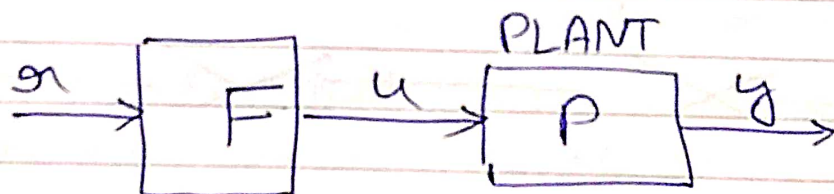


Lecture 2: Modeling

★ Control objective

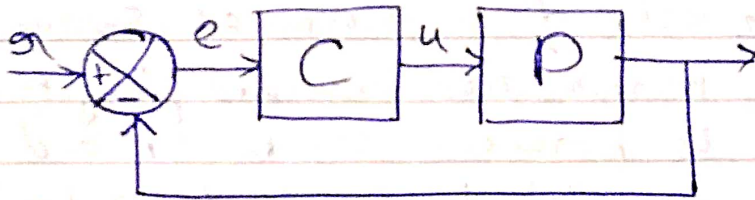
- ① Stabilization: Make sure the system stays within "normal" operating conditions.
- ② Regulation: Maintain a desired operating point in spite of disturbances.
- ③ Tracking: Follow a reference trajectory that changes over time, as closely as possible.

★ Basic control architectures: Feed Forward



- ⇒ Basic Idea: Given r , attempt to compute what should be the control input u that would make $y=r$.
- ⇒ Here F is 'inverse' of P .
- ⇒ This method relies on good knowledge of P .
 - ↳ Very sensitive to modeling error.

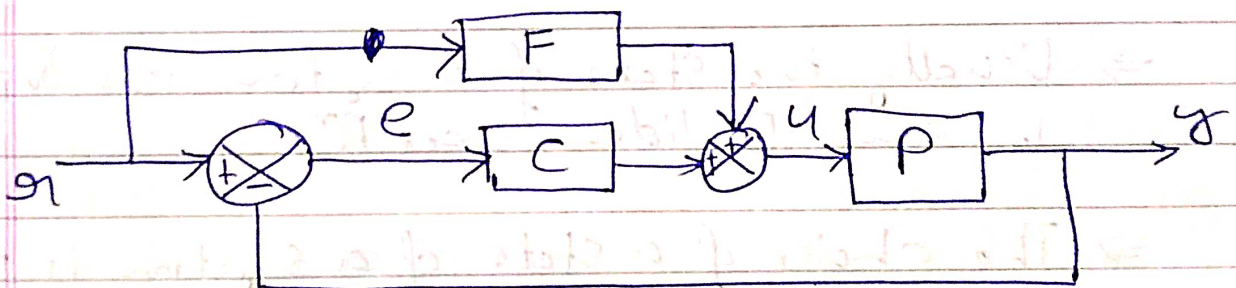
★ Basic Control architecture: Feedback



⇒ Basic Idea: Given the error $r - y$, compute u such that the error is small.

⇒ This method does not require a precise knowledge of P — robust to modeling errors.

★ Basic control architecture: Feed Forward & Feedback



⇒ Basic Idea: Given r , compute a guess for the control input u that would make $y = r$; correct the guess based on the measurement of the error e , so that the error is minimized.

⇒ Combines the main advantages of feed-forward & feedback architectures.

- ↳ Ensures stability & robustness
- ↳ Speed up tracking/regulation.

★ State of a System

“The state $x(t_1)$ of a causal system at time t_1 is the information needed, together with the input u between times t_1 and t_2 to uniquely predict the output at time t_2 for all $t_2 \geq t_1$.”

⇒ In other words, the state of the system at a given time t summarizes the whole history of the past input over $(-\infty, t)$, for the purpose of predicting the output at future times.

⇒ Usually, the state of a system is a vector in some Euclidean space \mathbb{R}^n .

⇒ The choice of a state of a system is not unique (In fact, there are infinite choices)

★ Dimension of a system (order of the system)

⇒ We define the dimension of a causal system as the minimal number of variables sufficient to describe the system's state:

⇒ We will mostly deal with finite-dimensional systems.

★ LTI State-space model

$$\begin{aligned}\dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t)\end{aligned}$$

⇒ The state x represents the memory of the system.

⇒ The state x is an internal variable that cannot be accessed directly, but only controlled through input u and observed through the output y .

★ Nonlinear systems

⇒ Finite-dimensional, time-invariant, causal nonlinear control systems, with input u and output y can typically be modeled using a set of differential equations as follows:

$$\begin{aligned}\dot{x}(t) &= f(x(t), u(t)) \\ y(t) &= g(x(t), u(t))\end{aligned}$$

★ Equilibrium point

A system described by an ODE $\dot{x}(t) = f(x(t), u(t))$ has an equilibrium point (x_e, u_e) if $f(x_e, u_e) = 0$.

* Jacobian Linearization Procedure

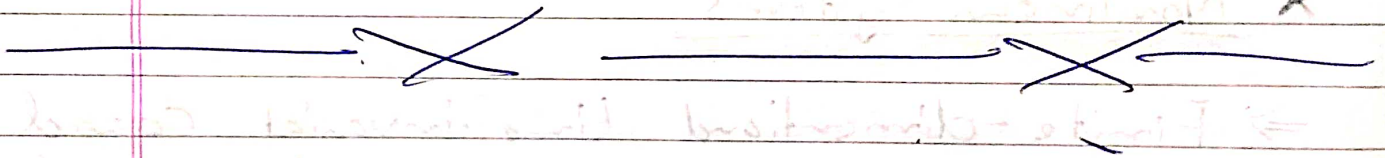
⇒ Given an equilibrium point (x_e, u_e) , with output $y_e = g(x_e, u_e)$, a linearized model is obtained by setting

$$x \leftarrow x_e + \delta x$$

$$u \leftarrow u_e + \delta u$$

$$y \leftarrow y_e + \delta y$$

and then neglecting all terms of second (or higher) order in δx , δu and δy in the (nonlinear) dynamics model.



$$\begin{bmatrix} \dot{x}(0) \\ y(0) \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$