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#### Introduction to State-Space

*or* "States... in... spaaace!"

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### **Previously on METR4202...**

- Robotics, kinematics, perception, oh my!
- You built robot arms to drive an end-effector to a specific point in space some of you did well!
- ...and many of you discovered that that's hard to do accurately!

### What you already know\*

- Signals can be represented by transfer functions in the s-domain
- Roots of a transfer function's denominator (poles) indicate the stability of the system
- Poles move around under feedback control
   Feedback can stabilise an unstable system

\*If you have no idea what I'm talking about, now is the time to mention it.

### A quick recap

• Differential equations are used to represent the dynamics of systems in time:

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

• For linear systems, we use the Laplace transform to represent differential operators:  $sx = \mathcal{L}{f(x, t)}$ 

### A quick recap

• For SISO LTI\* systems, output y is a linear function of input u in the Laplace domain: y = Hu

*H* is the 'transfer function' relating *y* and *u* 

• We use block diagrams to represent such systems in convenient graphical form:



\*Single Input Single Output, Linear Time Invariant

#### State-space lolwut?

• A 'clean' way of representing systems

• Easy implementation in matrix algebra

• Simplifies understanding Multi-Input-Multi-Output (MIMO) systems

#### Introduction to states\*

- Introductory brain-teaser:
  - If you have a step response model of a system with integration, how do you represent nonzero initial conditions?

Eg. how would you setup a simulation of a step response, mid-step?



7 \*Not-insubstantial portions of these slides are based on Franklin, Powel and Enami-Naeni

#### Store the values

- The time-history of dynamic systems can be encapsulated by 'states'.
- A state is any previous value upon which future outputs depend:
  - Eg. velocities, altitude, displacement, charge potential, stored magnetic fields, etc.

All the state values of a system are stored in a single column vector *x*, which is collectively termed "the system's state".



### Finding states

• Linear systems can be written as networks of simple dynamic elements:



### Finding states

- We can identify the nodes in the system
  - These nodes contain the integrated time-history values of the system response
  - We call them "states"



#### Linear system equations

• We can represent the dynamic relationship between the states with a linear system:

$$\dot{x_1} = -7x_1 - 12x_2 + u$$
  
$$\dot{x_2} = x_1 + 0x_2 + 0u$$
  
$$y = x_1 + 2x_2 + 0u$$



#### State-space representation

• We can write linear systems in matrix form:

$$\dot{\boldsymbol{x}} = \begin{bmatrix} -7 & 12\\ 1 & 0 \end{bmatrix} \boldsymbol{x} + \begin{bmatrix} 1\\ 0 \end{bmatrix} \boldsymbol{u}$$
$$\boldsymbol{y} = \begin{bmatrix} 1 & 2 \end{bmatrix} \boldsymbol{x} + 0\boldsymbol{u}$$

Or, more generally:  $\dot{x} = Ax + Bu$  "State-space y = Cx + Du equations"

### Why "State-space"?

• State vector entries can be thought of as coordinates in a space; hence 'state-space'



#### State-space representation

- State-space matrices are not necessarily unique representations of a dynamic system
   There are several common forms (here's two)
- <u>Control canonical form</u>
  - Each node each entry in x represents a state of the system (each order of *s* maps to a state)
- Modal form
  - Diagonals of the state matrix A are the poles ("modes") of the transfer function

#### Other forms

- There are other representations are useful for other purposes:
  - Observer canonical form
  - Phase variable canonical form
  - Jordan canonical form

– Etc.

But you don't need to know about those for this course, but check out http://www.ece.rutgers.edu/~gajic/psfiles/canonicalforms.pdf if you're interested!

For now, let's focus on CCF and MF

### Control canonical form

• CCF matrix representations have the following structure:





#### Modal form

• MF matrix representations have the following structure:





#### EXAMPLE TIME





#### **BREAK TIME**



#### Cool robotics share

The McDonnell Douglas DC-X "Delta Clipper" was a demonstrator developed to explore vertical rocket landing and reusable single-stage to orbit technology. The DC-X used nonlinear state-space control in its flight attitude regulation avionics. Testing of the prototype was carried out by NASA, but the design was a competitor for NASA's own X-33 craft. Despite a rigorous testing schedule, the public success of the DC-X (compared to the embarrassing delays plaguing the X-33) led to its continued development. However, punishing deadlines and burned-out ground crew eventually resulted in a landing accident that severely damaged the craft – the program was quickly cancelled.





### DC-X Delta Clipper

# DC-X Flight #8 July 7, 1995 The "Swan Dive" Test

#### State variable transformation

- Important note!
  - The states of a control canonical form system are not the same as the modal states
  - They represent the same dynamics, and give the same output, but the vector values are different!
- However we can convert between them:

– Consider state representations, x and q where

$$x = Tq$$

**T** is a "transformation matrix"

#### State variable transformation

• Two homologous representations:

 $\dot{x} = \mathbf{A}x + \mathbf{B}u$  and  $\dot{q} = \mathbf{F}q + \mathbf{G}u$  $y = \mathbf{C}x + Du$   $y = \mathbf{H}q + Ju$ 

#### We can write:

$$\dot{x} = \mathbf{T}\dot{q} = \mathbf{A}\mathbf{T}q + \mathbf{B}u$$
$$\dot{q} = \mathbf{T}^{-1}\mathbf{A}\mathbf{T}q + \mathbf{T}^{-1}\mathbf{B}u$$
Therefore,  $\mathbf{F} = \mathbf{T}^{-1}\mathbf{A}\mathbf{T}$  and  $\mathbf{G} = \mathbf{T}\mathbf{B}$ Similarly,  $\mathbf{C} = \mathbf{H}\mathbf{T}$  and  $D = J$ 

#### Consider...

• What if we try to turn an arbitrary state description into control canonical form?

- We expect that for  $AT^{-1} = T^{-1}F$ :

 $\begin{bmatrix} -a_1 & -a_2 & \dots & -an \\ 1 & 0 & & 0 \\ 0 & \ddots & & \vdots \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} t_1 \\ t_2 \\ \vdots \\ t_n \end{bmatrix} = \begin{bmatrix} t_1 \mathbf{F} \\ t_2 \mathbf{F} \\ \vdots \\ t_n \mathbf{F} \end{bmatrix}$ where  $t_i$  are the rows of  $\mathbf{T}^{-1}$ 

Then,  $t_2 = t_3 \mathbf{F}$ , and  $t_1 = t_2 \mathbf{F} = t_3 \mathbf{F}^2$ , etc...

Note:  $t_n$  is a row and  $t_n \mathbf{F}$  yields a row

#### Consider...

•  $\mathbf{T^{-1}G} = \mathbf{B}$ , in control canonical form yields

$$\begin{bmatrix} t_1 G \\ t_2 G \\ \vdots \\ t_n G \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$$

. . .

The two results together give:

$$t_n \mathbf{G} = 0$$

$$t_2 \mathbf{G} = t_n \mathbf{F}^{n-1} \mathbf{G} = 0$$
$$t_1 \mathbf{G} = t_n \mathbf{F}^n \mathbf{G} = 1$$

#### Consider...

• Look at the last entry for  $t_3...$ 

- We can write this as:  $t_3[G \ FG \ ... \ F^nG] = [0 \ 0 \ ... \ 1]$ Or

$$t_3 = \begin{bmatrix} 0 & 0 & \dots & 1 \end{bmatrix} \mathcal{C}^{-1}$$
  
where  $\mathcal{C} = \begin{bmatrix} \mathbf{G} & \mathbf{F}\mathbf{G} & \mathbf{F}^2\mathbf{G} & \cdots & \mathbf{F}^{n-1}\mathbf{G} \end{bmatrix}$ 

This is called the "controllability matrix"

### Controllability matrix

To convert an arbitrary state representation in F, G, H and J to control canonical form A, B, C and D, the controllability matrix C = [G FG F<sup>2</sup>G ··· F<sup>n-1</sup>G] must be invertible (i.e. full rank).

>deep think<

Why is it called the "controllability" matrix?

### Controllability matrix

• If you can write it in CCF, then the system equations must be linearly independent.

• Transformation by any invertible matrix preserves the controllability of the system.

• Thus, a invertible controllability matrix means *x* can be driven to any value.

#### Kind of awesome

• The controllability of a system depends on the particular set of states you chose

• You can't tell just from a transfer function whether all the states of *x* are controllable

• System poles are the Eigenvalues of  $\mathbf{F}$ ,  $(p_i)$ 



#### State evolution

• Consider the system matrix relation:

$$\dot{\mathbf{x}} = \mathbf{F}\mathbf{x} + \mathbf{G}u$$
$$y = \mathbf{H}\mathbf{x} + Ju$$

The time solution of this system is:  $F(t-t_{0}) = \int_{-\infty}^{t} F(t-\tau) \mathbf{c} d\tau$ 

$$\mathbf{x}(t) = e^{\mathbf{F}(t-t_0)} \mathbf{x}(t_0) + \int_{t_0} e^{\mathbf{F}(t-\tau)} \mathbf{G}u(\tau) d\tau$$

If you didn't know, the matrix exponential is:  $e^{\mathbf{K}t} = \mathbf{I} + \mathbf{K}t + \frac{1}{2!}\mathbf{K}^{2}t^{2} + \frac{1}{3!}\mathbf{K}^{3}t^{3} + \cdots$ 



#### Stability

• We can solve for the natural response to initial conditions  $x_0$ :

$$\boldsymbol{x}(t) = e^{p_i t} \boldsymbol{x}_0$$
  
$$\therefore \, \dot{\boldsymbol{x}}(t) = p_i e^{p_i t} \boldsymbol{x}_0 = \mathbf{F} e^{p_i t} \boldsymbol{x}_0$$

homogenous

Clearly, a system will be stable provided eig(F) < 0

#### Characteristic polynomial

• From this, we can see  $\mathbf{F}\mathbf{x}_0 = p_i\mathbf{x}_0$ or,  $(p_i\mathbf{I} - \mathbf{F})\mathbf{x}_0 = 0$ which is true only when  $\det(p_i\mathbf{I} - \mathbf{F})\mathbf{x}_0 = 0$ Aka. the characteristic equation!

• We can reconstruct the CP in *s* by writing:  $det(sI - F)x_0 = 0$ 



#### Great, so how about control?

• Now that we have a state space model, how do we make the system stable, or converge to desired states?

Easy: Feedback!

• Given  $\dot{x} = Fx + Gu$ , if we know F and G, we can design a controller u = -Kx such that eig(F - GK) < 0

#### Full state feedback

• If we have full measurement and control of the states of *x*, we can position poles of the closed-loop system in arbitrary locations

– Modal form makes this straight forward:

$$\mathbf{F} - \mathbf{G}\mathbf{K} = \begin{bmatrix} -p_1 - G \cdot K_{1j} & & \\ & -p_2 - G \cdot K_{2j} & \\ & & -p_3 - G \cdot K_{3j} \end{bmatrix}$$

Of course, this hardly ever happens in reality.

**Brain teaser: Why?** 

### Example: PID control

• Consider a system parameterised by three states:  $x_1, x_2, x_3$  where  $x_2 = \dot{x}_1$  and  $x_3 = \dot{x}_2$ 

$$\dot{\mathbf{x}} = \begin{bmatrix} 1 & & \\ & 1 & \\ & & -2 \end{bmatrix} \mathbf{x} - \mathbf{K}u$$
$$y = \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} \mathbf{x} + 0u$$

 $x_2$  is the output state of the system;  $x_1$  is the value of the integral;  $x_3$  is the velocity.



### Example: PID control

• We can choose **K** to move the eigenvalues of the system as desired:  $det \begin{bmatrix} 1 - K_1 \\ 1 - K_2 \\ -2 - K_3 \end{bmatrix} = \mathbf{0}$ 

All of these eigenvalues must be positive.

It's straightforward to see how adding derivative gain  $K_3$  can stabilise the system.

#### In reality...

- You can never measure or apply control action to all states directly.
  - The majority of system states will be hidden to the control engineer.

#### But we can pretend!

• We can design a controller as if we did, using an estimate – an educated guess.



#### Observers

• Observers (aka "estimators") are used to infer the hidden states of a system from measured outputs.



A controller is designed using estimates in lieu of full measurements

#### Observers

- The state estimate can be treated like a control system itself
  - Dynamics to update the estimate:  $\dot{\hat{x}} = \mathbf{F}\hat{x} + \mathbf{G}\mathbf{u}$
  - By measuring an 'error signal' from the difference between the real output measurement and the output estimate,  $\tilde{x} = x - \hat{x}$ , the state estimate can be shown to converge

#### Observers

• Just like you might expect:  $\dot{\hat{x}} = F\hat{x}+Gu+L(y-H\hat{x})$   $\therefore \dot{\tilde{x}} = (F-LH)\tilde{x}$ Choose L to make  $\tilde{x}$  converge to 0



#### Observability

- The ability to infer these values is called "Observability"
  - This is the dual of controllability; a system that is observable is also controllable and vice versa.
  - Observability matrix:

$$\boldsymbol{\mathcal{O}} = \begin{bmatrix} \mathbf{H} \\ \mathbf{HF} \\ \vdots \\ \mathbf{HF}^{n-1} \end{bmatrix}$$



### Just scratching the surface

• There is a lot of stuff to state-space control

• One lecture (or even two) can't possibly cover it all in depth

Go play with Matlab and check it out!

#### EXAMPLE TIME



:Ш:

### Quick plug\*

#### "Feedback Control of Dynamic Systems" by Franklin, Powell and Emami-Naeini. "Control System Design: An Introduction to State-Space Methods" by Friedland





\* No, they're not paying me – they're just really good books!



## Tune-in next time for...

# Control/Planning Under Uncertainty

Starring

#### Hanna Kurniawati!

Fun fact: Saying "eigenvalue" makes you feel smarter!



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# And now a word from our hideous sponsor!

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# EXAM TIPS

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Look at the examination information at www.uq.edu.au/myadvisor/examinations or email examinations@uq.edu.au

#### WHAT CAN I TAKE INTO THE EXAM?

#### Yourself



Student ID card

Writing implements and any authorised material

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### Discretisation FTW!

• We can use the time-domain representation to produce difference equations!

$$\boldsymbol{x}(kT+T) = e^{\mathbf{F}T} \boldsymbol{x}(kT) + \int_{kT}^{kT+T} e^{\mathbf{F}(kT+T-\tau)} \mathbf{G}\boldsymbol{u}(\tau) d\tau$$

Notice  $u(\tau)$  is not based on a discrete ZOH input, but rather an integrated time-series.

We can structure this by using the form:

 $u(\tau) = u(kT), \qquad kT \le \tau \le kT + T$ 



#### Discretisation FTW!

• Put this in the form of a new variable:  $\eta = kT + T - \tau$ 

Then:

$$\boldsymbol{x}(kT+T) = e^{\boldsymbol{F}T}\boldsymbol{x}(kT) + \left(\int_{kT}^{kT+T} e^{\boldsymbol{F}\eta}d\eta\right)\boldsymbol{G}\boldsymbol{u}(kT)$$

Let's rename  $\mathbf{\Phi} = e^{FT}$  and  $\mathbf{\Gamma} = \left(\int_{kT}^{kT+T} e^{F\eta} d\eta\right) \mathbf{G}$ 



#### Discrete state matrices

#### So,

$$\mathbf{x}(k+1) = \mathbf{\Phi}\mathbf{x}(k) + \mathbf{\Gamma}u(k)$$
$$y(k) = \mathbf{H}\mathbf{x}(k) + \mathbf{J}\mathbf{u}(k)$$

Again,  $\mathbf{x}(k+1)$  is shorthand for  $\mathbf{x}(kT+T)$ 

#### Note that we can also write $\Phi$ as: $\Phi = \mathbf{I} + \mathbf{F}T\Psi$

where

$$\Psi = \mathbf{I} + \frac{\mathbf{F}T}{2!} + \frac{\mathbf{F}^2 T^2}{3!} + \cdots$$

### Simplifying calculation

- We can also use  $\Psi$  to calculate  $\Gamma$ 
  - Note that:

$$\Gamma = \sum_{k=0}^{\infty} \frac{\mathbf{F}^k T^k}{(k+1)!} T\mathbf{G}$$
$$= \Psi T\mathbf{G}$$

 $\Psi \text{ itself can be evaluated with the series:} \\ \Psi \cong \mathbf{I} + \frac{\mathbf{F}T}{2} \left\{ \mathbf{I} + \frac{\mathbf{F}T}{3} \left[ \mathbf{I} + \cdots \frac{\mathbf{F}T}{n-1} \left( \mathbf{I} + \frac{\mathbf{F}T}{n} \right) \right] \right\}$ 



#### State-space z-transform

#### We can apply the z-transform to our system: $(z\mathbf{I} - \mathbf{\Phi})\mathbf{X}(z) = \mathbf{\Gamma}U(k)$ $Y(z) = \mathbf{H}X(z)$

## which yields the transfer function: $\frac{Y(z)}{X(z)} = G(z) = \mathbf{H}(z\mathbf{I} - \mathbf{\Phi})^{-1}\mathbf{\Gamma}$

#### State-space control design

• Design for discrete state-space systems is just like the continuous case.

– Apply linear state-variable feedback:

 $u = -\mathbf{K}\mathbf{x}$ 

such that  $det(z\mathbf{I} - \mathbf{\Phi} + \mathbf{\Gamma}\mathbf{K}) = \alpha_c(z)$ 

where  $\alpha_c(z)$  is the desired control characteristic equation

Predictably, this requires the system controllability matrix  $C = [\Gamma \quad \Phi \Gamma \quad \Phi^2 \Gamma \quad \cdots \quad \Phi^{n-1} \Gamma]$  to be full-rank.