delivery planning problem

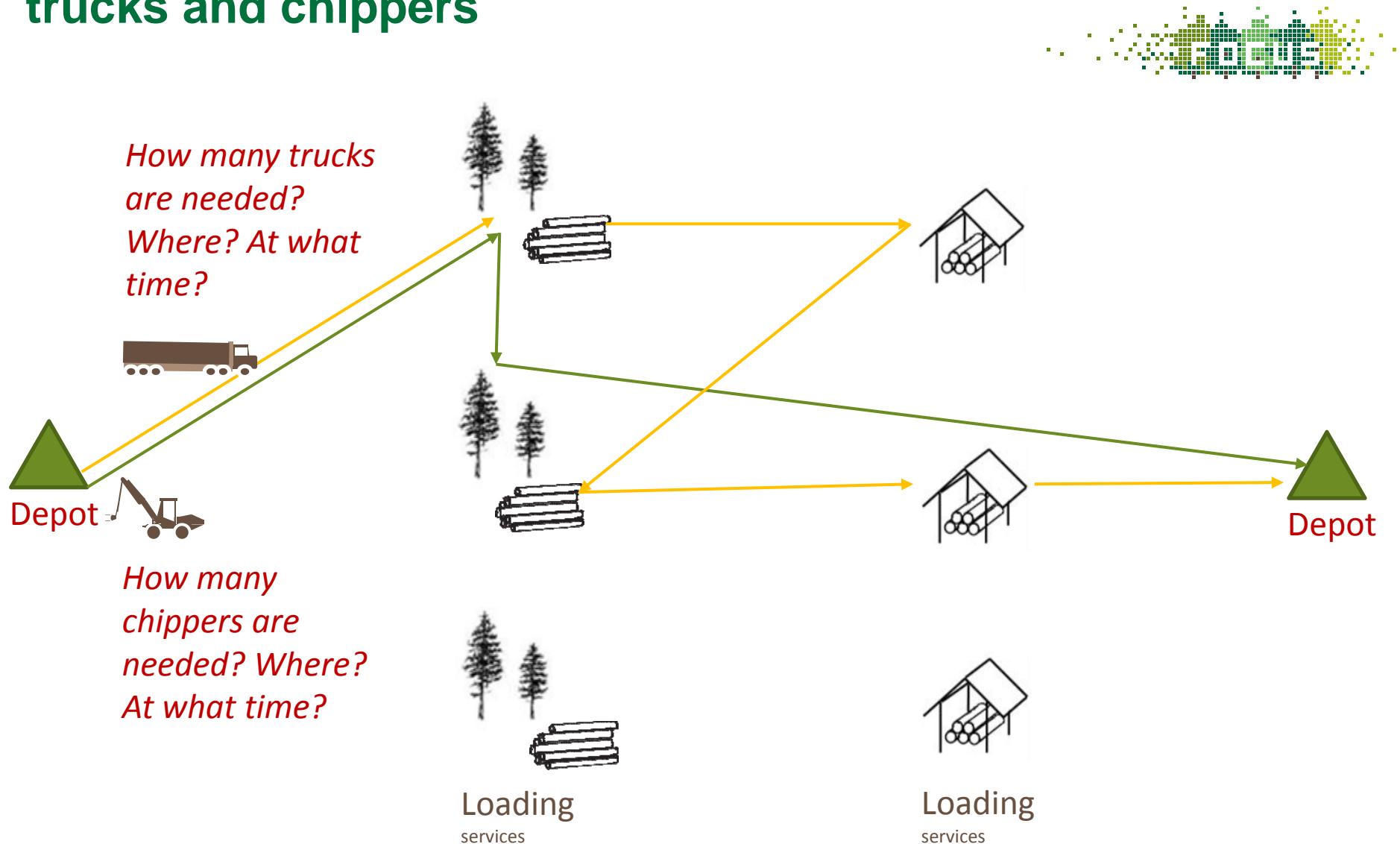
- max NPV
 - ✓ Allocation of raw materials to the mills;
 - ✓ Upper bound for truck fleet and chipper dimensioning

- min unproductive time in chipping and transport
 - ✓ Chippers routing and scheduling;
 - ✓ Trucks routing and scheduling;
 - ✓ Synchronization of vehicles;
-> Fleet size

- Check impact of events into the plan execution and expected energy content at the end of the day



Operational planning: Daily routing and scheduling of trucks and chippers



MIP for the Daily routing and scheduling of trucks and chippers



The problem can be modelled as two interconnected Vehicle Routing Problem, one for chippers and another for trucks, over the same network. They are interconnected through precedence and synchronization constraints:

Decision variables:

$$x_{ij}^v = \begin{cases} 1, & \text{if vehicle } v \in R \text{ transverses arc } (i,j) \in A \\ 0, & \text{otherwise} \end{cases}$$

$$w_i = \text{Time spent by a chipper in node } i \in P_i^{1*}. \text{ Note that, this includes the waiting time of a chipper in pile } i \in P \text{ (minutes)}$$

Objective function: Minimization of unproductive **waiting** and **deadhead times** (i.e. (empty truck travel times)

MIP for the Daily routing and scheduling of trucks and chippers



Constraints:

- Standard VRP constraints for trucks
- Standard VRP constraints for chippers
- One chippers per wood pile
- Maximum working time for chippers and trucks
- Precedence and synchronization between trucks and chippers

$$\sum_{j \in \delta_{G^+}^k} x_{G^+ j}^k = 1 \quad \forall k \in K$$

$$\sum_{i \in \delta_{G^-}^k} x_{i G^-}^k = 1 \quad \forall k \in K$$

$$\sum_{j \in \delta_i^+} x_{ij}^k - \sum_{j \in \delta_i^-} x_{ji}^k = 0 \quad \forall k \in K, i \in P^{1*}$$

$$\sum_{k \in K} D_i^k \leq \sum_{v \in V} D_i^v \quad \forall i \in P^{1*}$$

$$\sum_{v \in T} D_{P_s^{i*}}^v \leq \sum_{v \in V} D_{P_s^{j*}}^v \quad \forall s \in P; i, j \in P_s^* \wedge i < j$$

$$w_s \geq \sum_{v \in V} D_{P_s^{n*}}^v + st \sum_{v \in V} \sum_{j \in \delta_{P_s^{n*}}^+} x_{P_s^{n*} j}^v - \sum_{k \in K} D_{P_s^{1*}}^k \quad \forall s \in P$$

Chippers need to arrive first to any pile

Once start chipping, chipp until pile empty

Hard to solve for real-life instances -> need matheuristic

Mst



Precedence/synchronization constraints

$$\sum_{k \in K} D_i^k \leq \sum_{v \in V} D_i^v \quad \forall i \in P^{1*} \quad (17) \quad \text{Vehicle precedence}$$

$$\sum_{v \in T} D_{p_s^{i*}}^v \leq \sum_{v \in V} D_{p_s^{j*}}^v \quad \forall s \in P; i, j \in P_s^* \wedge i < j \quad (18) \quad \text{Service ordering}$$

$$w_s \geq \sum_{v \in V} D_{p_s^{n*}}^v + st \sum_{v \in V} \sum_{j \in \delta_{p_s^{n*}}^i} x_{p_s^{n*} j}^v - \sum_{k \in K} D_{p_s^{1*}}^k \quad \forall s \in P \quad (19) \quad \text{Waiting times}$$

$$x_{ij}^v \in \{0,1\}$$

$$D_s^v, w_l \geq 0$$

$$\forall (i,j) \in A, v \in R$$

$$\forall (i,j) \in A, s \in ND, l \in P$$

(20) Variables' nature
(21)

Given this formulation, **real-size instances are unable to be solved in viable computational time**

Solution method - MIP model

$$\text{Min } Z = \sum_{i \in P^{1*}} w_i + \sum_{(i,j) \in A: i \in M^*} \sum_{v \in R} t_{ij} x_{ji}^v$$

Minimization of unproductive waiting and deadhead times
(empty truck travel times)

s.t.

For the chippers

$$\sum_{j \in \delta_{O^+}^+} x_{O^+ j}^k = 1 \quad \forall k \in K$$

$$\sum_{i \in \delta_{O^-}^-} x_{i O^-}^k = 1 \quad \forall k \in K$$

$$\sum_{j \in \delta_i^+} x_{ij}^k - \sum_{j \in \delta_i^-} x_{ji}^k = 0 \quad \forall k \in K, i \in P^{1*}$$

$$\sum_{k \in K} \sum_{j \in \delta_i^+} x_{ij}^k = 1 \quad \forall i \in P^{1*}$$

$$D_{O^+}^k = 0 \quad \forall k \in K$$

$$D_i^k \leq \text{maxTime} \sum_{j \in \delta_i^+} x_{ij}^k \quad \forall i \in P^{1*} \cup O^-, k \in K$$

$$D_i^k + t_{ij} x_{ij}^k \leq D_j^k + \text{maxTime}(1 - x_{ij}^k) \quad \forall i = O^+, j \in \delta_i^+, k \in K$$

$$D_i^k + (w_s + t_{ij}) x_{ij}^k \leq D_j^k + \text{maxTime}(1 - x_{ij}^k) \quad \forall s \in P, i \in P_s^{1*}, j \in \delta_i^+, k \in K$$

(1)

(2)

(3)

(4)

Standard VRP const.

(5) Time initialization

(6) Maximum working time

(7a) Time elapsed constraints

(7b)

Solution method - MIP model

For the trucks

$$\sum_{j \in \delta_{0^+}^+} x_{0^+j}^v = 1 \quad \forall v \in V \quad (8)$$

$$\sum_{i \in \delta_{0^-}^-} x_{i0^-}^v = 1 \quad \forall v \in V \quad (9)$$

$$\sum_{j \in \delta_i^+} x_{ij}^v - \sum_{j \in \delta_i^-} x_{ji}^v = 0 \quad \forall v \in V, i \in ND \quad (10)$$

$$\sum_{v \in T} \sum_{j \in \delta_i^+} x_{ij}^v = 1 \quad \forall i \in M^* \quad (11)$$

$$\sum_{v \in T} \sum_{j \in \delta_i^+} x_{ij}^v = 1 \quad \forall i \in P^* \quad (12)$$

$$D_{0^+}^v = 0 \quad \forall v \in V \quad (13) \quad \text{Time initialization}$$

$$D_i^v \leq maxTime \sum_{j \in \delta_i^+} x_{ij}^v \quad \forall i \in ND \cup O^-, v \in V \quad (14) \quad \text{Maximum working time}$$

$$D_i^v + t_{ij} \leq D_j^v + maxTime(1 - x_{ij}^v) \quad \forall i = O^+, j \in \delta_i^+, v \in V \quad (15a)$$

$$D_i^v + st + t_{ij} \leq D_j^v + maxTime(1 - x_{ij}^v) \quad \forall i \in ND, j \in \delta_i^+, v \in V \quad (15b)$$

$$LTW_i \leq D_i^v \leq UTW_i \quad \forall i \in M^*, v \in T \quad (16) \quad \text{Time windows}$$



Standard VRP const.

(11)
(12)

(13) Time initialization

(14) Maximum working time

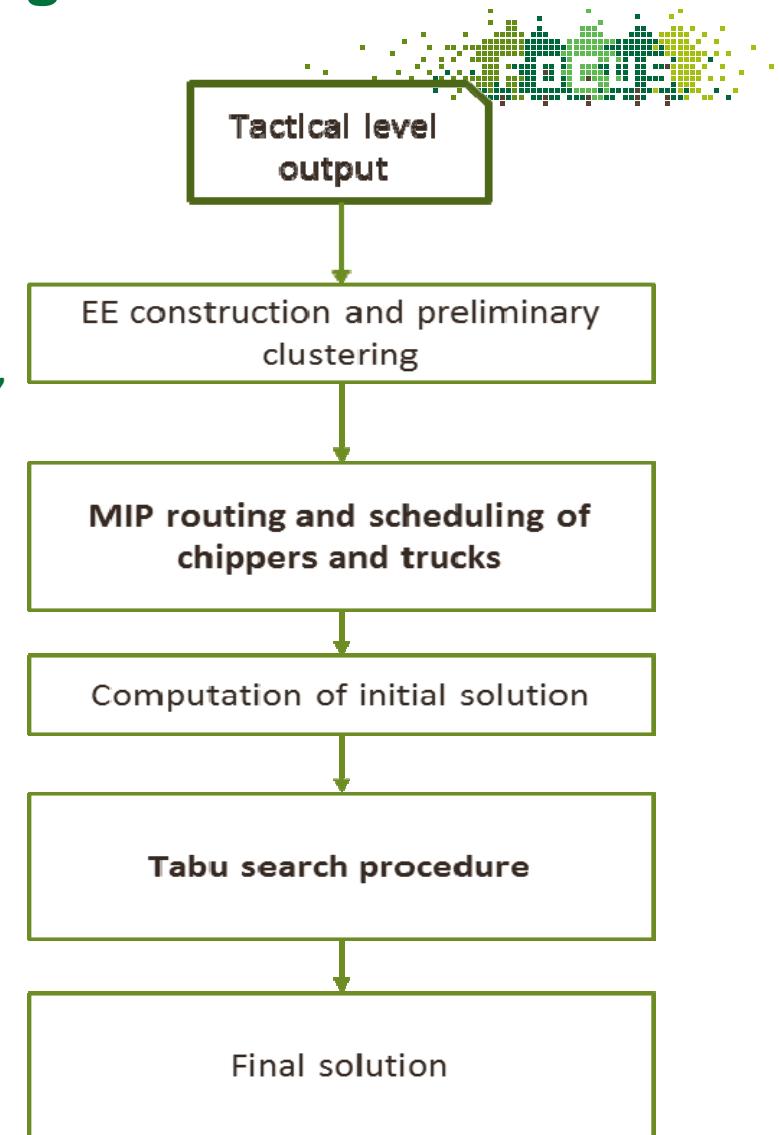
(15a)
(15b) Time elapsed constraints

(16) Time windows

Matheuristic for operational planning

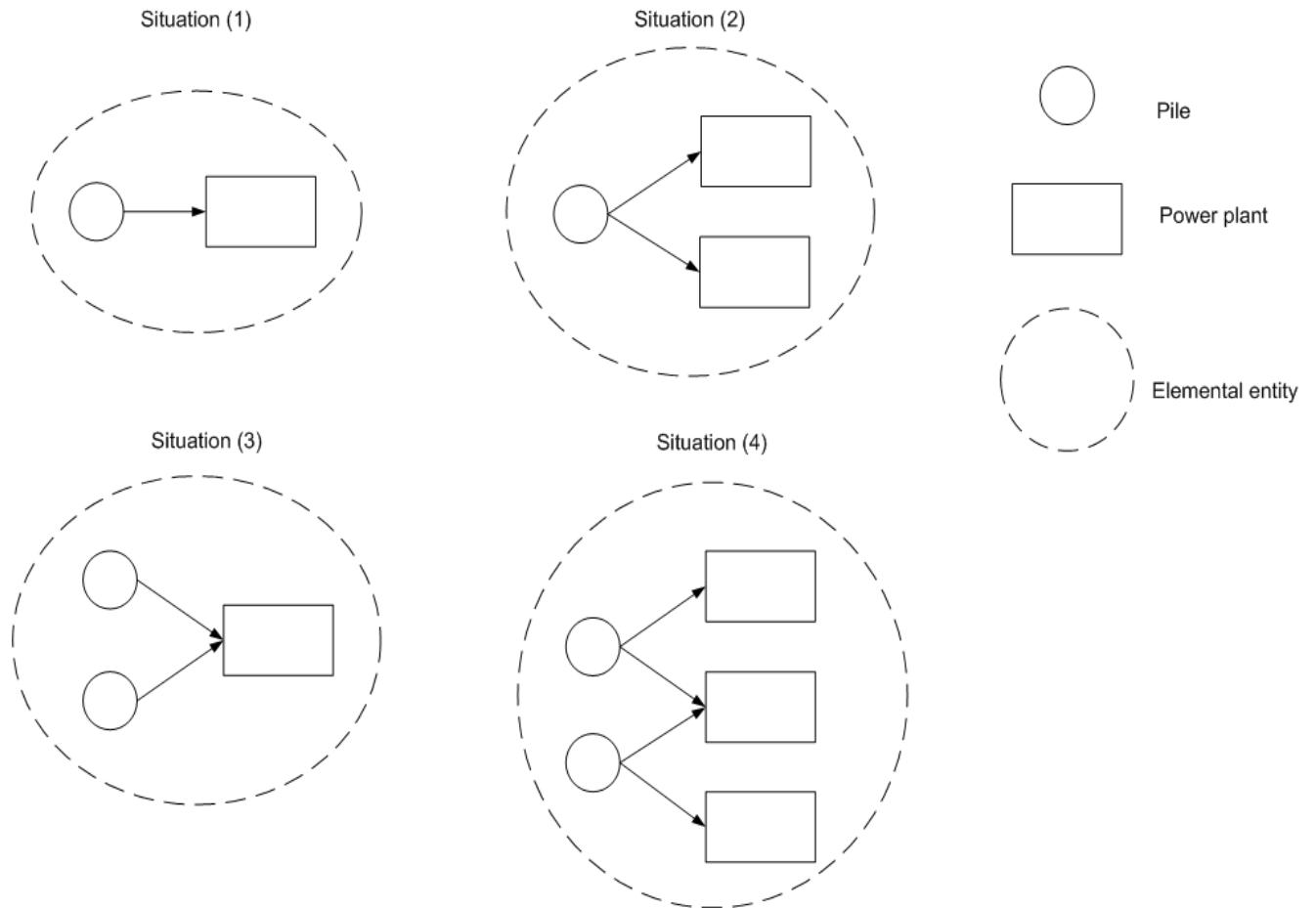
1. Problem decomposition methods (clustering);
2. MIP models for clustering and vehicles routing, scheduling and synchronization;
3. Meta-heuristic inspired in Tabu Search:

Doerner, K.F., Schmid, V., 2010.; Maniezzo, V., et al. 2010.



Matheuristic > Defining Elemental Entities based on output from tactical planning

An EE is composed of all piles and power plants so that the material flows are contained in that EE.



Matheuristic > MILP model for Clustering EE's



Decision variables

$$x_{ik} = \begin{cases} 1, & \text{if the Elemental Entity } i \in E \text{ is assigned to cluster } k \in K \\ 0, & \text{otherwise} \end{cases}$$

$$y_k = \begin{cases} 1, & \text{if cluster } k \in K \text{ is used} \\ 0, & \text{otherwise} \end{cases}$$

NC Number of clusters

Model

e.g. Barreto, S., et al. 2007.

$$\text{Min } Z = NC$$

s.t.

$$\sum_{k \in K} y_k \leq NC$$

(1) Number of clusters

$$x_{ik} \leq y_k \quad \forall i \in E, k \in K$$

(2) Linking constraints

$$\sum_{k \in K} x_{ik} = 1 \quad \forall i \in E$$

(3) An EE will only belong to one cluster

$$\sum_{i \in E} n_i x_{ik} \leq Max_{\#} \quad \forall k \in K$$

(4) Maximum number of dummy nodes

$$x_{il} \in \{0,1\} \quad \forall i \in EE, l \in K$$

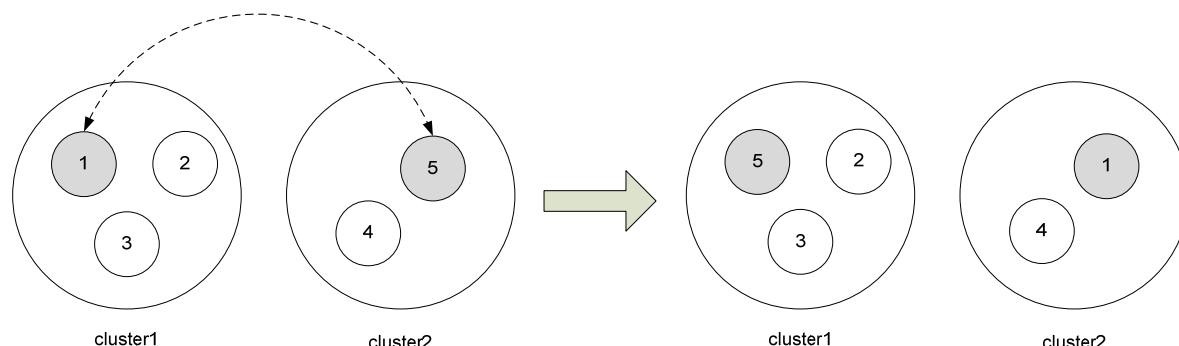
(5)

Matheuristic >Tabu search procedure



Move from one solution to the next within its Neighborhood...

... by swapping EE's



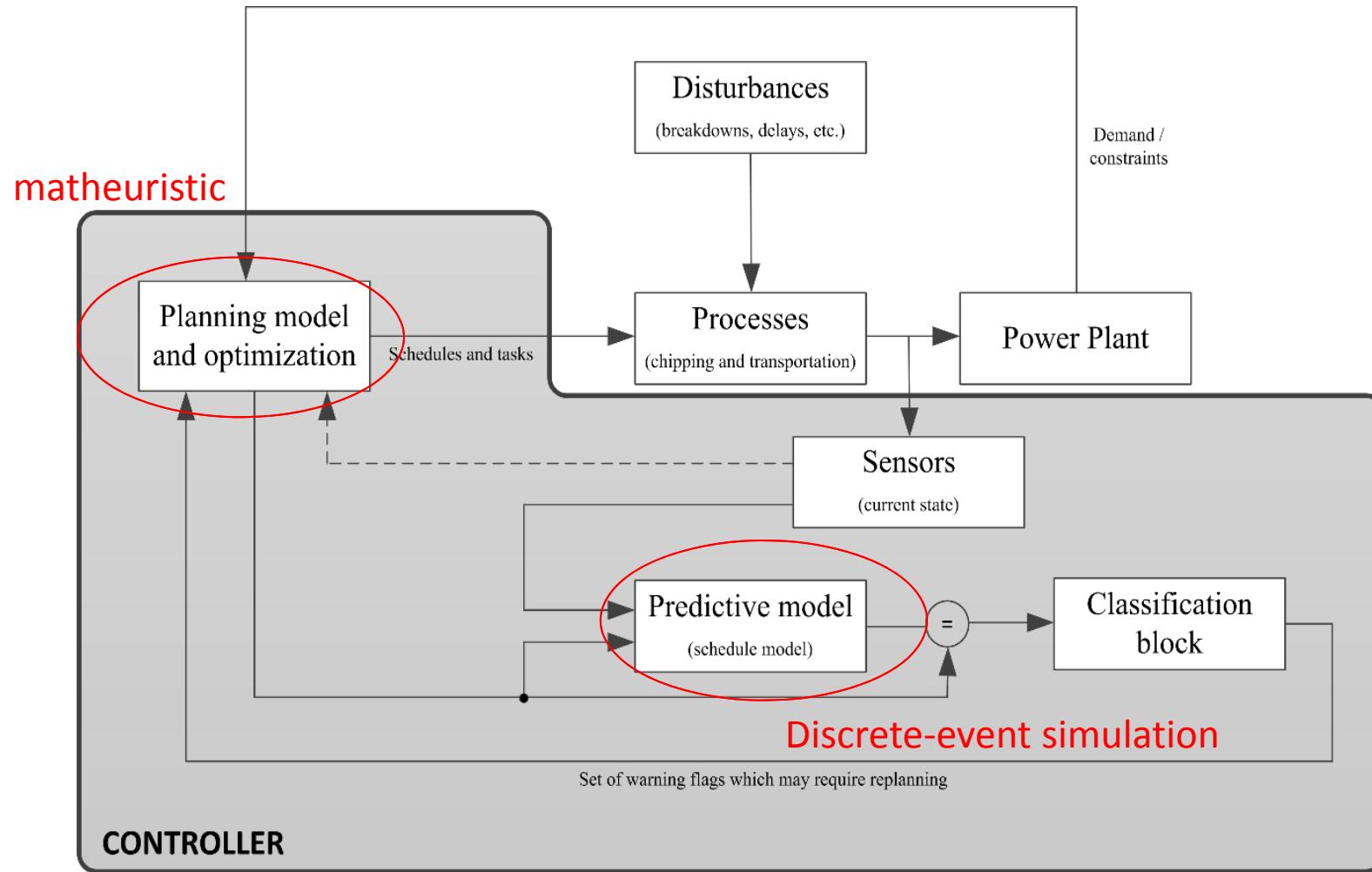
Move:
Cluster1 = 1
Cluster2 = 2
EE1 = 1
EE2 = 5

Tabu Move:
Cluster1 = 1
Cluster2 = 2
EE1 = 5
EE2 = 1

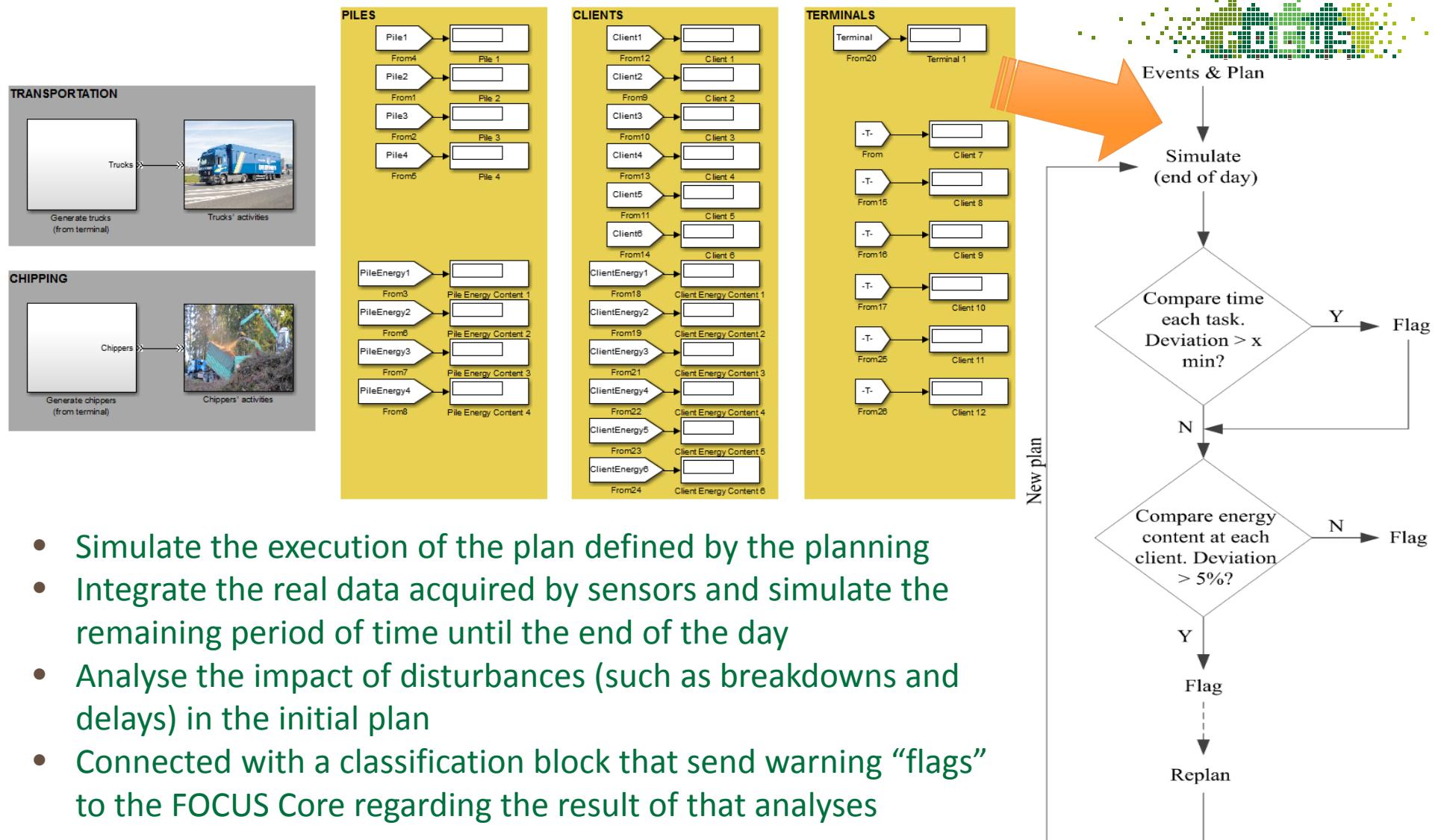
... evaluate new solution -> solve MILP within each cluster

Update tabu list

The Model Predictive Control approach



The discrete-event simulation model



“Control” phase



Vehicle \ Time	195	202	225	...	463	473	493	498	...	588	596	606	636
Vehicle													
Chipper 1		p4d1	p4d1	...	pld1		t						
Chipper 2	p2d1	p2d1	p2d1	...	p3d1	p3d1	p3d1	p3d1	...	t			
Truck 1		p4d1		...	pld4				...				
Truck 2			p2d2	...		c1d3			...	c6d4		t	
Truck 3	p2d1			...		c6d1			...		t		
Truck 4				...			p3d2		...				



Daily plan:

- Number of chippers / trucks
- Tasks and schedules of chippers and trucks for a day
- Expected amount of material in the power plants

Preliminary results

EE_id	#piles	#powerplants	# chippers	# trucks	Computational time
0	2	3	5	15	10 min
0	2	3	2	7	8.46 sec
1	1	2	5	15	4.95 sec.
1	1	2	1	4	0.39 sec.
2	1	1	5	15	0.67 sec.
2	1	1	1	2	0.22 sec.
3	1	1	5	15	0.58 sec.
3	1	1	1	4	0.22 sec.
4	1	1	5	15	0.55 sec.
4	1	1	1	2	0.14 sec.
5	1	1	5	15	0.66 sec.
5	1	1	1	4	0.09 sec.
6	2	1	5	15	0.65 sec.
6	2	1	1	2	0.34 sec.
7	1	1	5	15	0.75
7	1	1	1	4	0.11 sec.
8	1	1	5	15	0.86
8	1	1	1	2	0.26



For a given day:
9 Elemental entities

Grouped into 5
clusters

Preliminary results



New daily schedule for one of the clusters:

Vehicle \ Time	142	202	225	232	240	270	285	295	315	325	330	333	360	363						
Vehicle \ Time	373	403	413	430	433	443	463	473	493	498	503	523	536	543	553	566	576	588	596	
Chipper 1		p4d1	p4d1						Pile 4 - p4											
Chipper 2		p2d1						Pile 2 . P2												
Truck 1		p4d1		c4d1		p4d3					c4d3									
Truck 2			p2d2		c2d2															
Truck 3		p2d1		c2d1		p2d3					c2d3									
Truck 4			p4d2	c4d2			p2d4					c2d4								
Chipper 1		p1d1		Pile 1 - p1						depot										
Chipper 2			p3d1				Pile 3 - p3										depot			
Truck 1	p1d1		c1d1			p1d4		c1d4			depot									
Truck 2				p3d1			c1d3				p3d4						c6d4		depot	
Truck 3					c1d2		c6d1		p3d3		c6d3							depot		
Truck 4	p1d2					p3d2			c6d2			depot								

Concluding remarks



1. The proposed solution method efficiency is highly dependent on cluster creation and graph simplification techniques. Some are still to be implemented.
2. Problem decomposition (cluster creation) allows for events management to be addressed in a faster computational time as it only takes into account the part of the system in which it occurs.
3. Further testing will be conducted to properly parameterize the tabu search method governing the local search
4. Results of the matheuristic are promising. Some efficiency procedures are still to be implemented which we expect to improve computational results.
5. The bottleneck, as expected, is the efficiency of the exact method. MIP model resolution methods may be explored further on (Drexel, M. (2012)).
6. Additional testing is required for control and re-optimization

FOCUS in a nutshell

What?

7 FP SME-target collaborative RTD project
01-01-2014 to 30-06-2016 (30 months)



Why?

Need for integrated processing and control systems for sustainable production in farms and forests.

How?

New FOCUS technological platform that combines sensors and sophisticated software solutions for integrated control and planning of the whole forest-based value chain.

7 Work Packages encompassing specification, development of data collection tools as well as control and planning tools, integration; assessment of prototypes into 4 pilot cases. Covering the value chains in Europe of lumber, pulpwood, biomass and cork transformation; from forest planning to industrial processing.
Total budget of ~4M€ (~3M€ EC funding)

The goal of FOCUS is to improve sustainability, productivity, and product marketability of forest-based value chains through an innovative technological platform for integrated planning and control of the whole tree-to-product operations, used by forest-producers to industry players.

www.focusnet.eu

By whom?

Consortium of 6 SMEs and 6 RTDs from Portugal, Finland, Belgium, Switzerland, Austria and Germany, combining expertise in forestry, sensors, automation and software development.

References



VRPs with multiple synchronization constraints (VRPMSSs)

Bredstrom, D., Ronnqvist, M., 2008. Combined vehicle routing and scheduling with temporal precedence and synchronization constraints. *Eur. J. Oper. Res.* 191, 19–31. doi:10.1016/j.ejor.2007.07.033

Dohn, A., Rasmussen, M.S., Larsen, J., 2011. The vehicle routing problem with time windows and temporal dependencies. *Networks* 58, 273–289. doi:10.1002/net.20472

Drexel, M., 2012b. Synchronization in Vehicle Routing—A Survey of VRPs with Multiple Synchronization Constraints. *Transportation Science* 46, 297–316. doi:10.1287/trsc.1110.0400

Drexel, M., 2012a. Branch and Price for vehicle routing problems with multiple synchronization constraints. International workshop on column generation, Bomonnt.

El Hachemi, N., Gendreau, M., Rousseau, L.-M., 2011. A hybrid constraint programming approach to the log-truck scheduling problem. *Ann. Oper. Res.* 184, 163–178. doi:10.1007/s10479-010-0698-x

Clustering

Barreto, S., Ferreira, C., Paixão, J., Santos, B.S., 2007. Using clustering analysis in a capacitated location-routing problem. *European Journal of Operational Research* 179, 968–977. doi:10.1016/j.ejor.2005.06.074

References



Vehicle routing near-real time control

Bock, S., 2010. Real-time control of freight forwarder transportation networks by integrating multimodal transport chains. *European Journal of Operational Research* 200, 733–746. doi:10.1016/j.ejor.2009.01.046

Gendreau, M., Guertin, F., Potvin, J.-Y., Séguin, R., 2006. Neighborhood search heuristics for a dynamic vehicle dispatching problem with pick-ups and deliveries. *Transportation Research Part C: Emerging Technologies* 14, 157–174. doi:10.1016/j.trc.2006.03.002

Ghiani, G., Guerriero, F., Laporte, G., Musmanno, R., 2003. Real-time vehicle routing: Solution concepts, algorithms and parallel computing strategies. *European Journal of Operational Research* 151, 1–11. doi:10.1016/S0377-2217(02)00915-3

Sáez, D., Cortés, C.E., Núñez, A., 2008. Hybrid adaptive predictive control for the multi-vehicle dynamic pick-up and delivery problem based on genetic algorithms and fuzzy clustering. *Computers & Operations Research* 35, 3412–3438. doi:10.1016/j.cor.2007.01.025

References



Tabu search

Archetti, C., Speranza, M.G., Hertz, A., 2006. A Tabu Search Algorithm for the Split Delivery Vehicle Routing Problem. *Transportation Science* 40, 64–73. doi:10.1287/trsc.1040.0103

Berbotto, L., Garcia, S., Nogales, F.J., 2014. A Randomized Granular Tabu Search heuristic for the split delivery vehicle routing problem. *Annals of Operations Research* 222, 153–173.

Cordeau, J.-F., Laporte, G., 2005. Tabu search heuristics for the vehicle routing problem. Springer.

Zhang, Z., He, H., Luo, Z., Qin, H., Guo, S., 2015. An Efficient Forest-Based Tabu Search Algorithm for the Split-delivery Vehicle Routing Problem, in: Twenty-Ninth AAAI Conference on Artificial Intelligence.

Contacts



- Tatiana M. Pinho TATIANA.M.PINHO@UTAD.EU
- Bruno Oliveira BRUNO.C.OLIVEIRA@INESCPORTO.PT
- Dmitry Podkopaev DMITRY.PODKOPAEV@SIMOSOL.FI
- Jussi Rasinmäki JUSSI.RASINMAKI@SIMOSOL.FI
- Alexandra Marques ALEXANDRA.S.MARQUES@INESCPORTO.PT
- José Boaventura-Cunha JBOAVENT@UTAD.PT

www.focusnet.eu



Thank you!