

## COMPARISON OF OPTIMISATION ALGORITHMS TO DETERMINE CONTROL STRATEGIES IN UDS

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### KEYWORDS

RTC of UDS, Evaluation of Control Strategy, Determination of Control Strategy, Linear Optimisation, Non-linear Optimisation

### 1. Introduction

In combination with other sanitation measures RTC can improve the performances of UDS. Control strategies can be determined by optimisation algorithms. They have following characteristics:

- control objectives are expressed as a cost function, whose value is to be minimised
- static (i.e. maximal storage or transport capacity) and dynamic characteristics of the UDS are expressed in so called domain restrictions<sup>1</sup>

Because of limited computation capacity, a general formulation of constraints is incompatible with on-line implementation. Most of the previous studies refer to linear optimisation, where restrictions and cost functions are linear combinations of decision variables. Specific non-linear formulations can make the mathematical transposition more flexible. On the other hand, their cost coefficients are more difficult to determine and computation times increase. This paper describes a linear and non-linear optimisation module. Aim of the study is not to get the very best of each method in a specific case. It will rather be tried to specify the characteristics of each method by analysing and comparing their results.

### 2. Principles of optimisation

#### 2.1 Linear optimisation

A linear optimisation problem is expressed as follows; "minimise the cost function  $f_1(X)$  under following conditions:  $AX \leq B$  and  $X \geq 0$ "

$$f_1(X) = \sum (\sum \alpha_i X_{ik}), \text{ summation over } i=1, \dots, n \text{ and } k=1, \dots, N$$

- $n$  : number of decision variables and  $N$  : number of decision steps
- $\alpha_i$  : cost coefficients related to decision variable No  $i$
- $X$  :  $x_{ik}$  represent the decision variables at a given time step
- $A$  : matrix (does not depend on  $X$ ) and  $B$  : vector (does not depend on  $X$ )

The conversion of the practical control problem into the numerical one requires a specific representation of the UDS and its behaviour. Basic descriptive elements are nodes and connections.

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<sup>1</sup> In reality, it is also possible to express specific constraints in the direct formulation of the cost function

- Each connection corresponds to a transport capacity (i.e. collectors, pumps). It has a maximal flow capacity and an initial value. Besides, the flow process in each transport element is characterised by a constant transport time.
- Each node corresponds to a storage capacity (m<sup>3</sup>). It has a maximal volume capacity and an initial value. Besides, it satisfies the principle of continuity.

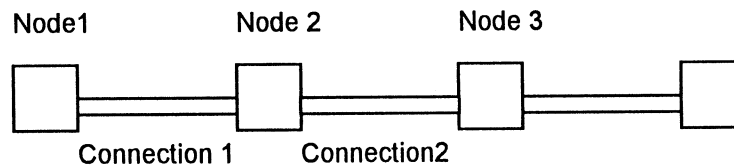


Fig. 1: simplified description of the UDS for optimisation

In a single calculation step for a given period of time  $[t, t+Ht]$ , optimisation algorithms determine the values of the decision variables, which minimise the costs.  $Ht$  is an integer multiple of the decision interval, called horizon of optimisation. Theoretically,  $Ht$  should include the whole rainfall event. But the number of restrictions and decision values increases linearly with  $Ht$ , so that the computer capacity is rapidly overtaxed. In most of the cases,  $Ht$  only includes a few  $\Delta t$ . The optimisation successively calculates values for shifted horizon  $Ht$ , till the whole event is simulated.

The determination of the minimum of  $f_1(X)$  is based on the simplex algorithm developed by Dantzig (1948). Its description can be found in almost every book dealing with linear optimisation.

## 2.2 Non linear optimisation

In the non-linear optimisation the formulation of the domain restrictions remains unchanged, but the formulation of the cost function  $f_2(X)$  is different.

$$f_2(X) = \sum (\sum \alpha_i X_{ik} \beta_i + \gamma_i (x_{iN} - x_{i1})) , \quad i=1, \dots, n , \quad k=1, \dots, N$$

with  $\alpha_i$ : cost coefficients,  $\beta_i$  form parameters and  $\gamma_i$  variability coefficients

The implemented algorithm is based on the "branch-and-bound-method" (see HORST, 1979 and ZOUTENDIJK, 1960). The advantage of the method is that it can cope with convex elementary functions ( $\alpha_i \geq 0, \beta_i \geq 1$ ) and concave elementary functions ( $\alpha_i \geq 0, \beta_i < 1$ ). The characteristics of the system and its control can be accounted for in a more flexible way.

## 3. Study case

### 3.1 Description and Representations of the UDS

An artificial UDS has been selected (see Fig. 2 and Tab. 1). It drains four identical catchments ( $E_i, i=1,4$ ) of 100 ha each (50% impervious surfaces). The pipes cross sections are circular and their diameters range from 1m to 1.5 m. Besides the pipes storage capacity, the UDS contains five identical retention basins ( $S_i, i=1,5$ ). Each of them has a maximal capacity of 5000 m<sup>3</sup>. There are five outlets.

Internal pump stations D2 (upstream/downstream), D4 (upstream/downstream), D5 (only downstream) control the transport. For every admissible flow direction, the maximal pumping capacity is 4 m<sup>3</sup>/s, so that every retention basin can be filled and emptied in less than 25 minutes.

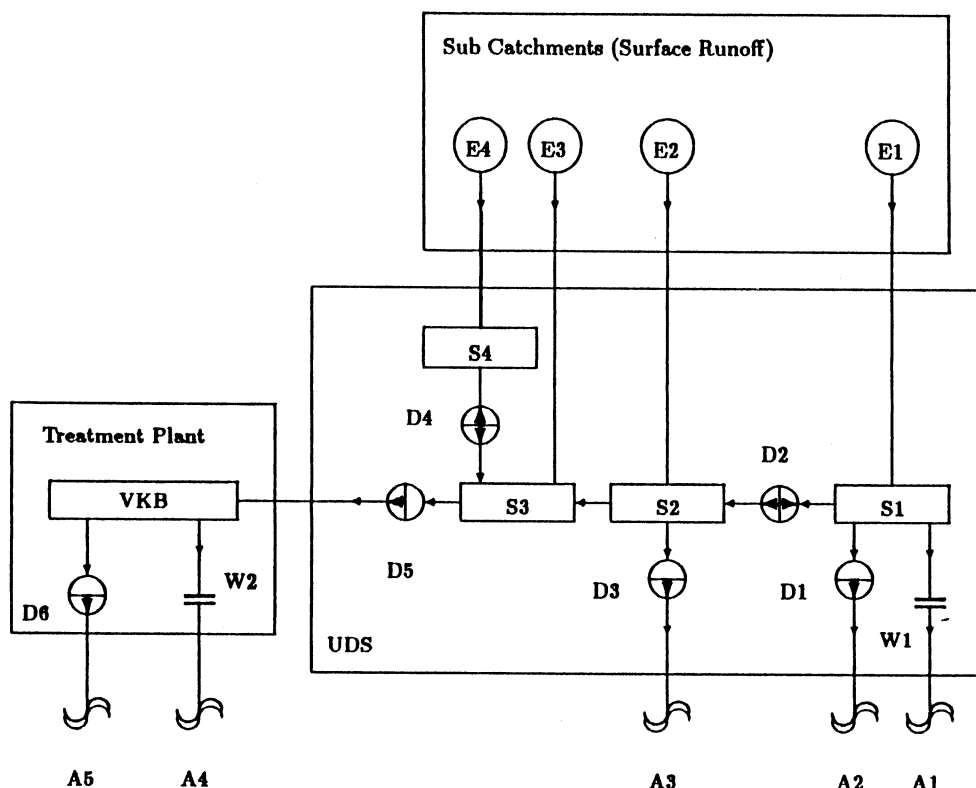


Fig. 2: System representation of the UDS

outlet	Type of connection	max. capacity Q	type of receiving waters
A1	Weir W1	> 60 m <sup>3</sup> /s	very sensitive to pollution loads
A2	Pump D1	4 m <sup>3</sup> /s	not very sensitive
A3	Pump D3	4 m <sup>3</sup> /s	not very sensitive
A4	Weir W2	> 110 m <sup>3</sup> /s	not very sensitive
A5	Pump D6	0.8 m <sup>3</sup> /s	Treatment plant

Tab. 1: System outlets

No	Notation	Type	stands for
1	EZ1	-	inflows from catchment EZ1
2	EZ23	-	inflows from catchment EZ2+EZ3
3	EZ4	-	inflows from catchment EZ4
1	KV1	internal node	Retention Bassin S1
2	KV23	internal node	Retention Bassins S2+S3
3	KV4	internal node	Retention Bassin S4
4	KV5	internal node	Retention Bassin S5
5	KENT11	external node	Receiving Waters F1
6	KENT12	external node	Receiving Waters F2
7	KENT21	external node	Receiving Waters F3
8	KENT51	external node	Receiving Waters F4
9	KENT52	external node	Outlet of the TP

Tab. 2: Description of the UDS for optimisation

No	Notation	Type	stands for
1	BP1	connection	CSO Pump
2	BP2	connection	internal Pump downstream
3	BP3	connection	internal Pump upstream
4	BP4	connection	CSO Pump
5	BP5	connection	internal Pump downstream
6	BP6	connection	internal Pump upstream
7	BP7	connection	internal Pump downstream
8	BP8	connection	TP treatment capacity
9	BW1	connection	Weir Outlet into F1
10	BW2	connection	Weir Outlet into F4

Tab. 2: Description of the UDS for optimisation

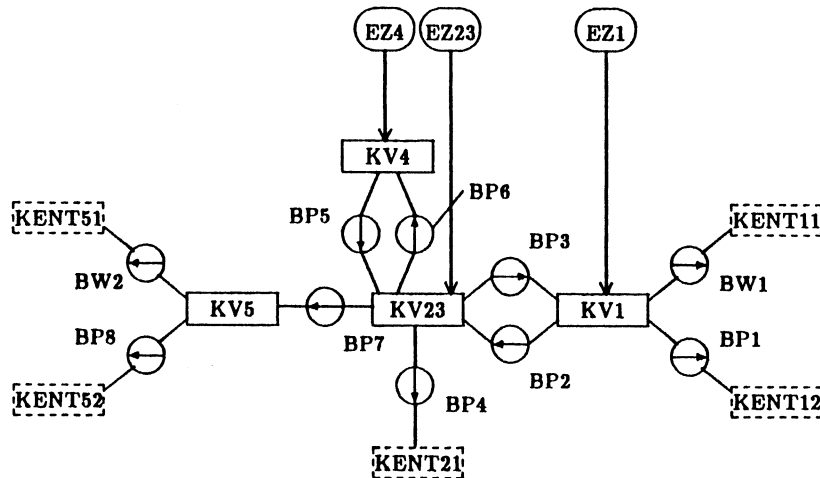


Fig. 3: Representation of the UDS for the optimisation programmes

### 3.2 The restrictions

Every decision variable must be non-negative. Supplementary restrictions are :

- non negative initial conditions in connections and nodes,
- non negative minimal and maximal capacity in connections ( $\text{m}^3/\text{s}$ ) (14 inequalities)
- non negative minimal and maximal capacity in nodes ( $\text{m}^3$ ) (10 inequalities)
- continuity equations in nodes (4 equalities + 5 equalities)

The total number of restrictions depends on  $H_t$  (see Tab. 3). A decision interval  $\Delta t$  is 5 min.

N	2	3	4	5
with external nodes	83	116	149	182
without external nodes	57	80	103	126

Tab. 3: Number of restrictions for various N (for the first optimisation horizon)

### 3.3 cost function ( Tab. 4)

The costs were first determined for the linear optimisation **f1** according to following priorities: no flooding > no CSO > no storage > no transport. The cost function **f2** was determined on the basis of **f1**, as follows:

- costs coefficients for connections:  $\alpha_i$  unchanged,  $\beta_i=2$
- for internal nodes:  $\beta_i=2$ , coefficients  $\alpha_i$  were calculated such that the storage costs remain higher than the transport costs
- for external nodes :  $\beta_i=0.2$  and coefficients  $\alpha_i$  were determined such that following inequality is verified;  $\alpha_{KV} \times KV^2 \geq \alpha_{KENT} * 1 \text{ m}^3/\Delta t$ , where KV designates the retention basin connected to the external node KENT

Element	max. cap. in m <sup>3</sup>	linear optim.		non linear optim			
		$\alpha_i$	max. costs	$\beta_i$	$\alpha_i$	$\gamma_i$	max. costs
BP1	1200	1.00	1200	2	1.00	100	$1.44 \cdot 10^6$
BW1	1950	1.00	1950	2	1.00	0	$3.80 \cdot 10^6$
BP2	1200	0.02	24	2	0.02	10	28800
BP3	1200	1.00	1200	2	1.00	10	$1.44 \cdot 10^6$
BP4	1200	1.00	1200	2	1.00	100	$1.44 \cdot 10^6$
BP5	1200	0.02	24	2	0.02	10	28800
BP6	1200	1.00	1200	2	1.00	10	$1.44 \cdot 10^6$
BP7	1200	0.01	12	2	0.01	10	14400
BP8	240	0	0	0	0	50	0
BW2	1950	1.00	1950	2	1.00	0	$3.8 \cdot 10^6$
KV1	5000	0.20	1000	2	0.08	0	$2.0 \cdot 10^6$
KV23	10000	0.18	1800	2	0.03	0	$3.0 \cdot 10^6$
KV4	5000	0.20	1000	2	0.08	0	$2.0 \cdot 10^6$
KV5	5000	0.10	500	2	0.02	0	$5.0 \cdot 10^5$
KENT11	10000	200.00	$2.0 \cdot 10^6$	1/5	$2.0 \cdot 10^6$	0	$1.26 \cdot 10^7$
KENT12	10000	70.00	$7.0 \cdot 10^5$	1/5	$1.7 \cdot 10^6$	0	$1.07 \cdot 10^7$
KENT21	10000	100.00	$1.0 \cdot 10^6$	1/5	$2.5 \cdot 10^6$	0	$1.57 \cdot 10^7$
KENT51	15000	50.00	$7.5 \cdot 10^5$	1/5	$0.6 \cdot 10^6$	0	$4.11 \cdot 10^7$
KENT52	35000	0.00	0	1/5	0	0	0

Tab. 4: cost functions for linear and non linear optimisations

Optimisations were also performed without external nodes (KENT). In this cases, the costs for corresponding external connections (BP resp. BW) were adjusted as follows;  $\alpha_{BP}=\alpha_{KENT}$ ,  $\beta_{BP}=1$ ,  $\gamma$  unchanged.

#### 4. Some results

For several rainfall-runoff events optimisation results have been obtained, here a single event (Tab. 5) and its results (Tab. 6) are presented.

No.	Rainfall height (mm)	Rainfall duration (min)	Runoff duration (min)	runoff volume (m <sup>3</sup> )	max. inflow rate (m <sup>3</sup> /5min)	$a_U$ (min)
1	15	30	104	30 000	2485	20

Tab. 5: rainfall and run-off characteristics of reference event

Obviously, something is wrong in the costs of the external nodes in the non-linear optimisation: the greater the optimisation horizon, the more CSO increases!. After the  $\alpha_i$  coefficient for KENT51 was doubled, the amount of CSO was reduced of 50%, but the deterioration of the strategy results for greater N remained. The concavity of the elementary cost function in KENT51 ( $\beta_i=0.2$ ), is responsible for this negative tendency.

If the external nodes are removed (=no concave elementary functions), no deterioration of the results occurs, when the horizon increases.

No.	Optim. method	external nodes	N	KENT51 m <sup>3</sup>	KENT51 duration in $\Delta t$ (-)	KENT51 max Q in m <sup>3</sup> /5min
1	non linear	yes	2	4786	37	893
2	non-linear	yes	3	6186	34	614
3	non-linear	yes	4	8686	32	682
4	non-linear	yes	5	10788	30	699
5	non-linear	no	2-5	2500	7	893
6	linear	no	2-5	2513	7	893

**Tab. 6:** results of various optimisations for the reference rainfall-runoff event<sup>2</sup>

The amounts of **CSO** are almost the same for both optimisation methods and independent of the optimisation horizon. This is all the more astonishing, if we compare the computation times (**Tab. 7**).

Optimisation method	External nodes	N=2 (h.min)	N=3 (h.min)	N=4 (h.min)	N=5 (h.min)
non linear	yes	2.07	5.43	12.14	21.53
non linear	no	0.11	0.27	0.54	1.37
linear	yes	0.04	0.09	0.18	0.33
linear	no	0.02	0.04	0.10	0.15

**Tab. 7:** Comparison of the computation times

The utilisation of the storage capacity is almost independent of N in both methods. In the non-linear optimisation the distribution of the storage volumes in the different basins is more even. This is due to the quadratic form of the elementary cost function related to storage. The pump flow rates are also more regular in the non-linear optimisation, as long as the optimisation horizon remains small. But when N increases, the non-linear optimisation generates more and more irregular flow patterns. In linear optimisation the contrary is true. The negative tendency in the non-linear optimisation is due to the fact that small improvements in the exploitation of the storage capacity must be paid by a more and more erratic control of the pumps.

## 5. Conclusions

A linear and a non linear optimisation algorithm to determine control strategies in UDS have been briefly presented and applied in a study case. The following conclusions can be made:

- optimisation horizon does not seem to be a very sensible parameter
- concave elementary function should be carefully used, if ever
- linear optimisation gives as good results as non-linear optimisation, provided the amount of **CSO** is the predominant parameter of evaluation. But a non linear optimisation achieves a better exploitation of the available storage and transport capacity.

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<sup>2</sup>For this event, no **CSO** in KENT11, KENT12 and KENT21 occurs.

