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FYS3240- 4240

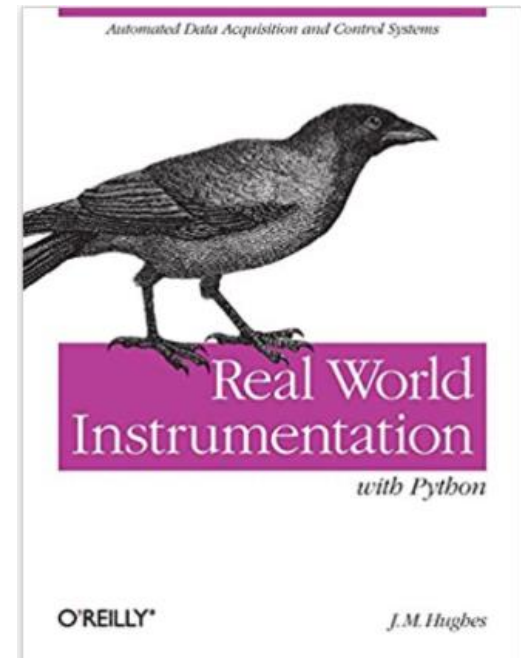
Data acquisition & control

# Control systems

Spring 2021 – Lecture #11



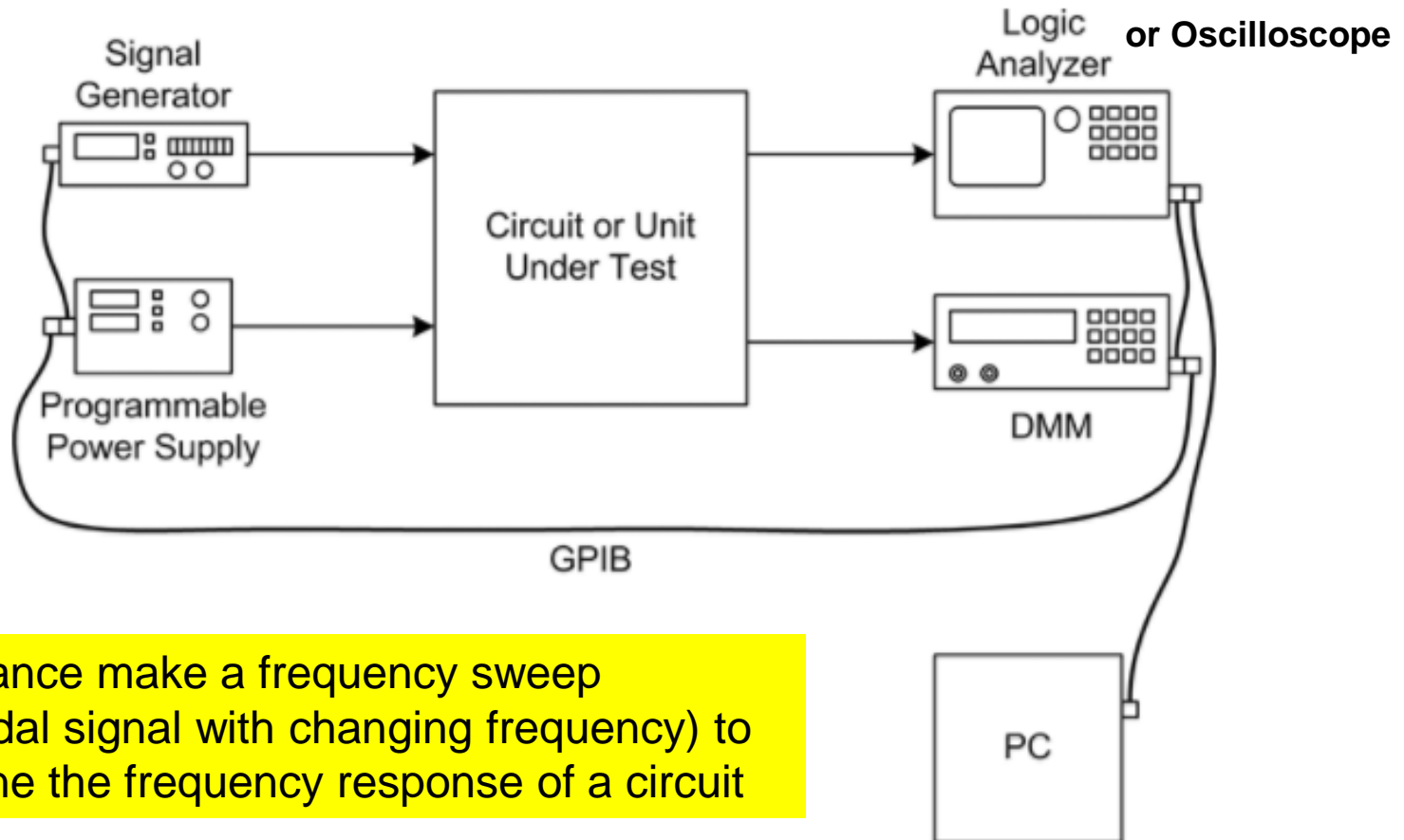
- Note: Some examples (Figures) from the book *Real World Instrumentation with Python: Automated Data Acquisition and Control Systems*
  - Ch. 9, page 303 – 339



# Topics

- Linear vs. nonlinear control examples
- Open loop vs. closed loop control
- Discrete-time closed loop system
- PID control
- Control system examples
  - Motor control
  - Water tank
  - Satellite control
  - Missile guidance and control

# PC-based automated lab test setup



For instance make a frequency sweep (sinusoidal signal with changing frequency) to determine the frequency response of a circuit

Figure from [Real World Instrumentation with Python \(oreilly.com\)](http://oreilly.com)

# Linear control systems

$$u(t) = K_p e(t) + P$$

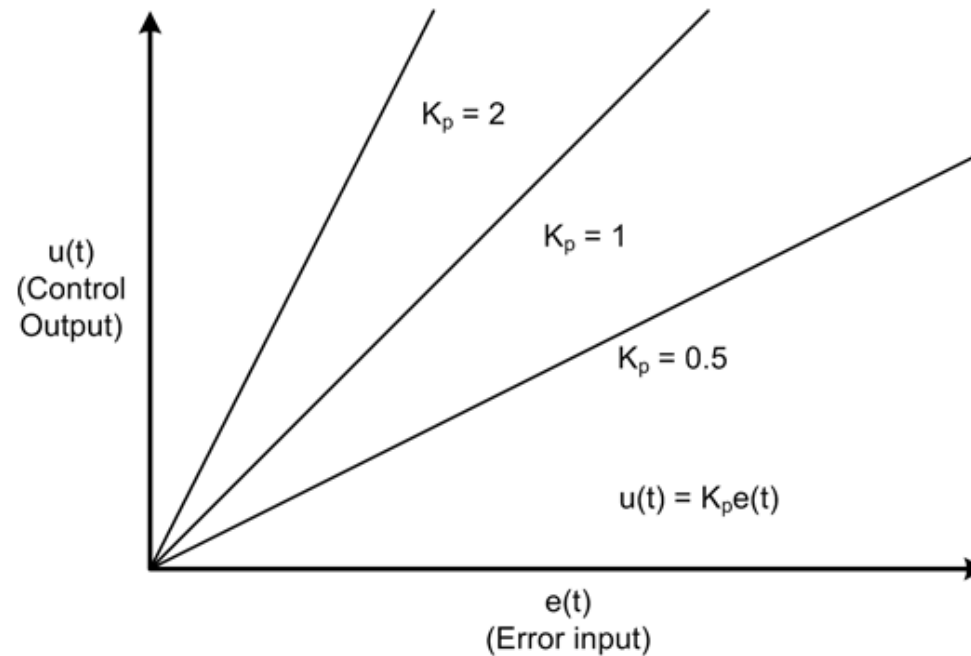
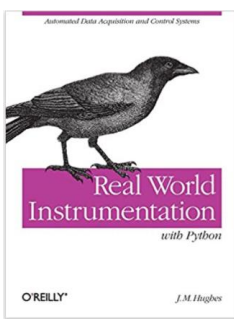


Figure 9-1. Linear control system proportional response



# Nonlinear control systems

Example 1 (on/off controller)

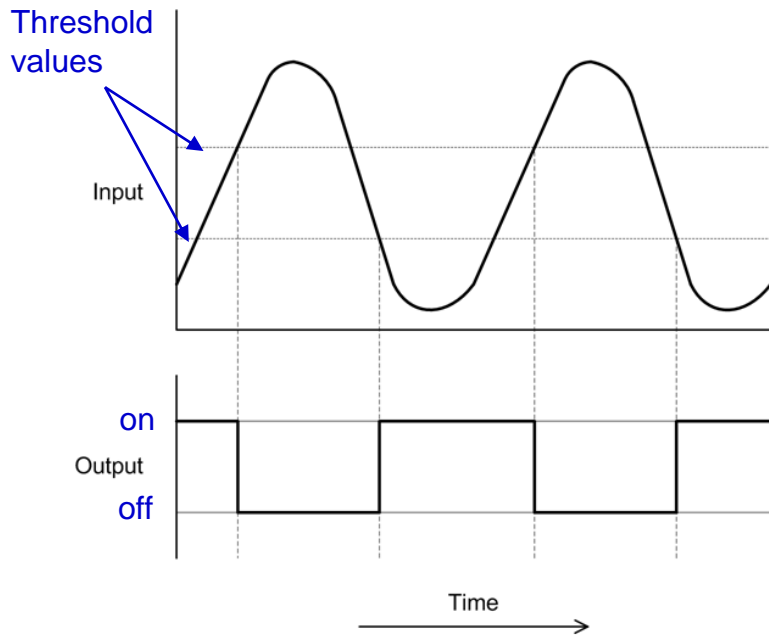


Figure 9-2. Nonlinear control system response

Example 2 (Pulse Width Modulation, PWM, controller)

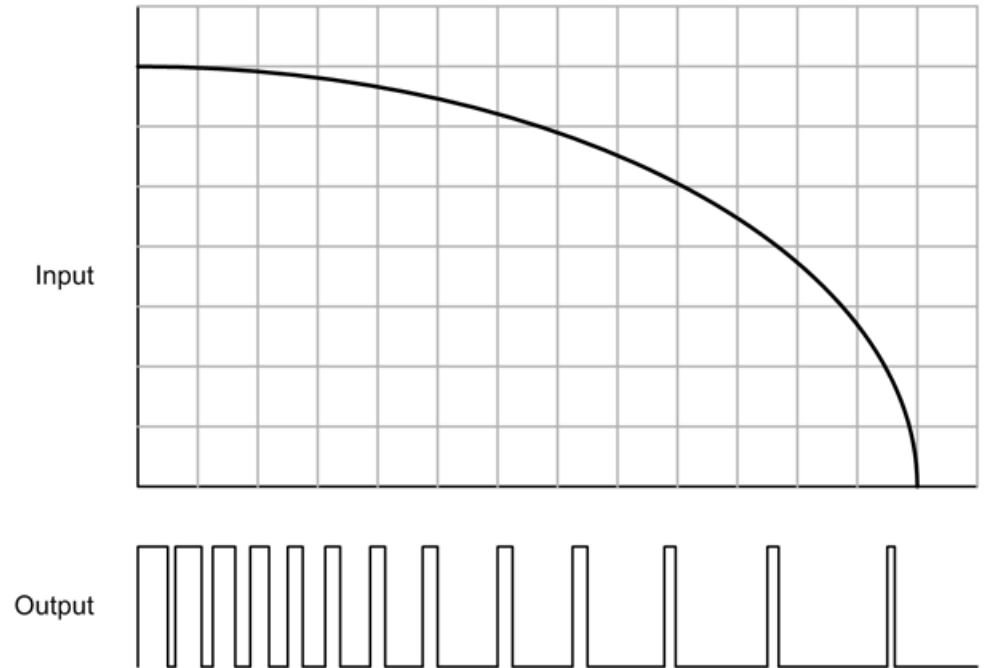
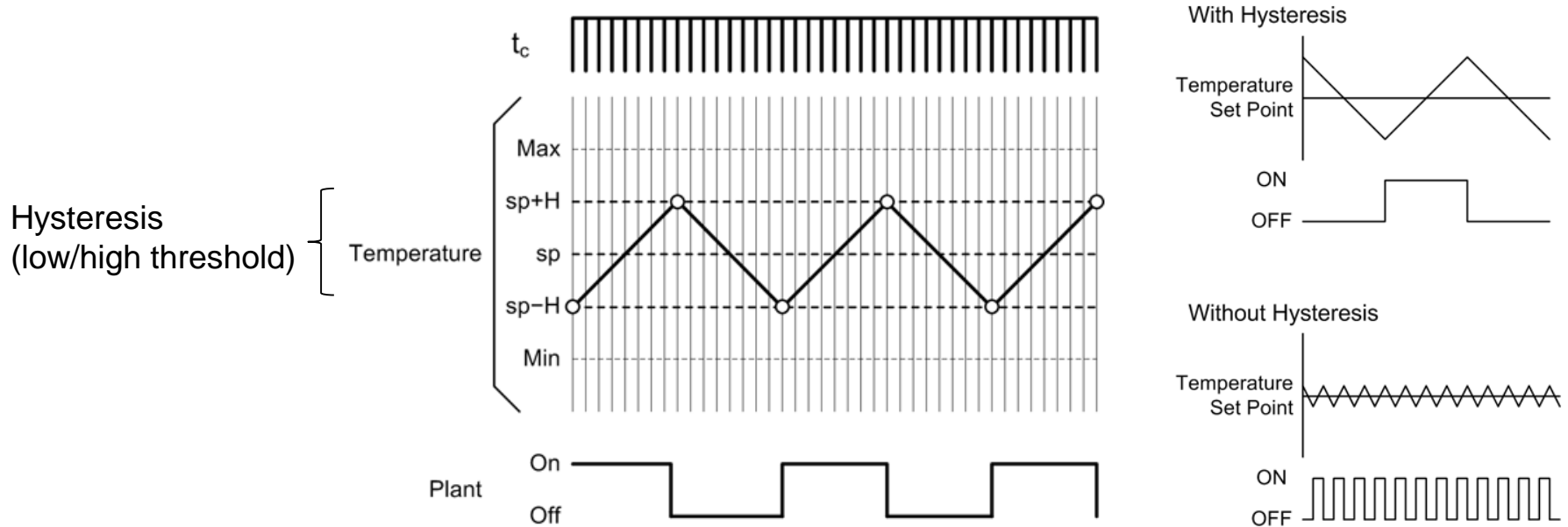
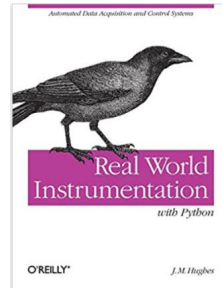


Figure 9-3. Nonlinear pulse control

# Nonlinear bang-bang (on/off) controllers

- On/off controller that switches between two states; either completely on or completely off.
  - Often used for temperature control.
  - Also used in old missiles for fin control (+/- full deflection)
- Often hysteresis is used
  - To avoid to frequent on/off switching.





# Sequential control systems

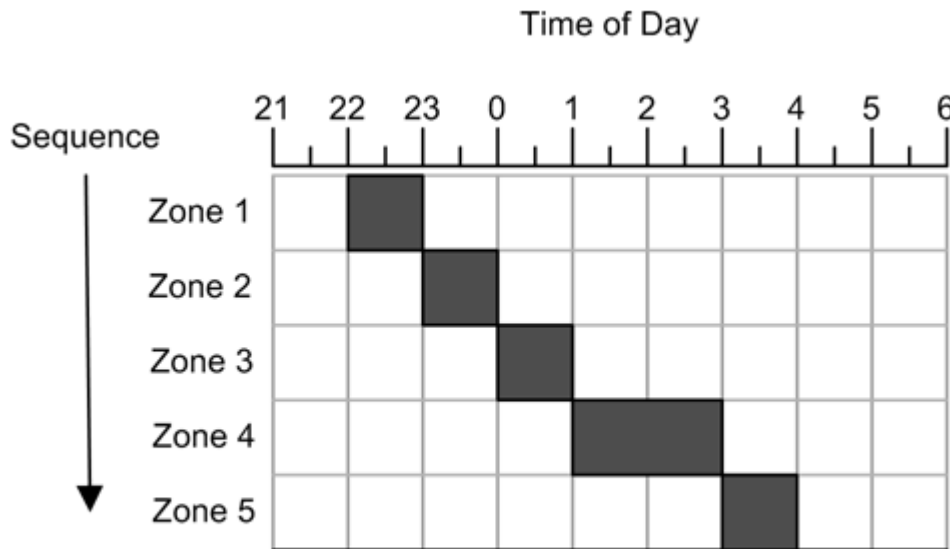


Figure 9-4. Sprinkler system sequential control

## Sequential power control

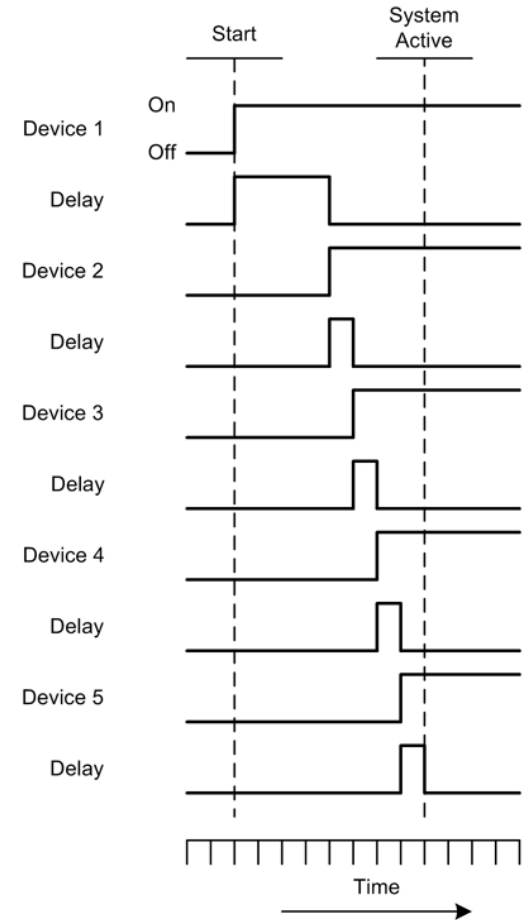
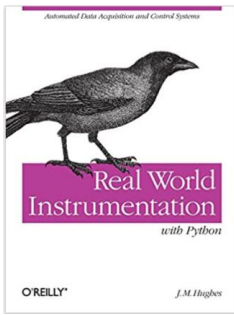


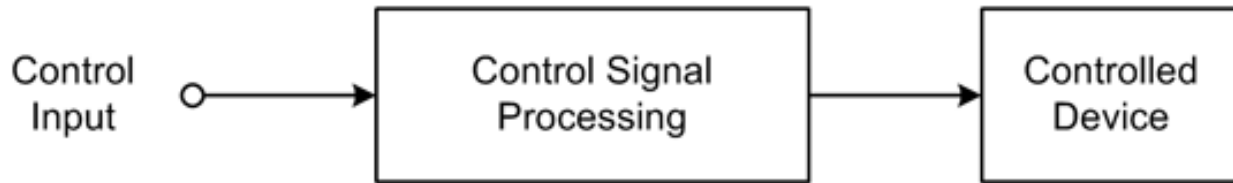
Figure 1-8. Sequential power control



# Open-loop control



# Open-loop control



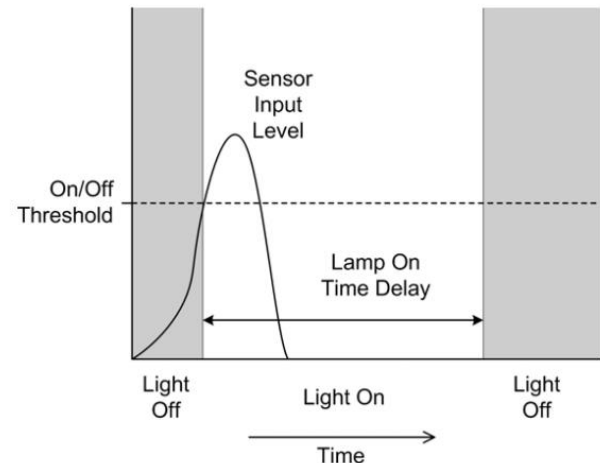
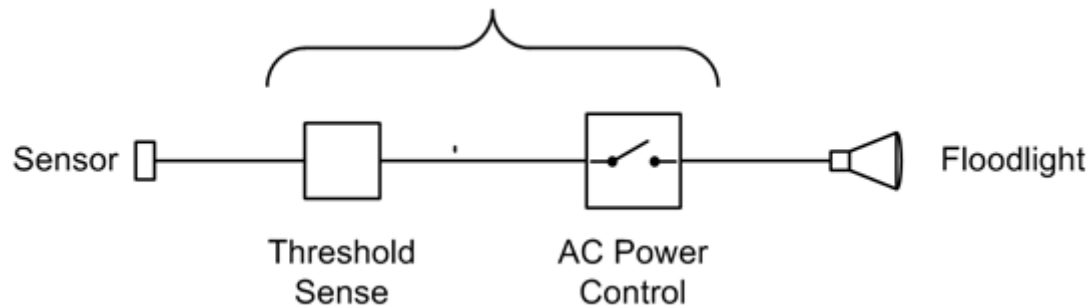
Possible Functions:

- Threshold (limit trip)
- Amplification
- Inversion
- Filtering
- Time Delay

Figure from [Real World Instrumentation with Python \(oreilly.com\)](https://oreil.ly/Real-World-Instrumentation-with-Python)

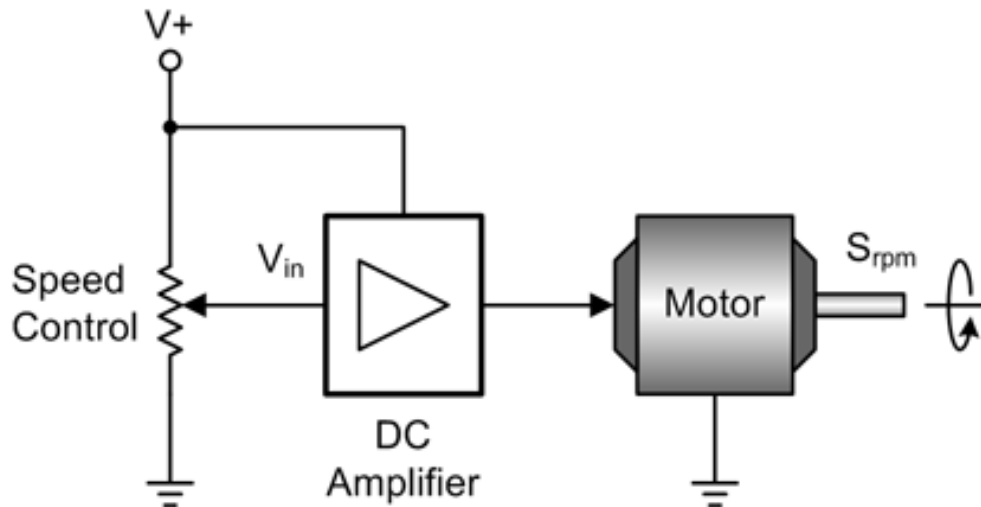
# Open-loop light control system

- Turn on a lamp on for a given time if a motion is detected.
- On/off control



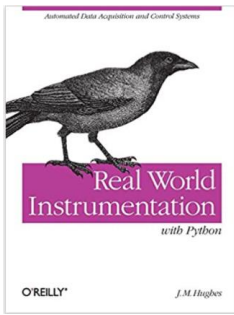
# Simple open-loop motor control

- Motor rotation rate will vary with load!
  - Not a good controller!

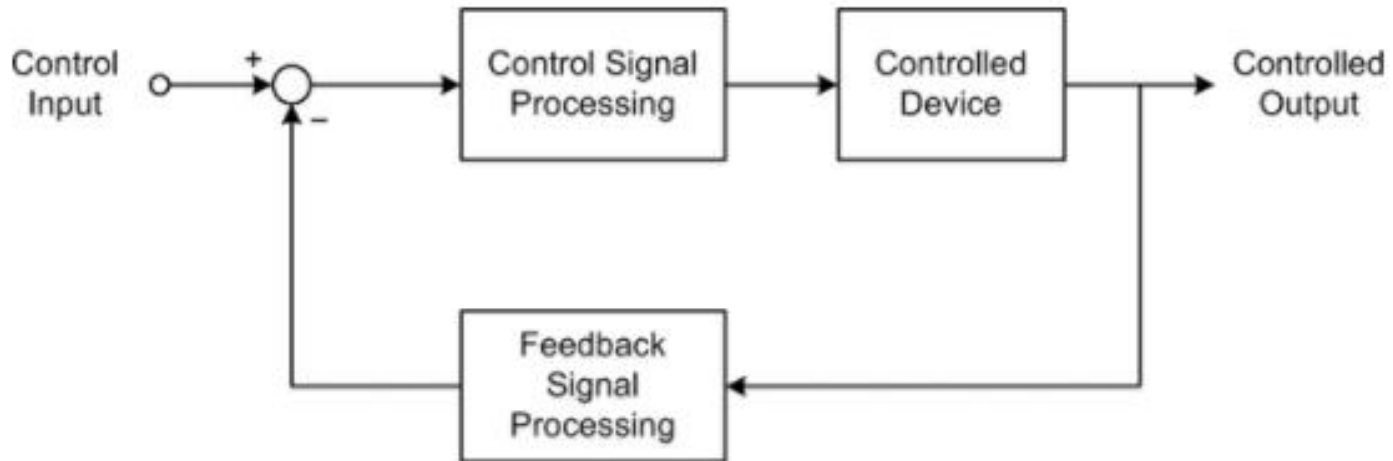


*Figure 9-13. Simple open-loop DC motor control*

# Closed-loop control



# Closed-loop control



*Figure 1-6. Closed-loop control*

# PWM motor speed control with feedback

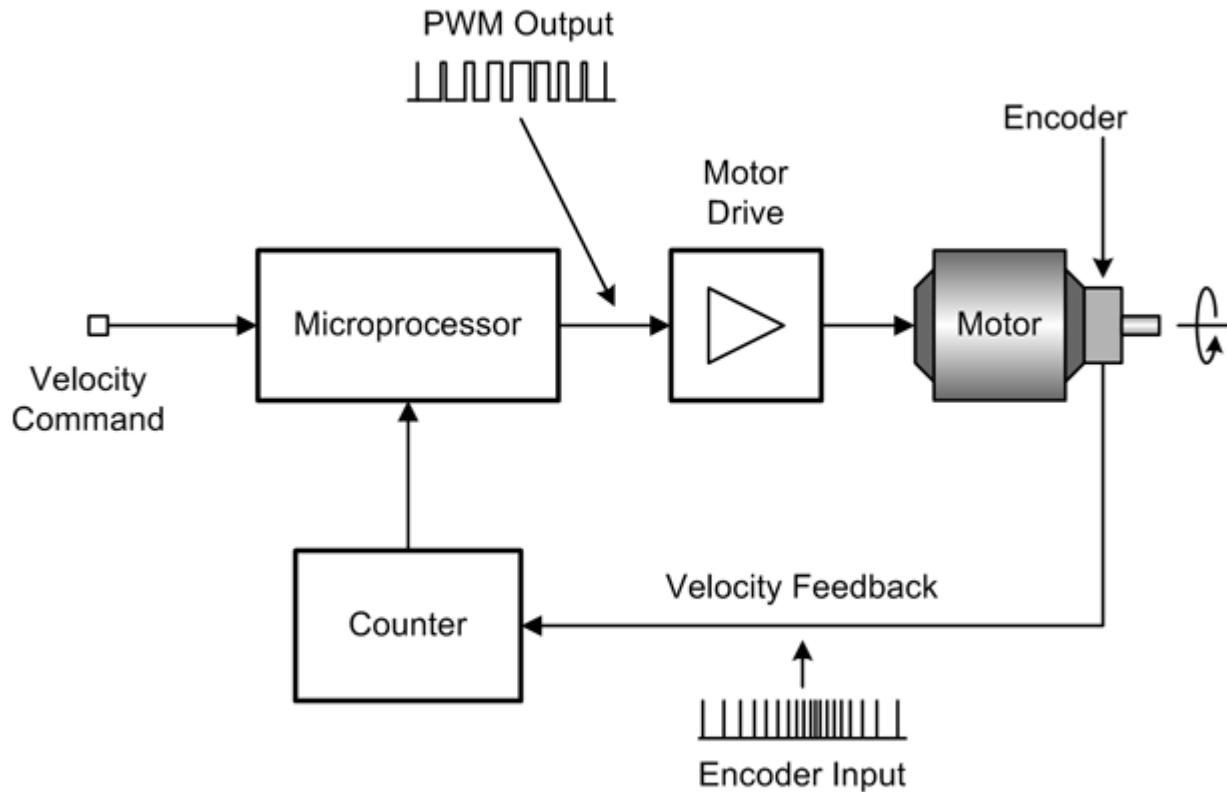


Figure from [Real World Instrumentation with Python \(oreilly.com\)](http://oreilly.com)

# Commercial DC motor controller with RPM feedback

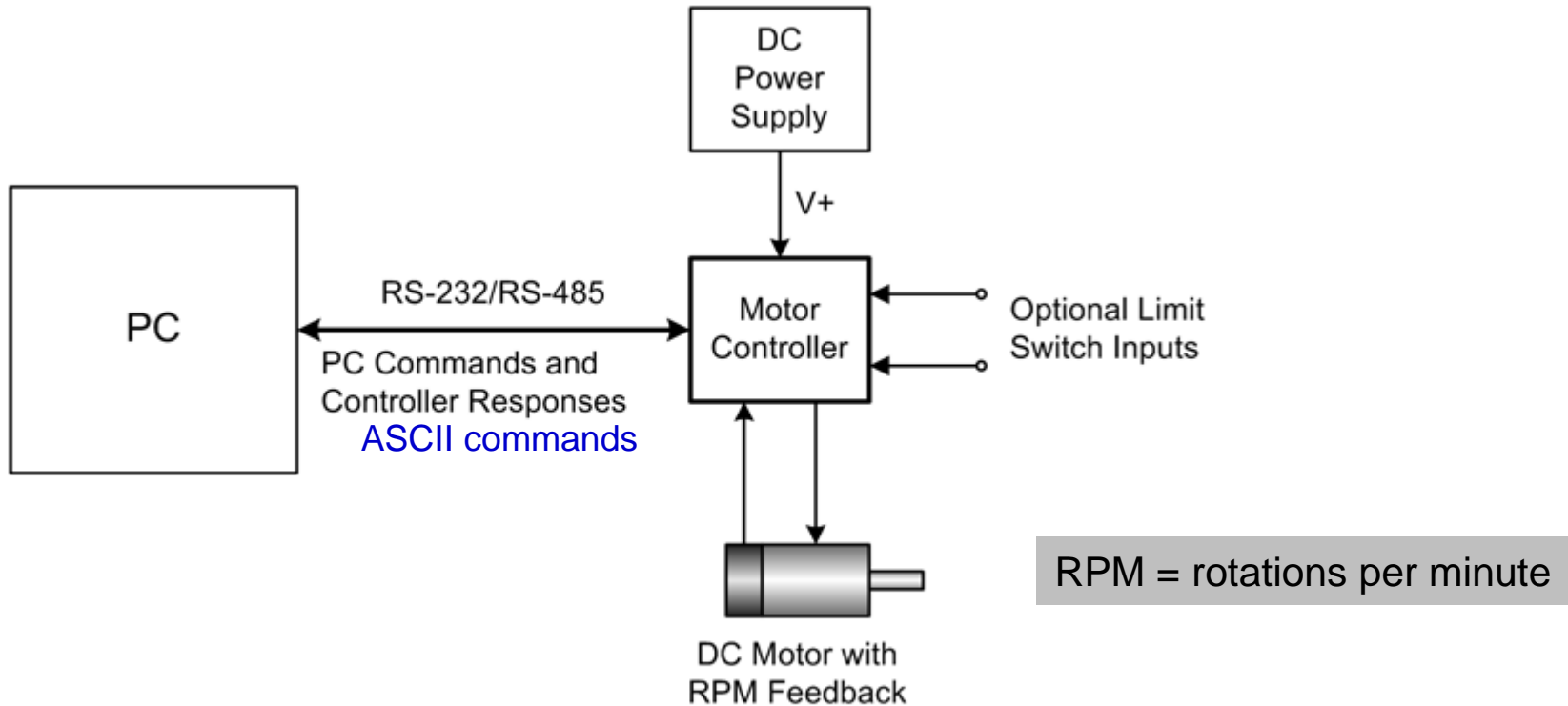


Figure 9-17. Commercial DC motor controller

Figure from [Real World Instrumentation with Python \(oreilly.com\)](http://oreilly.com)



# Closed-loop water tank control system

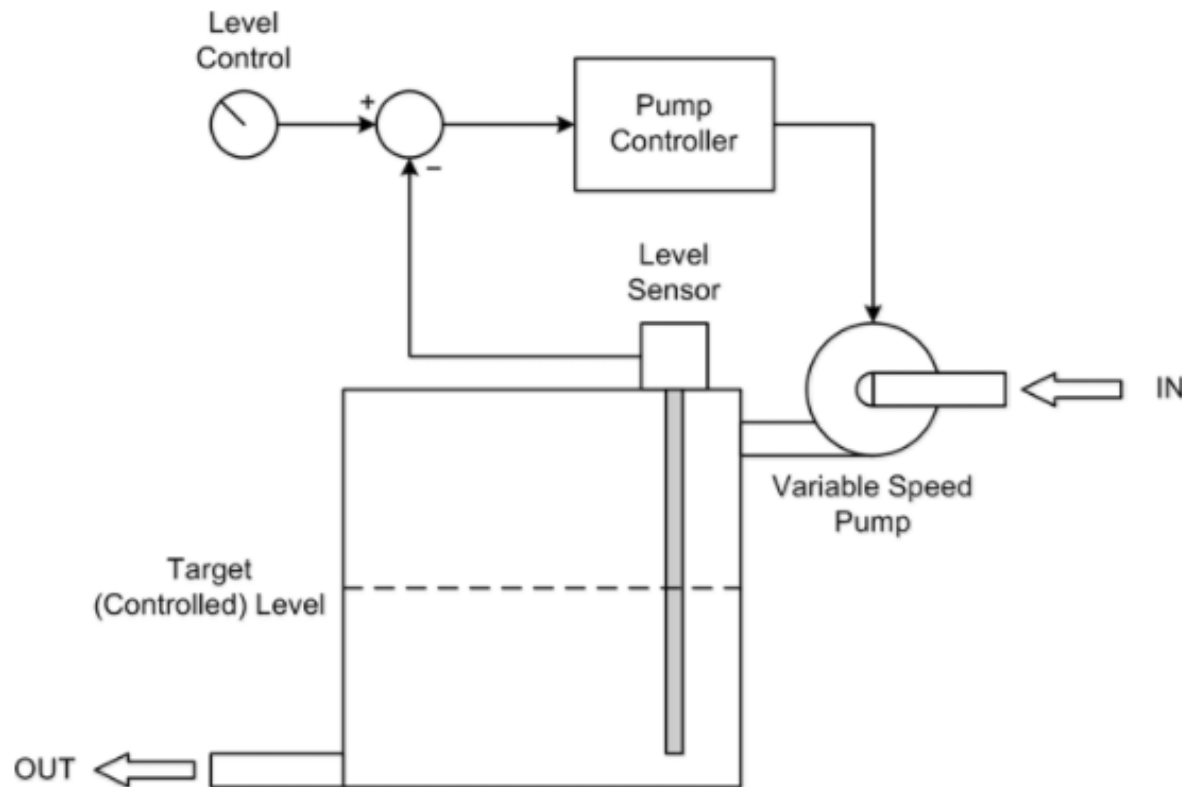
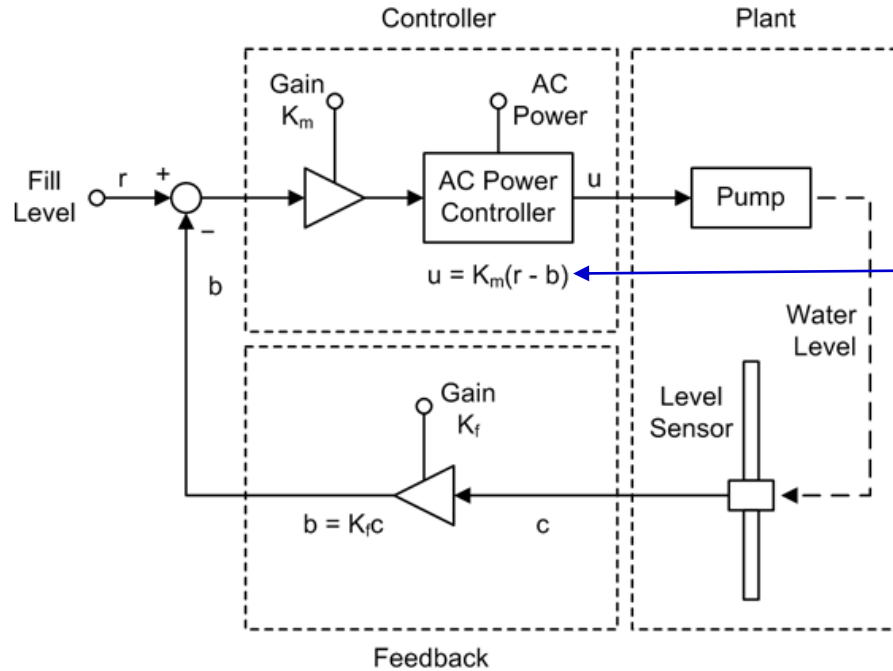


Figure from [Real World Instrumentation with Python \(oreilly.com\)](https://oreil.ly.com)

# Closed-loop water tank control system



Proportional control

Figure 9-11. Closed-loop water tank control system details

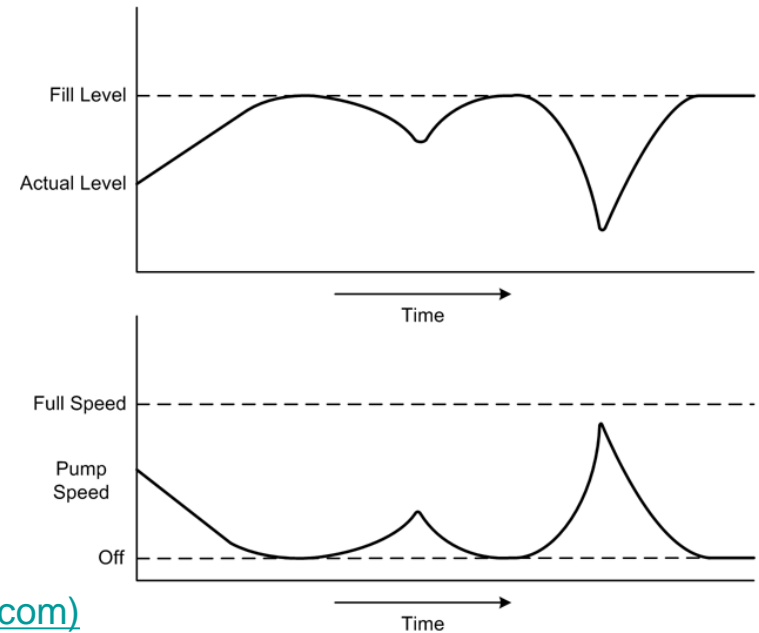


Figure 9-12. Water tank control system response graphs

# Discrete-time closed loop system

- ADC
- DAC
- Clock
  - For synchronization

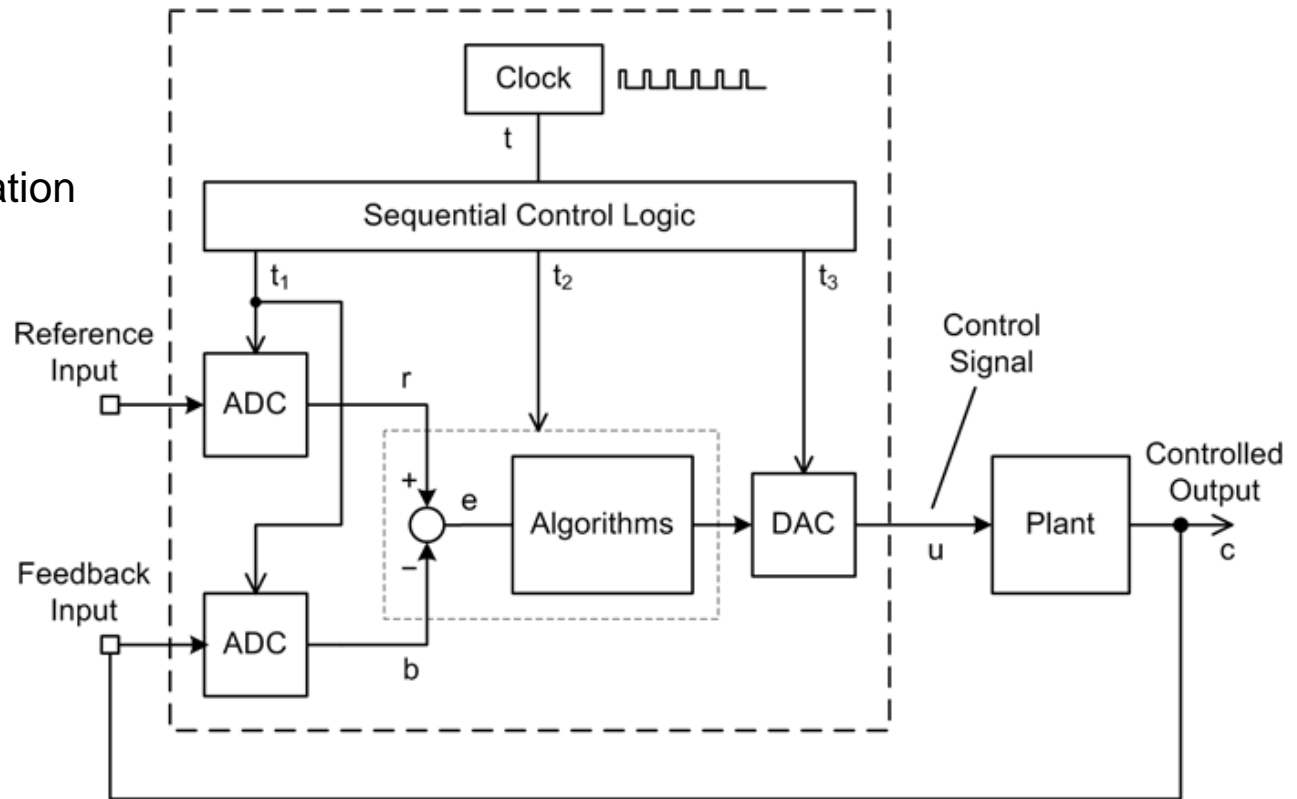
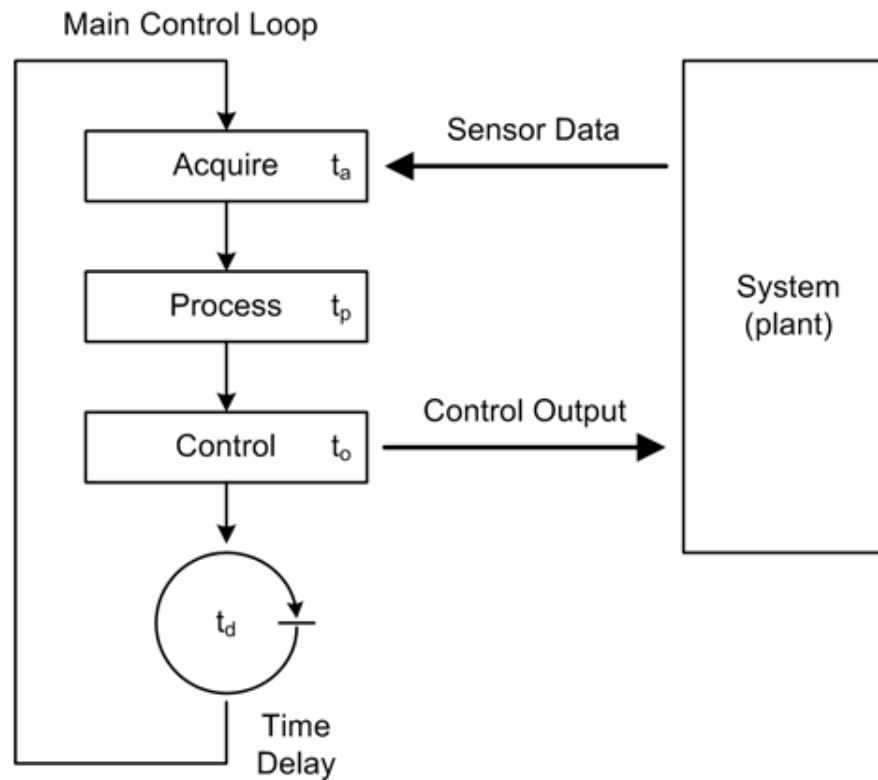


Figure from [Real World Instrumentation with Python \(oreilly.com\)](https://oreilly.com)

# Control software flow

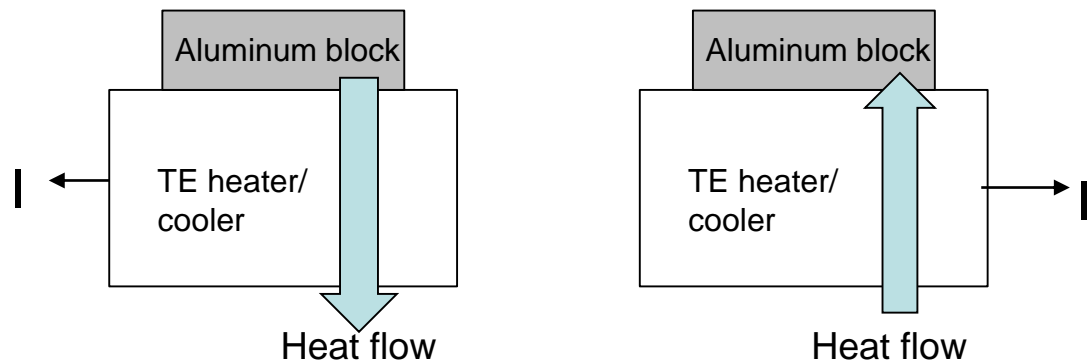
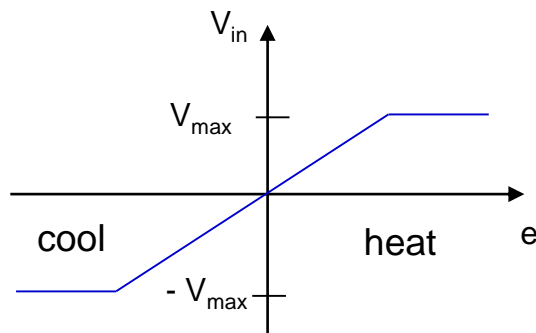


Figures from [Real World Instrumentation with Python \(oreilly.com\)](https://oreilly.com/catalog/errata/csp/bookerrata.php?isbn=9781492012122)

# PID control

# Temperature controller example

- We need to build a temperature controller that can control the temperature of an aluminum block to be close to a given set point  $T_{set\ point}$  which is higher than the surrounding temperature.
- We will read the temperature of the block,  $T_{block}$ , using a temperature sensor.
- Assume a Thermoelectric (TE) heater/cooler attached to the block, such that we can heat or cool the block by changing the current (I) direction.
- A voltage  $V_{in}$  is used to control the current flow
  - Using a voltage to current converter (driver) circuit.



# Proportional control of temperature

- The error  $e$  is defined to be:  $e = T_{set\ point} - T_{block}$
- The control voltage is set to  $V_{in} = K_p e$ , where  $K_p$  is a constant called the proportional gain.
- Problem:
  - When  $e = 0$  the power to the heater is turned of.
  - $\rightarrow$  The block will stabilize at a temperature below the set point, since the sample will lose heat to the surroundings.
- Solution:
  - Include a constant  $V_0$  that can counteract the heat loss to the surroundings

$$V_{in} = K_p e + V_0$$

# PI control of temperature

- Instead of trying to find the correct constant value for  $V_0$  we want to build some intelligence into the control system.
- We want to find the proper value of  $V_0$ , given the chosen set point, automatically.
  - We use an integral to construct the correct constant during runtime.

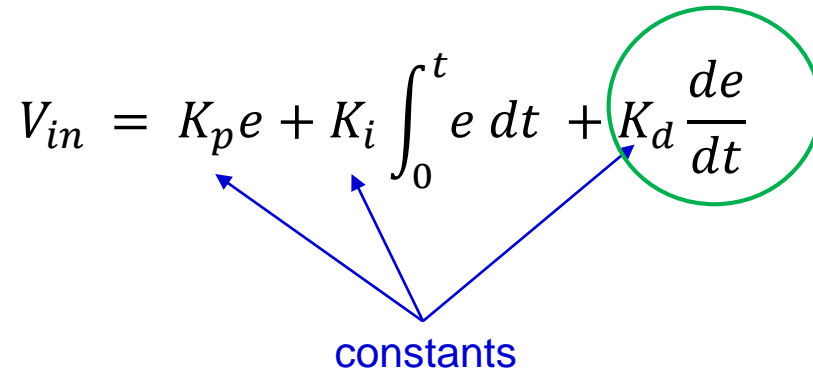
$$V_{in} = K_p e + K_i \int_0^t e dt$$

- The integral keeps a running sum of all the error values (up to time  $t$ ).
- $K_i$  is a constant that need to be selected.



# PID control of temperature

- A derivative term can be added to damp out oscillations

$$V_{in} = K_p e + K_i \int_0^t e dt + K_d \frac{de}{dt}$$


The diagram shows the PID control equation with three blue arrows pointing from the word "constants" to the gain terms  $K_p$ ,  $K_i$ , and  $K_d$ . The derivative term  $K_d \frac{de}{dt}$  is circled in green.

# Discrete-sampling PID control of temperature

- Sample interval of  $\Delta t$
- After  $n$  samples the continuous time equation can be approximated with:

$$V_{in} = K_p e_n + K_i \Delta t \underbrace{\sum_{m=0}^n e_m}_{\text{Sum over all error values since the control algorithm was turned on}} + \frac{K_d}{\Delta t} (e_n - e_{n-1})$$

Sum over all error values  
since the control algorithm  
was turned on

# PID controller summary

- Proportional-Integral-Derivative (PID) algorithm is the most common control algorithm
  - Used for heating and cooling systems, fluid level control, pressure control, ...
- Calculates a term **proportional to the error** - the P term.
- Calculates a term **proportional to the integral of the error** - the I term.
- Calculates a term **proportional to the derivative of the error** - the D term.
- The three terms - the P, I and D terms, are added together to produce a control signal that is applied to the system being controlled.
- Sometimes a PI-controller is used.

# PID controller – general terms

- A PID controller continuously calculates **an error value** as the difference between **a measured process variable** and a desired **set point**.
- The controller attempts to minimize the error  $e$  over time, by adjustment of a *control variable*  $u(t)$ , such as the position of a control valve.

$$u(t) = K_p e(t) + K_i \int_0^t e(t) dt + K_d \frac{de(t)}{dt}$$

- $P$  accounts for present values of the error.
- $I$  accounts for past values of the error, accumulates over time.
- $D$  accounts for possible future values of the error, based on its current rate of change.
- Must tune the coefficients  $K_p$ ,  $K_i$  and  $K_d$ .

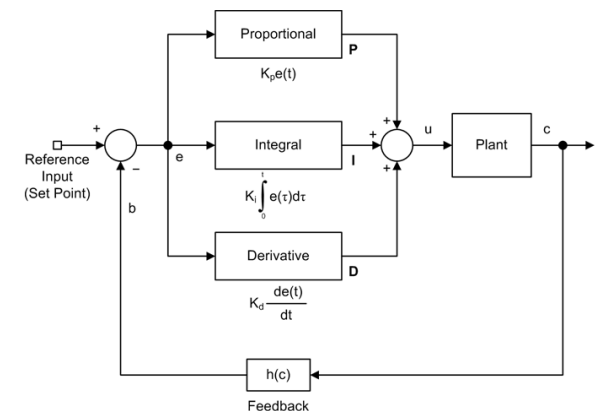


Figure 9-24. PID control block diagram

In general PID does not provide *optimal* control, since no modelling of the Plant/process is used.

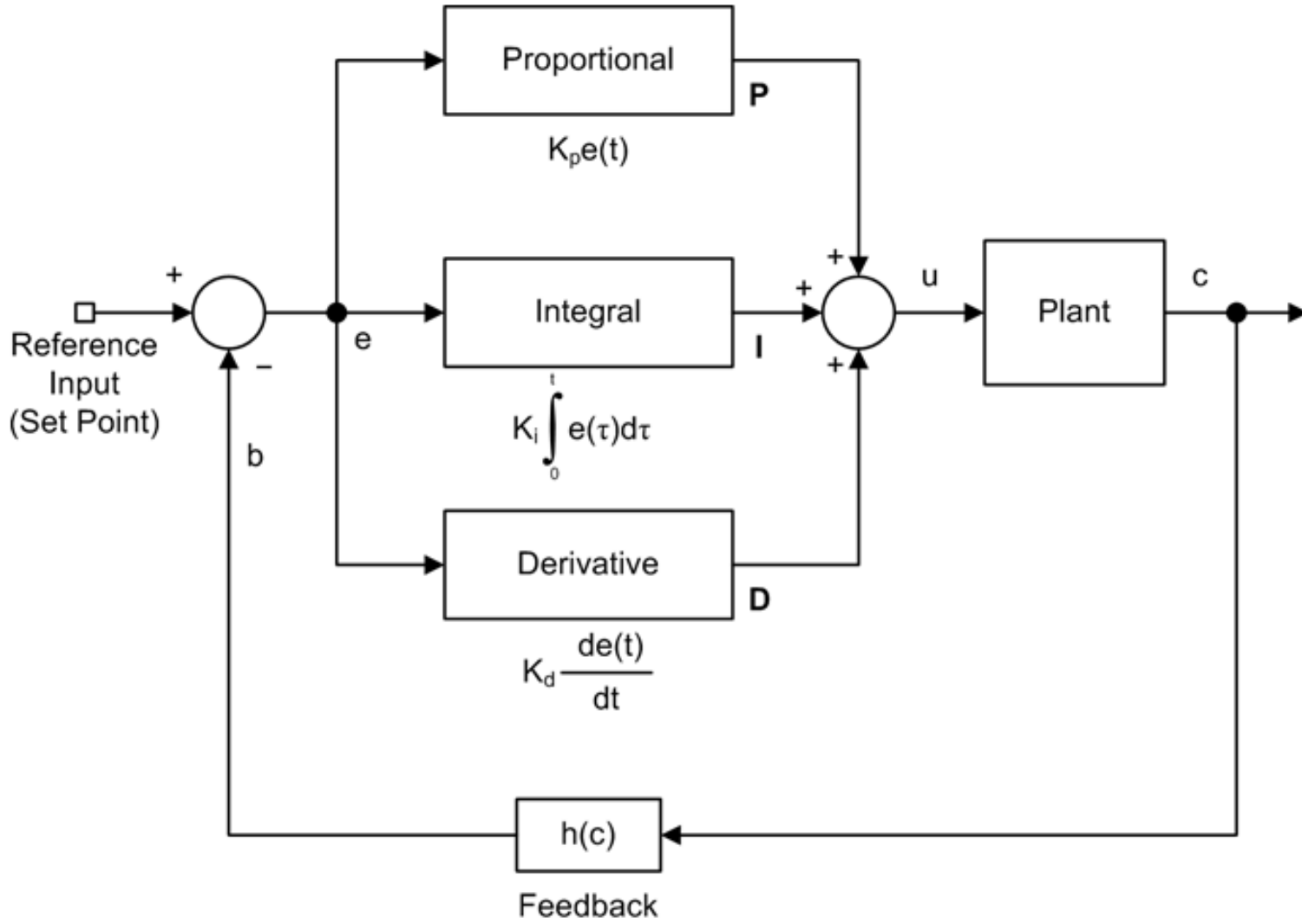
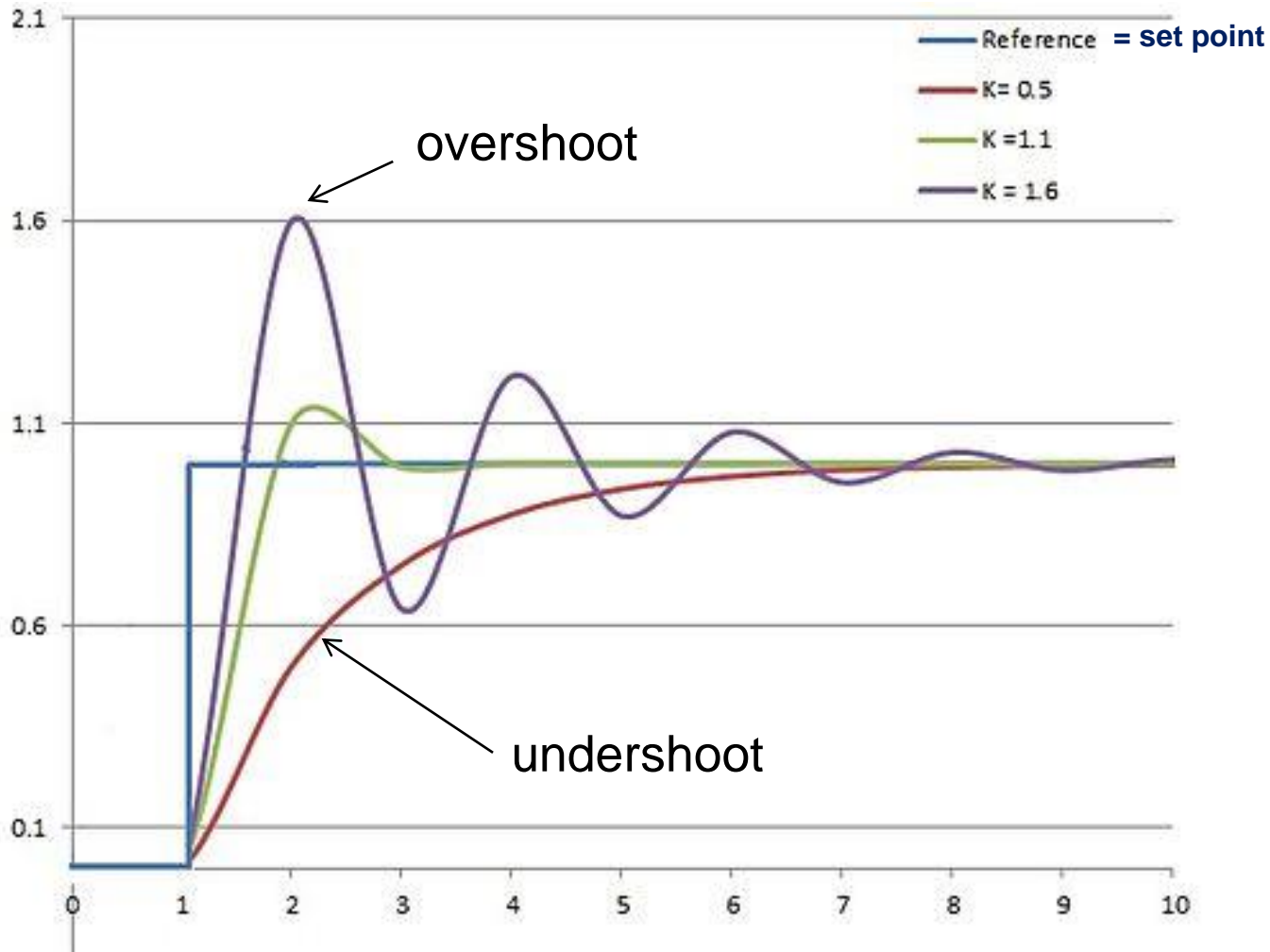


Figure 9-24. PID control block diagram

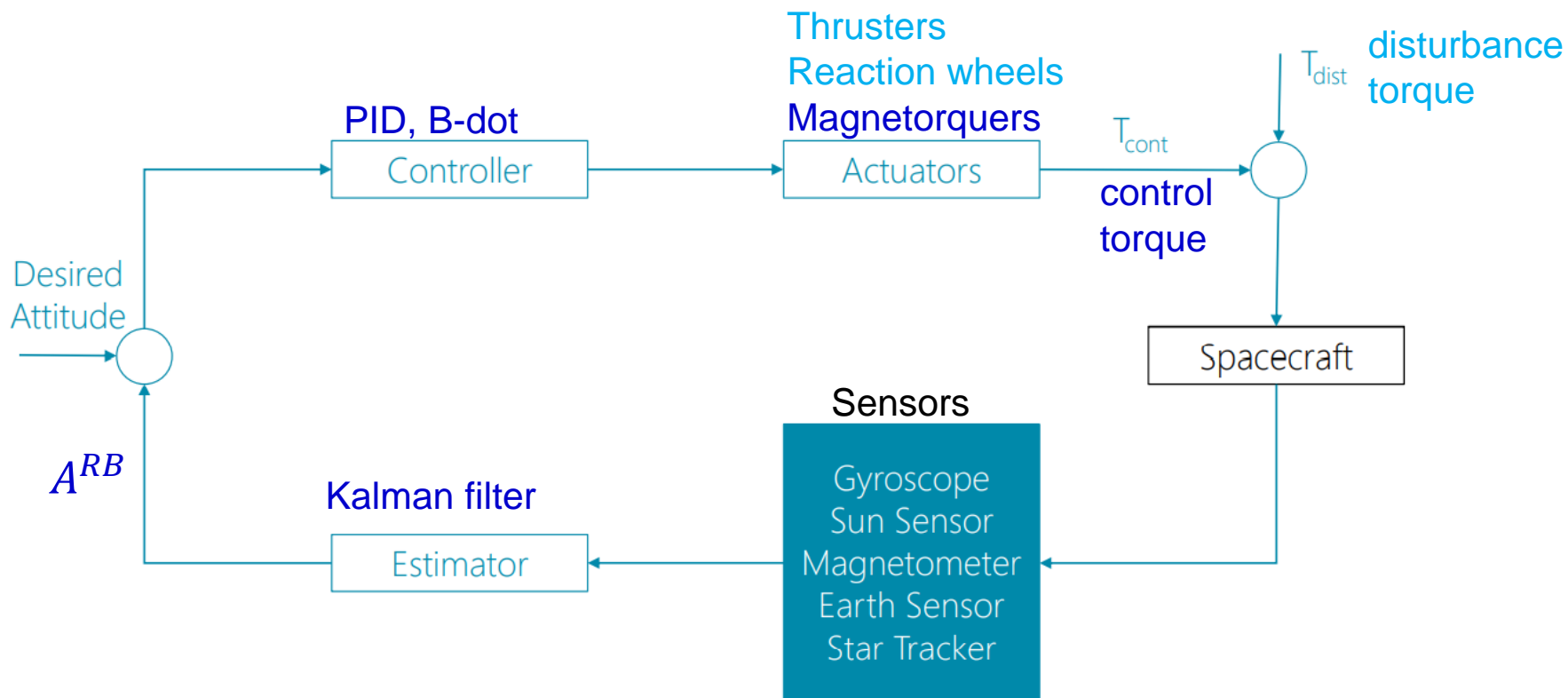
# PID controller tuning examples



# Satellite control

# Active attitude control

- With active attitude control, we estimate the spacecraft attitude and control actuators to actively change it to a desired attitude.





# Magnetic Torque Attitude control

- The attitude control is performed using actuator coils.
- Three coils (magnetorquers) are used to control the attitude, one for each axis.
- The coils generate a magnetic field that interacts with the Earth's magnetic field and creates a moment.

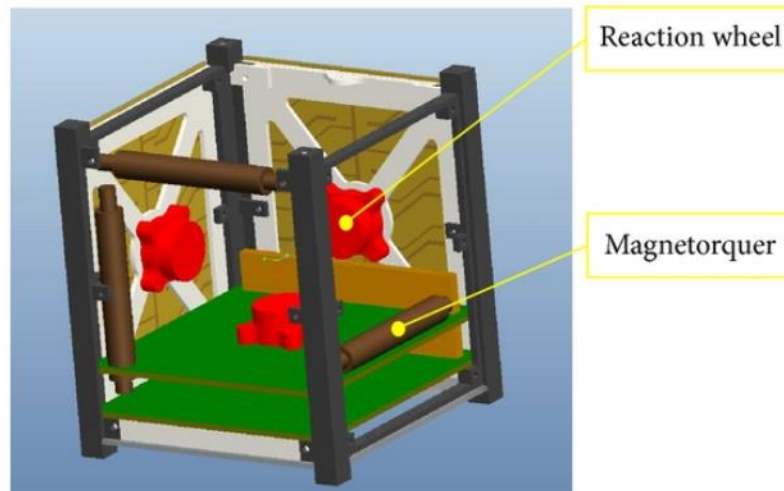


Figure from Hindawi



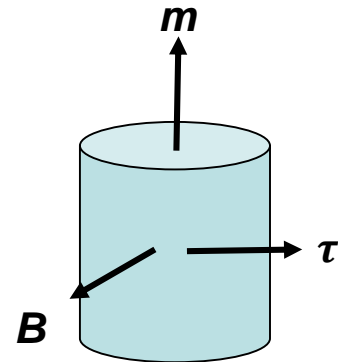
One possible control method – usually combined with another method

# Magnetic Torque Attitude control

- The moment  $\tau$  is given by  $\tau = \mathbf{m} \times \mathbf{B}$

commanded magnetic dipole moment

Earth's magnetic field vector  
(proportional to  $1/r^3$  with  $r$  the distance of the center of Earth to the spacecraft)

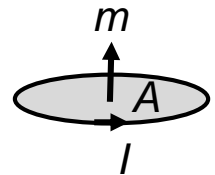


- Where  $m = nIA$

Number of turns  
(constant)

Current through the loop  
(direction and magnitude)

Coil/loop area  
(constant)



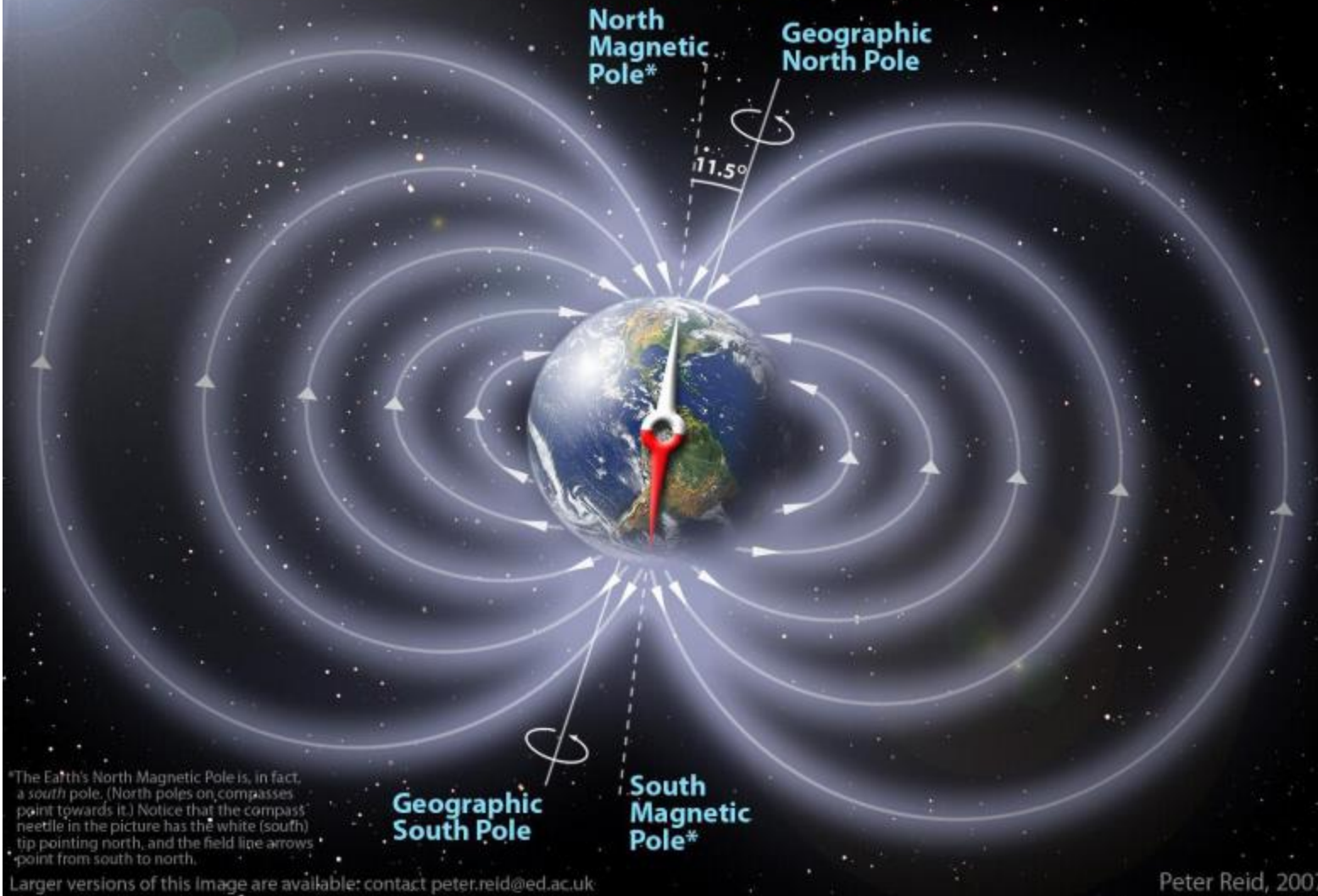
- Decomposed in the spacecraft body frame {b} we get:

$$\tau^b = \mathbf{m}^b \times R^{be} \mathbf{B}^e$$

Required dipole moment must be calculated!

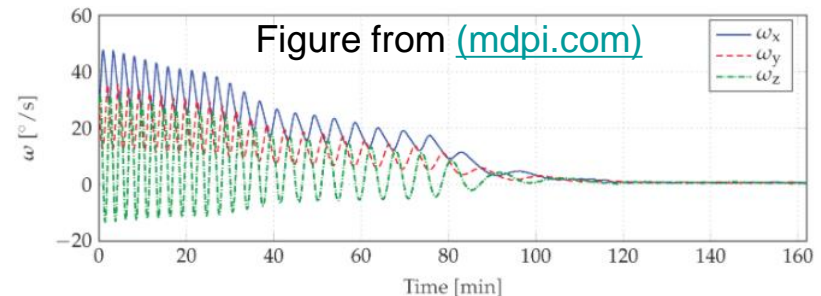
magnetic field vector in frame {e} given from a model

# The Earth's Magnetic Field



# Satellite control – detumbling

- The first task a spacecraft attitude control system must perform after separation from the launcher is to detumble the spacecraft, i.e., to bring it to a final condition with a sufficiently small angular velocity in all three axis.
- Will present a common control algorithm that is used to **detumble** (null the angular velocity) of a spacecraft.
  - Magnetic control has been used for decades to fulfill this task.
- Detumbling is necessary for satellites after orbital insertion
  - Also used on CubeSats.



# Detumbling using B-dot algorithm

- The principle of the B-dot algorithm relies on the usage of magnetorquers to generate a torque which is opposed to the “natural” rotation of the satellite, in order to reduce the angular rate.
- The control law creates a magnetic dipole in the opposite direction to the change in the magnetic field.
- The B-dot controller **uses a magnetometer to derive the angular rates** → No IMU / rate gyroscope is required!
- B-dot control law (represented in body frame):

$$\mathbf{m}^b = -k \dot{\mathbf{B}}^b$$

required magnetic dipole moment

a constant (to be tuned)

The time derivative of the B-field vector measured by the magnetometer

The diagram shows the equation  $\mathbf{m}^b = -k \dot{\mathbf{B}}^b$  in the center. Three blue arrows point from text labels below to the terms in the equation: one from 'required magnetic dipole moment' to  $\mathbf{m}^b$ , one from 'a constant (to be tuned)' to  $-k$ , and one from 'The time derivative of the B-field vector measured by the magnetometer' to  $\dot{\mathbf{B}}^b$ .

# Why the B-dot controller works

- The relation between the B-field vector in body frame {b} and an inertial frame {i}:  $\mathbf{B}^i = R^{ib} \mathbf{B}^b$

- Time derivation of the B vector gives:

$$\dot{\mathbf{B}}^i = R^{ib} \left( \dot{\mathbf{B}}^b + \Omega^{bi} \mathbf{B}^b \right)$$

*~ 0 during the short sampling interval* →  $\dot{\mathbf{B}}^i = R^{ib} \left( \dot{\mathbf{B}}^b + \boldsymbol{\omega}^{bi} \times \mathbf{B}^b \right)$

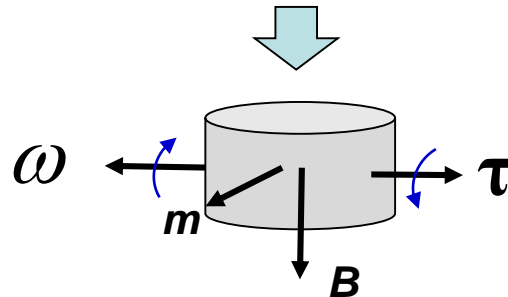
Remember:

$$\dot{R}^{AB} = R^{AB} \Omega^{BA}$$

$$\Omega^{bi} = \boldsymbol{\omega}^{bi} \times$$

$$\dot{\mathbf{B}}^b = -\boldsymbol{\omega}^{bi} \times \mathbf{B}^b$$

Measured by an IMU/rate gyroscope. So, it is possible to avoid derivation of the B-field ...



B-dot method generate a torque which is opposed to the rotation of the satellite

# Derivation of signals

- B-dot can be calculate from

$$\dot{B}_n = \frac{B_n - B_{n-1}}{\Delta t}$$

- Note that a derivation (finite difference) amplify the noise in the measurements
  - A filter should be used to lower the noise level.

# A practical implementation of the B-dot algorithm

- B-dot control is often implemented as a **bang-bang control law** (to avoid the difficulty of tuning the constant k).
- Assume that each magnetorquers can produce a maximum dipole moment of  $\pm m_i^{max}$  in each axis, where  $i = 1,2,3$ .
- Then the bang-bang B-dot detumbling control commands in each axis is given by

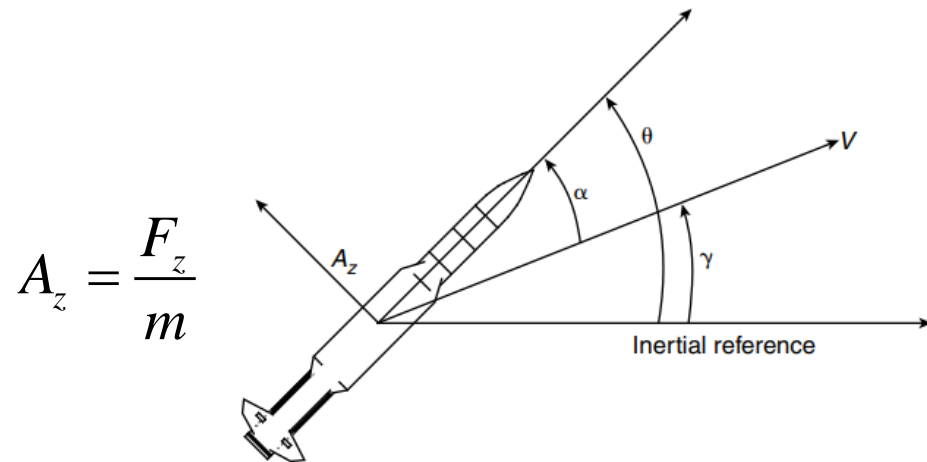
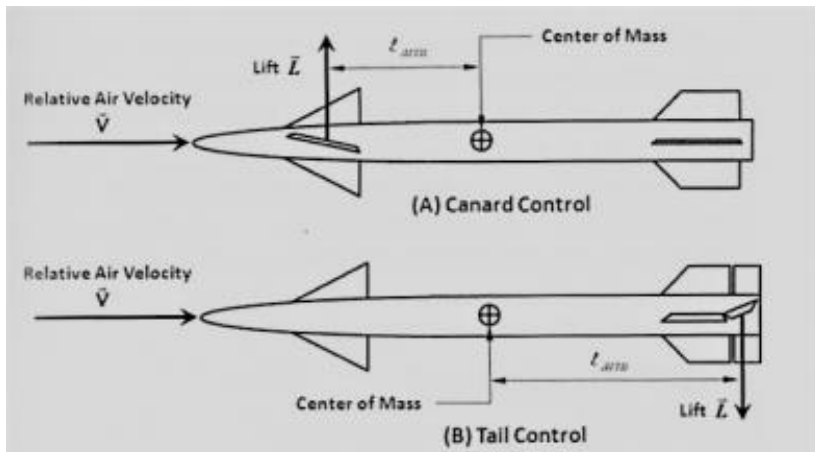
$$m_i = -m_i^{max} \underbrace{\text{sign}(\dot{B}_i)}_{\substack{1 \text{ if } \dot{B}_i > 0 \\ -1 \text{ if } \dot{B}_i < 0}}$$



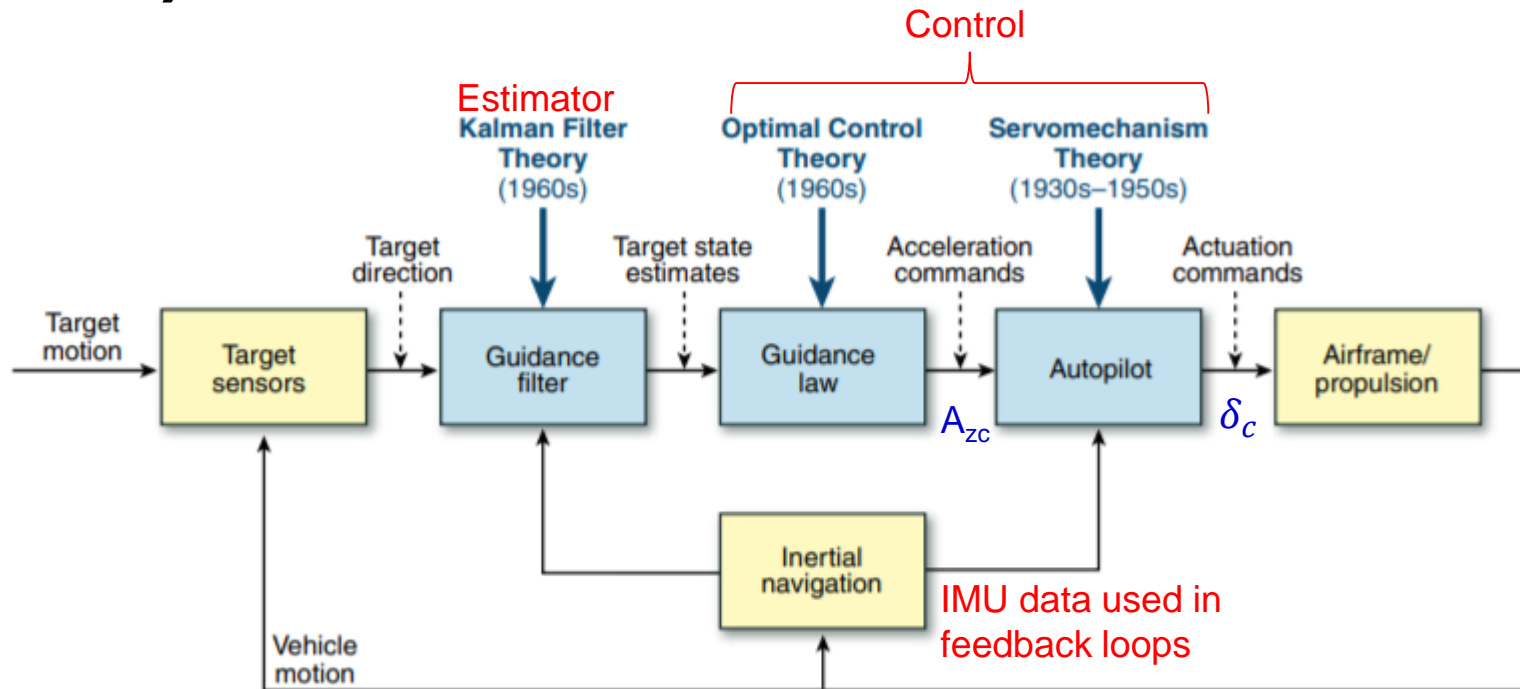
# Missile Guidance & Control

# Missile guidance

- A missile is guided towards a target by generating an acceleration  $A_z$  normal to the missile longitudinal axis
  - This force gives a change in the velocity vector  $V$ .
- (The required force is created by a lift force, by controlling aerodynamic surfaces / fins)



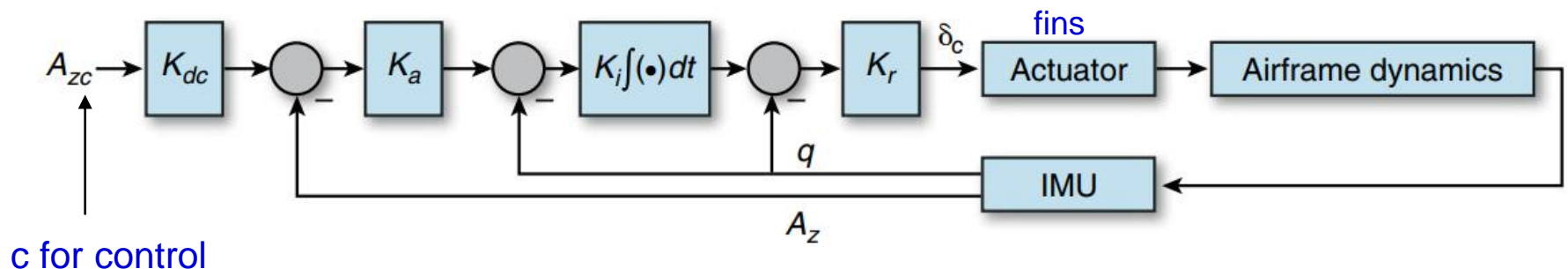
# Missile Guidance, navigation & control (GNC)



**Figure 3.** Traditional missile GNC topology. The traditional GNC topology for a guided missile comprises guidance filter, guidance law, autopilot, and inertial navigation components. Each component may be synthesized by using a variety of techniques, the most popular of which are indicated here in blue text.

# Missile acceleration control autopilot

- Used in all missiles.
- Classical approach to the design of an acceleration control autopilot:
  - The difference between the scaled input acceleration command  $A_{zc}$  and the measured acceleration  $A_z$  is multiplied by a gain  $K_a$  to effectively form a pitch rate command. The difference between the effective pitch rate command and the measured pitch rate  $q$  is multiplied by a gain  $K_i$  and integrated with respect to time. The resulting integral is differenced with the measured pitch rate  $q$  and multiplied by a third gain  $K_r$  to form the control command  $\delta_c$  such as desired fin-deflection angle



# But how do we calculate the required acceleration $A_z$ ?

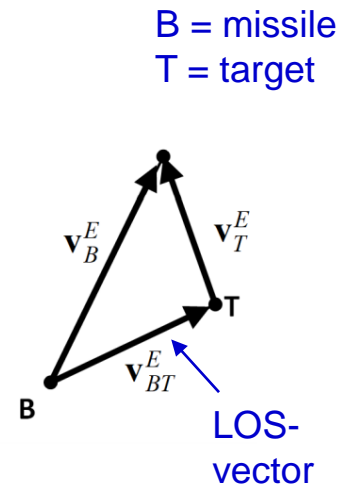
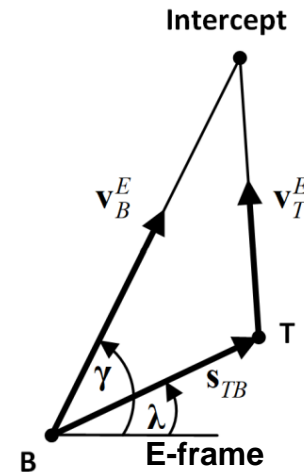
- Lets have a look in 2D (for simplification).
- Homing missiles (with a missile seeker in the nose) use a guidance method called **proportional navigation (PN)**, given by

$$A_z = NV \dot{\lambda}$$

Navigation constant (3 – 5)

Missile velocity magnitude  $|v^E|$  calculated from an IMU

Line-of-sight (LOS) rate towards the target, calculated by the seeker



- The missile is guided towards an intercept point.