

Feedback Controller Implementation



- **The sampling process**
- **Approximation of continuous time controllers**
- **Aliasing and new Frequencies: Anti-aliasing filters**
- **PID Control: Discretization, integrator windup**
- **Real time implementation**

Feedback control systems

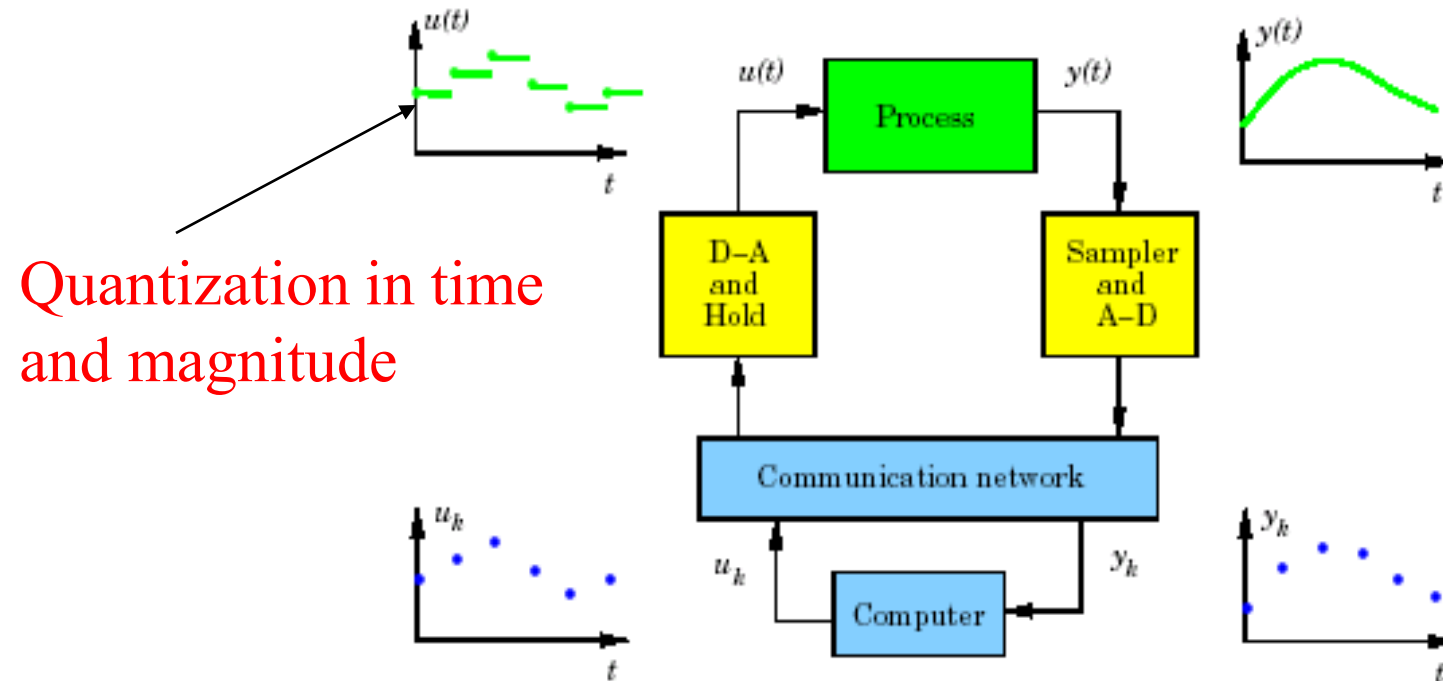


Figure 1 Schematic diagram of a computer-controlled system.

The sampling process



- ***Periodic sampling***
- ***Sampling instants***: times when the measured physical variables are converted into digital form
- ***Sampling period***: time between two sampling instants (denoted by h or T)

Sampling and Reconstruction

□ **Sampler:** A device that converts a continuous time signal into a sequence of numbers, e.g., A/D converter

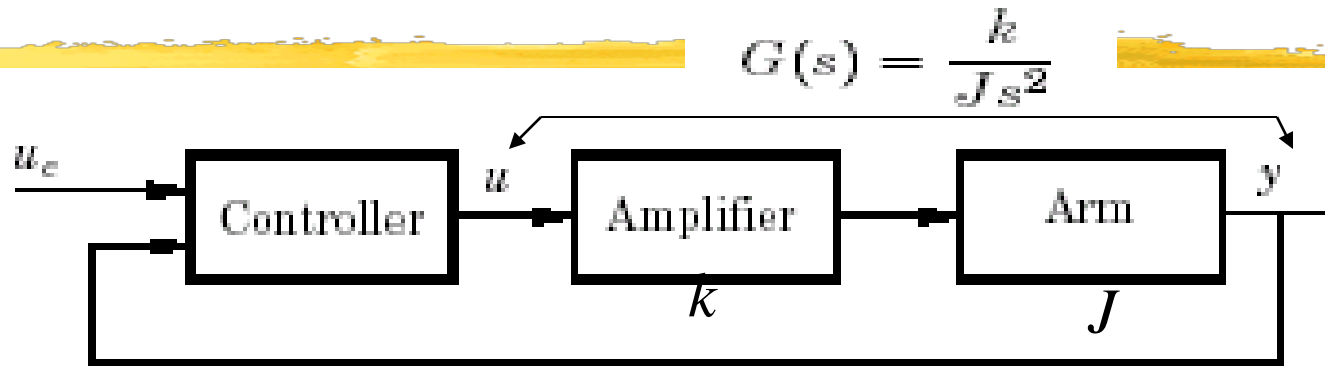
□ A/D: 8-16 bits → 2^8 to 2^{16} levels of quantization, normally higher than the precision of physical sensor

□ **Reconstructor:** Converts numbers from the computer to analog signals

□ D/A converter combined with a *hold circuit (Zero Order Hold)*

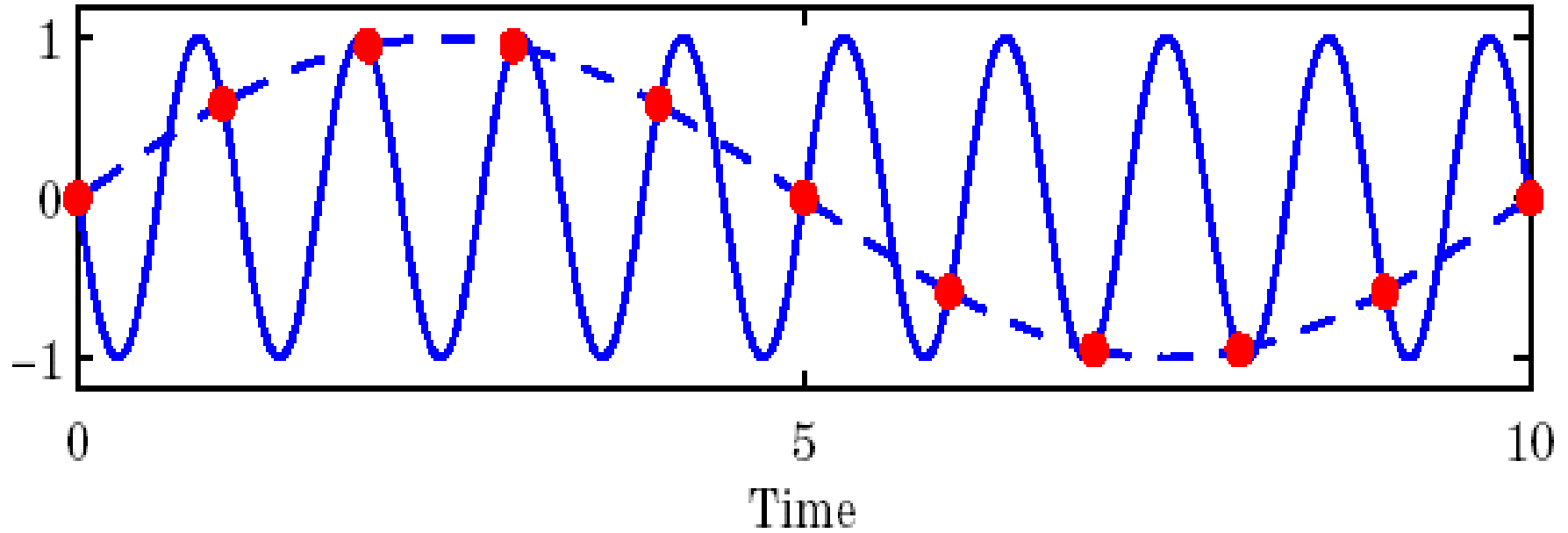
□ **PWM:** Plant responds to the average PWM signal

Discretization of continuous controllers



Disk drive system controller can be discretized by backward/forward differences

Sampling continuous signals



Aliasing and new frequencies

❑ Can information be lost by sampling a continuous time signal?

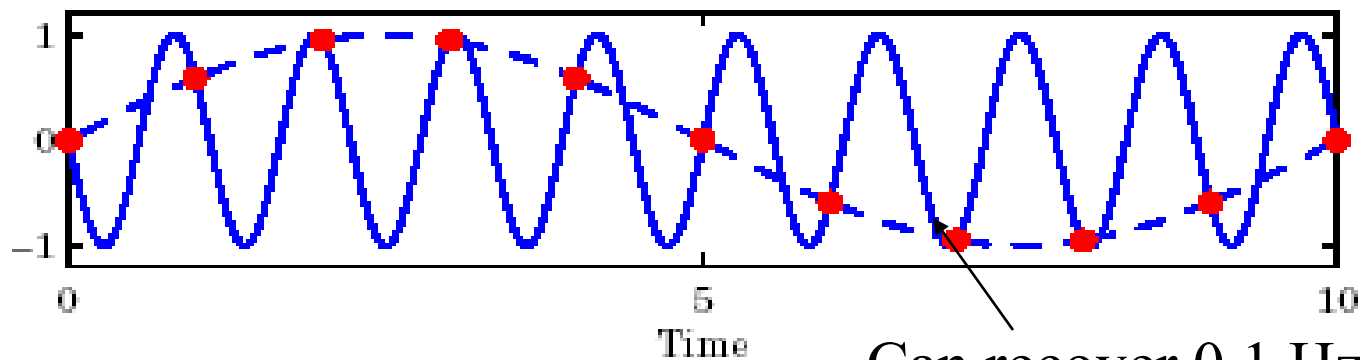
→ $\sin(2\pi 0.9 t)$ & $\sin(2\pi 0.1 t)$

❑ Are sampled values obtained from a low frequency signal or a high frequency signal?

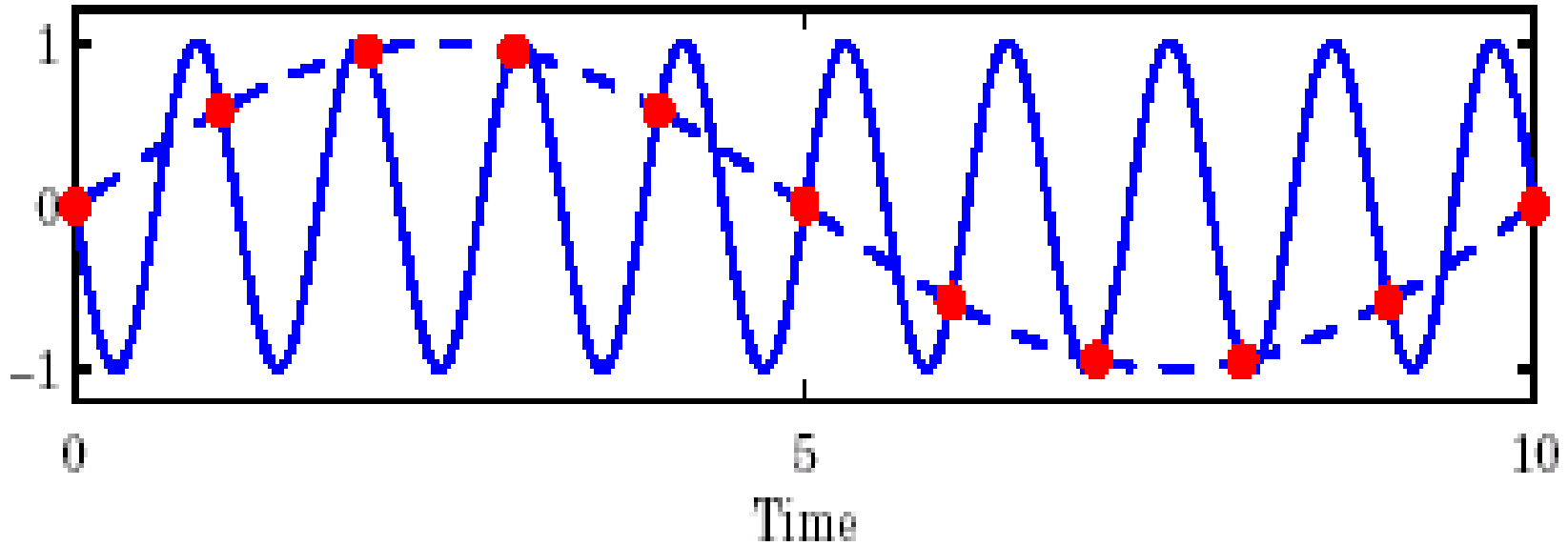
→ It may be possible that some continuous signal CANNOT be recovered from the sampled signal → May have loss of information

Shannon's Sampling Theorem (1949)

□ If a signal contains no frequencies above ω_0 , then the continuous time signal can be uniquely reconstructed from a periodically sampled sequence provided the sampling frequency is higher than $2\omega_0$ (Nyquist rate)



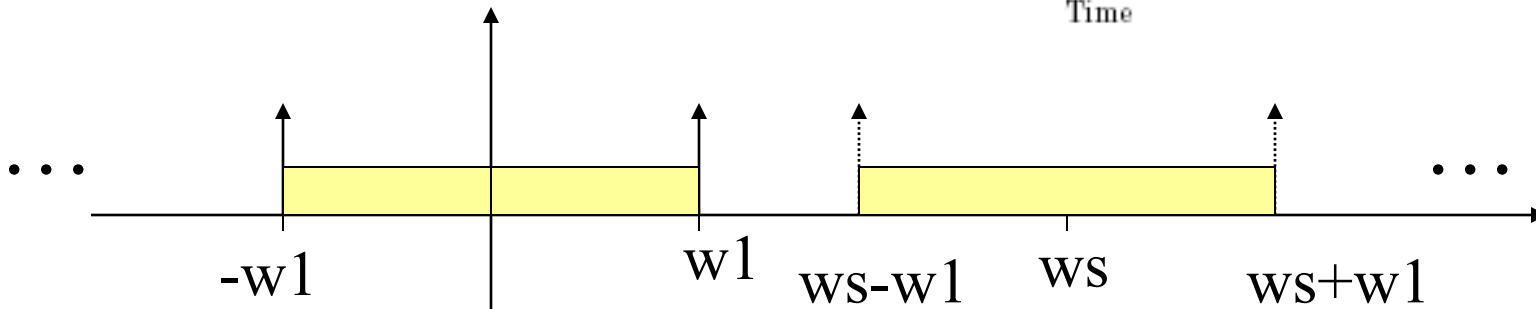
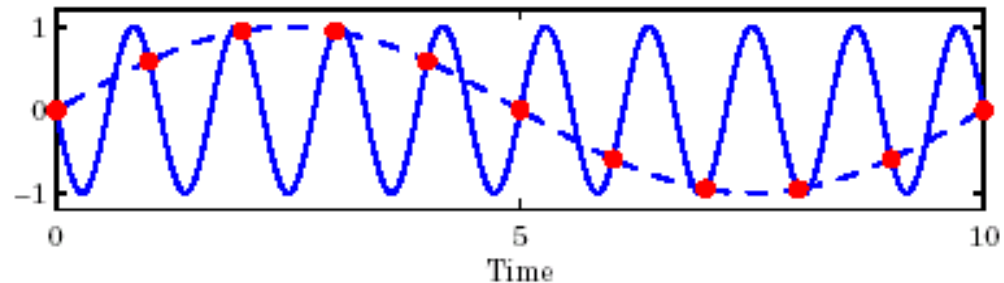
Sampling Theory- Example



- ❖ Can recover 0.1 Hz but not 0.9 Hz
- ❖ What should be the sampling frequency for recovering the 0.9 Hz sinusoidal?

Aliasing & Anti-aliasing Filters

□ Aliasing (or frequency folding): High frequency signal interpreted as a low frequency signal when sampled at lower than the Nyquist rate (the $0.1 = 1 - 0.9$ Hz signal)



... anti-aliasing filters

□ To prevent aliasing:

➤ Remove all frequencies above the Nyquist frequency before sampling the signal.

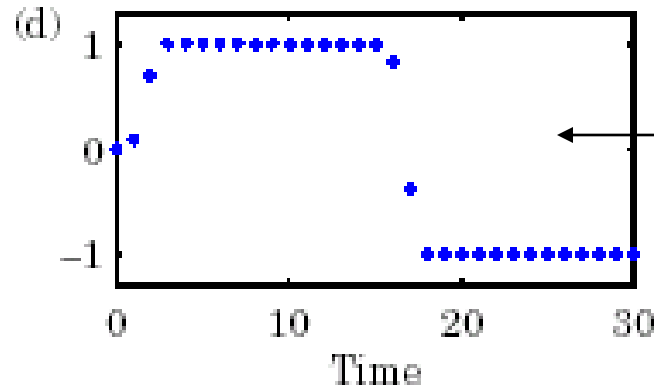
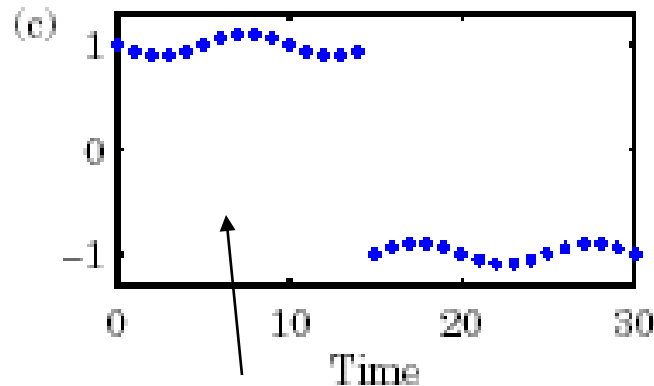
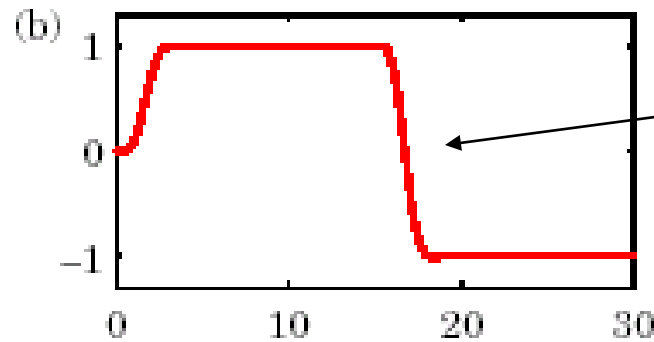
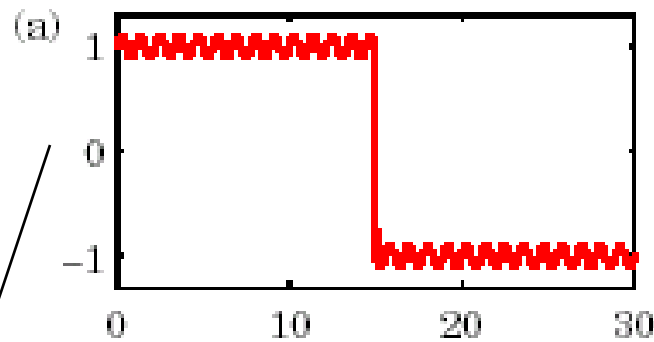
□ Antialiasing filter (low pass analog filter)

➤ High attenuation at and above Nyquist frequency

$$\frac{\omega^2}{(s/\omega_B)^2 + 2\zeta\omega(s/\omega_B) + \omega^2}$$

Example: Pre-filtering

□ Square wave + 0.9 Hz sine disturbance



Filtered by a 0.25 Hz LPF.

Sampled values

Sampled at 1 Hz (alias freq 0.1 Hz)

Summary



- ❑ Information can be lost through sampling if the signal contains frequencies higher than the Nyquist frequency.
- ❑ Sampling creates new frequencies (aliases).

... summary

- Aliasing makes it necessary to filter the signals before they are sampled.
- Effect of the anti-aliasing filters must be considered when designing controllers.
 - The dynamics can be neglected if the sampling period is sufficiently short.

A Typical Controller



- ⌘ PID Control: Discretization, saturation and windup
- ⌘ Other practical issues

PID Control



⌘ Most common controller:

$$u(t) = K \left(e(t) + \frac{1}{T_i} \int^t e(s) ds + T_d \frac{de(t)}{dt} \right)$$

⌘ e: error, K: gain, T_i : integration time, T_d : derivative time

⌘ System error ↓ as K & $1/T_i$ ↑, stability improves as T_d ↑, increasing $1/T_i$ reduces stability

⌘ Originally implemented using analog technology

PID Control & Discretization

⌘ Modifications to PID:

$$U(s) = K \left(bU_c(s) - Y(s) + \frac{1}{sT_i} \left(U_c(s) - Y(s) \right) - \frac{sT_d}{1 + sT_d/N} Y(s) \right)$$

⌘ Derivative term:

$$sT_d \approx \frac{sT_d}{1 + sT_d/N}$$

*N: gain at
high freq:
3-20*

⌘ Derivative not acting on the command signal

Discretization of PID Controller

⌘ Proportional part needs no approximation:

$$P(t) = K \left(bu_c(t) - y(t) \right)$$

⌘ Integral part: $I(t) = \frac{K}{T_i} \int^t e(s) ds$

→ can be approximated as

$$I(kh + h) = I(kh) + \frac{Kh}{T_i} e(kh)$$

.. Discretization of PID: **P+I+D**

⌘ Derivative part: $D(t)$

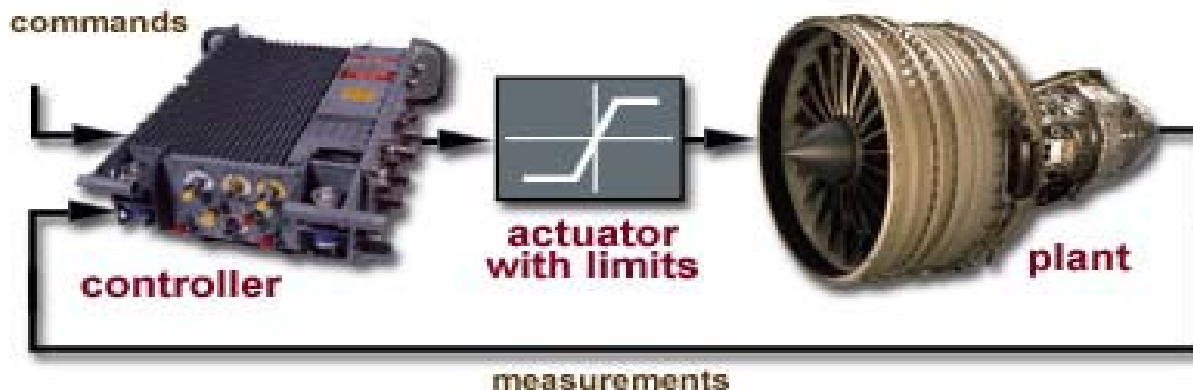
$$\frac{T_d}{N} \frac{dD}{dt} + D = -KT_d \frac{dy}{dt}$$

→ use backward differences for approximation:

$$D(kh) = \frac{T_d}{T_d + Nh} D(kh - h) - \frac{KT_d N}{T_d + Nh} \left(y(kh) - y(kh - h) \right)$$

Integrator windup

- ⌘ Actuators can become saturated due to large inputs
- ⌘ Examples of saturated actuators:
 - throttle position in an automobile
 - amplifier output voltage
 - aircraft control surfaces

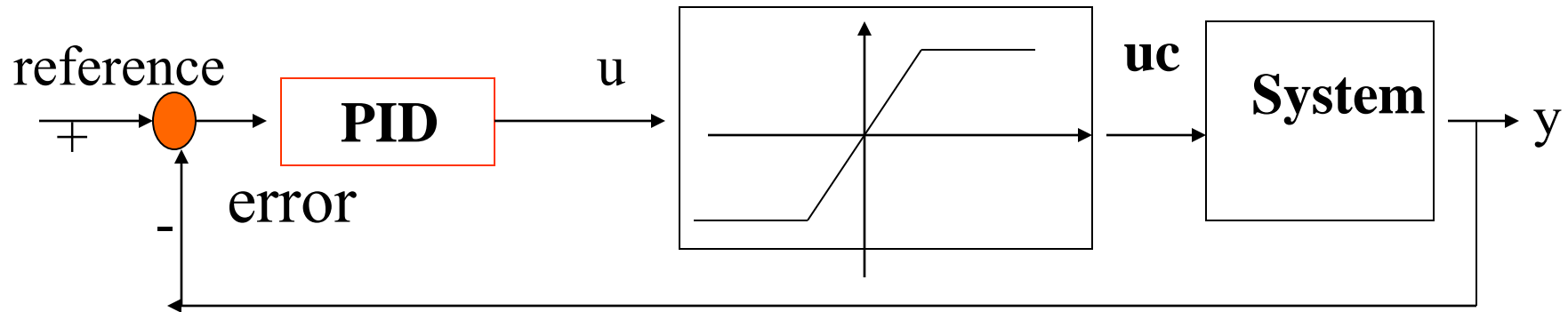


⌘ Pushing
actuator
limits



A taste of the practical world

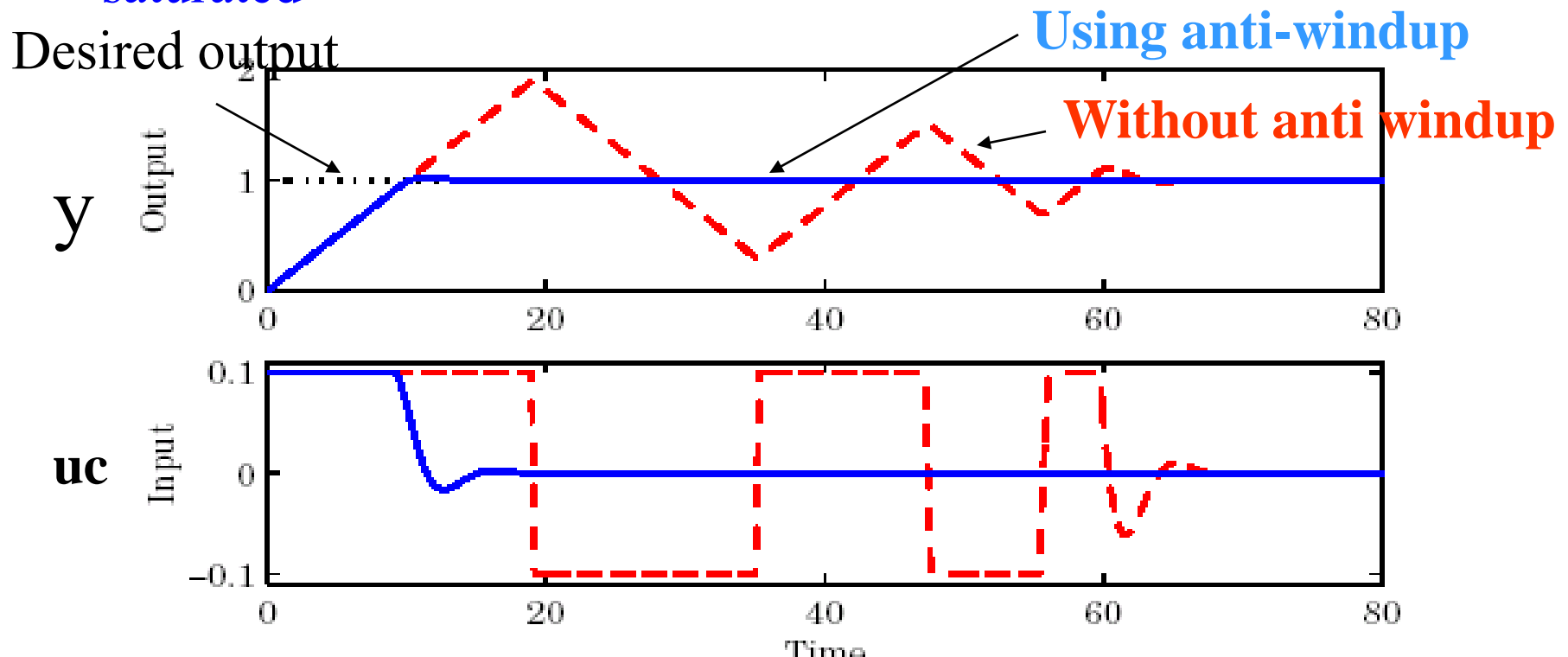
⌘ If error is large, Integral Control can saturate the actuator.



□ Integrator output can become large (windup) with no effect on the output (due to saturation).

... integrator windup

⌘ Integrator output *winds up* until the sign of $e(t)$ changes and the integration turns the other way around → Solution: Reset integrator when actuator is saturated



Summary



⌘ Things to consider:

- ☑ Filtering of the derivative term
- ☑ Updating the integrator
- ☑ Anti windup compensation

Other practical issues



⌘ **Numerical accuracy is improved if **more bits** are used:**

☑ **float coeff; vs double coeff; (64 bits)**

⌘ **A/D should be accurate but D/A can have much less accuracy**

☑ Typical resolutions: 8, 12, 16 bits

☑ Better to have a good resolution at the A/D converter than D/A

☒ A control system is less crucial for a quantized signal applied to the input of the plant

... practical issues



⌘ Nonlinearities such as quantization by A/D can result in oscillations

☒ Do simulations to find out if A/D, D/A resolutions can improve performance

⌘ Selection of the sampling period

☒ A common rule: $\omega T = 0.1 \text{ to } 0.6$,

ω : desired BW of the closed-loop system

... practical issues



⌘ Anti-aliasing filters

- ☑ Must provide attenuation at Nyquist frequency

⌘ Anti-reset windup is important to include in the controller.

Next: Design of a typical control system

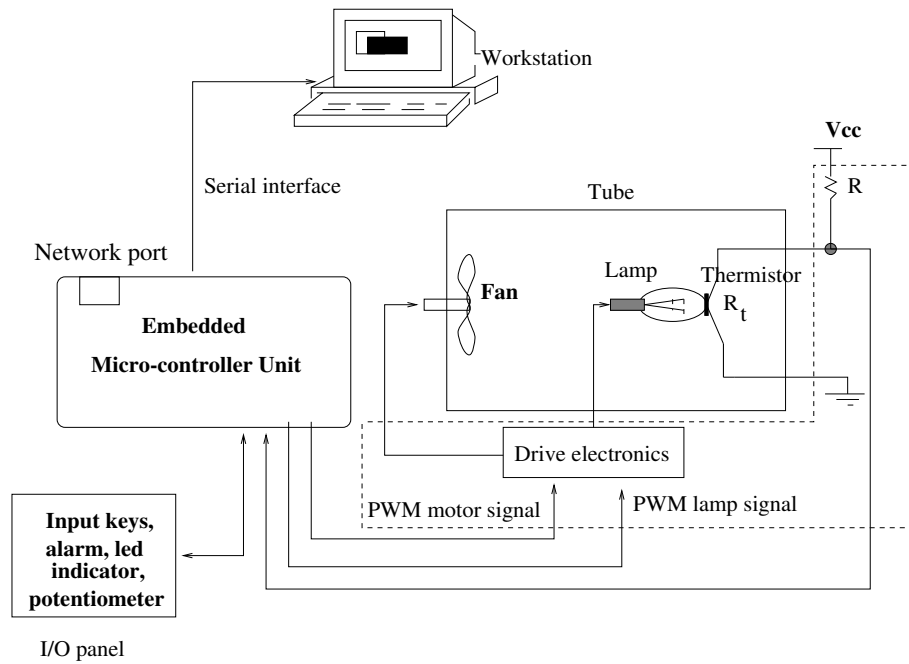


Figure 1: Components of the temperature control system.

Laboratory Setup

- Objective: Design the software and build a sensor-based feedback control system, using a PID controller and existing components

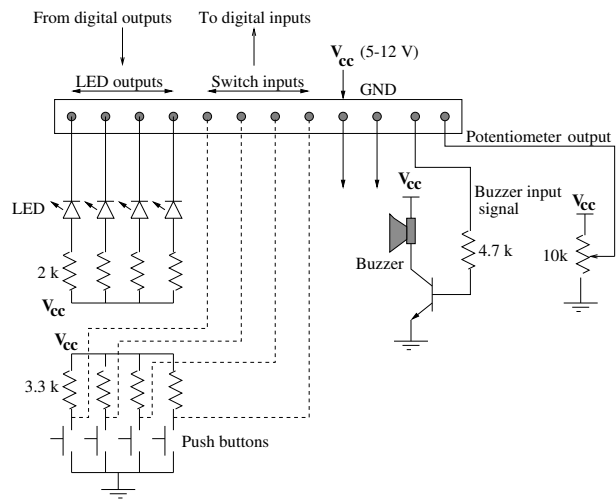


Figure 2: Circuit diagram of the I/O panel board.

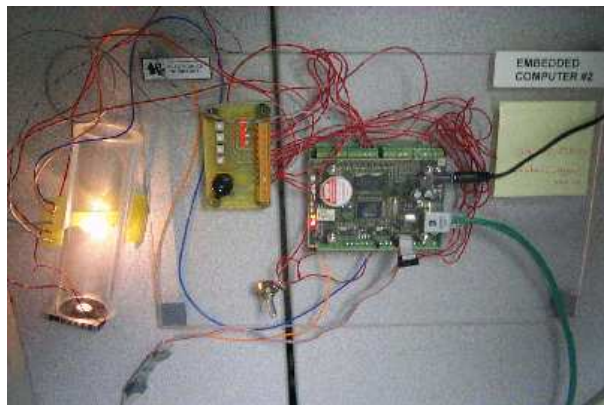


Figure 3: Embedded system hardware components.

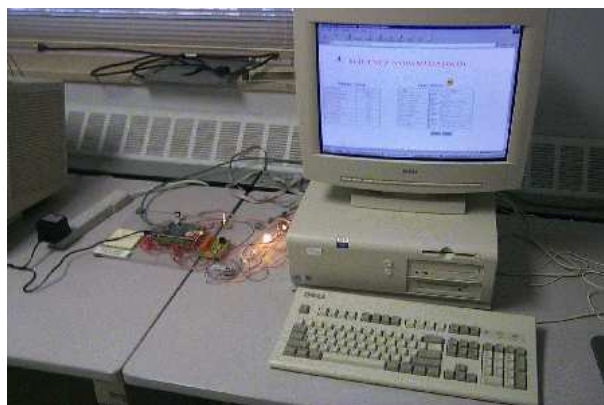


Figure 4: Embedded system controlled and monitored using an Internet browser.

- System components:
 - tube → DC fan, lamp, 10 $k\Omega$ thermistor, drive electronics
 - I/O panel → push button switches, buzzer, potentiometer,
 - microcontroller development kit
 - host-target configuration

Design of the Control System Software

Coding a feedback algorithm to work properly is not always sufficient from a product design point of view

- Software design must take into account:
 - functionality
 - user interface
 - safety and reliability
 - upgradability
 - power consumption
 - cost

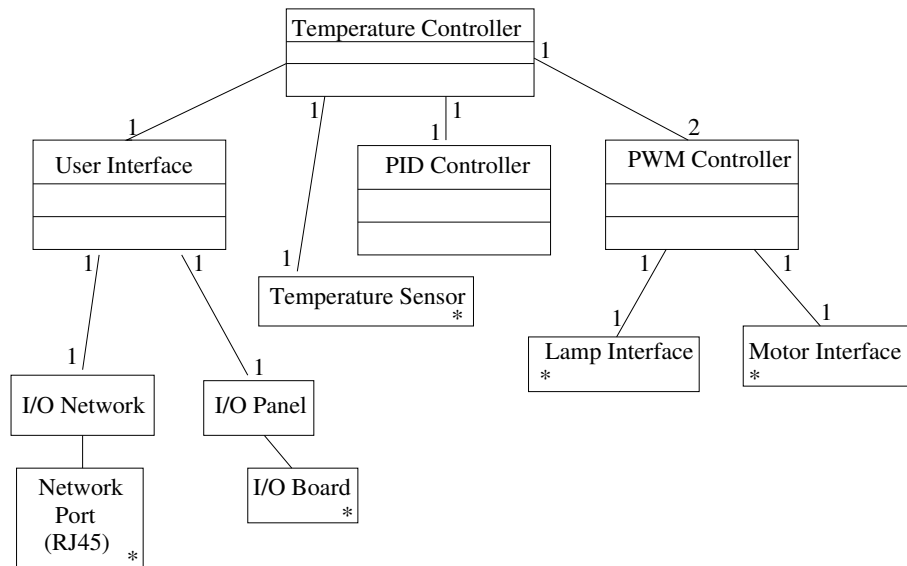


Figure 5: A Unified Modeling Language (UML) class diagram showing composition of subsystems.

- Design steps:
 - requirements, specifications, architecture, components, system integration
 - top-down, bottom-up approach
- Class diagram:
 - UML representation
 - classes: *Temperature Controller*, *User Interface*, *PID Controller*, *PWM Controller*
 - *Temperature Controller* is the main module → initializes other modules, links modules together, and runs systems' tasks

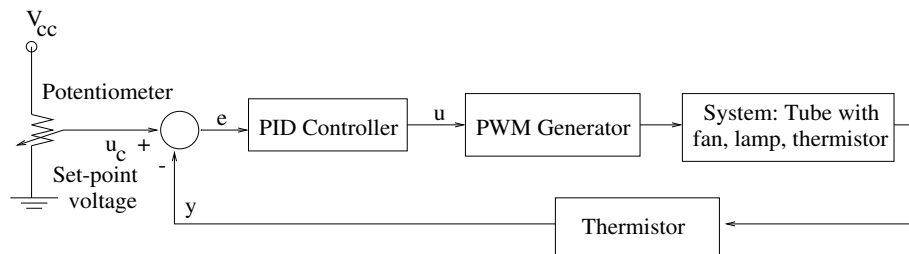


Figure 6: Block diagram of the temperature feedback control system.

PID Controller

- Modified PID (B. Wittenmark, *et. al*):

$$\begin{aligned}
 U(s) = & K(bU_c(s) - Y(s) + \frac{1}{sT_i}(U_c(s) \\
 & - Y(s)) - \frac{sT_d}{1 + sT_d/N}Y(s)) \quad (1)
 \end{aligned}$$

- Discretized algorithm:

$$u(kT) = P(kT) + I(kT) + D(kT) \quad (2)$$

where

$$\begin{aligned} P(kT) &= K(bu(kT) - y(kT)) \\ I(kT) &= I((k-1)T) + \frac{KT}{T_i}e((k-1)T) \\ D(kT) &= \frac{T_d}{T_d + NT}D((k-1)T) \\ &\quad - \frac{KT_d N}{T_d + NT}(y(kT) - y((k-1)T)). \end{aligned} \quad (3)$$

- Modular software in C \rightarrow Abstract Data Types (*ADT*)
— *PID Controller* module in Figure 5

```
typedef struct {
    float K, b, Ti, T, Td;
    unsigned short N; /* parameters */
    float i, ilast; /* integral term */
    float y, ylast; /* output */
    float d, dlast; /* derivative term */
    float uc; /* input */
    float u; /* output */
    float elast; /* error term */
}
```

```
float ulow, uhigh; /* low/high inputs
                    for windup compensation */
} pid_t;
```

— routines operating on the `pid_t` data type:

```
void initPid(pid_t *p, ...);
void updatePid(pid_t *p, ...);
void calcPid(pid_t *p, ...);
```

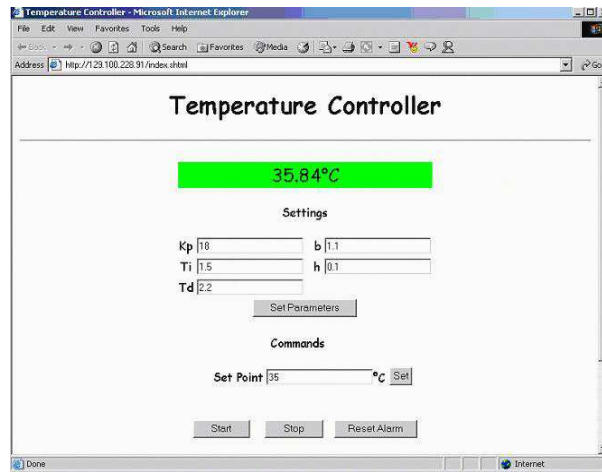


Figure 7: Web browser interface to the embedded system.

Other Modules

- *PWM Controller*
 - uses a digital output pin
- *User Interface*: Scans I/O panel for
 - new inputs from the push button switches and the potentiometer
 - provides outputs to the LED's and the buzzer
 - handles *I/O Network* (for devices with a network port): email, ftp, http

- *Temperature Sensor* consists of a thermistor in series with a resistor

$$T = \frac{\beta T_0}{\beta - T_0 \ln \left(\frac{v_t R}{(V_{cc} - v_t) R(T_0)} \right)}. \quad (4)$$

Execution of Tasks

- Concurrent programming
 - allows for separation of modules into separate tasks
 - a key element in modular software development
 - schedulability analysis is required
- Cooperative multitasking

```

main ( ) {
/* variable declarations */
for (; ;){
    costate Task_1 {
        waitfor(delay(period of task 1));
/* code for task 1 */
    }
    costate Task_2 {
        waitfor(delay(period of task 2));
/* code for task 2 */
    }
        .....
    costate Task_N {
        waitfor(delay(period of task N));
/* code for task N */
    }
}
}

```

- *costates* are blocks of code that can suspend themselves
- timing analysis can be performed on the system

Conclusion

- Control systems are one of the main application domains for embedded computing
- There is a growing need to train control engineers who are aware of the design procedures and challenges in embedded computing technology
- An experimental testbed to serve the above purpose was introduced

M. Moallem, "A Laboratory Testbed for Embedded Computer Control," *IEEE Transactions on Education*, Vol. 47, No. 3, pp.340-347, 2004.