

# **Control, Observation and Feedback: a State Space View**

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# Systems as Physical Objects

- Have inputs, states and outputs
- Can be interconnected
- Distinguish as plants, controllers, filters, reference signal generators, data converters, etc
- Wide variety of technological, biological, logistical and economic examples fit this perspective

# Systems as Mathematical Objects

- A family of transformations (depending on input signals) of states
- A family of read-out maps (depending on input signals)

Causal models derived from ordinary and partial differential equations.

$$\dot{x}(t) = f(x(t), u(t))$$

$$y(t) = h(x(t), u(t))$$

$x(t) \in X$  state space

$u(t) \in U$  input space

$y(t) \in Y$  output space

# Systems as Mathematical Objects

Structure in models

$$\dot{x}(t) = f_0(x(t)) + \sum_{i=1}^m u_i(t) f_i(x(t))$$

$$y(t) = h(x(t))$$

Here  $f_i$ ,  $i = 0, 1, 2, \dots, m$  are vector fields on state space

Linearity:

$$\dot{x}(t) = A(t)x(t) + B(t)u(t)$$

$$y(t) = C(t)x(t) + D(t)u(t)$$

Stationarity:  $A, B, C, D$  time invariant.

## Descriptions of Systems (Internal)

Parameters:  $A, B, C, D$

Input-state response:

$$x(t) = \Phi_A(t, t_0)x(t_0) + \int_{t_0}^t \Phi_A(t, \sigma)B(\sigma)u(\sigma)d\sigma,$$

where  $\Phi_A(\cdot, \cdot)$ , the transition matrix, satisfies

$$\dot{\Phi}_A(t, t_0) = A(t)\Phi_A(t, t_0)$$

and  $\Phi_A(t_0, t_0) = \mathbb{1}$ , the identity matrix.

Easy to verify

$$\begin{aligned} \Phi_A(t, t_0) &= \mathbb{1} + \int_{t_0}^t A(\sigma)d\sigma \\ &+ \int_{t_0}^t \int_{t_0}^{\sigma_1} A(\sigma_1)A(\sigma_2)d\sigma_2d\sigma_1 + \dots \end{aligned}$$

## Descriptions of Systems (External)

Input-Output response:

$$\begin{aligned}y(t) &= C(t)\Phi_A(t, t_0)x(t_0) \\ &\quad + \int_{t_0}^t C(t)\Phi_A(t, \sigma)B(\sigma)u(\sigma)d\sigma \\ &\quad + D(t)u(t) \\ &= y_0(t) + \int_{t_0}^t W(t, \sigma)u(\sigma)d\sigma \\ &\quad + D(t)u(t)\end{aligned}$$

Weighting pattern:

$$W(t, \sigma) = C(t)\Phi_A(t, \sigma)B(\sigma).$$

$$\text{Drift} = y_0(t)$$

depends only on initial conditions.

## Specializing to Time-invariant Setting

$$\Phi_A(t, t_0) = e^{A \cdot (t - t_0)}$$

$$W(t, t_0) = C e^{A \cdot (t - t_0)} B$$

Impulse response

$$= C e^{At} B + D \delta(t)$$

Transfer function

$$G(s) = C(s\mathbb{1} - A)^{-1} B + D$$

# Descriptions of Systems (Internal vs External)

Change of variables

$$z(t) = P(t)x(t)$$

changes the internal description but not the external one.

In the time-invariant setting, two internal descriptions with parameters  $[A, B, C, D]$  and  $[PAP^{-1}, PB, CP^{-1}, D]$  have the same transfer function.

Transfer functions are proper, and if  $D = 0$ , strictly proper (i.e.,  $G(s) \rightarrow 0$  as  $s \rightarrow \infty$ )



## Reachability Problem

Given  $x_0 =$  state at time  $t_0$ , does there exist a control  $u(\cdot)$  defined on the time interval  $[t_0, t_1]$  that drives the system to  $x_1$  at time  $t_1$ ?

Define  $R_{(x_0, t_0)}^{t_1}$  set of such  $x_1$ .

$$R_{(x_0, t_0)} = \bigcup_{t_1 > 0} R_{(x_0, t_0)}^{t_1}$$

We say that the system is reachable from  $(x_0, t_0)$  if  $R_{(x_0, t_0)} =$  state space

## Reachability Problem

For linear systems with  $X = \mathbb{R}^n$ ,  $U = \mathbb{R}^m$ ,  
 $Y = \mathbb{R}^p$  solve

$$\begin{aligned}(x_0 - \Phi_A(t_0, t_1)x_1) &= - \int_{t_0}^{t_1} \Phi_A(t_0, \sigma)B(\sigma)u(\sigma)d\sigma \\ &= L(u)\end{aligned}$$

System is reachable from  $(x_0, t_0)$  if  
 $\mathcal{R}(L) = \text{range space of } L = \mathbb{R}^n$

Define reachability gramian  $W(t_0, t_1) = LL^*$ ,  
where

$$\begin{aligned}L^* : \mathbb{R}^n &\rightarrow C^m[t_0, t_1] \\ \eta &\mapsto -B'(\cdot)\Phi'_A(t_0, \cdot)\eta,\end{aligned}$$

where  $'$  denotes transpose of a matrix.  
Then  $\mathcal{R}(L) = \mathcal{R}(W)$

## Reachability Problem

Suppose there exists  $\eta \in \mathbb{R}^n$  such that

$$(x_0 - \Phi_A(t_0, t_1)x_1) = W(t_0, t_1)\eta.$$

Then, control defined by

$$u_0(t) = -B'(t)\Phi'_A(t_0, t)\eta$$

drives the system from  $(x_0, t_0)$  to  $(x_1, t_1)$ .

System is reachable iff  $W$  is invertible.

If  $u$  is any other control that drives  $(x_0, t_0)$  to  $(x_1, t_1)$ , then

$$\int_{t_0}^{t_1} u'(\sigma)u(\sigma)d\sigma \geq \int_{t_0}^{t_1} u'_0(\sigma)u_0(\sigma)d\sigma$$

## Observability Problem

Is it possible to determine the initial state  $x(t_0)$  from an input-output pair known over a time interval  $[t_0, t_1]$ ? If yes, we say the system is observable. For linear systems, define the drift map

$$\begin{aligned} P : \mathbb{R}^n &\rightarrow C^p[t_0, t_1] \\ x_0 &\mapsto C(\cdot)\Phi_A(\cdot, t_0)x_0 \end{aligned}$$

Define observability gramian

$$\begin{aligned} M(t_0, t_1) &= P^*P \\ &= \int_{t_0}^{t_1} \Phi'_A(\sigma, t_0)C'(\sigma)C(\sigma)\Phi_A(\sigma, t_0)d\sigma \end{aligned}$$

Since null space  $\mathcal{N}(P) = \mathcal{N}(M(t_0, t_1))$ , it follows that the system is observable iff  $M(t_0, t_1)$  is invertible.

# Gramians and the Time-invariant Setting

For time-invariant linear systems

$W(t_0, t_1)$  is invertible for  $t_1 > t_0$

$\Leftrightarrow [B, AB, \dots, A^{n-1}B]$  has rank  $n$

$\Leftrightarrow [s\mathbb{1} - A|B]$  has constant rank  $n$

for all  $s \in \mathbb{C}$

$M(t_0, t_1)$  is invertible for  $t_1 > t_0$

$$\begin{aligned}
&\Leftrightarrow \begin{bmatrix} C \\ CA \\ \vdots \\ CA^{n-1} \end{bmatrix} \text{ has rank } n \\
&\Leftrightarrow \begin{bmatrix} C \\ \hline s\mathbb{1} - A \end{bmatrix} \text{ has constant rank } n \\
&\text{for all } s \in \mathbb{C}
\end{aligned}$$

## Realization Problem

We have already seen that internal representations are not uniquely defined by an external representation – the change of variables idea.

It can be shown that any weighting pattern  $W(t, \sigma)$  that is factorizable in the form

$$W(t, \sigma) = Q(t)R(\sigma)$$

where  $Q(t)$  is  $p \times n_1$  and  $R(\sigma)$  is  $n_1 \times m$ , admits a finite dimensional representation

$$W(t, \sigma) = C(t)\Phi_A(t, \sigma)B(\sigma)$$

## Realization Problem

The finite dimensional representation (realization) above, is of the lowest possible (state space) dimension =  $n$ , iff it is both reachable and observable.

All minimal state space realizations are related by the (possibly time-dependent) change of variables formula.

In the time-invariant setting realizability is equivalent to the condition that the transfer function  $G(s)$  is a  $p \times m$  matrix of strictly proper, rational functions.



## Realization Problem

For a time-invariant linear system, let the transfer function be given by a Laurent series

$$G(s) = \sum_{i=0}^{\infty} \frac{L_i}{s^{i+1}}.$$

$L_i$  are called Markov parameters.

Clearly  $G(s)$  is strictly proper. It is rational iff the infinite Hankel matrix

$$\begin{bmatrix} L_0 & L_1 & L_2 & \cdots \\ L_1 & L_2 & L_3 & \cdots \\ L_2 & L_3 & L_4 & \cdots \\ \vdots & & & \end{bmatrix}$$

is of finite rank  $= n$ , called the McMillan degree.

A classic realization algorithm (Ho-Kalman), constructs a minimal realization from Markov parameter data.

# Minimality, Poles and Zeros

For  $m = p = 1$  a strictly proper

$$\begin{aligned} G(s) &= \frac{q(s)}{p(s)} \\ &= \frac{q_{n-1}s^{n-1} + \dots + q_0}{s^n + p_{n-1}s^{n-1} + \dots + p_0} \end{aligned}$$

$q(s)$  and  $p(s)$  relatively prime,

Poles  $\{G(s)\} =$  roots of  $p(s)$

Zeros  $\{G(s)\} =$  roots of  $q(s)$

If  $G(s) = C(s\mathbb{1} - A)^{-1}B$ , then

Poles  $\{G(s)\} \subset$  spectrum  $(A)$ .

## Minimality, Poles and Zeros, cont'd

Poles  $\{G(s)\} = \text{spectrum}(A)$   
iff  $[A, B, C]$  is minimal.

In that case

McMillan degree = degree of  $p(s)$   
= dimension of state space

## Remark on System Identification

The realization problem (passing from the sequence  $\{L_i\}_{i=1}^{\infty}$  to a minimal triple  $[A, B, C]$ ) was viewed as an idealized form of the identification problem.

In practice, pre-processing of data from input-output experiments into the sequence  $\{L_i\}_{i=0}^{\infty}$  is not the preferred way. There are alternatives founded on statistical methodologies, e.g., the canonical correlation analysis dues to Akaike (1977).

## Remark on System Identification

System identification algorithms may be viewed as dynamical systems on spaces of transfer functions. For  $m = p = 1$ , the space  $Rat(n)$  of strictly proper rational functions of McMillan degree  $n$  was identified as an object of study by Brockett(1976).  $Rat(n)$  has very interesting topological and geometric structures. It admits interesting dynamics - e.g., a flow equivalent to the famous integrable system of Toda.

## Gramians and Reduction

For time-invariant linear systems  $[A, B, C]$ , with spectrum  $(A) \subseteq \mathbb{C}^-$ , let

$$W_c = \int_0^\infty \exp(tA)BB' \exp(tA')dt$$

and

$$W_o = \int_0^\infty \exp(tA')C'C \exp(tA)dt$$

A minimal triple  $[A, B, C]$  is balanced iff  $W_c = W_o = \Sigma = \text{diag}(\sigma_1, \sigma_2, \dots, \sigma_n)$  where the Hankel singular values

$$\sigma_i = (\lambda_i(W_c W_o))^{1/2} \quad i = 1, 2, \dots, n$$

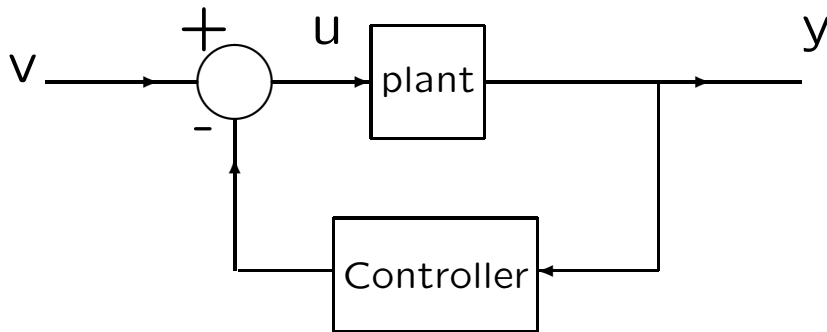
## Gramians and Reduction, cont'd

and are ordered such that

$$\sigma_i \geq \sigma_{i+1}$$

Balancing + truncation  $\rightarrow$  reduction.  
Connections to PCA.

## Closing the Loop



$$[A, B, C, ] \rightarrow [A - BKC, B, C]$$

$$\begin{array}{l} \dot{x} = Ax + Bu \\ y = Cx \end{array} \rightarrow \begin{array}{l} \dot{x} = (A - BKC)x + Bv \\ y = Cx \end{array}$$

$$W(t, \sigma) \rightarrow W^f(t, \sigma)$$

where

$$W^f(t, t_0) = W(t, t_0) - \int_{t_0}^t W(t, \sigma) K(\sigma) W^f(\sigma, t_0) d\sigma$$



## Closing the Loop, cont'd

In the time-invariant case

$$G^f(s) = G(s) - G(s)KG^f(s)$$

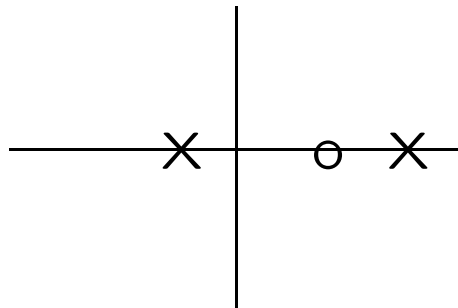
$$\Leftrightarrow G^f = (\mathbb{1} + GK)^{-1}G$$

## Closing the Loop

Feedback alters the system response. But there are severe limits to how much can be done with a constant gain controller.

Example:

$m = p = 1$   
 $g(s)$  with pole-zero pattern



Output feedback cannot move the r.h.p. pole to the left.

Root-locus calculations tell us why.  
Dynamic compensators help.

## State Feedback

Consider the linear time invariant system

$$\dot{x} = Ax + Bu$$

$$y = Cx$$

Under state feedback

$$u = -Kx + v$$

$$[A, B, C] \rightarrow [A - BK, B, C].$$

State feedback preserves reachability properties (set of attainable state trajectories is unchanged), but not observability properties.

## State Feedback

In the single input setting, there is a change of variables such that

$$A = \begin{bmatrix} 0 & 1 & 0 & \cdots & 0 \\ 0 & \cdot & & & \\ 0 & \cdot & \cdot & \cdot & 1 \\ -p_0 & -p_1 & \cdots & \cdots & -p_{n-1} \end{bmatrix}$$

$$B = \begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$$

the control canonical form. In this case letting  $\chi_A(s)$  denote the characteristic polynomial of  $A$ , there is  $K = (k_0, k_1, \dots, k_{n-1})$  s.t.

$$\begin{aligned} \chi_{A-BL}(s) &= s^n + \beta_{n-1}s^{n-1} + \cdots + \beta_0 \\ &= \beta(s) \end{aligned}$$

for any polynomial  $\beta$ .

# State Feedback

Essentially, the same idea can be used to show

## **Theorem** (Pole Placement)

Let  $[A, B]$  be a controllable pair. Then there exists a state feedback  $K$  *s.t.*

$$\chi_{A-BK}(s) = \beta(s)$$

for any desired polynomial  $\beta(s)$  of degree  $n =$  dimension of state space.

**Remark** There is a famous canonical form associated with P. Brunovsky for controllable pairs  $[A, B]$ , under the feedback group

$$\begin{array}{ll} A \rightarrow PAP^{-1} & B \rightarrow PB \\ A \rightarrow A - BK & B \rightarrow B \\ A \rightarrow A & B \rightarrow BQ \end{array}$$

## State Feedback

One application of the pole placement theorem is to find a feedback law  $u = -Kx + v$  such that all eigenvalues of  $(A - BK)$  are in  $\mathbb{C}^-$  the open *l.h.p.* The theorem guarantees such a  $K$ . There are a number of approaches to finding such stabilizing feedback laws. See lectures of Khaneja.

## A Difficulty with State Feedback

- State variables are typically not directly measurable.
- A set of linear combinations of state variables may be all that is available.
- Is there a way to stabilize the system?

**Idea:** Use output variables to estimate the state (asymptotically). Then substitute estimates for actual state.

# Observer/Estimator

Consider

$$\dot{x} = Ax + Bu$$

$$y = Cx$$

a time invariant system.

Assume that  $[A, B, C]$  is minimal. Consider the associated system

$$\dot{\hat{x}} = (A - \Gamma C)\hat{x} + Bu + \Gamma y$$

(This new system accepts as inputs, the original inputs  $u$  together with the outputs of the original system.)



## Observer/Estimator, cont'd

Let  $e = x - \hat{x}$ .

Then  $\dot{e} = (A - \Gamma C)e$ .

Observability of  $[A, C]$

$\leftrightarrow$  Reachability of  $[A', C']$

$\leftrightarrow$  spectrum assignability of  $(A' - C'\Gamma')$

$\leftrightarrow$  spectrum assignability of  $(A - \Gamma C)$

(by pole placement theorem)

## Observer/Estimator

Choose  $\Gamma$  so that all eigenvalues of  $(A - \Gamma C)$  are in  $\mathbb{C}^-$ .

Then, as  $t \rightarrow \infty$ ,

$$e(t) \rightarrow 0$$

$$x(t) \rightarrow \hat{x}(t)$$

Thus the state  $\hat{x}(t)$  asymptotically estimates  $x(t)$ .

Suppose  $K$  is such that  $(A - BK)$  has spectrum  $\subset \mathbb{C}^-$ . Consider the closed loop system

$$\dot{x} = Ax + B(-K\hat{x} + v)$$

obtained by using the state estimate  $\hat{x}$  in place of  $x$ .

## Observer/Estimator/Controller

Then the combination of observer, plant and controller takes the form

$$\begin{aligned}\dot{x} &= Ax - BK(x - e) + Bv \\ &= (A - BK)x + BKe + Bv \\ \dot{e} &= (A - \Gamma C)e\end{aligned}$$

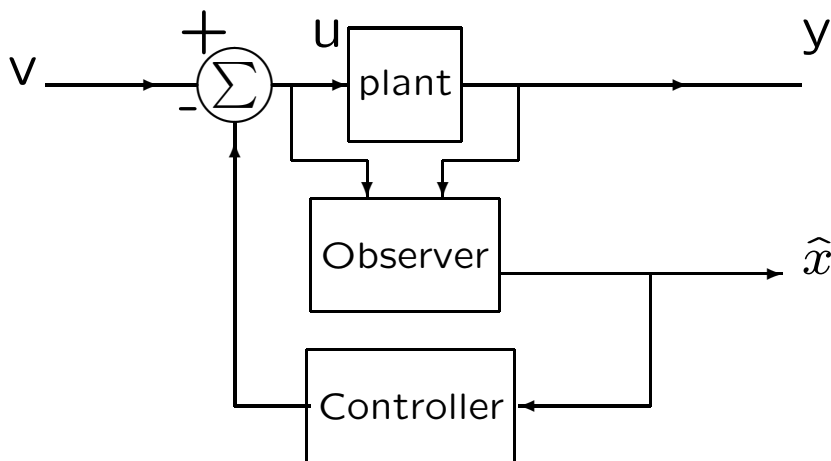
The *overall* dynamics is governed by the matrix

$$A_{overall} = \begin{bmatrix} A - BK & BK \\ 0 & A - \Gamma C \end{bmatrix}$$

choice of  $K$  and  $\Gamma$  as above ensures that spectrum  $(A_{overall}) \subseteq \mathbb{C}^-$

## Separation Theorem

The control system structure above takes the form



This is a powerful prototype for control system design in which the choices of controller and observer parameters can be made in a separated/independent manner.

## Separation Theorem

We return to this idea in later lectures of Khaneja and James from a variety of perspectives.