# Innovation and Creativity

Data reconciliation and optimal operation . With applications to refinery processes

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### Introduction

- Data reconciliation and optimal operation
- Started December 1998
- Funded by Statoil
- Ph.D advisor professor Ph.D.Sigurd Skogestad
- Statoil advisor Ph.D. Stig Strand



# Thesis

- 1. Introduction
- 2. Steady state models for effective on-line applications
- 3. Data reconciliation
- 4. Data reconciliation and optimal operation of a catalytic naphtha reformer
- 5. On-line optimization of a crude unit heat exchanger network
- 6. Implementation issues for real time optimization of a crude unit heat exchanger network.
- 7. Conclusions and further work



# Data reconciliation and optimal operation



#### **Preferred properties**

- Open-equation formulation, f(z) = 0
- Equations written as unit models
- Standardization of equations
- Scaling
- Analytical first order derivatives
- Automatic generation of initial values
- Reuse of models



#### Models for effective on-line applications Data reconciliation

min

s.t. 
$$f(z) = 0$$
  
 $A_r z = b_r$   
 $z_{r \min} \le z \le z_{r \max}$ 

where 
$$J_r(z) = (Uz - y_m)^{\mathrm{T}} \Sigma^{-1} (Uz - y_m)$$



#### Models for effective on-line applications Optimization

 $\underset{Z}{\min} J_{opt}(Z)$ 

s.t. 
$$f(z) = 0$$
  
 $A_{opt}z = b_{opt}$   
 $Z_{opt \min} \le z \le Z_{opt \max}$ 

where  $J_{opt}(z) = -p^{T}z$  and  $b_{opt} = A_{opt}z_{r}$ 



 $\frac{\min}{z} J_s(z)$ 

s.t. 
$$f(z) = 0$$
  
 $A_s z = b_s$ 

where  $J_s(z) = 0$ 



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Unit model





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#### **Process model**





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Variables and equations

$$z = \begin{bmatrix} S_1 \\ S_2 \\ \vdots \\ S_7 \\ \Theta_1 \\ \Theta_2 \\ \Theta_3 \end{bmatrix}$$

$$r_i = f_i(z)$$



#### Variables and equations

$$r = f(z) = \begin{bmatrix} f_1(z) \\ f_2(z) \\ f_3(z) \end{bmatrix}$$

$$\frac{\partial f(z)}{\partial z} = \begin{bmatrix} \frac{\partial f_1(z)}{\partial z} \\ \frac{\partial f_2(z)}{\partial z} \\ \frac{\partial f_3(z)}{\partial z} \end{bmatrix}$$



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#### Models for effective on-line applications The scaled process model

 $\widetilde{f}(\widetilde{z}) = 0$  $\widetilde{A}_s \widetilde{z} = \widetilde{b}$ 

where  $\tilde{z} = S_v^{-1} z$ .

$$\begin{split} \widetilde{f}(\widetilde{z}) &= S_f f(S_v \widetilde{z}) \\ \widetilde{A}_s &= S_l A_s \\ \widetilde{J}(\widetilde{z}) &= S_o J(S_v \widetilde{z}) \end{split}$$

where  $\tilde{b} = S_l b_s$ .  $S_l$ ,  $S_f$  and  $S_v$  are fixed diagonal scaling matrices and  $S_o$  is a fixed factor.



- 1. Make a pairing of equations and variables.
- Scale all variables such that the scaled variable has a value close to one
- 3. Scale all equations such that the absolute value of the elements of the first order derivatives, corresponding to the equation and variable pairing, is close to one.
- 4. Scale the objective function such that the largest element of the first order derivative  $\tilde{J}(\tilde{z})$  has an absolute value close to one.



P(i,j) = 1 if variable number *j* is paired with equation number *i* 

$$\tilde{z} = S_v^{-1} z \approx 1$$

$$S_{f_i} = \left| \left[ I \times \left( \frac{\partial f_i(z)}{\partial z} S_v P_{n_i}^{\mathrm{T}} \right) \right]^{-1} \right|$$
$$S_o = 1/\max \left| \frac{\partial J(z)}{\partial z} S_v \right|$$
(1)





	Cond.no.	Number	of iterations		Active	
	Ĥ	Rec.	Opt.	d <sub>est</sub>	Jopt	Constraints
Unscaled	5.0E+09	23	11	2.99	-9.34	$x_{6}(1)$
Method 1	4.8E+05	14	7	0.30	-9.85	$P_4, x_6(1), Q_{HT}$
Method 2	7.0E+09	9	3	3.49	-5.84	Q <sub>HT</sub>
Method 3	4.0E+03	28	10	0.30	-9.85	$P_{4}, x_{6}(1), Q_{HT}$
New method	5.1E+01	12	5	0.30	-9.85	$P_{4}, x_{6}(1), Q_{HT}$



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#### Optimal operation of a naphtha reformer





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#### Optimal operation of a naphtha reformer

#### Octane number

The octane number is a measure of the autoignition resistance of gasoline.





#### Optimal operation of a naphtha reformer Process model





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#### Optimal operation of a naphtha reformer Nominal operation



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#### Optimal operation of a naphtha reformer

#### Data reconciliation results













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#### Optimal operation of a naphtha reformer

#### Data reconciliation results





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#### Optimal operation of a naphtha reformer Optimal operation

Description	Variable	Unit	Min.	Max.	Rec.	Case 1	Case 2	Price
Feed flow	$\tilde{F}_1$	t/h			89.2	95.6	84.1	-60.0
Reformate flow	$\tilde{F}_{52}$	t/h			84.2	90.6	79.7	**100.0
Gas flow (LPG)	$\widetilde{F}_{53}(G)$	t/h			1.2	1.0	0.9	50.0
H <sub>2</sub> mass flow	$\widetilde{F}_{53}(H)$	t/h	3.5		3.8	4.0	*3.5	
Reformat octane	RON		103.0		103.9	* 103.0	* 103.0	
Reactor 1 temp.	$T_5$	K		810.0	794.0	790.7	794.1	
Reactor 2 temp.	$T_{16}$	K		810.0	788.6	782.7	788.8	
Reactor 3 temp.	T <sub>27</sub>	K		810.0	801.2	799.9	798.8	
Reactor 4 temp.	T <sub>38</sub>	K		810.0	799.6	791.6	780.4	
Heater 1 duty	$Q_1$	MW		9.5	9.3	* 9.5	8.6	-0.015
Heater 2 duty	$Q_2$	MW		13.0	12.7	* 13.0	12.2	-0.015
Heater 3 duty	$Q_3$	MW		13.0	12.1	* 13.0	11.3	-0.015
Heater 4 duty	$Q_4$	MW		10.0	10.0	* 10.0	7.6	-0.015
Compressor duty	Ŵ	MW			0.88	0.48	0.39	-0.015
Feed H <sub>2</sub> /HC ratio	H2/HC		3.0		5.0	* 3.0	* 3.0	
Separator pres.	P <sub>53</sub>	bar	8.0	10.0	8.0	* 10.0	* 10.0	
Profit		\$/h			2638	2883	-249	

(\* = active constraint, \* \* = in case 2 the price of reformate is 65\$/t)



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## Optimal operation of a naphtha reformer

#### Control

Manipulated	Controlled variables	Controlled variables
variables	Case 1	Case 2
Feed flow	Reformate RON	H <sub>2</sub> product flow
Heater 1 duty	Maximum*	Reformate RON
Heater 2 duty	Maximum*	$T_{R1_i} - T_{R2_i} (=0)$
Heater 3 duty	Maximum*	$T_{R2_i} - T_{R3_i} (=0)$
Heater 4 duty	Maximum*	$T_{R3_i} - T_{R4_i} (=0)$
Pressure	Maximum*	Maximum*
Compressor work	H <sub>2</sub> /HC ratio	H <sub>2</sub> /HC ratio

(\* Manipulated variable fixed at maximum value)



### Summary

- Modeling framework simplifies the development of on-line models.
- Proposed scaling shows promising results, also for larger models.
- Data reconciliation and problem analysis gives useful knowledge of the measurements and process behavior.
- Proper selection of controlled variables simplifies the implementation of the optimal result.



#### Crude unit heat exchanger network





**Method 1.** Scaling based on variable bounds and initial equation residual

$$\begin{aligned} S_{v_{jj}} &= 2^{a_j} & \text{where } a_j = \inf[\log_2(z_{\max_i} - z_{\min_i})] & (2) \\ S_{n_{ji}} &= 2^{-a_i} & \text{where } a_i = \inf[\log_2(|f(z_0)|_i)] & (3) \\ S_{l_{ji}} &= 2^{-a_i} & \text{where } a_{=i} \inf[\log_2(|A_s z_0 - b_s|_i)] & (4) \end{aligned}$$

where  $z_0$  is the initial value. The equation scaling factor is limited to some maximum value in case the equation residual is close to zero.

Method 2. Scaling based on first order derivatives

$$\mathbb{C} = \begin{bmatrix} \frac{\partial f(z_0)}{\partial z} \\ A_s \end{bmatrix}$$
(5)  

$$S_{v_{jj}} = ||\mathbb{C}_j||_2^{-1} \text{ where } j = 1...n_z$$
(6)  

$$S_{n_{ij}} = ||\mathbb{C}_i||_2^{-1} \text{ where } i = 1...n_f$$
(7)  

$$S_{l_{ij}} = ||\mathbb{C}_i||_2^{-1} \text{ where } j = n_f + 1...n_f + n_s$$
(8)

where  $\mathbb{C}_i$  and  $\mathbb{C}_i$  denotes the columns and rows of  $\mathbb{C}$  respectively.



Method 3. Scaling based on order of magnitude

$$S_{v_{ij}} = 10^{-a_j} \text{ where } a_j = \text{int}[\log_{10}(z_0)_j]$$

The equation scaling factor is the reciprocal of an integer power of 10 of the value of a given term or group of terms, normally related to the scale factor of a relevant variable.

As an example, let a typical value of a mass balance term  $x_i F$  be  $0.5 \cdot 0.3 = 0.15$ . The scaling factor for the mass balance equation is then  $10^{(-int(\log_{10}(0.15)))} = 10$ . The objective scaling factor is divided by an integer power of 10 close to its typical value.

