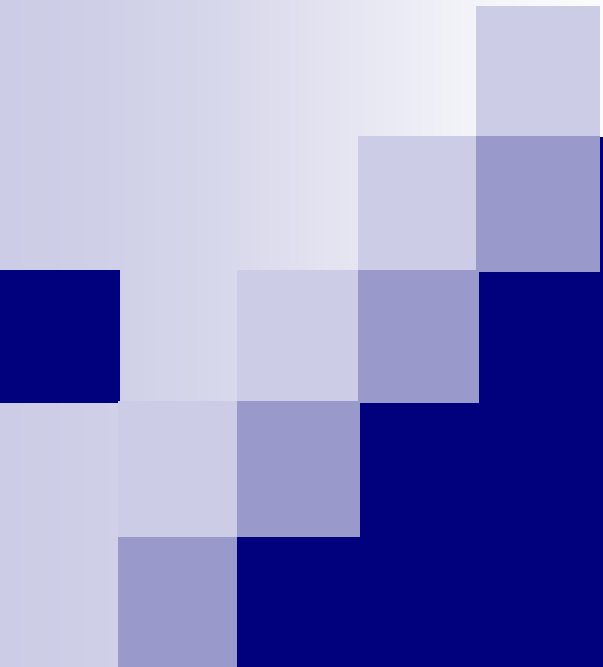


*Laboratoire d'Analyse
Appliquée et Optimisation*



Observation and Identification for Nonlinear Systems

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Introduction

$$\Sigma : \begin{cases} \frac{dx}{dt} & = f(x, u(t), \varphi \circ \pi(x(t))) \\ y & = h(x, u(t), \varphi \circ \pi(x(t))) \end{cases}$$

where $\varphi \circ \pi : X \rightarrow Z \rightarrow I \subset \mathbf{R}$
 $x \rightarrow z = \pi(x) \rightarrow \varphi(\pi(x))$

and $P_{\Sigma} : X \times L^{\infty}[U] \times L^{\infty}[I] \rightarrow L^{\infty}[\mathbf{R}^{d_y}]$
 $(x_0, u(\cdot), \hat{\varphi}(\cdot)) \rightarrow y(\cdot)$

φ is an unknown function of $\pi(x)$

$\hat{\varphi}$ is a function of time

P_{Σ} is the input/output function of Σ

Identifiability

Definition 1: Σ is identifiable at

$$(u(\cdot), y(\cdot)) \in L^\infty[U] \times L^\infty[\mathbf{R}^{d_y}]$$

if there is at most a single couple

$$(x_0, \hat{\varphi}) \in X \times L^\infty(I)$$

such that for almost all t

$$P_\Sigma(x_0, u, \hat{\varphi})(t) = y(t)$$

and $\hat{\varphi}(t) = \varphi \circ \pi(x(t))$ for some smooth function $\varphi : Z \rightarrow I$.

Σ is identifiable if it is identifiable at any admissible $(u(\cdot), y(\cdot))$.

Infinitesimal Identifiability

Definition 2:

$$T\Sigma : \begin{cases} \frac{d\xi}{dt} = T_{x,\varphi}f(x, u, \varphi; \xi, \eta) \\ \hat{y} = d_{x,\varphi}h(x, u, \varphi; \xi, \eta) \end{cases}$$

where $(\xi, \eta) \in T_x X \times T_\varphi I$, we set

$$\begin{aligned} P_{T\Sigma}^t(\xi_0, \eta) &= d_{x,\varphi}h(x, u, \hat{\varphi}; T_{x,\varphi}\phi_t(x, u, \hat{\varphi}; \xi_0, \eta), \eta) \\ &= T_{x,\varphi}P_\Sigma^t(\xi_0, \eta) \end{aligned}$$

Σ is infinitesimally identifiable at $(x_0, u, \hat{\varphi}) \in X \times L^\infty[U] \times L^\infty[I]$ if $P_{T\Sigma}^t$ is injective $\forall t > 0$

Σ is uniformly infinitesimally identifiable if this is true at all $(x_0, u, \hat{\varphi})$

Differential Identifiability

Let $D_k \Phi = X \times (U \times \mathbf{R}^{(k-1)d_u}) \times (I \times \mathbf{R}^{k-1})$ be the space of k -jets of the system Σ
 $(j^k(u) = (u(0), u'(0), \dots, u^{(k-1)}(0)))$, we set

$$\begin{aligned} \Phi_k^\Sigma : \quad & D_k \Phi && \rightarrow \mathbf{R}^{kd_y} \\ & (x_0, j^k(u), j^k(\hat{\varphi})) && \rightarrow j^k(y) \end{aligned}$$

$$\begin{aligned} \Phi_{k,2}^{\Sigma,*} : \quad & D_k \Phi \times D_k \Phi && \rightarrow \mathbf{R}^{kd_y} \times \mathbf{R}^{kd_y} \\ & (z_1, z_2) && \rightarrow (\Phi_k^\Sigma(z_1), \Phi_k^\Sigma(z_2)) \end{aligned}$$

Definition 3: Σ is differentially identifiable of order k if

$$\Phi_{k,2}^{\Sigma,*}(z_1, z_2) \in \Delta_k \Rightarrow (x_1, \hat{\varphi}_1(0)) = (x_2, \hat{\varphi}_2(0))$$

Genericity (without control)

Proposition. Differential Identifiability \Rightarrow Identifiability

Theorem 1.

- If $d_y \geq 3$, differential identifiability of order $2n + 1$ is a **generic property** in the class of C^∞ systems.
- If $d_y < 3$, differential identifiability **is not** a generic property.

Proof of genericity 1/2

$$Z_i = \left(x_i, \varphi_i, \varphi'_i, \dots, \varphi_i^k, j_{\Sigma}^k(x_i, \varphi_i) \right), \quad i = 1, 2$$

$$Z = (Z_1, Z_2)$$

$$\Phi(Z) = \Phi_{\Sigma}^k(Z_1) - \Phi_{\Sigma}^k(Z_2) \in R^{k d_y},$$

$$k = 2n + 1, \quad d_y \geq 3$$

Let us suppose that Φ is a submersion

$$\text{codim} \Phi^{-1}(0) = k d_y$$

$$\text{Let } \Pi \Phi^{-1}(0) = \left(x_i, \varphi_i, j_{\Sigma}^k(x_i, \varphi_i) \right)_{i=1,2}$$

$$\begin{aligned} \text{codim} \Pi \Phi^{-1}(0) &\geq k d_y - 2(k - 1) = k(d_y - 2) + 2 \\ &\geq k + 2 \geq 2n + 3 \end{aligned}$$

Proof of genericity 2/2

$$\begin{aligned} \rho_\Sigma : (X \times I)^2 \setminus \Delta &\rightarrow (J_\Sigma^k)^2 \\ (x_1, \varphi_1, x_2, \varphi_2) &\rightarrow (x_i, \varphi_i, j_\Sigma^k(x_i, \varphi_i))_{i=1,2} \end{aligned}$$

Multijet transversality theorem: the set of Σ such that ρ_Σ is transversal to $\Pi\Phi^{-1}(0)$ is residual.

$$\begin{aligned} \dim (X \times I)^2 \setminus \Delta &= 2n + 2 \\ &\Downarrow \\ \text{generically, } \rho_\Sigma &\text{ avoids } \Pi\Phi^{-1}(0) \end{aligned}$$

Single-output case

Theorem 2. If Σ is uniformly infinitesimally identifiable then

i) $\frac{\partial}{\partial \varphi} \{h, L_{f_\varphi} h, \dots, (L_{f_\varphi})^{n-1} h\} \equiv 0$

ii) $\frac{\partial}{\partial \varphi} L_{f_\varphi}^n h \neq 0$

iii) $d_x h \wedge \dots \wedge d_x L_{f_\varphi}^{n-1} h \neq 0,$

Therefore, locally, the system can be written

$$\begin{cases} \dot{x}_1 & = & x_2 \\ & \vdots & \\ \dot{x}_{n-1} & = & x_n \\ \dot{x}_n & = & \psi(x, \varphi) \\ y & = & x_1 \end{cases} \quad \text{and} \quad \frac{\partial}{\partial \varphi} \psi(x, \varphi) \neq 0$$

Single-output case, pseudo-converse

Theorem 3. If Σ meets the following conditions,

$$\text{i) } \frac{\partial}{\partial \varphi} \left\{ h, L_{f_\varphi} h, \dots, (L_{f_\varphi})^{n-1} h \right\} \equiv 0$$

$$\text{ii) } \frac{\partial}{\partial \varphi} L_{f_\varphi}^n h \neq 0$$

$$\text{iii) } dxh \wedge \dots \wedge dx L_{f_\varphi}^{n-1} h \neq 0,$$

then Σ is

- 1) locally identifiable,
- 2) loc. unif. infinitesimally identifiable,
- 3) loc. diff. identifiable of order $n + 1$.

Proof of the single-output case 1/2

Let $k < n$ be the first k such that $d_\varphi L_f^k h \neq 0$:

$$\Sigma \quad \begin{cases} y & = x_1 \\ \dot{x}_1 & = x_2 \cdots \\ \dot{x}_{k-1} & = x_k \\ \dot{x}_k & = L_f^k(x, \varphi) = f_k(x, \varphi) \cdots \\ \dot{x}_n & = f_n(x, \varphi) \end{cases}$$
$$T\Sigma \quad \begin{cases} \dot{x} & = f(x, \varphi) \\ \hat{y} & = \xi_1 \\ \dot{\xi}_1 & = \xi_2 \cdots \\ \dot{\xi}_{k-1} & = \xi_k \\ \dot{\xi}_k & = d_x f_k(x, \varphi) \xi + d_\varphi f_k(x, \varphi) \eta \end{cases}$$

Proof of the single-output case 2/2

A feedback $\eta = -\frac{d_x f_k(x, \varphi_0) \xi}{d_\varphi f_k(x, \varphi_0)}$ in φ_0 s.t. $d_\varphi f_k(x, \varphi_0) \neq$

0 gives $\frac{d\xi_k}{dt} = 0$ which contradict observability.

If $\frac{\partial}{\partial \varphi} L_{f_\varphi}^n h = 0$ at (x, φ)

$$X \times I \supset E = \{(x, \varphi); d_\varphi L_{f_\varphi}^n h = 0\}$$

$$\downarrow \Pi$$

$$X \supset \Pi E$$

Hardt's theorem $\Rightarrow \exists \hat{\varphi}$

$$\begin{cases} y = x_1, & \dot{x}_1 = x_2, & \dots & \dot{x}_n = \psi(x, \hat{\varphi}(x)) \\ \hat{y} = \xi_1, & \dot{\xi}_1 = \xi_2, & \dots & \dot{\xi}_n = d_x \psi(x, \hat{\varphi}(x)) + 0 \end{cases}$$

Two output case: definitions of k and r

Define $E_l = \{d_x h_i, d_x L_{f_\varphi} h_i, \dots, d_x L_{f_\varphi}^{l-1} h_i, i = 1, 2\}$
 and $N(l) = \text{rank}(E_l)$ at a generic point:

k is defined by

$N(0)$	$N(1)$	\dots	$N(k-1)$	$N(k)$	$N(k+1)$	\dots	$N(k+m)$
0	2		$2k-2$	$2k$	$2k+1$		$2k+m$

$$(2k + m \leq n)$$

The **order** of the system is the first integer r
 such that $d_\varphi L_{f_\varphi}^r (h_1, h_2) \neq 0$.

Classification

Lemma: If Σ is uniformly infinitesimally identifiable then (1) $2k + m = n$

$$(2) \quad r \leq k + m$$

Proof:

$$(1) \quad \varphi = \varphi_0 = \text{cte} \quad \begin{cases} \dot{x} = f(x, \varphi_0) \\ \dot{\xi} = g(x, \xi, \varphi_0) \\ y = h(x, \varphi_0) \end{cases} \quad \begin{array}{l} \text{contradict} \\ \text{observability} \end{array}$$

$$(2) \quad \begin{cases} \dot{x} = f(x) \\ y = h(x) \end{cases} \quad \text{contradict identifiability}$$

Définition 5. A system Σ is regular if (1) and (2) holds.

Type 3: $r=k$ and $n=2k$

$$\left\{ \begin{array}{ll} y_1 & = x_1 \\ \dot{x}_1 & = x_3 \\ & \vdots \\ \dot{x}_{n-3} & = x_{n-1} \\ \dot{x}_{n-1} & = f_{n-1}(x, \varphi) \end{array} \quad \begin{array}{ll} y_2 & = x_2 \\ \dot{x}_2 & = x_4 \\ & \vdots \\ \dot{x}_{n-2} & = x_n \\ \dot{x}_n & = f_n(x, \varphi) \end{array} \right.$$

with $\frac{\partial}{\partial \varphi}(f_{n-1}, f_n) \neq 0$

$N(l)$ increases by steps of 2 until the last derivative and apparition of φ .

Type 1: $r > k$

$$\left\{ \begin{array}{llll} y_1 & = & x_1 & y_2 & = & x_2 \\ \dot{x}_1 & = & x_3 & \dot{x}_2 & = & x_4 \\ & & \vdots & & & \vdots \\ \dot{x}_{2k-3} & = & x_{2k-1} & \dot{x}_{2k-2} & = & x_{2k} \\ \dot{x}_{2k-1} & = & f_{2k-1}(x_1, \dots, x_{2k+1}) & & & \\ \dot{x}_{2k} & = & x_{2k+1} & & & \\ & & \vdots & & & \\ \dot{x}_{n-1} & = & x_n & & & \\ \dot{x}_n & = & f_n(x, \varphi) & & & \end{array} \right.$$

with $\frac{\partial f_n}{\partial \varphi} \neq 0$.

$N(l)$ increases by steps of 1 when φ appears for the first time, \simeq single-output case.

Type 2: $r < k$

$$\begin{array}{llll} y_1 & = & x_1 & y_2 & = & x_2 \\ \dot{x}_1 & = & x_3 & \dot{x}_2 & = & x_4 \\ & & \vdots & & & \vdots \\ \dot{x}_{2r-3} & = & x_{2r-1} & \dot{x}_{2r-2} & = & x_{2r} \\ \dot{x}_{2r-1} & = & \psi(x, \varphi) & \dot{x}_{2r} & = & F_{2r}(x_1, \dots, x_{2r+1}, \psi(x, \varphi)) \\ & & & \dot{x}_{2r+1} & = & F_{2r+1}(x_1, \dots, x_{2r+2}, \psi(x, \varphi)) \\ & & & & & \vdots \\ & & & \dot{x}_{n-1} & = & F_{n-1}(x, \psi(x, \varphi)) \\ & & & \dot{x}_n & = & F_n(x, \varphi) \end{array}$$

with $\frac{\partial \psi}{\partial \varphi} \neq 0, \frac{\partial F_{2r}}{\partial x_{2r+1}} \neq 0, \dots, \frac{\partial F_{n-1}}{\partial x_n} \neq 0$

φ appears when $N(l)$ increases by steps of 2.

$r=k$ and $2k < n$

If $r = k$ before the last derivative:

$$d_x h_1 \wedge \cdots \wedge d_x L_{f_\varphi}^{k-1} h_1 \wedge d_x L_{f_\varphi}^{k-1} h_2 \wedge d_x L_{f_\varphi}^k h_2 \neq 0$$

If $d_\varphi L_{f_\varphi}^k h_1 \neq 0$, we obtain φ using y_1 and x_{2k}, \dots, x_n using y_2


→ Type 2

If $d_\varphi L_{f_\varphi}^k h_1 \equiv 0$, we obtain φ using y_2

→ Type 1

Canonical form for observer construction

$$\left\{ \begin{array}{l} \dot{x}_1 = F_1(x_1, x_2, u) \quad \frac{\partial F_1}{\partial x_2} \neq 0 \\ \dot{x}_2 = F_2(x_1, x_2, x_3, u) \quad \frac{\partial F_2}{\partial x_3} \neq 0 \\ \vdots \\ \dot{x}_n = F_n(x, u) \end{array} \right. \quad \begin{array}{l} \xi_1 = y = x_1, \quad \xi_2 = F_1(x_1, x_2, u) \\ \xi_3 = \frac{\partial F_1}{\partial x_2} F_2(x_1, x_2, u), \dots \\ \xi_{i+1} = \frac{\partial F_1}{\partial x_2} \dots \frac{\partial F_{i-1}}{\partial x_i} F_i(x_1, \dots, x_{i+1}, u) \end{array}$$



$$\left\{ \begin{array}{l} \dot{\xi}_1 = \xi_2 \\ \dot{\xi}_2 = \xi_3 + \frac{\partial F_1}{\partial x_1} \dot{x}_1 + \frac{\partial F_1}{\partial u} \dot{u} \\ \vdots \\ \dot{\xi}_n = G(x, u, \dot{u}) \end{array} \right.$$

Ref. H. Hammouri, M. Farza, *Nonlinear observers for local uniform observable systems*, to appear

Canonical form of observability

$$\begin{cases} \frac{dx}{dt} = A(t)x + b(x, u) \\ y = C(t)x \end{cases}$$

$$A(t) = \begin{pmatrix} 0 & a_2(t) & 0 & \cdots & 0 \\ & & a_3(t) & \ddots & \vdots \\ \vdots & & \ddots & \ddots & 0 \\ & & & & a_n(t) \\ 0 & \cdots & & & 0 \end{pmatrix}$$

$$C(t) = \begin{pmatrix} a_1(t) & 0 & \cdots & 0 \end{pmatrix}$$

$$0 < a_m \leq a_i(t) \leq a_M$$

$$b(x, u) = b_1(x_1, u) \frac{\partial}{\partial x_1} + b_2(x_1, x_2, u) \frac{\partial}{\partial x_2} + \cdots + b_n(x_1, \dots, x_n, u) \frac{\partial}{\partial x_n}$$

Modified Extended Kalman filter

$$\begin{aligned}\frac{dz}{dt} &= A(t)z + b(z, u) - S(t)^{-1}C(t)'r^{-1}(C(t)z - y(t)) \\ \frac{dS}{dt} &= -(A(t) + b^*(z, u))'S - S(A(t) + b^*(z, u)) \\ &\quad + C(t)'r^{-1}C(t) - SQ_{\theta}S\end{aligned}$$

$$\Delta = \begin{pmatrix} 1 & & & \\ & \frac{1}{\theta} & & \\ & & \dots & \\ & & & (\frac{1}{\theta})^{n-1} \end{pmatrix} \quad Q_{\theta} = \theta^2 \Delta^{-1} Q \Delta^{-1}$$

If θ is large, high-gain observer (HGEKF)

If $\theta \approx 1$, Classical Extended Kalman filter (EKF)

Modified Extended Kalman filter

$$\begin{aligned}\frac{dz}{dt} &= A(t)z + b(z, u) - S(t)^{-1}C(t)'r^{-1}(C(t)z - y(t)) \\ \frac{dS}{dt} &= -(A(t) + b^*(z, u))'S - S(A(t) + b^*(z, u)) \\ &\quad + C(t)'r^{-1}C(t) - SQ_{\theta}S\end{aligned}$$

$$\frac{d\theta}{dt} = \lambda(1 - \theta)$$

$$\Delta = \begin{pmatrix} 1 & & & \\ & \frac{1}{\theta} & & \\ & & \dots & \\ & & & (\frac{1}{\theta})^{n-1} \end{pmatrix}$$

$$Q_{\theta} = \theta^2 \Delta^{-1} Q \Delta^{-1}$$

If θ is large, high-gain observer (HGEKF)

If $\theta \approx 1$, Classical Extended Kalman filter (EKF)

Theorem

There exist $\lambda_0 > 0$ such that for any $0 \leq \lambda \leq \lambda_0$, there exist θ_0 such that for any $\theta(0) > \theta_0$, for any $S(0) \geq c Id$, for any compact $K \subset \mathbf{R}^n$, for any $z(0) \in K$ then if we set $\varepsilon(t) = z(t) - x(t)$ for any $t \geq 0$

$$\|\varepsilon(t)\|^2 \leq R(\lambda, c)e^{-at} \Lambda(\theta(0), t, \lambda) \|\varepsilon(0)\|^2 \quad (1)$$

where

$$\Lambda(\theta(0), t, \lambda) = \theta(0)^{2(n-1) + \frac{a}{\lambda}} e^{-\frac{a}{\lambda}\theta(0)(1-e^{-\lambda t})}$$

and a is a positive constant and $R(\lambda, c)$ is a decreasing function of c .

Proof

$$\text{Change of variables } \begin{cases} \tilde{x} & = \Delta x \\ \tilde{P} & = \frac{1}{\theta} \Delta P \Delta \end{cases} \quad (P = S^{-1})$$

+ time change $d\tau = \theta(t) dt$

We set $\varepsilon = z - x = \text{error}$ then we calculate $T_{\varepsilon}(\tau) S(\tau) \varepsilon(\tau)$.

Observability give us $\alpha I \leq S(\tau) \leq \beta I$ then

$$T_{\varepsilon}(\tau) S(\tau) \varepsilon(\tau) \longrightarrow 0 \iff \varepsilon(\tau) \longrightarrow 0$$

When $\tau \leq T$

$$\|\varepsilon(\tau)\|^2 \leq \theta(\tau)^{2(n-1)} H(c) e^{-(a_1\theta(T)-a_2)\tau} \|\varepsilon(0)\|^2$$

Parallel high-gain and non-high-gain EKF

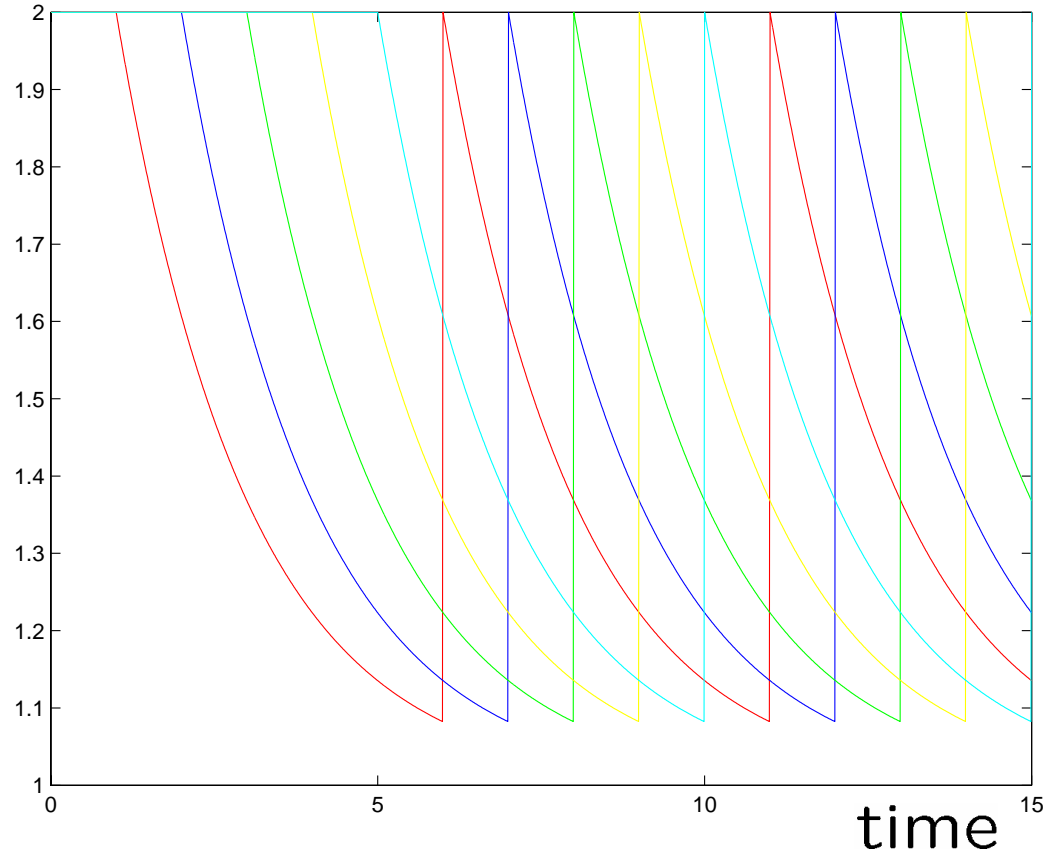
We use N observers in parallel. At times kT :

- a new observer is initialized with $\theta(kT) = \theta_0$,
- the older observer is killed.

Therefore, at any time t , we have N observers initialized at times $kT, (k-1)T \dots (k-N+1)T$ where $k = \lfloor \frac{t}{T} \rfloor$.

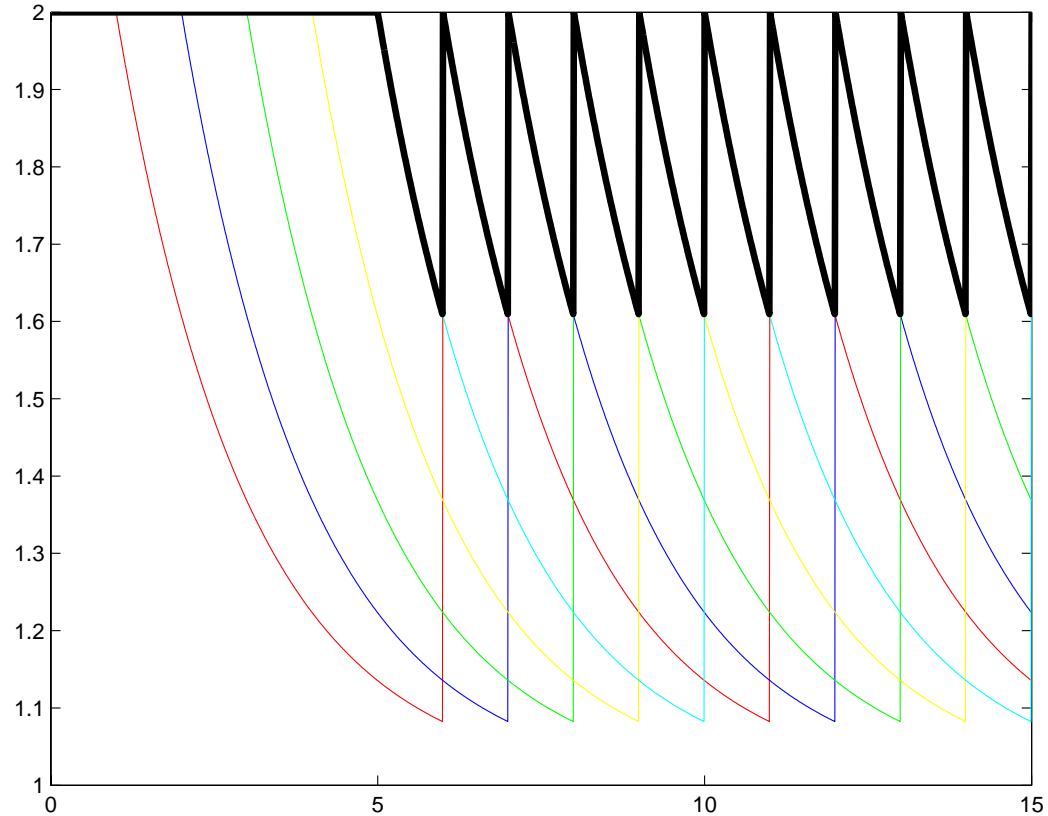
State estimation: the estimation given by the observer with smallest innovation $\|y - C\hat{x}\|$.

Parallel Extended Kalman Filter



θ for 5 observers

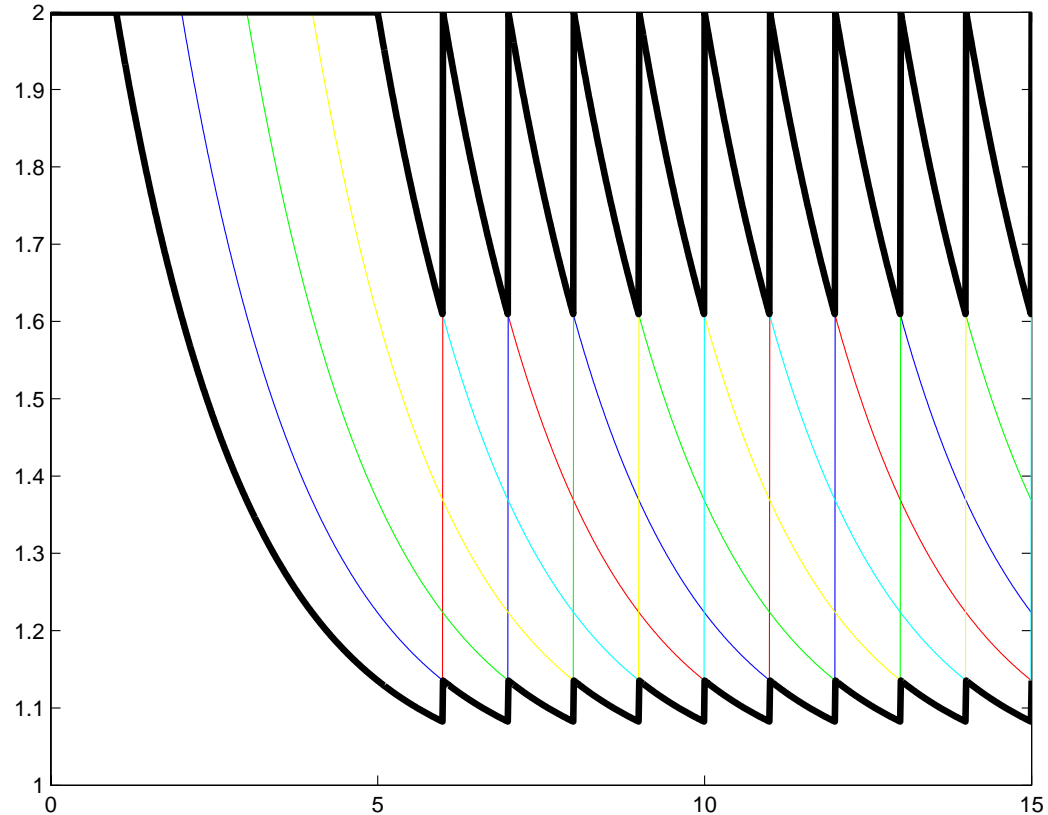
High-gain Extended Kalman Filter



θ for the youngest observer is

$$1 + e^{-\lambda(t-kT)} (\theta_0 - 1) \geq 1 + e^{-\lambda T} (\theta_0 - 1)$$

Standard Extended Kalman Filter



θ for the oldest observer is

$$1 + e^{-\lambda(t-kT+(N+1)T)} (\theta_0 - 1) \approx 1$$

Extended Kalman filtering equations

$$\frac{dx}{dt} = F(x, u)$$

Our diffeomorphism $\xi = \varphi(x, u)$ depend on u supposed to be smooth, hence:

$$\begin{aligned} \frac{d\xi}{dt} &= D_{\varphi}(\varphi_u^{-1}(\xi)) f(\varphi_u^{-1}(\xi), u) + \frac{\partial \varphi(\varphi_u^{-1}(\xi), u)}{\partial u} \dot{u} \\ &= F(\xi, u, \dot{u}) \end{aligned}$$

$$\begin{cases} \frac{d\hat{\xi}}{dt} = F(\hat{\xi}, u, \dot{u}) + PC^T R^{-1} (y - C\hat{\xi}) \\ \frac{dP}{dt} = F^*(\hat{\xi}, u, \dot{u}) P + PF^*(\hat{\xi}, u, \dot{u}) + Q_{\theta} - PC^T R^{-1} CP \end{cases}$$

In the original coordinates

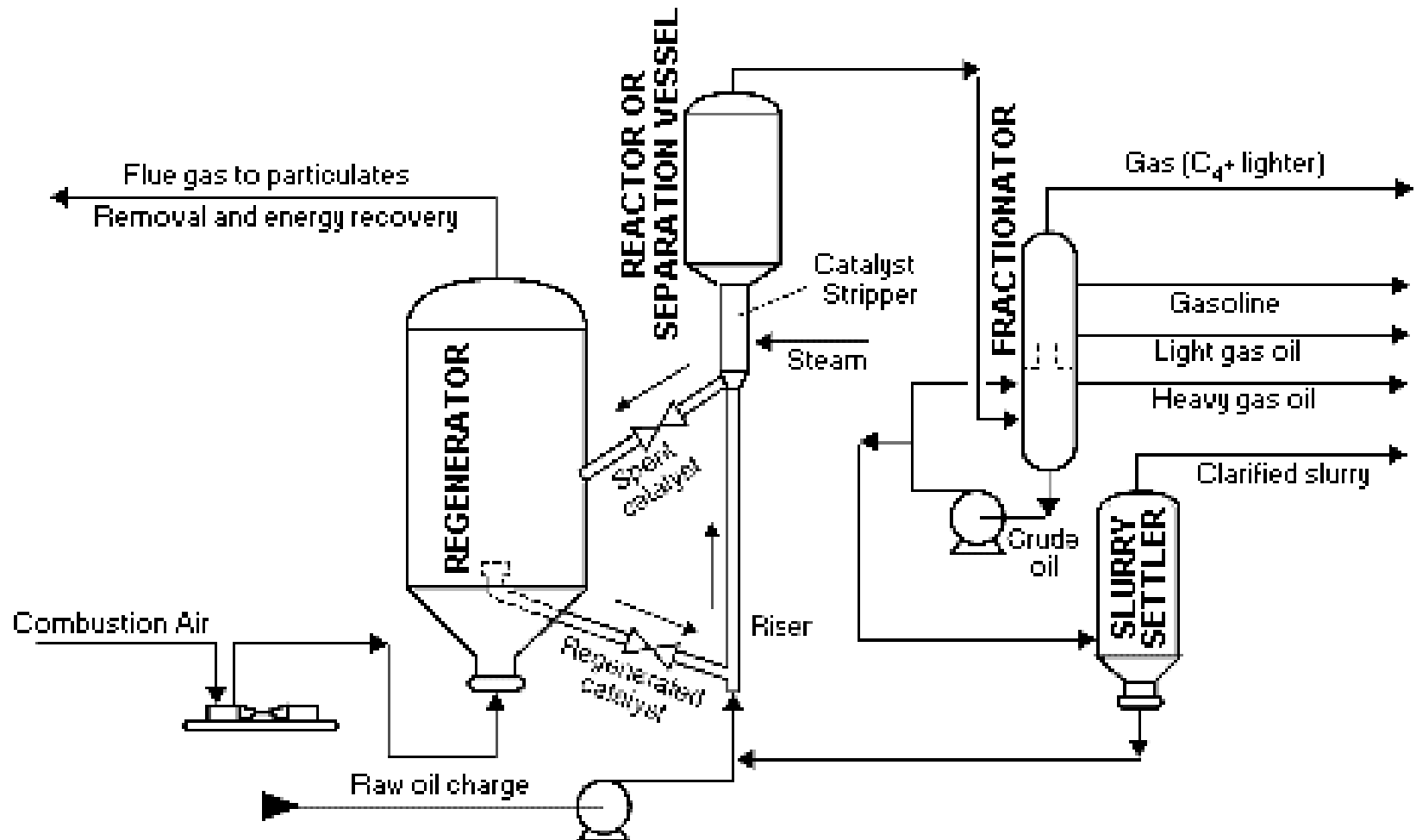
Since $C\varphi_u(x) = Cx$, equations are those of a modified extended Kalman filter

$$\left\{ \begin{array}{l} \frac{d\hat{x}}{dt} = f(\hat{x}, u) + pC^T R^{-1} (y - C\hat{x}) \\ \frac{dp}{dt} = f^*(\hat{x}, u)p + pf^*(\hat{x}, u)^T + q_\theta(\hat{x}) \\ \quad - ph^*(\hat{x}, u)^T R^{-1} h^*(\hat{x}, u)p \\ \quad + D_{\psi_u}^{-1}(\hat{x}) D_{\psi_u}^2 \cdot \left(ph^*(\hat{x}, u)^T R^{-1} (h(\hat{x}, u) - y) \right) p \\ \quad + p D_{\psi_u}^2 \cdot \left(ph^*(\hat{x}, u)^T R^{-1} (h(\hat{x}, u) - y) \right) D_{\psi_u}^{-1}(\hat{x})^T \end{array} \right.$$

where $q_\theta(\hat{x}) = D_{\varphi_u}(\hat{x})^{-1} Q_\theta \left(D_{\varphi_u}(\hat{x})^{-1} \right)^T$

The two last lines (transposed) correspond to the change of coordinate.

Application: Fluid Catalytic Cracker (FCC)



Reactor model

$$S_c H_{ra} \dot{T}_{ra} = S_c R_c (T_{rg} - T_{ra}) + S_{tf} R_{tf} (T_{tf} - T_{ra}) - \Delta H_{fv} R_{tf} - \Delta H_{cr} R_{tf} C_{tf}$$

$$C_{tf} = \frac{1}{1 + \frac{R_{tf}}{R_{cr}}} \quad R_{cr} = K_{cr} P_{ra} H_{ra}$$

$$C_{cat}^2 = \frac{100 P_{ra} H_{ra}}{R_c C_{rc}^{0.06}} k_{cc} \exp\left(-\frac{A_{cc}}{RT_{ra}}\right) \quad K_{cr} = \frac{k_{cr}}{C_{cat} C_{rc}^{0.15}} \exp\left(-\frac{A_{cr}}{RT_{ra}}\right)$$

$$H_{ra} \dot{C}_{sc} = R_c (C_{rc} - C_{sc}) + R_{cf}$$

$$R_{cf} = R_{cc} + R_{ad} \quad R_{ad} = F_{cf} R_{tf} \quad R_{cc} = K_{cc} P_{ra} H_{ra}$$

$$K_{cc} = \frac{k_{cc}}{C_{cat} C_{rc}^{0.06}} \exp\left(-\frac{A_{cc}}{RT_{ra}}\right)$$

Regenerator model

$$S_c H_{rg} \dot{T}_{rg} = S_c R_c (T_{ra} - T_{rg}) + S_a R_{ai} (T_{ai} - T_{rg}) + \Delta H_{rg} R_{cb}$$

$$R_{cb} = \frac{R_{ai}}{242} (21 - O_{fg}) \quad O_{fg} = 21 \exp \left(\frac{-\frac{P_{rg} H_{rg}}{R_{ai}}}{\frac{1}{K_{od}} + \frac{1}{K_{or} C_{rc}}} \right)$$

K_{or} = unknown function of T_{rg} .

$$K_{od} = 6.34 \cdot 10^{-9} R_{ai}^2$$

$$H_{rg} \dot{C}_{rc} = R_c (C_{sc} - C_{rc}) - R_{cb}$$

New form of the system

$$\begin{aligned}\varphi(x; u) &= \varphi(T_{rg}, T_{ra}, C_{rc}, C_{sc}, F_{cf}; R_{ai}) \\ &= \left(T_{rg}, T_{ra}, C_{tf}(C_{rc}, T_{ra}), \frac{C_{sc}}{C_{rc}}, \frac{F_{cf}}{C_{rc}} \right) = \xi\end{aligned}$$

$$\left\{ \begin{array}{l} \dot{x}_1 = \dot{T}_{rg} = \psi(x, \varphi(x_1), u) \\ \dot{x}_2 = \dot{T}_{ra} = a_3(t)x_3 + f_2(x_1, x_2) \\ \dot{x}_3 \simeq \dot{C}_{rc} = a_4(t)x_4 \\ \quad \quad \quad + f_3(x_1, x_2, x_3, \psi(x, \varphi(x_1), u), u, \dot{u}) \\ \dot{x}_4 \simeq \dot{C}_{sc} = a_5(t)x_5 + f_4(x_1, x_2, x_3, x_4) \\ \dot{x}_5 \simeq \dot{F}_{cf} = F(x) \end{array} \right.$$

Type 2

Here, $\psi = R_{cb}$, $\varphi = K_{or}$ and $\pi(x) = T_{rg} = x_1$.
 $u = (R_{ai}, P_{ra})$

Tuning

We use a second order system to estimate K_{or}

$$\text{i.e. } \frac{d^3 K_{or}}{dt^3} = 0$$

We use three parallel extended Kalman filters such that

- $\theta_0 = 3$ (starting value for each observers)
- $\theta_{HG} = 2$ (minimal value of θ ensuring high-gain)
- Time between two consecutive initializations:
2 hours

EKF with two outputs

At last,

$$\xi = \left(T_{rg}, R_{cb}, \dot{K}_{or}, \ddot{K}_{or}, T_{ra}, C_{tf}, \frac{C_{sc}}{C_{rc}}, \frac{F_{cf}}{C_{rc}} \right)$$

and

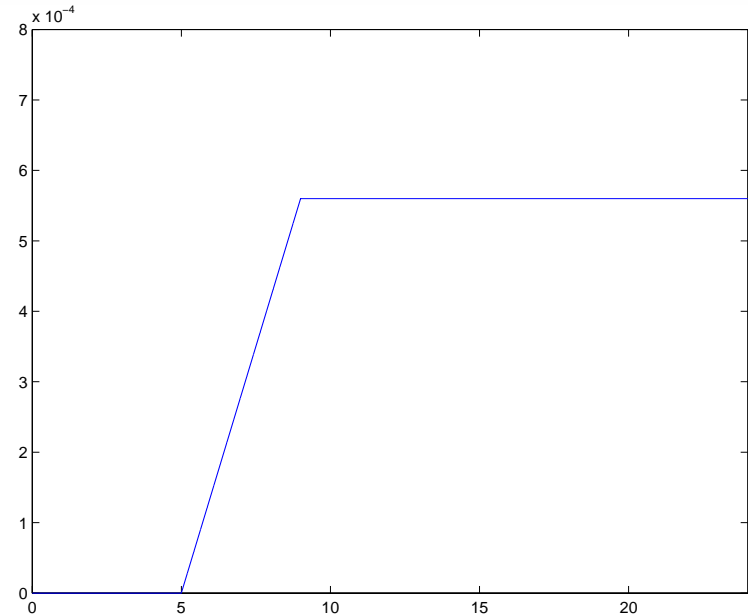
$$\Delta^{-1} = \text{diag} \left(1, \theta, \theta^2, \theta^3, 1, \theta, \theta^2, \theta^3 \right)$$

with $Q_\theta = \theta^2 \Delta^{-1} Q \Delta^{-1}$

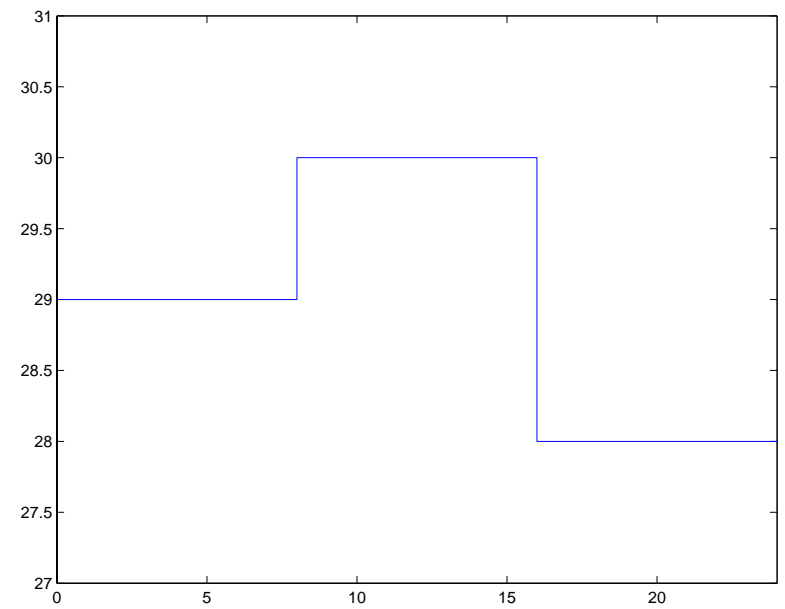
and $R_\theta = \left(C \Delta^{-1} C' \right) R \left(C \Delta^{-1} C' \right)$

Simulations

Colored noise (Ornstein–Uhlenbeck process) on both measured variables T_{rg} and T_{ra} .



Unknown parameter F_{cf}



Control variable R_{ai}

$$K_{or} = 1.16 \cdot 10^{-5} \exp \left(\frac{A_{or}}{R \left(\frac{1}{866.7} - \frac{1}{T_{rg}} \right)} \right)$$

Constants of the model

Reactor operating conditions

$$H_{ra} = 1.85 \cdot 10^{-4}, P_{ra} = 211.7,$$

Feed properties

$$R_{tf} = 41, T_{tf} = 492.8, S_{tf} = 3140,$$

Cat.recirculation

$$R_c = 290, S_c = 1047,$$

Heat constants

$$\Delta H_{cr} = 4.65 \cdot 10^5, \Delta H_{fv} = 1.74 \cdot 10^5,$$

$$\Delta H_{rg} = 3.02 \cdot 10^7, R = 8.314$$

$$k_{cr} = 8.31 \cdot 10^{-2}, A_{cr} = 6.28 \cdot 10^4,$$

$$k_{cc} = 2.66 \cdot 10^{-4}, A_{cc} = 4.18 \cdot 10^4$$

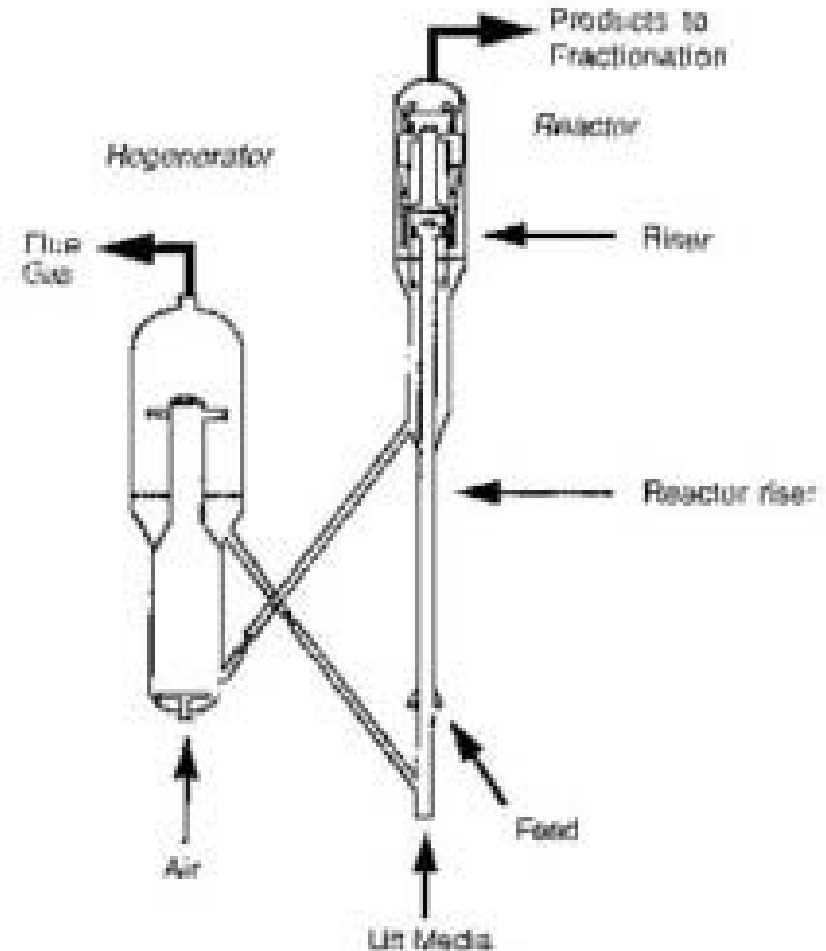
Regenerator operating conditions

$$H_{rg} = 1.53 \cdot 10^5, P_{rg} = 254.4,$$

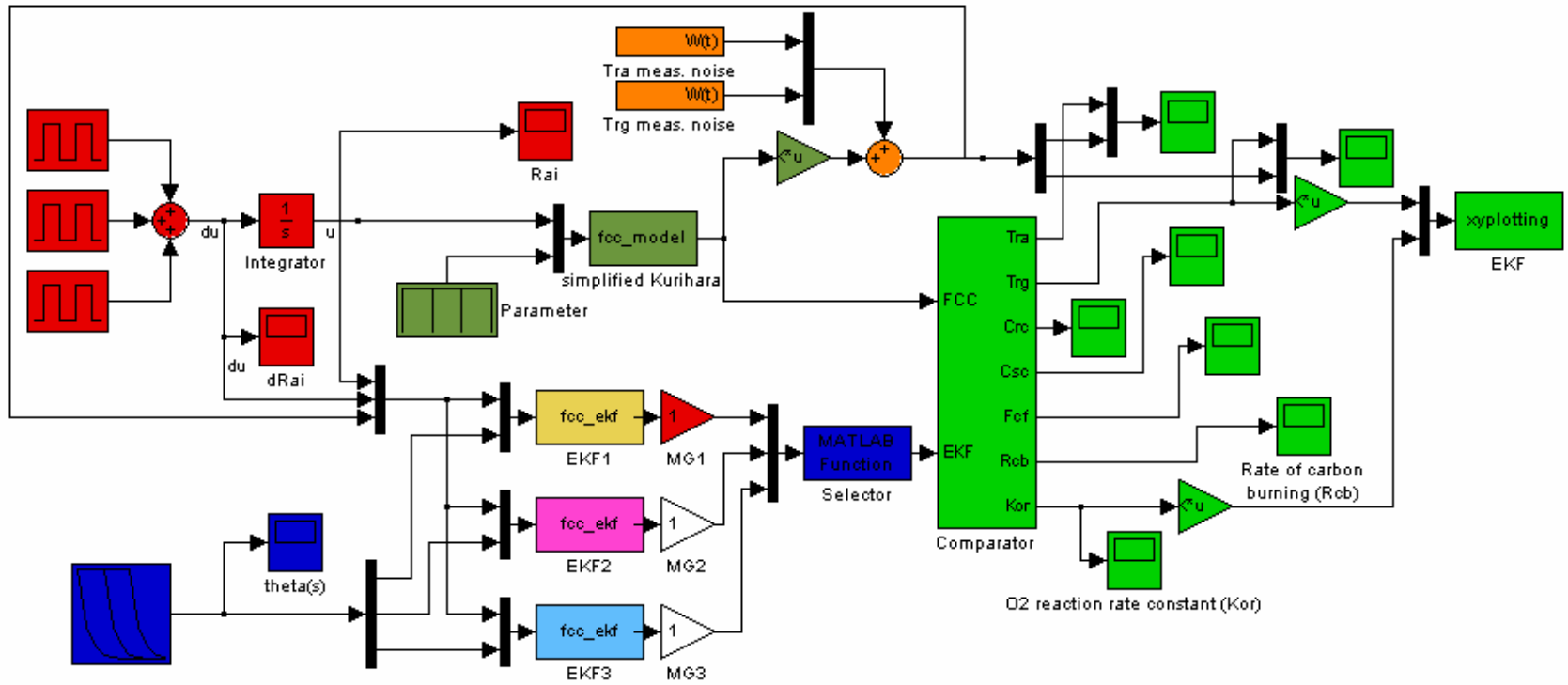
Air properties

$$R_{ai} = 26, T_{ai} = 394, S_{ai} = 1130$$

$$A_{or} = 1.47 \cdot 10^5$$



Simulation using Matlab/Simulink



to be continued...