



# A NEW LOOK AT OBSERVERS

**Jan C. Willems, K.U. Leuven, Belgium**

**Seminar, Southampton**

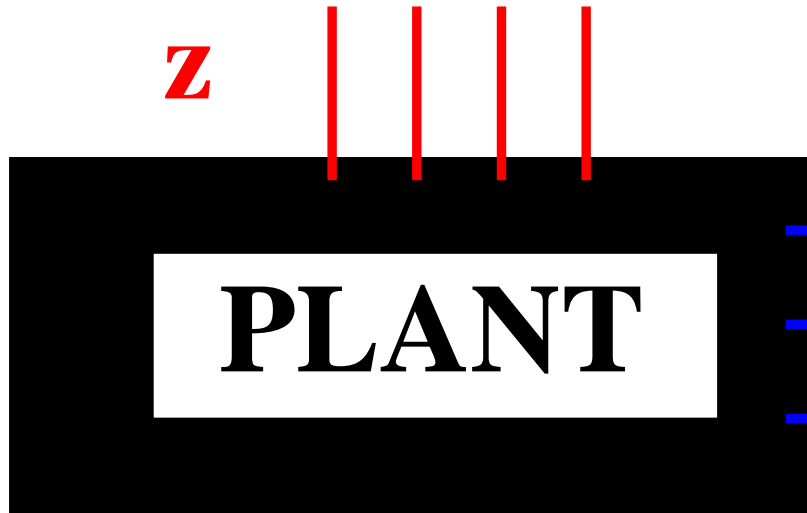
**November 22, 2007**

# Introduction

**Theme**

**to-be-estimated  
variables**

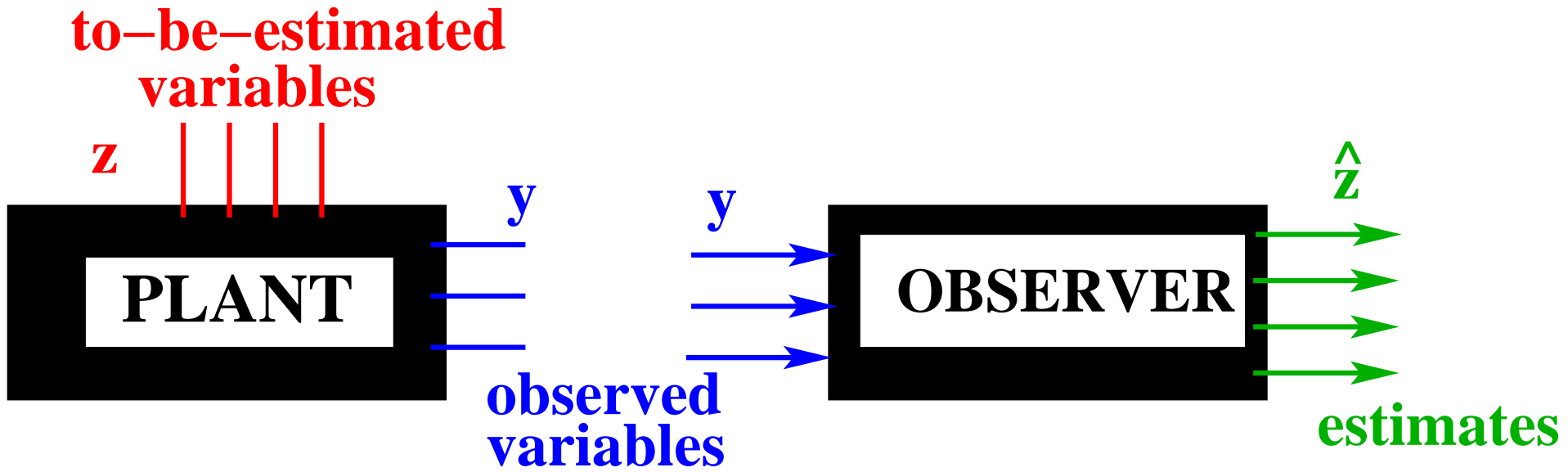
**$z$**



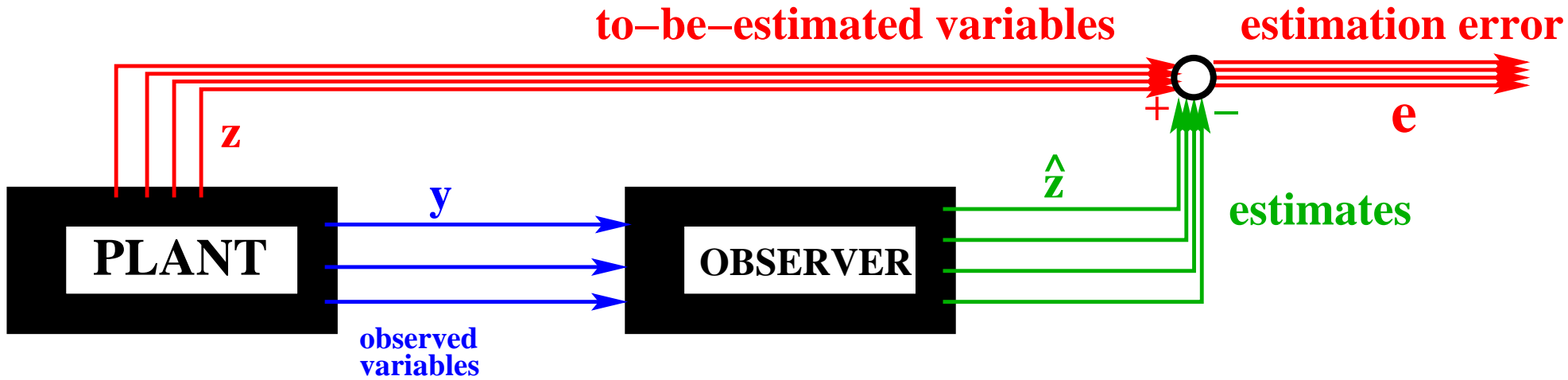
**$y$**

**observed  
variables**

**Theme**



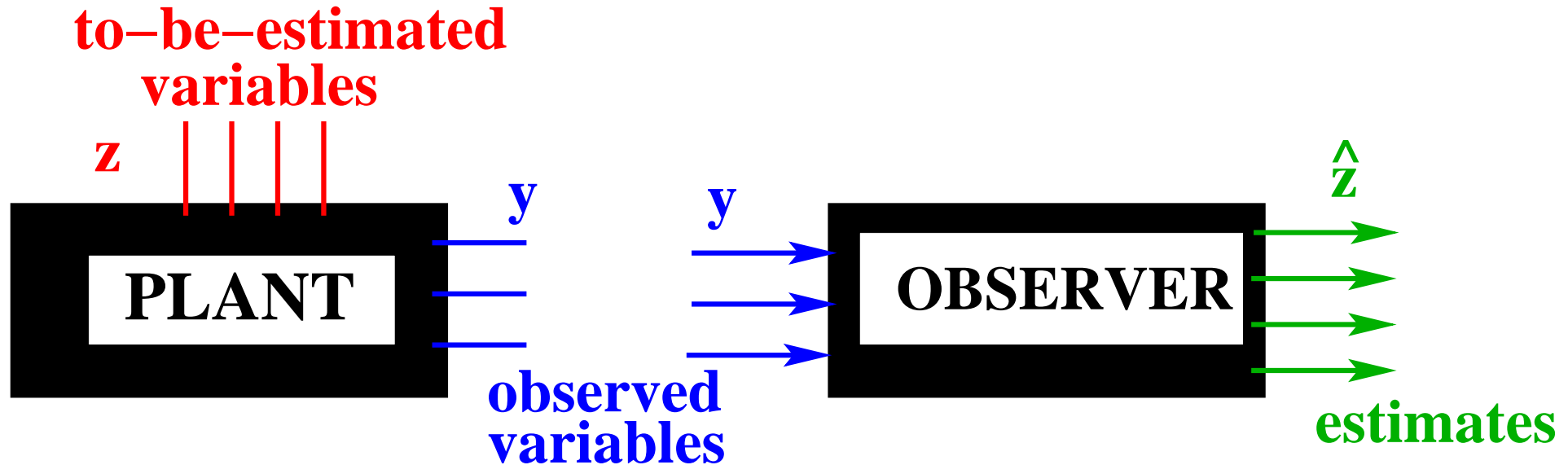
# Theme



**!! Keep estimation error**

**small, zero, convergent to zero, ... !!**

## Theme



- What is the model that relates the observed with the to-be-estimated variables ?
- Find the observer/filter algorithm !

## Joint Work with



**Maria Elena Valcher**  
**Università di Padova**

## Joint Work with



**Jochen Trumpf**  
**Australian National University**



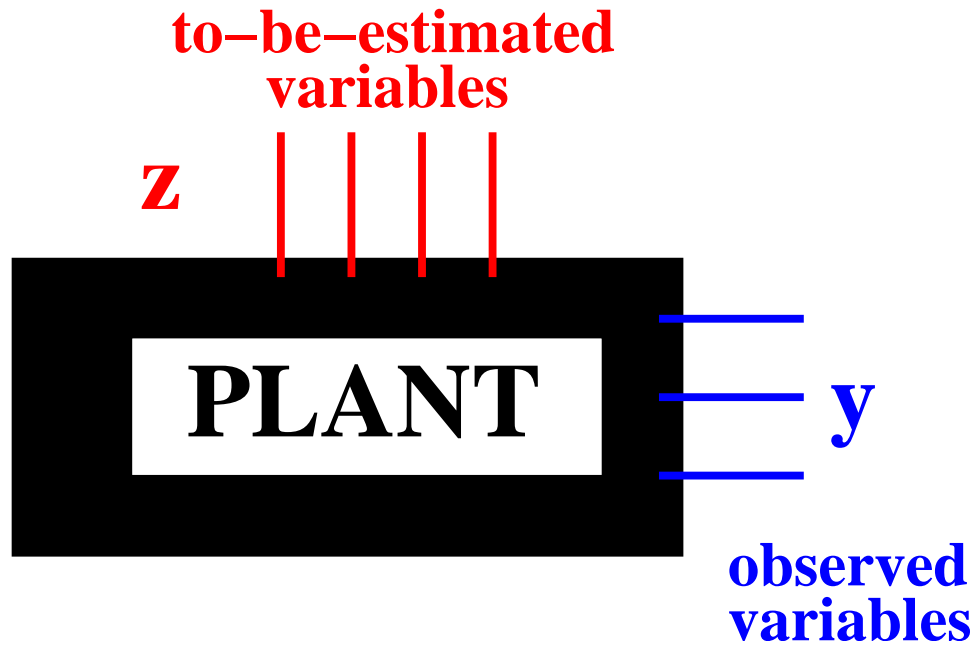
## Message

**Observers mean more**

**Controllers mean less**

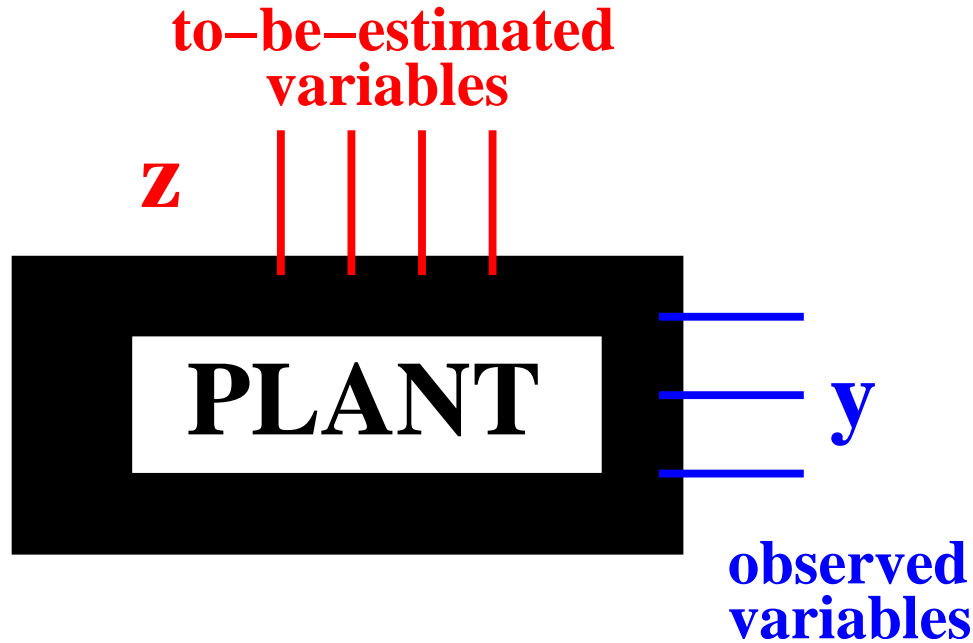
# History

# Wiener Filtering



**Modeling** : Assume  $\begin{bmatrix} y \\ z \end{bmatrix}$  is a **stochastic process**

# Wiener Filtering

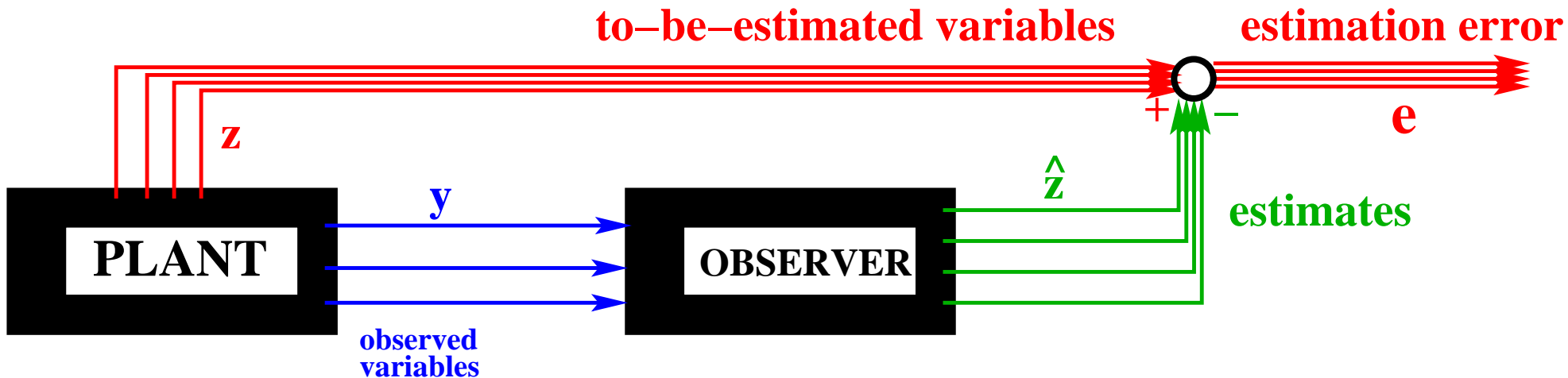


**Modeling**: Assume  $\begin{bmatrix} y \\ z \end{bmatrix}$  is a **stochastic process**

Henceforth time-set  $\mathbb{R}$ , stationary processes, normal, zero mean

# Wiener Filtering

**Modeling** : Assume  $\begin{bmatrix} y \\ z \end{bmatrix}$  is a **stochastic process**



**Estimation criterion:**  $\mathbb{E} (|e(t)|^2) \rightsquigarrow$  **algorithms**

## Wiener Filtering

Algorithms are easy to obtain if for the estimate  $\hat{z}(t)$ ,  
the observations  $y(t')$

are available **for all**  $t' \in \mathbb{R}$ .

Much much harder if the observations  $y(t')$

are available **only for**  $t' \leq t$

~> **non-anticipating filter**

## Wiener Filtering

Algorithms are easy to obtain if for the estimate  $\hat{z}(t)$ ,  
the observations  $y(t')$

are available **for all**  $t' \in \mathbb{R}$ .

Much much harder if the observations  $y(t')$

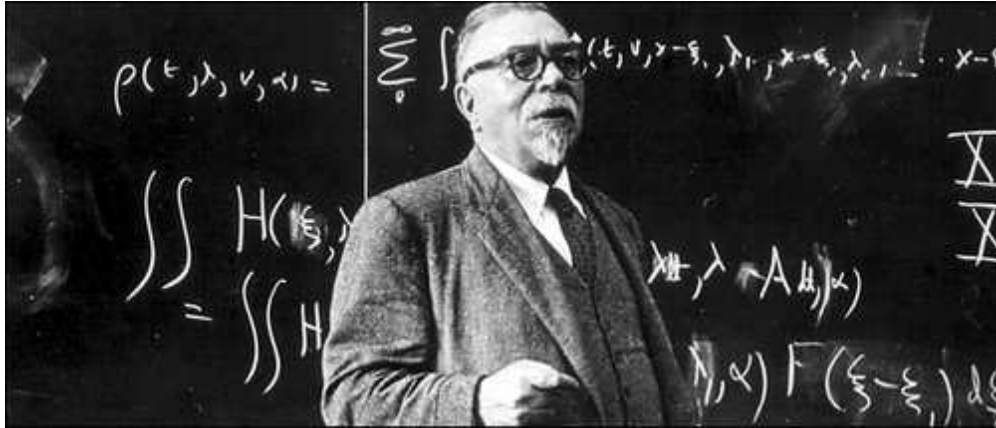
are available **only for**  $t' \leq t$

This is the problem Wiener solved in 1942,

in the **yellow peril**

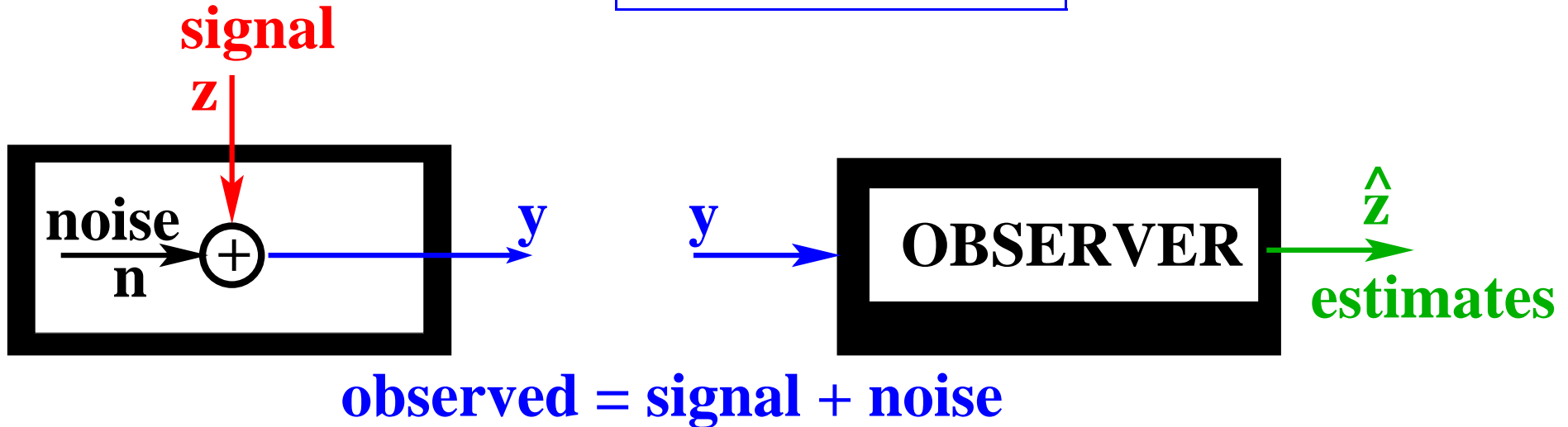
↪ **non-anticipating filter , a.k.a. the Wiener filter**

# Wiener Filtering





## The Wiener Filter



Signal  $\perp$  noise

Signal spectral density  $S_z(s)$

Noise white, intensity  $\rho_n^2$

Filters according to the signal-to-noise ratio.

## The Wiener Filter

**Knowledge of  $y(t) \forall t \in \mathbb{R} \rightsquigarrow$  tf. fn.**

$$y \mapsto \hat{z} = \boxed{1 - \frac{1}{1 + \frac{S_z(s)}{\rho_n^2}}} y$$

**Knowledge of  $y(t)$  in past, Wiener filter  $\rightsquigarrow$  tf. fn.**

$$y \mapsto \hat{z} = \boxed{1 - \frac{1}{\left[1 + \frac{S_z(s)}{\rho_n^2}\right]_+}} y$$

$$1 + \frac{S_z(s)}{\rho_n^2} = \left[1 + \frac{S_z(s)}{\rho_n^2}\right]_+ \left[1 + \frac{S_z(s)}{\rho_n^2}\right]_-$$

$\left[ \right]_+$  poles & zeros in LHP ‘spectral factorization’

## The Kalman Filter

By taking another **representation** of the stochastic process  $\begin{bmatrix} y \\ z \end{bmatrix}$ , the optimal non-anticipating filter becomes much easier to compute.

$$\frac{d}{dt}x = Ax + n_1, y = Cx + n_2, z = Hx, \begin{bmatrix} n_1 \\ n_2 \end{bmatrix} \text{ white}$$

This is the representation used by Kalman (1960)

## The Kalman Filter

$$\frac{d}{dt}x = Ax + Bw, \quad y = Cx + n, \quad z = Hx$$

$w \perp n$ , both white, intensities  $I$

$$\frac{d}{dt}\hat{x} = A\hat{x} + \Sigma C^\top (y - C\hat{x}), \quad \hat{z} = H\hat{x}$$

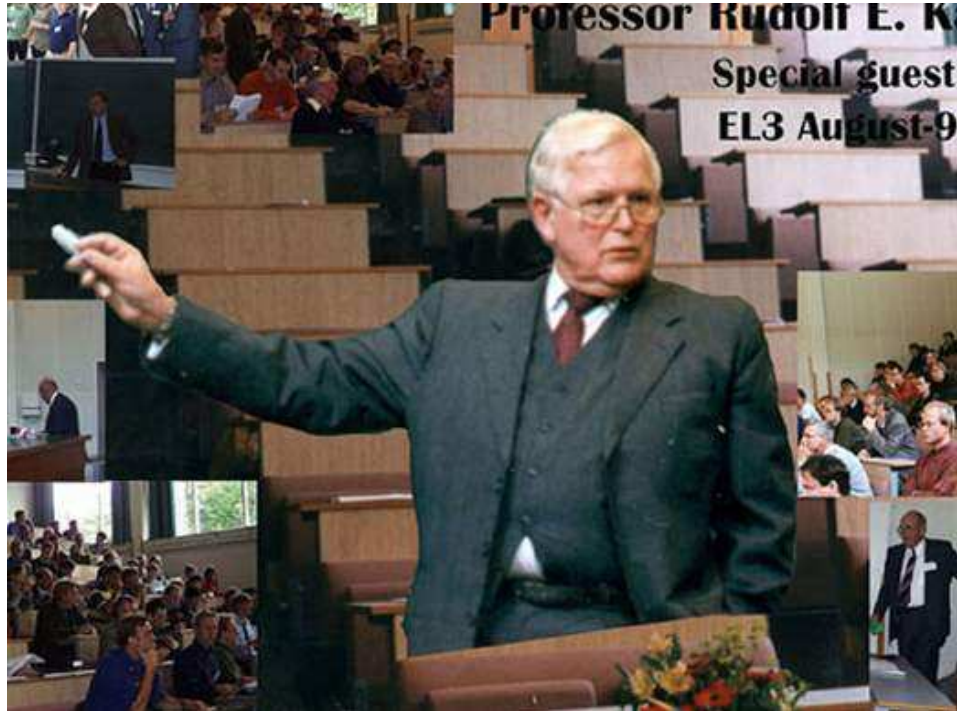
$\Sigma$  suitable solution of the ARE

$$A\Sigma + \Sigma A^\top - \Sigma C^\top C\Sigma + BB^\top = 0$$

*Exactly* the Wiener filter, but in a form that is  
*recursive, algorithmic, generalizable*

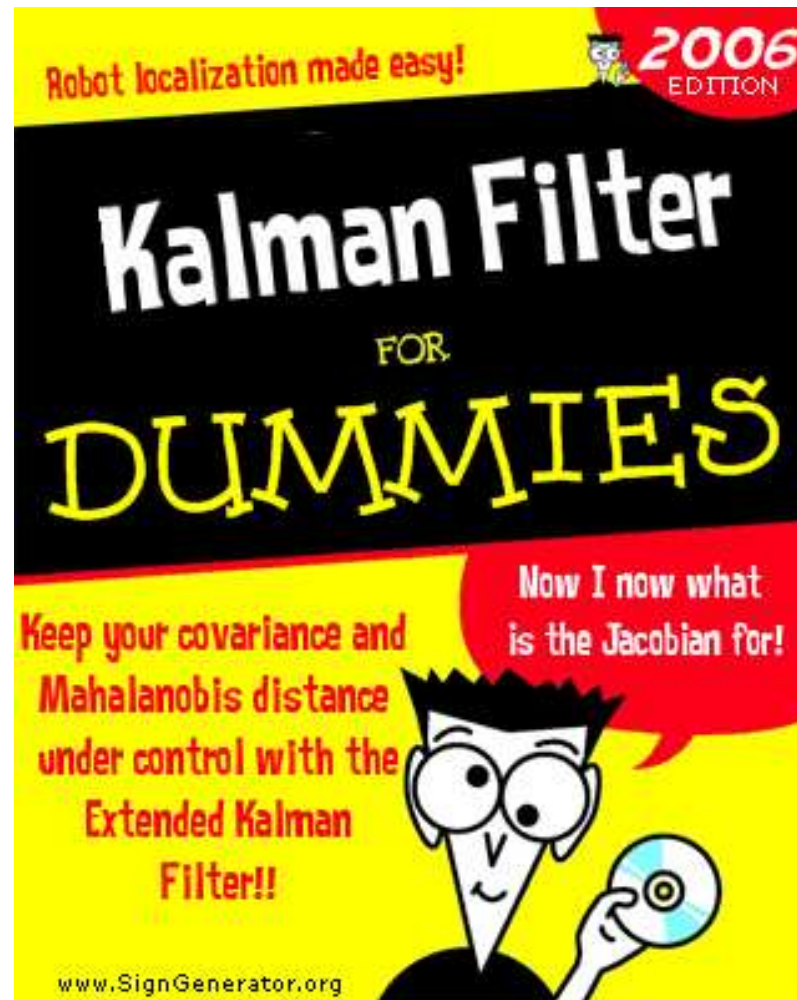
*(finite time, time-varying, nonlinear) ...*

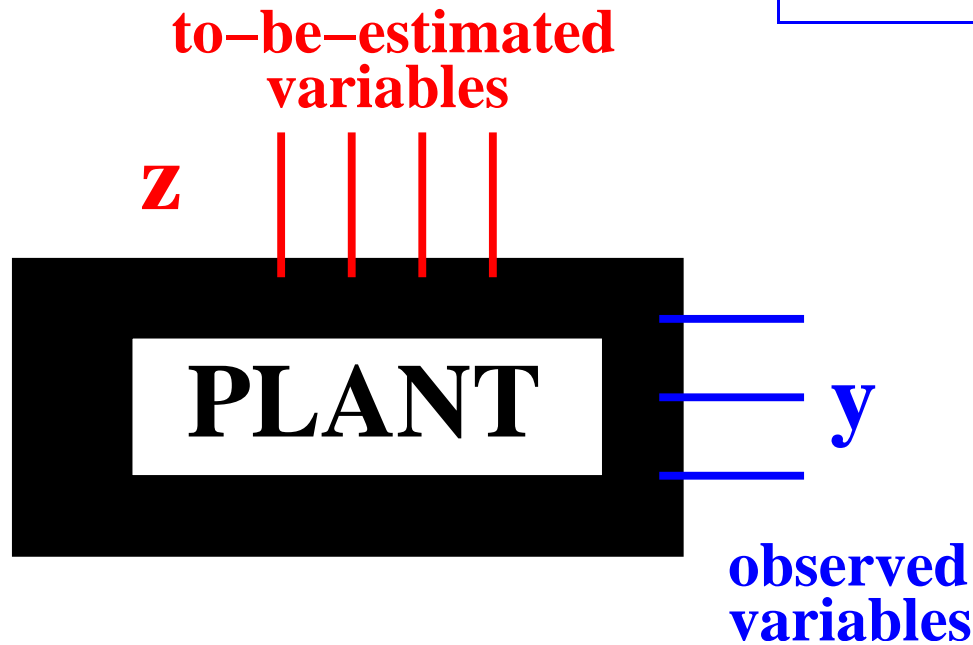
# The Kalman Filter



# The Kalman Filter

**The Kalman filter had a tremendous impact !**

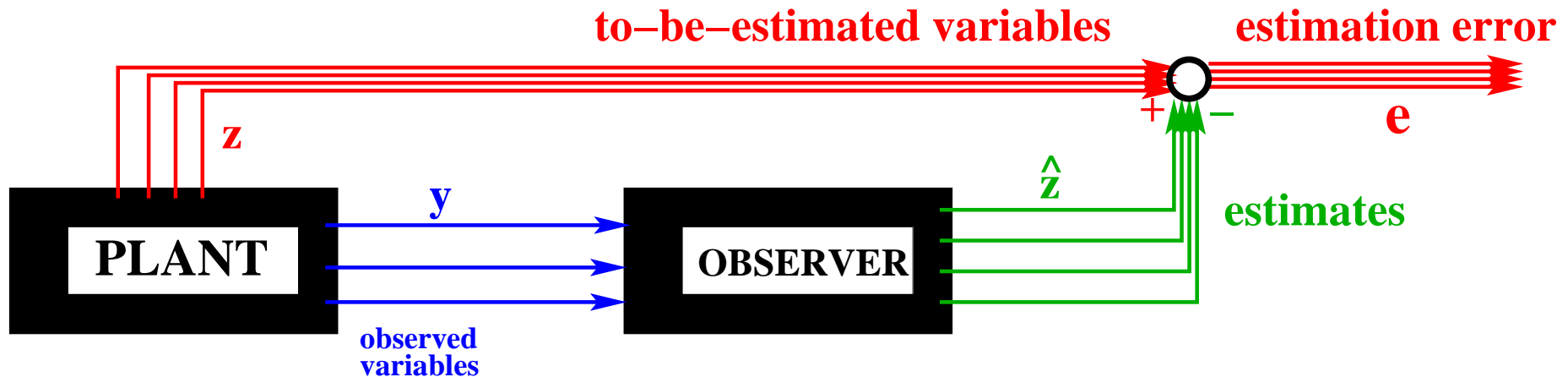




**Modeling**:  $\begin{bmatrix} y \\ z \end{bmatrix}$  input/output of a linear system

$$\frac{d}{dt}x = Ax + Bu, \quad y = \begin{bmatrix} u \\ Cx \end{bmatrix}, \quad z = Hx.$$

## History



**Plant:**

$$\frac{d}{dt}x = Ax + Bu, \quad y = \begin{bmatrix} u \\ Cx \end{bmatrix}, \quad z = Hx$$

**Observer:**

$$\frac{d}{dt}\hat{x} = A\hat{x} + Bu + L(Cx - C\hat{x}), \quad \hat{z} = H\hat{x}$$

**Error:**

$$\frac{d}{dt}e_x = (A - LC)e_x, \quad e = He_x$$



## History

*State observer*, first proposed in 1963 by Luenberger.



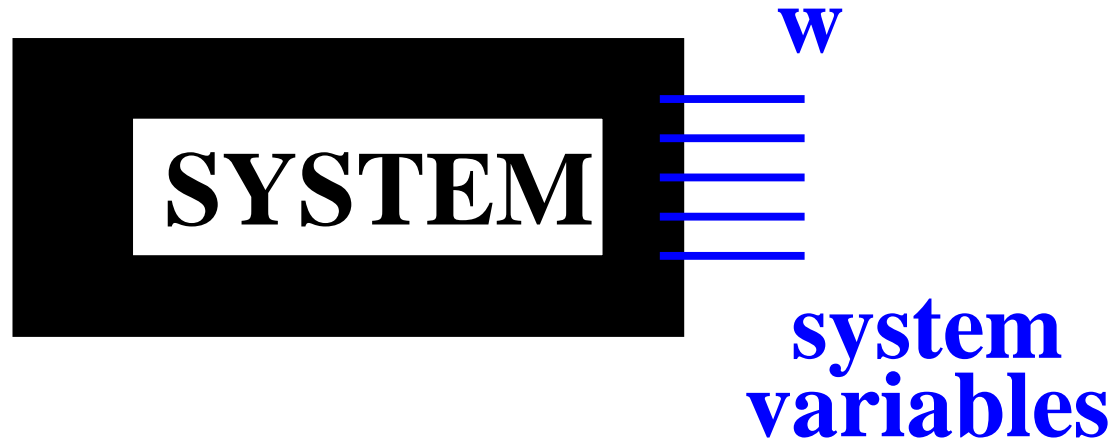
**Many variations (reduced order, dead-beat, ...)**

**Structure inspired by the ‘optimal’ Kalman filter.**

**No stochastic assumptions !**

# **Systems & Their Properties**

## Behaviors



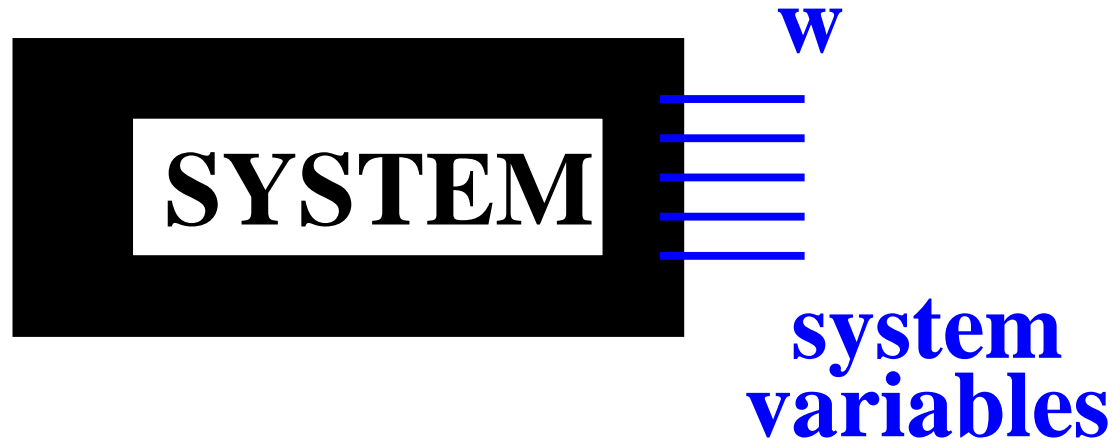
A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$  'time-set'

$\mathbb{W}$  'signal space'

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$  'behavior'

## Behaviors



A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$  'time-set'

$\mathbb{W}$  'signal space'

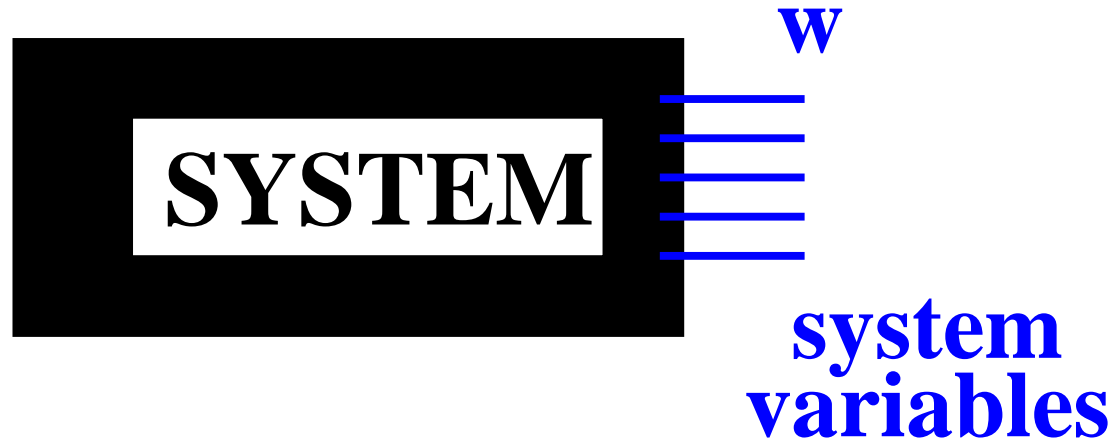
$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$  'behavior'

Consider  $w : \mathbb{T} \rightarrow \mathbb{W}$

$w \in \mathfrak{B}$  the model **allows** the trajectory  $w$

$w \notin \mathfrak{B}$  the model **forbids** the trajectory  $w$

## Behaviors



A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$       ‘time-set’      today,  $\mathbb{T} = \mathbb{R}$

$\mathbb{W}$       ‘signal space’      today,  $\mathbb{W} = \mathbb{R}^w$

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$       ‘behavior’      today, **LTIDS**

## Behaviors

A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$  ‘time-set’ today,  $\mathbb{T} = \mathbb{R}$

$\mathbb{W}$  ‘signal space’ today,  $\mathbb{W} = \mathbb{R}^w$

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$  ‘behavior’ today, **LTIDS**

**Linear time-invariant differential system (LTIDS):**

$\mathfrak{B} =$  all solutions of

$$R \left( \frac{d}{dt} \right) w = 0$$

where  $R \in \mathbb{R} [\xi]^{\bullet \times w}$

‘**kernel representation**’ (numerous other repr.)

## Behaviors

A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$     ‘time-set’    today,  $\mathbb{T} = \mathbb{R}$

$\mathbb{W}$     ‘signal space’    today,  $\mathbb{W} = \mathbb{R}^w$

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$     ‘behavior’    today, **LTIDS**

**Linear time-invariant differential system (LTIDS):**

$\mathfrak{B} =$  all  $\mathcal{C}^{\infty}(\mathbb{R}, \mathbb{R}^w)$ - solutions of

$$R \left( \frac{d}{dt} \right) w = 0$$

## Behaviors

A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$     ‘time-set’    today,  $\mathbb{T} = \mathbb{R}$

$\mathbb{W}$     ‘signal space’    today,  $\mathbb{W} = \mathbb{R}^w$

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$     ‘behavior’    today, **LTIDS**

**Linear time-invariant differential system (LTIDS):**

$\mathfrak{B} =$  all  $\mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^w)$ - solutions of

$$R_0 w + R_1 \frac{d}{dt} w + \cdots + R_n \frac{d^n}{dt^n} w = 0$$



## Behaviors

A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$  ‘time-set’ today,  $\mathbb{T} = \mathbb{R}$

$\mathbb{W}$  ‘signal space’ today,  $\mathbb{W} = \mathbb{R}^w$

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$  ‘behavior’ today, **LTIDS**

**Linear time-invariant differential system (LTIDS):**

$\mathfrak{B} = \text{all } \mathcal{C}^{\infty}(\mathbb{R}, \mathbb{R}^w)\text{-solutions of}$

**The behavior is all there is !**

**Representations, properties (controllability, observability, symmetries) in terms of behavior**

## Behaviors

A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$     ‘time-set’    today,  $\mathbb{T} = \mathbb{R}$

$\mathbb{W}$     ‘signal space’    today,  $\mathbb{W} = \mathbb{R}^w$

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$     ‘behavior’    today, **LTIDS**

**Linear time-invariant differential system (LTIDS):**

$\mathfrak{B} =$  all  $\mathcal{C}^{\infty}(\mathbb{R}, \mathbb{R}^w)$ - solutions of

**The behavior is all there is !**

**SYSID refers to behavior,**

**control = restricting behavior, ...**

## Behaviors

A *dynamical system* is  $\Sigma = (\mathbb{T}, \mathbb{W}, \mathfrak{B})$

$\mathbb{T} \subseteq \mathbb{R}$     ‘time-set’    today,  $\mathbb{T} = \mathbb{R}$

$\mathbb{W}$     ‘signal space’    today,  $\mathbb{W} = \mathbb{R}^w$

$\mathfrak{B} \subseteq \mathbb{W}^{\mathbb{T}}$     ‘behavior’    today, **LTIDS**

**Linear time-invariant differential system (LTIDS):**

$\mathfrak{B} = \text{all } \mathcal{C}^{\infty}(\mathbb{R}, \mathbb{R}^w)\text{-solutions of}$

**The behavior is all there is !**

**Physical models specify the behavior !**

## LTIDSs

$\mathcal{L}^\bullet$ : the LTIDSs.  $\mathcal{L}^\bullet$  is closed under projection

## LTIDSs

$\mathcal{L}^\bullet$ : the LTIDSs.  $\mathcal{L}^\bullet$  is **closed under projection**

$$R_1 \left( \frac{d}{dt} \right) w_1 = R_2 \left( \frac{d}{dt} \right) w_2 \quad (*)$$

$$R_1, R_2 \in \mathbb{R} [\xi]^{\bullet \times \bullet}.$$

## LTIDSs

$\mathcal{L}^\bullet$ : the LTIDSs.  $\mathcal{L}^\bullet$  is **closed under projection**

$$R_1 \left( \frac{d}{dt} \right) w_1 = R_2 \left( \frac{d}{dt} \right) w_2 \quad (*)$$

$R_1, R_2 \in \mathbb{R} [\xi]^{\bullet \times \bullet}$ . Define

$$\mathcal{B}_1 := \{ w_1 \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^\bullet) \mid \exists w_2 \text{ such that } (*) \}$$

## LTIDSs

$\mathcal{L}^\bullet$ : the LTIDSs.  $\mathcal{L}^\bullet$  is **closed under projection**

$$R_1 \left( \frac{d}{dt} \right) w_1 = R_2 \left( \frac{d}{dt} \right) w_2 \quad (*)$$

$R_1, R_2 \in \mathbb{R} [\xi]^{\bullet \times \bullet}$ . Define

$$\mathcal{B}_1 := \{ w_1 \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R}^\bullet) \mid \exists w_2 \text{ such that } (*) \}$$

**Thm:  $\mathcal{B}_1 \in \mathcal{L}^\bullet$**

$$\rightsquigarrow R \left( \frac{d}{dt} \right) w_1 = 0$$

$\exists$  algorithms

$$(R_1, R_2) \mapsto R$$

## LTIDSs

$\mathcal{L}^\bullet$ : the LTIDSs.  $\mathcal{L}^\bullet$  is **closed under projection**

$$F \left( \frac{d}{dt} \right) w = 0, F \in \mathbb{R} [\xi]^{\bullet \times w}$$

is a **consequence** of

$$R \left( \frac{d}{dt} \right) w = 0, R \in \mathbb{R} [\xi]^{\bullet \times w} \quad : \Leftrightarrow$$

$$\llbracket R \left( \frac{d}{dt} \right) w = 0 \rrbracket \Rightarrow \llbracket F \left( \frac{d}{dt} \right) w = 0 \rrbracket$$



## LTIDSs

$\mathcal{L}^\bullet$ : the LTIDSs.  $\mathcal{L}^\bullet$  is closed under projection

$$F \left( \frac{d}{dt} \right) w = 0, F \in \mathbb{R} [\xi]^{\bullet \times w}$$

is a consequence of

$$R \left( \frac{d}{dt} \right) w = 0, R \in \mathbb{R} [\xi]^{\bullet \times w} \quad : \Leftrightarrow$$

$$\llbracket R \left( \frac{d}{dt} \right) w = 0 \rrbracket \Rightarrow \llbracket F \left( \frac{d}{dt} \right) w = 0 \rrbracket$$

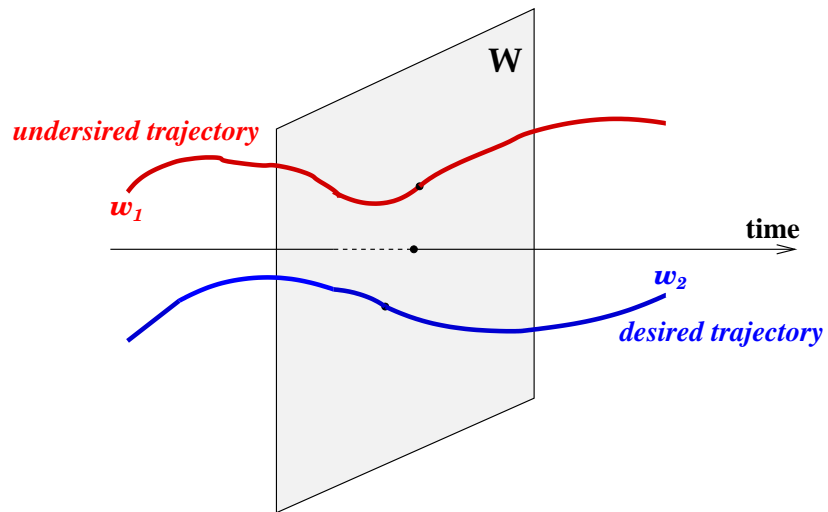
**Thm: Consequence  $\Leftrightarrow F = F' R$**

## System Properties

- **Controllable**
- **Stabilizable**
- **Autonomous**
- **Stable**
- **Observable**
- **Detectable**

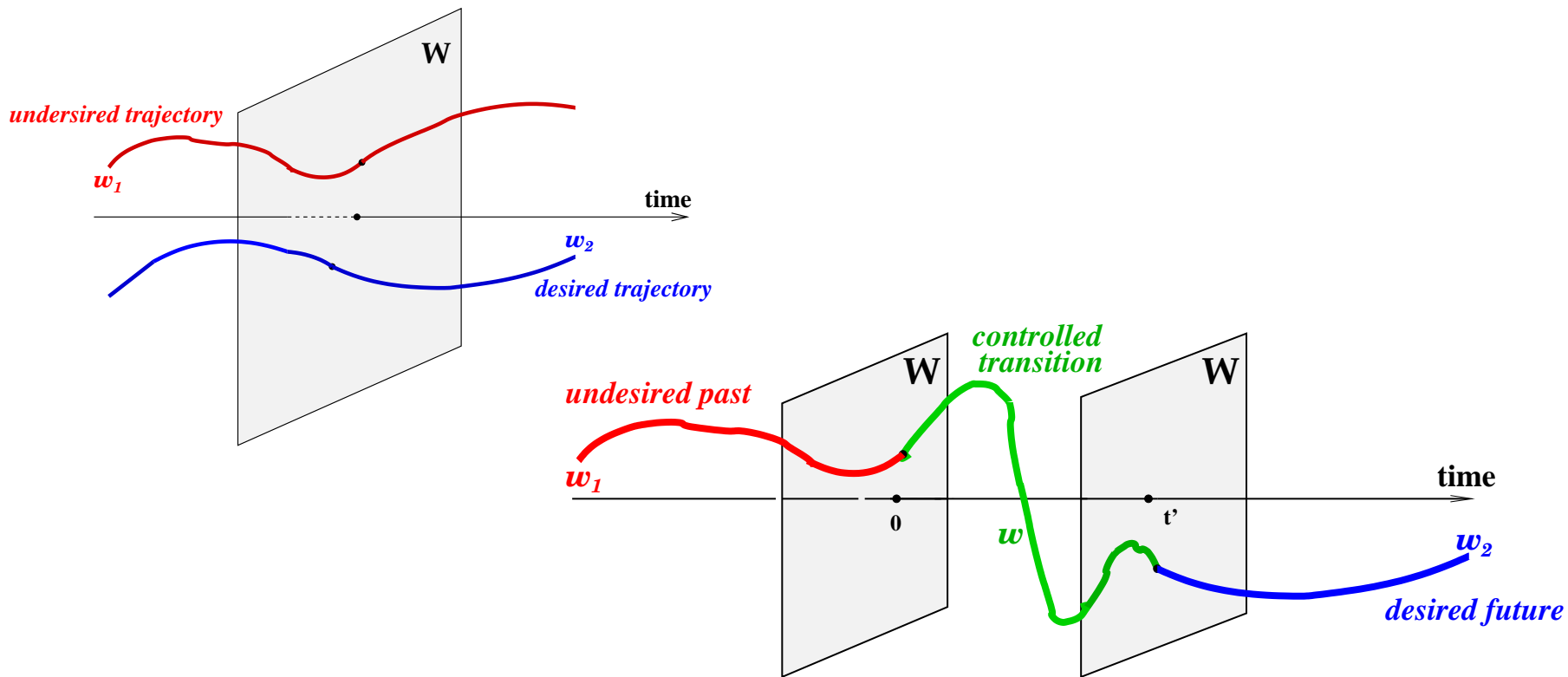
# System Properties

$\Sigma = (\mathbb{R}, \mathcal{W}, \mathcal{B})$  is **controllable** : $\Leftrightarrow$



# System Properties

$\Sigma = (\mathbb{R}, \mathcal{W}, \mathcal{B})$  is **controllable** : $\Leftrightarrow$



**Behavioral controllability of a dynamical system**

## System Properties

$\Sigma = (\mathbb{R}, \mathbb{W}, \mathfrak{B})$  is **autonomous**  $:\Leftrightarrow$

$\llbracket w_1, w_2 \in \mathfrak{B} \text{ and } w_1(t) = w_2(t) \text{ for } t < 0 \rrbracket$

$\Rightarrow \llbracket w_1(t) = w_2(t) \text{ for } t \geq 0 \rrbracket$

‘past implies future’

**stable**  $:\Leftrightarrow \llbracket w \in \mathfrak{B} \rrbracket \Rightarrow \llbracket w(t) \rightarrow 0 \text{ as } t \rightarrow \infty \rrbracket$

## System Properties

$$R \left( \frac{d}{dt} \right) w = 0$$

**defines a controllable system iff**

**rank  $(R(\lambda))$  is the same for all  $\lambda \in \mathbb{C}$**

**a stabilizable one ...  $\lambda \in$  the closed RHP**

## System Properties

$$R \left( \frac{d}{dt} \right) w = 0$$

**defines an autonomous system iff**

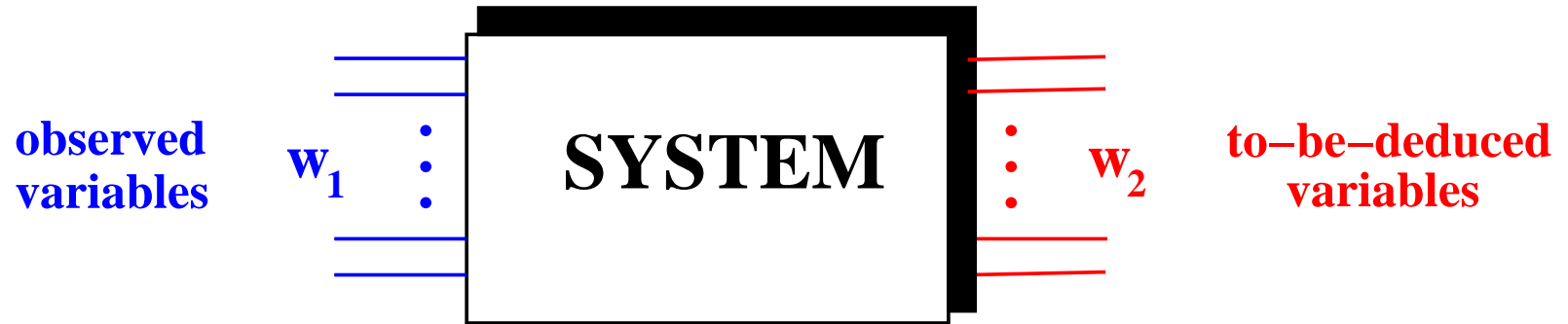
**$R(\lambda)$  full column rank  $\forall$  but finite number  $\lambda \in \mathbb{C}$**

**$\exists$  kernel repr. with  $R$  square and  $\det(R) \neq 0$ .**

**a stable one ...  $\lambda \in$  the closed LHP**

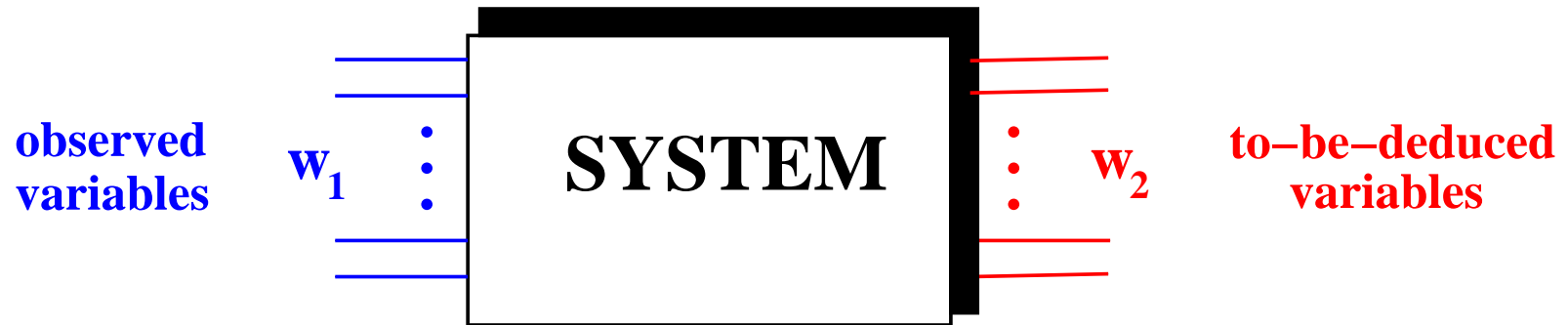
**$\exists R$  ‘Hurwitz’**

# Properties Involving Relations Among Variables





# Properties Involving Relations Among Variables



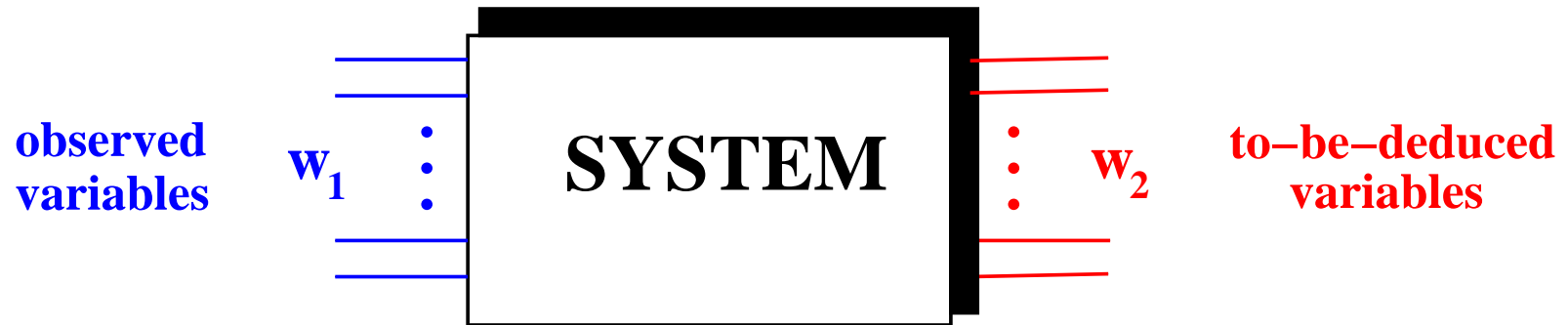
$w_1$  is **observable** from  $w_2$  in  $\Sigma = (\mathbb{T}, \mathbb{W}_1 \times \mathbb{W}_2, \mathfrak{B}) : \Leftrightarrow$

$$\llbracket (w_1, w'_2), (w_1, w''_2) \in \mathfrak{B} \rrbracket$$

$$\Rightarrow \llbracket w'_2 = w''_2 \rrbracket$$

**Observed trajectory implies the to-be-deduced one**

# Properties Involving Relations Among Variables



$w_1$  is **detectable** from  $w_2$  in  $\Sigma = (\mathbb{T}, \mathbb{W}_1 \times \mathbb{W}_2, \mathfrak{B}) : \Leftrightarrow$

$$\llbracket (w_1, w'_2), (w_1, w''_2) \in \mathfrak{B} \rrbracket$$

$$\Rightarrow \llbracket w'_2(t) - w''_2(t) \rightarrow 0 \text{ as } t \rightarrow \infty \rrbracket$$

**Observed trajectory implies the to-be-deduced one asymptotically**

## Tests for Observability and Detectability

$$R_1 \left( \frac{d}{dt} \right) w_1 = R_2 \left( \frac{d}{dt} \right) w_2$$

defines an **observable** system iff

$$R_2 (\lambda) \text{ has full column rank } \forall \lambda \in \mathbb{C}$$

defines a **detectable** system iff

... $\forall$  but finite number  $\lambda \in$  closed RHP

## Tests for Observability and Detectability

$$R_1 \left( \frac{d}{dt} \right) w_1 = R_2 \left( \frac{d}{dt} \right) w_2$$

**observable** iff there are ‘consequences’

$$w_2 = F \left( \frac{d}{dt} \right) w_1$$

$$\rightsquigarrow R \left( \frac{d}{dt} \right) w_1 = 0, \quad w_2 = F \left( \frac{d}{dt} \right) w_1$$

$\exists$  algorithms ...

## Tests for Observability and Detectability

$$R_1 \left( \frac{d}{dt} \right) w_1 = R_2 \left( \frac{d}{dt} \right) w_2$$

**detectable** iff there are ‘consequences’

$$H \left( \frac{d}{dt} \right) w_2 = F \left( \frac{d}{dt} \right) w_1, \text{ with } H \text{ ‘Hurwitz’}$$

$$\leadsto R \left( \frac{d}{dt} \right) w_1 = 0, \quad H \left( \frac{d}{dt} \right) w_2 = F \left( \frac{d}{dt} \right) w_1$$

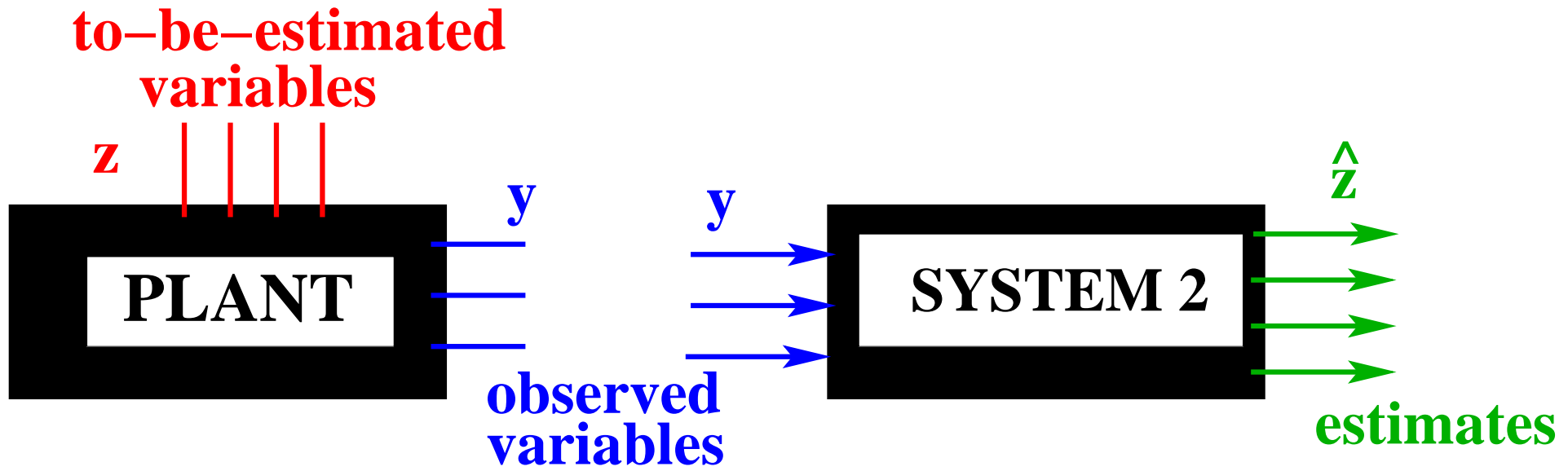
$\exists$  algorithms ...

**System properties ought to hold beyond the state space setting,**

**they ought to be representation independent**

**What is an observer ?**

## Observers



Consider two LTIDS systems.

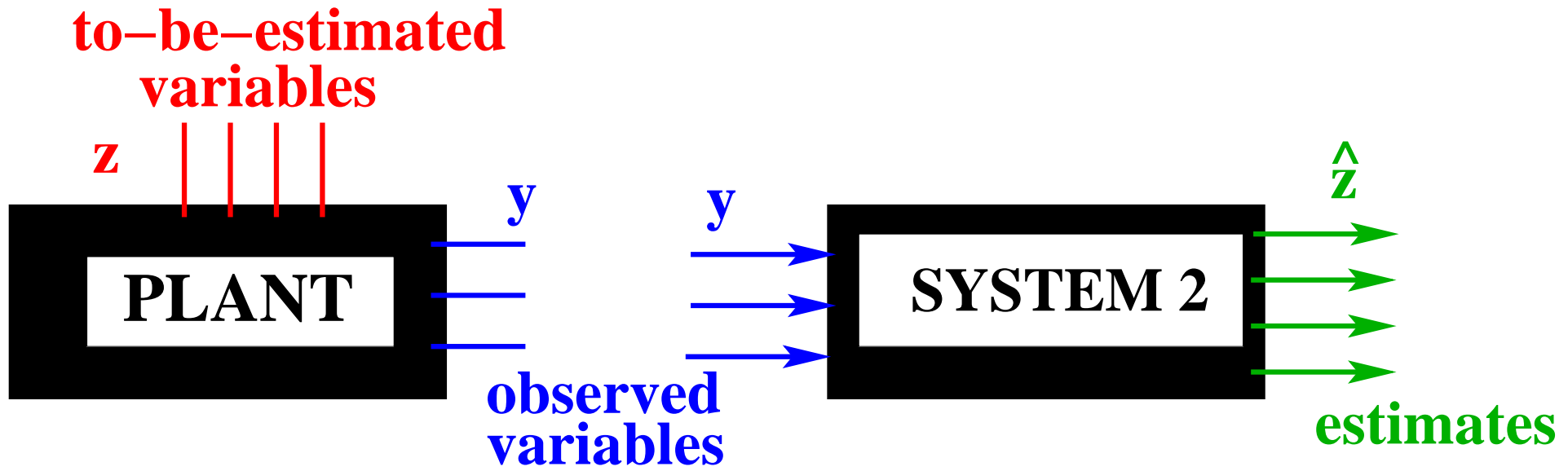
When is system 2 an observer for the plant?

Denote their behavior by

$$\mathcal{B}_{\text{plant}} \quad \text{and} \quad \hat{\mathcal{B}}$$



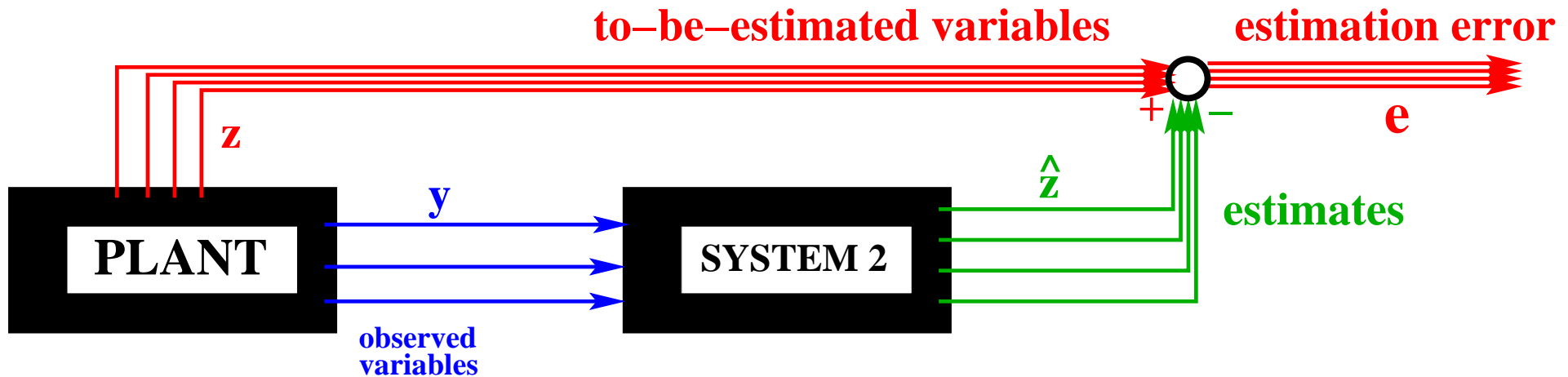
## Observers



Condition 1: System 2 **simulates** the plant, that is

$$\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}}$$

# Observers



Condition 1: System 2 **simulates** the plant, that is

$$\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}}$$

Condition 2: Error behavior,  $\mathcal{B}_{\text{error}}$ , is **autonomous**

$\mathcal{B}_{\text{error}} = \{0\}$ , exact observer

$\mathcal{B}_{\text{error}}$  nilpotent, dead-beat (discr. time)

$\mathcal{B}_{\text{error}}$  stable, asymptotic observer

## Observers

**Condition 1: System 2 **simulates** the plant, that is**

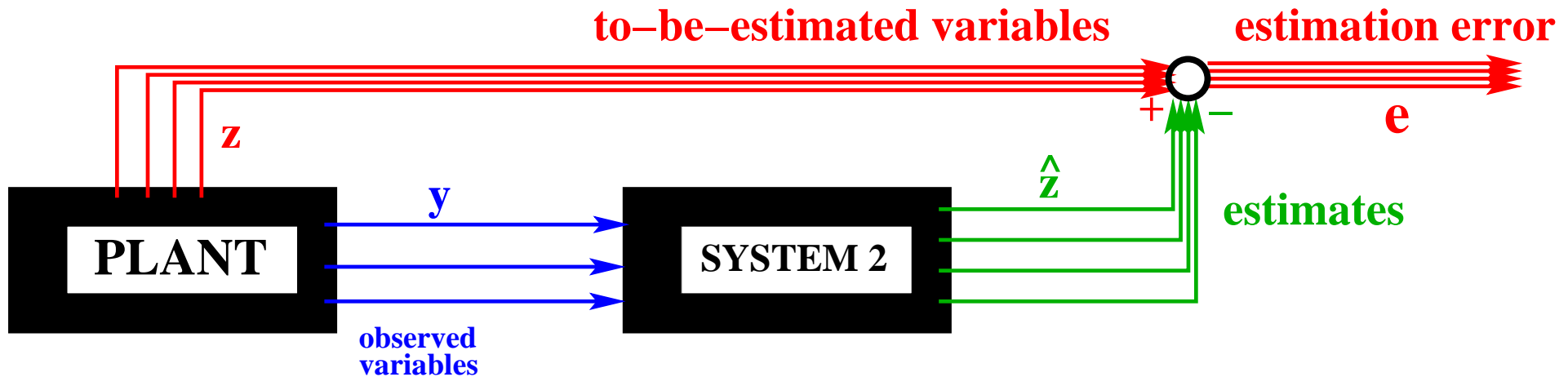
$$\mathfrak{B}_{\text{plant}} \subseteq \hat{\mathfrak{B}}$$

**Condition 2: Error behavior,  $\mathfrak{B}_{\text{error}}$ , is **autonomous****

**These conditions imply that**

- 1. it is possible to follow  $z$  through  $y$ ,**
- 2. once  $z(t') = \hat{z}(t')$  for  $t' \in [T - \varepsilon, T]$ ,  $\varepsilon > 0$ ,  
there holds  $z(t) = \hat{z}(t)$  for  $t > T$ .**

## Observers



Condition 1: System 2 **simulates** the plant, that is

$$\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}}$$

Condition 2: Error behavior,  $\mathcal{B}_{\text{error}}$ , is **autonomous**

Condition 3: WLOG, add  $y$  is **free** ('input') in  $\hat{\mathcal{B}}$ ,  
 $y$  is 'processed' in  $\hat{\mathcal{B}}$

## Observers

**Condition 1: System 2 **simulates** the plant, that is**

$$\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}}$$

**Condition 2: Error behavior,  $\mathcal{B}_{\text{error}}$ , is **autonomous****

**Condition 3: WLOG, add  $y$  is **free** ('input') in  $\hat{\mathcal{B}}$ ,  
 $y$  is 'processed' in  $\hat{\mathcal{B}}$**

**These conditions are not independent.**

**1 + 3 ( $y$  input) +  $\hat{z}$  output  $\Rightarrow$  2**

**controllability of plant + 2 + 3  $\Rightarrow$  1**

**Assume contr. & 3. Then  $\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}} \Leftrightarrow$  **observer****

## Observers

Condition 1: System 2 **simulates** the plant, that is

$$\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}}$$

Condition 2: Error behavior,  $\mathcal{B}_{\text{error}}$ , is **autonomous**

Condition 3: WLOG, add  $y$  is **free** ('input') in  $\hat{\mathcal{B}}$ ,  
 $y$  is 'processed' in  $\hat{\mathcal{B}}$

Theorem: An observer exists if and only if

**$\{(z, y) \in \mathcal{B}_{\text{plant}} \mid y = 0\}$  is autonomous**

## Observers

Condition 1: System 2 **simulates** the plant, that is

$$\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}}$$

Condition 2: Error behavior,  $\mathcal{B}_{\text{error}}$ , is **autonomous**

Condition 3: WLOG, add  $y$  is **free** ('input') in  $\hat{\mathcal{B}}$ ,  
 $y$  is 'processed' in  $\hat{\mathcal{B}}$

Roughly, observer design  $\cong$  finding a cover

$$\mathcal{B}_{\text{plant}} \subseteq \hat{\mathcal{B}}$$

# Observer Design



## Covers

**Essential condition:**

$$\mathfrak{B}_{\text{plant}} \subseteq \hat{\mathfrak{B}}$$

**Easy to find a supsystem,  $\mathfrak{B}' \supseteq \mathfrak{B}$ , for a given LTIDS  $\mathfrak{B}$ . For example, from ‘kernel representation’**

$$R \left( \frac{d}{dt} \right) w = 0$$

**Then  $\mathfrak{B}' \supseteq \mathfrak{B}$  iff  $\mathfrak{B}'$  has kernel representation**

$$F \left( \frac{d}{dt} \right) R \left( \frac{d}{dt} \right) w = 0$$

**for some  $F \in \mathbb{R}[\xi]^{\bullet \times \bullet}$ .**

## Covers

**Plant:**

$$\mathbf{Z} \left( \frac{d}{dt} \right) \mathbf{z} = \mathbf{Y} \left( \frac{d}{dt} \right) \mathbf{y}$$

**Observer therefore**

$$\mathbf{F} \left( \frac{d}{dt} \right) \mathbf{Z} \left( \frac{d}{dt} \right) \hat{\mathbf{z}} = \mathbf{F} \left( \frac{d}{dt} \right) \mathbf{Y} \left( \frac{d}{dt} \right) \mathbf{y}$$

**Error dynamics**

$$\mathbf{F} \left( \frac{d}{dt} \right) \mathbf{Z} \left( \frac{d}{dt} \right) \mathbf{e} = \mathbf{0}$$

**Observer conditions:  $\mathbf{FZ}$  square and non-singular.**

## Covers

Given  $Z, Y \in \mathbb{R} [\xi]^{\bullet \times \bullet}$ , what can be achieved by  $F \in \mathbb{R} [\xi]^{\bullet \times \bullet}$   $(Z, Y) \mapsto (FZ, FY)$  ?

Achievable error dynamics

$$F \left( \frac{d}{dt} \right) Z \left( \frac{d}{dt} \right) e = 0$$

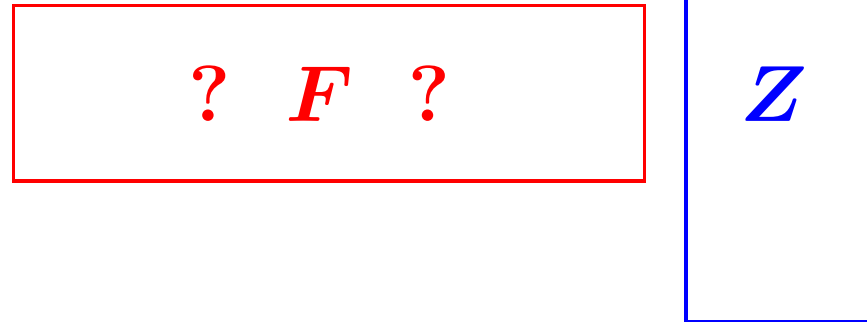
Can the observer be made **smoothing** ?

$$F \left( \frac{d}{dt} \right) Z \left( \frac{d}{dt} \right) \hat{z} = F \left( \frac{d}{dt} \right) Y \left( \frac{d}{dt} \right) y$$

transfer function  $(FZ)^{-1}(FY)$

proper, strictly proper, high-frequency roll-off, ...

## Covers



taking into consideration roll-off of  $(FZ)^{-1}(FY)$

## Error Dynamics

Assume that in the plant  $z$  is **observable** from  $y$ .

Then  $\forall r \in \mathbb{R}[\xi]$ , monic,  $\exists F$  such that

$$\det(FZ) = r$$

$r = 1 \quad \rightsquigarrow$  exact observer

$r$  Hurwitz  $\rightsquigarrow$  asymptotic observer

$r(\xi) = \xi^d \rightsquigarrow$  dead-beat observer (discr.-time)

Combinable with proper, high-frequency roll-off,  
provided  $\text{degree}(r)$  sufficiently large.

## Error Dynamics

Assume  $z$  is **detectable** from  $y$ . Then for any  $r \in \mathbb{R} [\xi]$ , monic, with a **given Hurwitz factor** (representing the unobservable modes) there exists  $F$  such that

$$\det(FZ) = r$$

$r$  Hurwitz  $\leadsto$  asymptotic observer

Combinable with proper, high-frequency roll-off,  
provided  $\text{degree}(r)$  sufficiently large.

## Example

**Autonomous system,  $z, y$  scalar:**

$$R \left( \frac{d}{dt} \right) \begin{bmatrix} z \\ y \end{bmatrix} = 0$$

$$\det (R) \neq 0.$$

## Example

**Autonomous system,  $z, y$  scalar:**

$$R \left( \frac{d}{dt} \right) \begin{bmatrix} z \\ y \end{bmatrix} = 0$$

**$\det(R) \neq 0$ . Observability  $\Rightarrow$  representation**

$$Y \left( \frac{d}{dt} \right) y = 0, z = Z \left( \frac{d}{dt} \right) y$$

$$Y, Z \in \mathbb{R} [\xi]$$



## Example

$$Y \left( \frac{d}{dt} \right) y = 0, z = Z \left( \frac{d}{dt} \right) y$$

**Observer:**

$$\pi_1 \left( \frac{d}{dt} \right) \hat{z} = \left[ \pi_1 \left( \frac{d}{dt} \right) Z \left( \frac{d}{dt} \right) + \pi_2 \left( \frac{d}{dt} \right) Y \left( \frac{d}{dt} \right) \right] y$$

$\pi_1$  given, sufficiently high degree, roots arbitrary  
arbitrary high roll-off by choosing  $\pi_2$

$\rightsquigarrow$  simple polynomial algebra.

## Example

$$Y \left( \frac{d}{dt} \right) y = 0, z = Z \left( \frac{d}{dt} \right) y$$

**Observer:**

$$\pi_1 \left( \frac{d}{dt} \right) \hat{z} = \left[ \pi_1 \left( \frac{d}{dt} \right) Z \left( \frac{d}{dt} \right) + \pi_2 \left( \frac{d}{dt} \right) Y \left( \frac{d}{dt} \right) \right] y$$

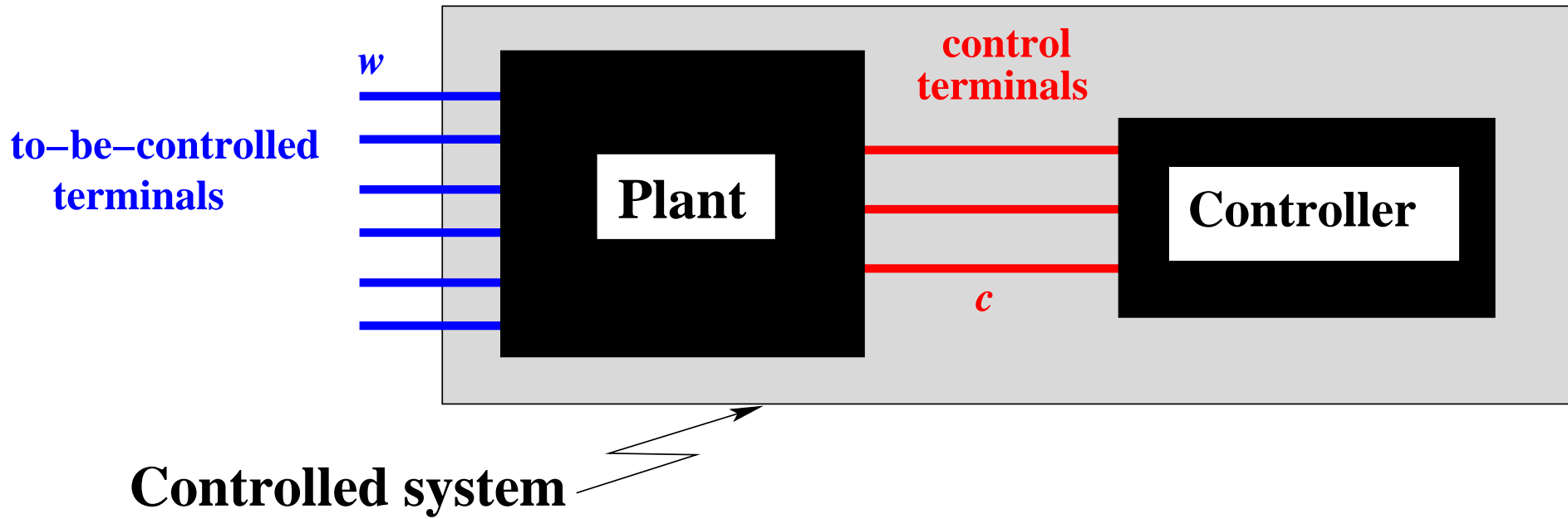
$\pi_1$  given, sufficiently high degree, roots arbitrary  
arbitrary high roll-off by choosing  $\pi_2$

$\rightsquigarrow$  simple polynomial algebra.

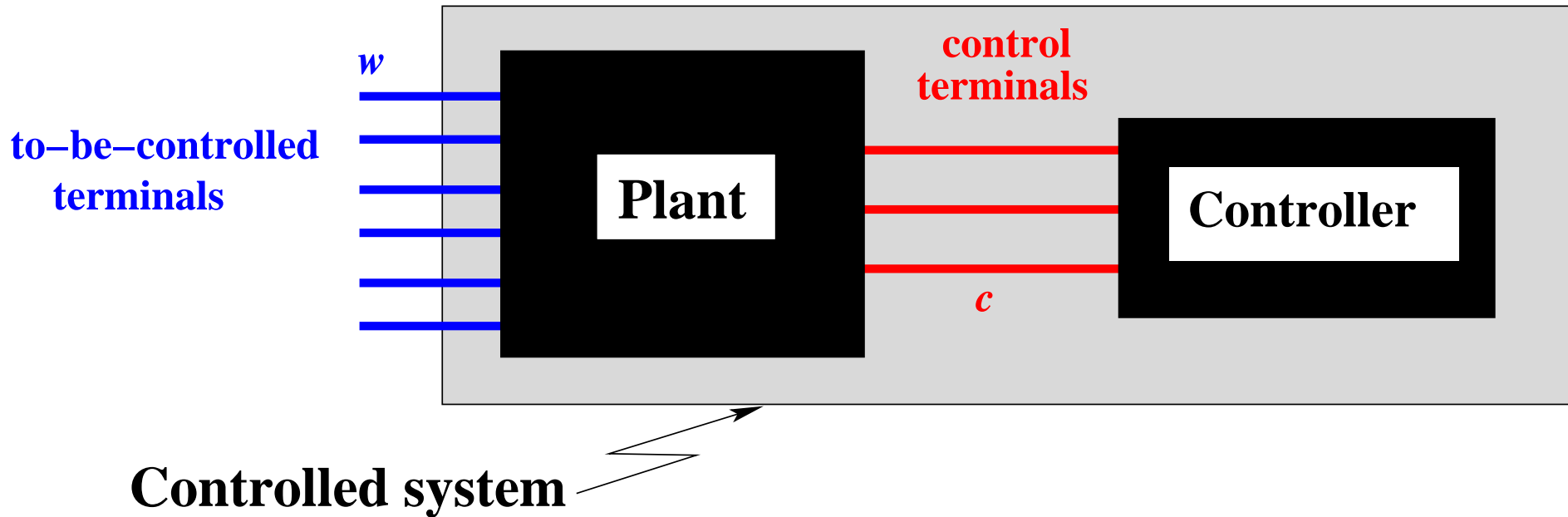
**When plant is autonomous, the pole placement  
combinable with arbitrary roll-off**

# Duality with Control

# Control in a Behavioral Setting

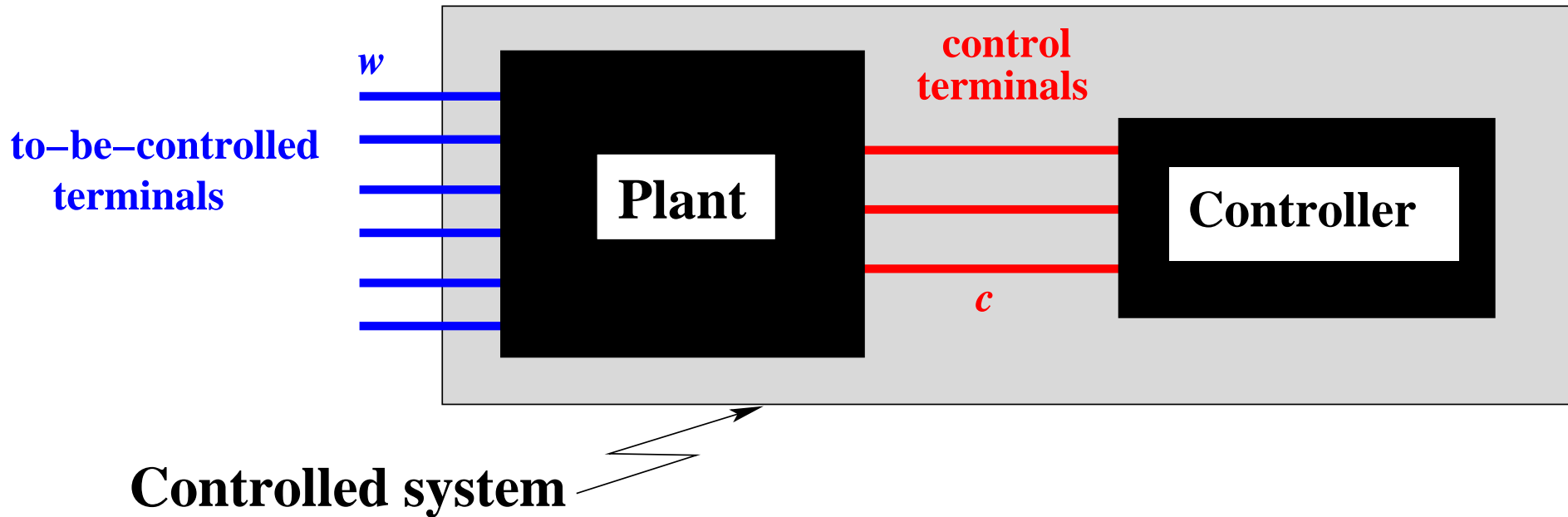


# Control in a Behavioral Setting



**Behavior of to-be-controlled variables, before  
controller is applied:  $\mathfrak{B}_{\text{plant}}$ , after:  $\mathfrak{B}_{\text{controlled}}$**

# Control in a Behavioral Setting

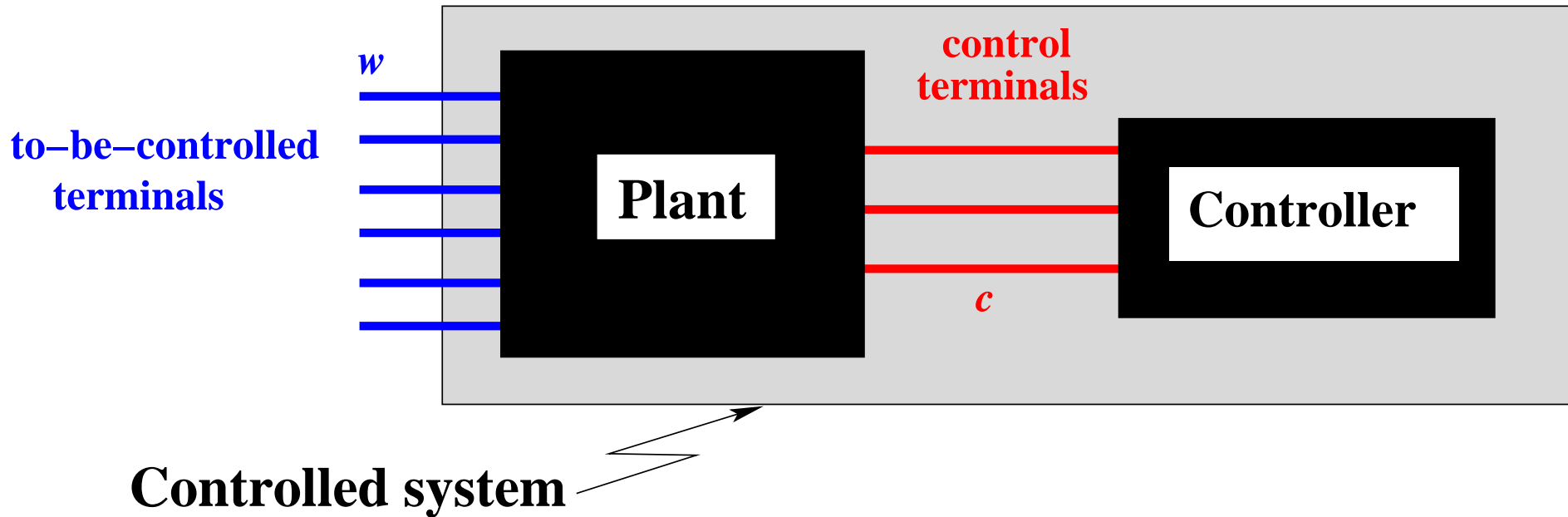


Behavior of to-be-controlled variables, before controller is applied:  $\mathcal{B}_{\text{plant}}$ , after:  $\mathcal{B}_{\text{controlled}}$

Obviously,

$$\mathcal{B}_{\text{controlled}} \subseteq \mathcal{B}_{\text{plant}}$$

## Control in a Behavioral Setting



Behavior of to-be-controlled variables, before controller is applied:  $\mathcal{B}_{\text{plant}}$ , after:  $\mathcal{B}_{\text{controlled}}$

Obviously,  $\mathcal{B}_{\text{controlled}} \subseteq \mathcal{B}_{\text{plant}}$

If  $w$  is observable from  $c$  in the plant, then every such  $\mathcal{B}_{\text{controlled}}$  is implementable.

## Duality

**Given  $\mathcal{B}_{\text{plant}}$ , LTIDS**



## Duality

Given  $\mathfrak{B}_{\text{plant}}$ , LTIDS

Control  $\rightsquigarrow$  find a subsystem

$$\mathfrak{B} \subseteq \mathfrak{B}_{\text{plant}}$$

that meets controller specs.

Given

$$R \left( \frac{d}{dt} \right) w = 0$$

$$C \left( \frac{d}{dt} \right) w = 0$$

‘Squaring up’  $R$  to  $\begin{bmatrix} R \\ C \end{bmatrix}$

## Duality

Given  $\mathfrak{B}_{\text{plant}}$ , LTIDS

Control  $\rightsquigarrow$  find a subsystem

$$\mathfrak{B} \subseteq \mathfrak{B}_{\text{plant}}$$

that meets controller specs.

Observer  $\rightsquigarrow$  find a supsystem

$$\mathfrak{B} \supseteq \mathfrak{B}_{\text{plant}}$$

that meets observer specs.

## Duality

Given  $\mathfrak{B}_{\text{plant}}$ , LTIDS

Control  $\rightsquigarrow$  find a subsystem

$$\mathfrak{B} \subseteq \mathfrak{B}_{\text{plant}}$$

that meets controller specs.

Observer  $\rightsquigarrow$  find a supsystem

$$\mathfrak{B} \supseteq \mathfrak{B}_{\text{plant}}$$

that meets observer specs.

Controllers mean less, **Observers mean more**

## Extensions

- **Systems defined by rational (rather than polynomial) ‘symbols’**
- **Least squares,  $\mathcal{H}_\infty$ , ...**
- **nD systems, PDEs**

**Details & copies of frames are available from/at**

Jan.Willems@esat.kuleuven.be

<http://www.esat.kuleuven.be/~jwillems>

**Thank you**

**Thank you**

**Thank you**

**Thank you**

**Thank you**

**Thank you**

**Thank you**