Definition of systems

- Systems are used to process signals and to modify or extract information.
- Physical systems characterized by their input-output relationships.
  - E.g. electrical systems are characterized by voltage-current relationships for their components and the laws of interconnections (i.e. Kirchhoff’s laws).
- We can derive a mathematical model for a system represented by (a so called) “black box”.

![System Model Diagram]
Classification of systems

Systems may be classified into:

1. Linear and non-linear systems
2. Constant parameter and time-varying-parameter systems
3. Instantaneous (memoryless) and dynamic (with memory) systems
4. Causal and non-causal systems
5. Continuous-time and discrete-time systems
6. Analogue and digital systems
7. Invertible and non-invertible systems
8. Stable and unstable systems
A linear system exhibits the additivity property. This means that if input $x_i(t)$, $i = 1,2$ produces output $y_i(t)$, $i = 1,2$ then input $x_1(t) + x_2(t)$ produces output $y_1(t) + y_2(t)$.

The system must also satisfy the homogeneity or scaling property. This means that if input $x$ produces output $y$ then input $kx$ produces output $ky$.

The above two properties can be combined into the property of superposition. This means that if input $x_i(t)$, $i = 1,2$ produces output $y_i(t)$, $i = 1,2$ then input $k_1 x_1(t) + k_2 x_2(t)$ produces output $k_1 y_1(t) + k_2 y_2(t)$.

A non-linear system is one that is not linear, i.e., it does not obey the principle of superposition.
• Show that the system described by the following differential equation is linear.

\[ \frac{dy(t)}{dt} + 3y(t) = x(t) \quad (1) \]

• Let’s assume that input \( x_i(t), i = 1, 2 \) produces output \( y_i(t), i = 1, 2 \). Then:

\[ \frac{dy_1(t)}{dt} + 3y_1(t) = x_1(t) \Rightarrow k_1 \left[ \frac{dy_1(t)}{dt} + 3y_1(t) \right] = k_1 x_1(t) \Rightarrow \]

\[ \frac{dk_1 y_1(t)}{dt} + 3k_1 y_1(t) = k_1 x_1(t) \quad (2) \]

• Similarly, we have:

\[ \frac{dk_2 y_1(t)}{dt} + 3k_2 y_1(t) = k_2 x_2(t) \quad (3) \]

• \( k_1 \) and \( k_2 \) are constants.
• By adding equations (2) and (3) we obtain equation (1) with:
  - \( x(t) = k_1 x_1(t) + k_2 x_2(t) \)
  - \( y(t) = k_1 y_1(t) + k_2 y_2(t) \)
Almost all systems become non-linear when large enough signals are applied.

Non-linear systems can be approximated by linear systems for small-signal analysis – this approach greatly simplifies the problem.

Once superposition applies, we analyse a system by decomposition into zero-input and zero-state components. You will know soon what these are.

Later we will see that it is very important to represent $x(t)$ as a sum of simpler functions (pulses or steps).
Time-invariant systems

Time-invariant system is one whose parameters do not change with time:

![Diagram showing time-invariant system with inputs and outputs](image)

Linear time-invariant (LTI) systems are one of the main topics for both this course and the Control course in 2nd year.

Terminology by Lathi: LTIC = LTI continuous, LTID = LTI discrete
In general, a system’s output at time $t$ depends on the entire past input. Such a system is a dynamic (with memory) system.
- It is analogous to a state machine in a digital system.

A system whose response at $t$ is completely determined by the input signals over the past $T$ units of time only is a finite-memory system.
- It is analogous to a finite-state machine in a digital system.

Networks containing inductors and capacitors are infinite memory dynamic systems.

If the system’s past history is irrelevant in determining its response, it is an instantaneous or memoryless system.
- This is analogous to a combinatorial circuit in a digital system.
Consider the following simple RC circuit. Its output is:

\[ y(t) = Rx(t) + v_c(t) = Rx(t) + \frac{1}{C} \int_{-\infty}^{t} x(\tau)d\tau \]

[Note that the current \( x(t) \) is considered to be the input of the system and the voltage \( y(t) \) is considered to be the output of the system.]

We can break the integral and obtain:

\[ y(t) = Rx(t) + \frac{1}{C} \int_{0}^{\infty} x(\tau)d\tau + \frac{1}{C} \int_{0}^{t} x(\tau)d\tau \]

The second term is \( v_c(0) \). Therefore,

\[ y(t) = Rx(t) + v_c(0) + \frac{1}{C} \int_{0}^{t} x(\tau)d\tau, \ t \geq 0 \]

This is a single-input, single-output (SISO) system. In general, a system can be multiple-input, multiple-output (MIMO).

Analog circuits are linear systems.
Analog circuits are linear systems

- A system’s output for $t \geq 0$ is the result of 2 independent causes:
  1. Initial conditions when $t = 0$ (zero-input response)
  2. Input $x(t)$ for $t \geq 0$ (zero-state response)

- The above statement is summarized in the Decomposition Property:
  \[
  \text{total response} = \text{zero-input response} + \text{zero-state response}
  \]

- In the previous example we had:
  \[
  y(t) = v_C(0) + Rx(t) + \frac{1}{c} \int_0^t x(\tau)d\tau, \ t \geq 0
  \]

\[x(t) \xrightarrow{\text{system}} y(t) = y_0(t) + x(t) \xrightarrow{\text{system}} y_s(t)\]
Causal and non-causal systems

• In a causal system the output at $t_0$ depends only on $x(t), t \leq t_0$.
• The above statement implies that the present output depends only on the past and present inputs, not on future inputs.
• Any practical real-time system must be causal.

• Non-causal systems are also important because:
  1. The can still be realizable when the independent variable is something other than “time” (e.g. space).
  2. For temporal systems, as for example video, we can pre-record the data and process them in a non-real time fashion. In that case the systems will be non-causal.
Continuous-time and discrete-time systems

- In contrast to continuous-time systems, discrete-time systems process data samples instead of continuous-time signals.
- Data samples normally arise from continuous-time signals when these are regularly sampled every $T$ units of time.
- Continuous-time input and output are $x(t)$ and $y(t)$. Discrete-time input and output samples are $x[nT]$ and $y[nT]$ where $-\infty \leq n \leq \infty$. 
• In discrete-time signals the samples are discrete in time, but are continuous in amplitude.
• Most modern systems are both digital in amplitude and discrete in time. Consider for example the internal circuits of most current electronic devices, as for example, the MP3 player.
Invertible and non-invertible systems

- Let a system $S$ produce an output $y(t)$ with input $x(t)$. If there exists another system $S_i$, which produces $x(t)$ from $y(t)$, then $S$ is an invertible system.
- It is essential that there is a one-to-one mapping between input and output so that a system is invertible.
- For example if $S$ is an amplifier with gain $G$, it is invertible and $S_i$ is an attenuator with gain $1/G$.
- Apply $S_i$ following $S$ gives a combined system which has no impact in the input (i.e., the input $x(t)$ is not changed).
- We will see in detail later which system is this!
Stable and unstable systems

- Externally stable systems are the ones in which a bounded input results in a bounded output (the system is said to be stable in the BIBO sense).
- Stability of a system will be discussed after introducing Fourier and Laplace transforms.
- More detailed analysis of stability is covered in the Control course.
In many systems in electrical and mechanical engineering the input \( x(t) \) and the output \( y(t) \) are related by differential equations.

For example the RLC circuit:

\[
\begin{align*}
 v_L(t) + v_R(t) + v_C(t) &= x(t) \\
 \frac{dy(t)}{dt} + 3y(t) + 2 \int_{-\infty}^{t} y(\tau)d\tau &= x(t) \\
 \frac{d^2 y(t)}{dt^2} + 3 \frac{dy(t)}{dt} + 2y(t) &= \frac{dx(t)}{dt}
\end{align*}
\]

Or the so called dashpot; \( x(t) \) is a force and \( y(t) \) is motion.

\[
x(t) = B \dot{y}(t) = B \frac{dy(t)}{dt}
\]
In general, the relationship between $x(t)$ and $y(t)$ in a linear time-invariant (LTI) differential system is given by the following relationship where all coefficients $a_i$ and $b_i$ are constants.

$$\frac{d^N y(t)}{dt^N} + a_1 \frac{d^{N-1} y(t)}{dt^{N-1}} + \cdots + a_{N-1} \frac{dy(t)}{dt} + a_N y(t)$$

$$= b_{N-M} \frac{d^M x(t)}{dt^M} + b_{N-M+1} \frac{d^{M-1} x(t)}{dt^{M-1}} + \cdots + b_{N-1} \frac{dx(t)}{dt} + b_N x(t)$$

We use the compact notation $D$ for the operator $d/dt$, i.e., $\frac{dy(t)}{dt} \equiv Dy(t)$, $\frac{d^2 y(t)}{dt^2} \equiv D^2 y(t)$ etc. We then get:

$$(D^N + a_1 D^{N-1} + \cdots + a_{N-1} D + a_N) y(t)$$

$$= (b_{N-M} D^M + b_{N-M+1} D^{M-1} + \cdots + b_{N-1} D + b_N) x(t)$$

Alternatively $Q(D)y(t) = P(D)x(t)$ with

$$Q(D) = D^N + a_1 D^{N-1} + \cdots + a_{N-1} D + a_N$$

$$P(D) = b_{N-M} D^M + b_{N-M+1} D^{M-1} + \cdots + b_{N-1} D + b_N$$
Let us consider the previous example again:
The system’s equation is:
\[
\frac{d^2 y(t)}{dt^2} + 3 \frac{dy(t)}{dt} + 2y(t) = \frac{dx(t)}{dt}
\]
This can be re-written as:
\[
(D^2 + 3D + 2)y(t) = Dx(t)
\]
\[
\begin{align*}
Q(D) & \quad \quad P(D)
\end{align*}
\]
For this system,
\[N = 2, M = 1, a_1 = 3, a_2 = 2, b_1 = 1, b_2 = 0.\]
For practical systems, \(M \leq N\). It can be shown that if \(M > N\), an LTI differential system acts as an \((M - N)\)th-order differentiator.
A differentiator is an unstable system because bounded input (e.g. a step input) results in an unbounded output (a Dirac impulse \(\delta(t)\)).
The principle of superposition and circuit analysis using differential equations was taught in 1st year circuit courses.

- Key conceptual difference: the problem is now tackled from a “black box” approach.

System modelling is the key not only for the analysis of circuits, but other types of systems as well (financial, mechanical and others).

The course overlaps with 2nd year Control course, but the emphasis is different.

- Equations from last two slides look similar to transfer functions of systems using Laplace Transforms, but they are actually different. Here we remain in time domain so far. Transfer function analysis is in a new domain (the s-domain). This will be taught later both in this course and in the Control course.

<table>
<thead>
<tr>
<th>Time-domain</th>
<th>s-domain</th>
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| \[
\frac{d^2y}{dt^2} + 3 \frac{dy}{dt} + 2y(t) = \frac{dx}{dt}
\]
\[
\Rightarrow (D^2 + 3D + 2)y(t) = Dx(t)
\] | \[
(s^2 + 3s + 2)Y(s) = sX(s)
\] |
| \[
\Rightarrow H(s) = \frac{Y(s)}{X(s)} = \frac{s}{(s^2 + 3s + 2)}
\] |