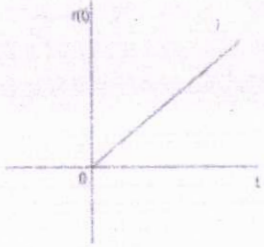


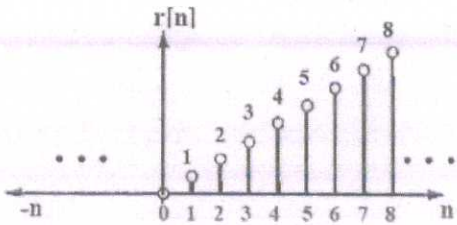
SCORING INDICATORS

COURSE NAME: SIGNALS AND SYSTEMS

COURSE CODE: 5201

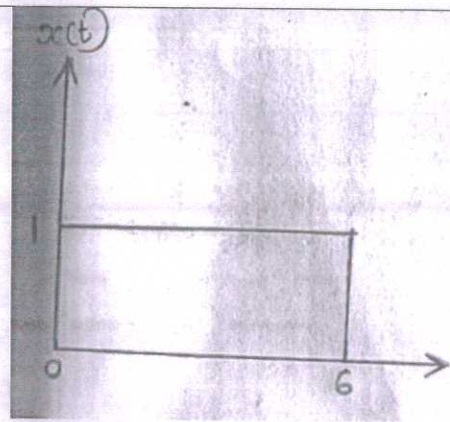
QID: 2109240233

QNo	Scoring Indicator	Split score	Sub Total	Total score
	PARTA	1	1	9
I.1	$x(t) = A\sin(\omega t + \phi)$	1	1	
I.2	A random signal is a signal that varies unpredictably over time.	1	1	
I.3	An odd signal is a signal that satisfies the condition $x(-t) = -x(t)$. For discrete-time signals, it satisfies $x[-n] = -x[n]$.	1	1	
I.4	A differential equation representing a continuous-time system may take the form: $a_n \frac{d^n y(t)}{dt^n} + \dots + a_1 \frac{dy(t)}{dt} + a_0 y(t) = b_m \frac{d^m x(t)}{dt^m} + \dots + b_0 x(t)$	1	1	
I.5	A system is time-invariant if its behavior and characteristics do not change over time, i.e., the output is shifted in time if the input is shifted.	1	1	
I.6	The Nyquist rate is twice the maximum frequency component, so for a 10KHz signal, the Nyquist rate is $2 \times 10\text{KHz} = 20\text{KHz}$	1	1	
I.7	Fourier Series (FS) is used to evaluate the frequency domain representation of a continuous-time periodic signal.	1	1	
I.8	The Laplace Transform of the unit step function $u(t)$ is $\frac{1}{s}$	1	1	
I.9	The inverse Laplace transform of $\frac{2}{s^2+4}$ is $\sin(2t)$	1	1	
	PART B			24
II.1	Continuous-time unit ramp signal: $r(t) = \begin{cases} t, & t \geq 0 \\ 0, & t < 0 \end{cases}$ 	1.5*2	3	

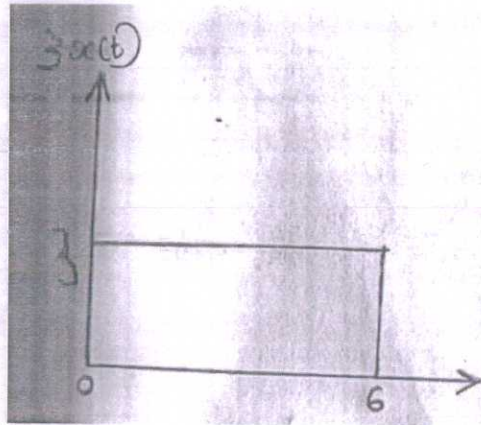
	<p>Discrete-time unit ramp signal: $r[n] = \begin{cases} n, & n \geq 0 \\ 0, & n < 0 \end{cases}$</p> 			
II.2	<p>Real signal: A signal is real if its value is a real number for all time. Example: $x(t) = \cos(t)$</p> <p>Imaginary signal: A signal is imaginary if its value is an imaginary number for all time. Example: $x(t) = j\sin(t)$</p>	1.5*2	3	
II.3	<p>Odd signal: $x(-t) = -x(t)$</p> <p>Example: $x(t) = \sin(t)$</p> <p>Even signal: $x(-t) = x(t)$.</p> <p>Example: $x(t) = \cos(t)$. Odd signals exhibit symmetry about the origin, while even signals exhibit symmetry about the vertical axis.</p>	1.5*2	3	
II.4	<p>Time shifting is the process of delaying or advancing a signal along the time axis. It alters the position of the signal without changing its shape.</p> <ul style="list-style-type: none"> Right shift (time delay): If $x(t)$ is the original signal, a right shift (delaying the signal) by t_0 is represented as $x(t - t_0)$. This shifts the signal to the right on the time axis, delaying it by t_0 units. <p>Effect: Each point on the signal occurs t_0 units later than it would in the original signal.</p> <ul style="list-style-type: none"> Left shift (time advance): A left shift (advancing the signal) by t_0 is represented as $x(t + t_0)$. This shifts the signal to the left, advancing it by t_0 units. <p>Effect: Each point on the signal occurs t_0 units earlier than it would in the original signal.</p>	<p>Definition - 1</p> <p>Explanation - 2</p>	3	

	<p>Continuous-Time Systems:</p> <ol style="list-style-type: none"> 1. Definition: Process signals that are defined for all continuous values of time t. 2. Signal Representation: Represented as $x(t)$, where t is a continuous variable. 3. Applications: Used in analog systems like electrical circuits and communication systems. 			
II.5	<p>Discrete-Time Systems:</p> <ol style="list-style-type: none"> 1. Definition: Process signals defined only at discrete intervals of time n. 2. Signal Representation: Represented as $x[n]$, where n is an integer representing time instants. 3. Applications: Used in digital systems like digital filters, signal processors, and computers. 	1+1+1	3	
II.6	The impulse response characterizes a continuous-time linear time-invariant (LTI) system completely. It provides the output of the system when a unit impulse is applied at the input and can be used to compute the output for any arbitrary input using convolution.	3	3	
II.7	<p>a) Time variant: $y(t) = x^2(t)$ is time-variant because a time shift changes the form of the output.</p> <p>b) Time-invariant: $y(n) = x(n + 1)$ is time-invariant since shifting the input results in a similar shift in the output.</p>	1.5*2	3	
II.8	<p>Fourier transform of unit impulse signal, $\delta(t) \rightarrow 1$</p> <p>Fourier transform of unit step signal $u(t) \rightarrow \frac{1}{j\omega} + \pi\delta(\omega)$</p> <p>Fourier transform of cosine signal $\cos(\omega_0 t) \rightarrow \pi[\delta(\omega - \omega_0) + \delta(\omega + \omega_0)]$</p>	1*3	3	
II.9	<p>a) Right-sided exponential signal: ROC is $\text{Re}(s) > \alpha$.</p> <p>b) Left-sided exponential signal: ROC is $\text{Re}(s) < \alpha$.</p> <p>c) Two-sided exponential signal: ROC is $\alpha_1 < \text{Re}(s) < \alpha_2$</p>		3	
II.10	The linearity property of the Laplace Transform states that if $x_1(t) \rightarrow X_1(s)$ and $x_2(t) \rightarrow X_2(s)$, then for any constants a and b , $ax_1(t) + bx_2(t) \rightarrow aX_1(s) + bX_2(s)$	1*3	3	
	PART C			42
III	• Energy Signals:	Energy signals-3.5	7	

	<ul style="list-style-type: none"> • Definition: A signal is classified as an energy signal if its total energy is finite over all time. These signals typically decay to zero as $t \rightarrow \infty$ • Example: The exponentially decaying signal $x(t) = e^{-t}u(t)$ has finite energy, making it an energy signal. The integral of $x(t) ^2$ over time converges to a finite value. <p>• Power Signals:</p> <ul style="list-style-type: none"> • Definition: A signal is classified as a power signal if it has infinite energy but finite average power. These signals usually oscillate or repeat over time without decaying. • Example: The sinusoidal signal $x(t) = \cos(\omega t)$ oscillates indefinitely but has finite average power. <p>• Example Calculations:</p> <ul style="list-style-type: none"> • For energy signals, the total energy is calculated by integrating the square of the signal over all time. For the signal $x(t) = e^{-t}u(t)$, the energy converges to $\frac{1}{2}$. $E = \int_{-\infty}^{\infty} x(t) ^2 dt$ $E = \int_0^{\infty} e^{-2t} dt = \frac{1}{2}$ <ul style="list-style-type: none"> • For power signals, the average power is calculated as the mean square value over one period. For the signal $x(t) = \cos(\omega t)$, the power is $\frac{1}{2}$, constant over time. $P = \lim_{T \rightarrow \infty} \frac{1}{2T} \int_{-T}^T x(t) ^2 dt$ $P = \frac{1}{2\pi} \int_0^{2\pi} \cos^2(\omega t) dt = \frac{1}{2}$	Power signals-3.5 Definition-2 Explanation-2 Calculation-3		
--	---	---	--	--



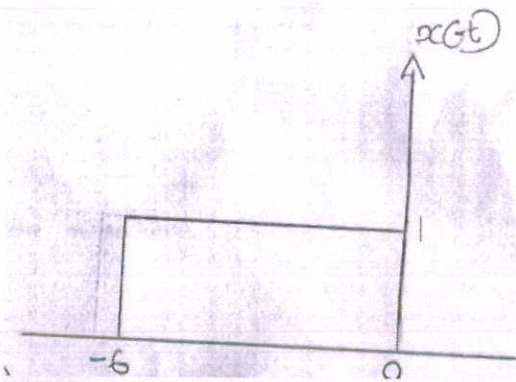
a)



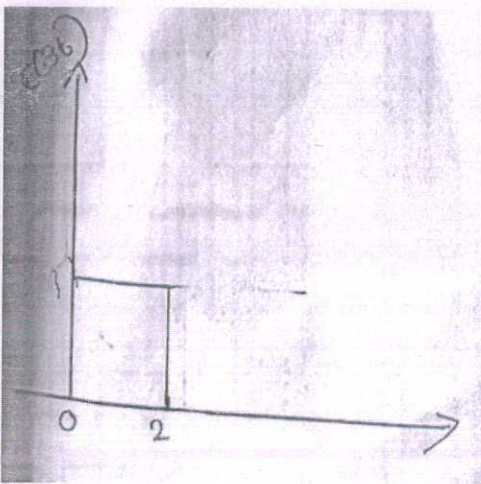
b)

IV

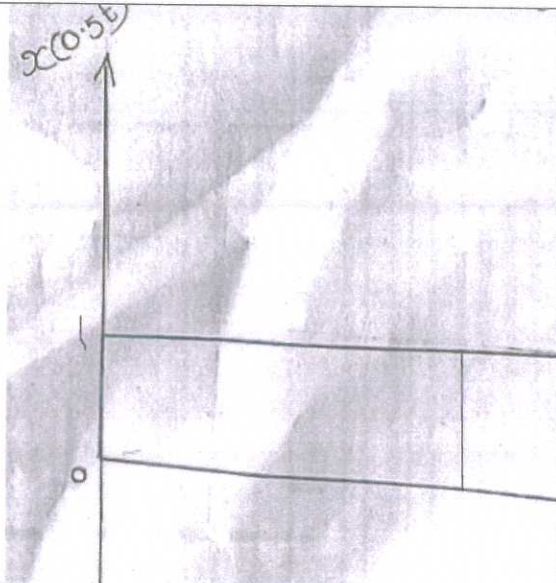
1+1+1+2+2 - 7



c)



d)

	 <p>e)</p>			
V	<p>a) $y(t) = x(t) + 3$</p> <ul style="list-style-type: none"> • Linearity Test: A system is linear if it satisfies both additivity and homogeneity. • Additivity: If the input is $x_1(t) + x_2(t)$, the system output becomes: $y(t) = (x_1(t) + x_2(t)) + 3 = x_1(t) + x_2(t) + 3$ <p>This is not equal $y_1(t) + y_2(t)$, so the system fails additivity.</p> <ul style="list-style-type: none"> • Homogeneity: For a scalar a, if the input is $a \cdot x(t)$, the system output becomes: $y(t) = a \cdot x(t) + 3$ <p>This is not equal $a \cdot y(t) = a \cdot (x(t) + 3)$, so the system fails homogeneity.</p> <ul style="list-style-type: none"> • Conclusion: The system is non-linear. <p>b) $y(t) = \frac{d}{dt} x(t)$</p> <ul style="list-style-type: none"> • Additivity: The derivative of a sum of inputs is: $y(t) = \frac{d}{dt} (x_1(t) + x_2(t)) = \frac{d}{dt} x_1(t) + \frac{d}{dt} x_2(t) = y_1(t) + y_2(t)$ Thus, additivity holds. • Homogeneity: For a scalar a, the derivative of a scaled input is: $y(t) = \frac{d}{dt} (a \cdot x(t)) = a \cdot \frac{d}{dt} x(t) = a \cdot y(t)$ Thus, homogeneity holds. • Conclusion: The system is linear. <p>c) $y(t) = e^{x(t)}$</p> <ul style="list-style-type: none"> • Additivity: For $y(t) = e^{x_1(t)+x_2(t)}$, we see: $y(t) = e^{x_1(t)+x_2(t)} \neq e^{x_1(t)} + e^{x_2(t)}$ 	1+2+2+2	7	

	<p>Thus, the system fails additivity.</p> <ul style="list-style-type: none"> Homogeneity: For a scalar a, the output becomes: $y(t) = e^{a \cdot x(t)} \neq a \cdot e^{x(t)}$ <p>Thus, the system fails homogeneity.</p> <ul style="list-style-type: none"> Conclusion: The system is non-linear. <p>d) $y(t) = \cos(x(t))$</p> <ul style="list-style-type: none"> Additivity: $y(t) = \cos(x_1(t) + x_2(t))$, we see: $y(t) = \cos(x_1(t) + x_2(t)) \neq \cos(x_1(t)) + \cos(x_2(t))$ <p>Thus, the system fails additivity.</p> <ul style="list-style-type: none"> Homogeneity: For a scalar a, the output becomes: $y(t) = \cos(a \cdot x(t)) \neq a \cdot \cos(x(t))$. Thus, the system fails homogeneity. <p>Conclusion: The system is non-linear.</p>			
VI	<ul style="list-style-type: none"> Linearity: <ul style="list-style-type: none"> A system is linear if it satisfies additivity (output for $x_1(t) + x_2(t)$ is the sum of individual outputs) and homogeneity (output for $a \cdot x(t)$ is $a \cdot y(t)$). Example: $y(t) = 3x(t)$ is linear, but $y(t) = x^2(t)$ is non-linear. Time-Invariance: <ul style="list-style-type: none"> A system is time-invariant if a time shift in the input results in an identical time shift in the output. Example: $y(t) = 2x(t)$ is time-invariant, while $y(t) = x(2t)$ is time-variant. Causality: <ul style="list-style-type: none"> A system is causal if the output depends only on present and past inputs, not future values. Example: $y(t) = x(t) + x(t - 1)$ is causal, but $y(t) = x(t) + x(t + 1)$ is non-causal. Stability: <ul style="list-style-type: none"> A system is stable if a bounded input results in a bounded output (BIBO Stability). Example: $y(t) = 0.5x(t)$ is stable, while $y(t) =$ 	<p>Linearity-1 Time invariance-2 Causality-2 Stability-2 (Equation-1, Explanati on-1)</p>	7	

	$2^t x(t)$ is unstable.			
VII	<ul style="list-style-type: none"> Sampling Theorem: <ul style="list-style-type: none"> The Sampling Theorem states that a signal can be exactly reconstructed from its samples if it is sampled at a rate greater than twice the maximum frequency of the signal, known as the Nyquist rate. Nyquist Rate: $f_s \geq 2f_{max}$, where f_s is the sampling rate and f_{max} is the highest frequency in the signal. Aliasing: <ul style="list-style-type: none"> Aliasing occurs when the sampling rate is below the Nyquist rate, causing different signals to become indistinguishable or "fold" into each other. This results in high-frequency components appearing as lower frequencies in the sampled signal. <ul style="list-style-type: none"> Impact of Incorrect Sampling: <ul style="list-style-type: none"> If the signal is under sampled, it leads to loss of information and distortion in the reconstructed signal, making it impossible to recover the original signal accurately. Example: A 10 kHz signal sampled below 20 kHz will exhibit aliasing, where the high-frequency components will appear incorrectly in the sampled data. 	Sampling Theorem-2 Aliasing-4(fig-2+explanati on-2) Impact- 2	7	
VIII	<p>a) Linearity:</p> <ul style="list-style-type: none"> The DTFT is linear, meaning if $x_1[n]$ and $x_2[n]$ are two signals and a and b are constants, then: 		7	

	<p>DTFT{$ax_1[n] + bx_2[n]$} = $aX_1(e^{j\omega}) + bX_2(e^{j\omega})$</p> <ul style="list-style-type: none"> • Illustration: If $x_1[n] = \delta[n]$ and $x_2[n] = u[n]$, their combined transform is a sum of individual transforms. <p>b) Time Shifting:</p> <ul style="list-style-type: none"> • If a signal $x[n]$ is shifted by n_0, the DTFT of the shifted signal is: $\text{DTFT}x[n - n_0] = X(e^{j\omega})e^{-j\omega n_0}$ <p>Illustration: A time shift introduces a phase shift in the frequency domain.</p> <p>c) Frequency Shifting:</p> <ul style="list-style-type: none"> • Multiplying a signal by a complex exponential in the time domain shifts the frequency components: DTFT{$x[n]e^{j\omega_0 n}$} = $X(e^{j(\omega - \omega_0)})$ • Illustration: Shifting the frequency of a signal modifies its spectrum. <p>d) Convolution Property:</p> <ul style="list-style-type: none"> • The convolution of two signals in the time domain corresponds to the multiplication of their DTFTs: DTFT$x_1[n] * x_2[n] = X_1(e^{j\omega}) \cdot X_2(e^{j\omega})$ • Illustration: Useful for filtering, where convolving a signal with a filter's impulse response corresponds to multiplying in the frequency domain 			
IX	<p>a) Linearity:</p> <ul style="list-style-type: none"> • The Fourier series is linear, meaning if two signals and $x_2(t)$ are combined with constants a and b, then: $ax_1(t) + bx_2(t) \Rightarrow aC_1 + bC_2$ <p>Example: Superposition of signals results in the superposition of their Fourier coefficients.</p> <p>b) Time Shift:</p> <ul style="list-style-type: none"> • Shifting a signal in time by t_0 introduces a phase shift in the Fourier series coefficients: $x(t - t_0) \Rightarrow C_n e^{-jn\omega_0 t_0}$ <p>Example: A time shift causes a corresponding phase</p>	1+2+2+2 (equation-1 +explanatio n-1 for each)	7	

	<p>shift in the frequency domain.</p> <p>c) Frequency Shift:</p> <ul style="list-style-type: none"> Multiplying a signal by a complex exponential in the time domain results in a frequency shift in the Fourier series $x(t)e^{j\omega_0 t} \Rightarrow C_{n-k}$ <p>Example: Shifting a signal's frequency changes the position of its Fourier coefficients.</p> <p>d) Multiplication in Time Domain:</p> <ul style="list-style-type: none"> Multiplying two signals in the time domain results in the convolution of their Fourier coefficients: $x_1(t) \cdot x_2(t) \Rightarrow C_1 * C_2$ <ul style="list-style-type: none"> Example: This property is useful in modulation, where time-domain multiplication spreads the frequency content. 			
X	<p>a) Fourier Transform of the Dirac Delta Function $\delta(t)$:</p> <ul style="list-style-type: none"> The Fourier Transform is given by: $\mathcal{F}\{\delta(t)\} = \int_{-\infty}^{\infty} \delta(t)e^{-j\omega t} dt = 1$ Explanation: The delta function $\delta(t)$ only exists at $t=0$, so the integral evaluates to 1. This means the Dirac delta function contains all frequencies equally, resulting in a flat spectrum in the frequency domain. <p>b) Fourier Transform of $e^{-a t }$ (Exponential Decay Function):</p> <ul style="list-style-type: none"> The Fourier Transform is: $\mathcal{F}\{e^{-a t }\} = \int_{-\infty}^{\infty} e^{-a t }e^{-j\omega t} dt = \frac{2a}{a^2 + \omega^2}$ Explanation: This is an even, symmetric function, decaying on both sides of $t=0$. The Fourier transform results in a rational function, indicating that the frequency response of the signal is low-pass, meaning high-frequency components are attenuated. <p>c) Fourier Transform of $\sin(\omega_0 t)$ (Sine Function):</p> <ul style="list-style-type: none"> The Fourier Transform is: $\mathcal{F}\sin(\omega_0 t) = \int_{-\infty}^{\infty} \sin(\omega_0 t)e^{-j\omega t} dt =$ 	<p>Dirac delta function - 1 (calculation)</p> <p>Exponential decay function - 3 (eqn-1 + calculation-2)</p> <p>Sine function - 3 (eqn-1 + calculation-2)</p>	7	

	$\frac{j}{2} [\delta(\omega - \omega_0) - \delta(\omega + \omega_0)]$ <ul style="list-style-type: none"> Explanation: The sine function results in two delta functions in the frequency domain, located at $\pm\omega_0$. This means the signal has energy concentrated at these two frequencies, corresponding to its oscillatory nature. 				
XI	<ul style="list-style-type: none"> Unit Step Function $u(t)$: <ul style="list-style-type: none"> $\mathcal{L}\{u(t)\} = \frac{1}{s}, \text{ for } \Re(s) > 0$ Exponential Function e^{at}: $\mathcal{L}\{e^{at}\} = \frac{1}{s - a}, \text{ for } \Re(s) > a$ <ul style="list-style-type: none"> Sine Function $\sin(\omega t)$ <ul style="list-style-type: none"> $\mathcal{L}\{\sin(\omega t)\} = \frac{\omega}{s^2 + \omega^2}, \text{ for } \Re(s) > 0$ Cosine Function $\cos(\omega t)$ $\mathcal{L}\{\cos(\omega t)\} = \frac{s}{s^2 + \omega^2}, \text{ for } \Re(s) > 0$ Ramp Function t: $\mathcal{L}\{t\} = \frac{1}{s^2}, \text{ for } \Re(s) > 0$ Dirac Delta Function $\delta(t)$ $\mathcal{L}\{\delta(t)\} = 1$ Exponential Decay $e^{-at}u(t)$ $\mathcal{L}\{e^{-at}u(t)\} = \frac{1}{s + a}, \text{ for } \Re(s) > -a$ 	1*7 (1 mark for each signal)		7	
XII	<p>a) Inverse Laplace Transform of $\frac{1}{(s+1)(s+2)}$:</p> <ul style="list-style-type: none"> Use partial fraction decomposition: $\frac{1}{(s+1)(s+2)} = \frac{A}{s+1} + \frac{B}{s+2}$ Solving for A and B, we get: $\frac{1}{(s+1)(s+2)} = \frac{1}{s+1} - \frac{1}{s+2}$ <p>Inverse Laplace Transform:</p>			3+1+3	7

	$\mathcal{L}^{-1}\left\{\frac{1}{s+1} - \frac{1}{s+2}\right\} = e^{-t} - e^{-2t}$ <p>b) Inverse Laplace Transform of $\frac{2s}{s^2+1}$:</p> <ul style="list-style-type: none"> Recognize this as the Laplace transform of a cosine function: $\mathcal{L}^{-1}\left\{\frac{2s}{s^2+1}\right\} = 2\cos(t)$ <p>c) $X(s) = \frac{s+3}{s^2+6s+13}$</p> $s^2 + 6s + 13 = (s + 3)^2 + 4$ $\frac{s + 3}{(s + 3)^2 + 4} = \frac{s + 3}{(s + 3)^2 + 2^2}$ <p>By applying frequency shifting property,</p> $\mathcal{L}^{-1}\left\{\frac{s + 3}{(s + 3)^2 + 4}\right\} = e^{-3t}\cos(2t)$			
XIII	<ul style="list-style-type: none"> The ROC of the Laplace transform is the range of values of s (in the complex plane) for which the Laplace transform integral converges. It defines where the Laplace transform is valid. <ul style="list-style-type: none"> ROC for Stable Systems: <ul style="list-style-type: none"> A system is stable if the ROC includes the imaginary axis $\Re(s) = 0$. Stability means that the system produces bounded output for any bounded input. Example: For $e^{-at}u(t)$, the ROC is $\Re(s) > -a$, and for the system to be stable, the imaginary axis must be within the ROC. ROC for Causal Systems: <ul style="list-style-type: none"> A system is causal if the ROC is to the right of the rightmost pole. 	Definition of ROC-2 ROC for Stable Systems-3(explanati on-2+example -1) ROC for Causal Systems-3(explanati on-2+example -1)	7	

	<ul style="list-style-type: none"> • Causal systems depend only on present and past inputs (not future). • Example: For $e^{-at}u(t)$, the ROC is $\Re(s) > -a$, ensuring causality. 			
XIV	<p>a) Time Reversal:</p> <ul style="list-style-type: none"> • Property: If $x(t)$ has a Laplace transform $X(s)$, then the Laplace transform of $x(-t)$ is $X(-s)$. • Explanation: Reversing the signal in time corresponds to negating the s-variable in the Laplace domain. <p>b) Time Shifting:</p> <ul style="list-style-type: none"> • Property: If $x(t)$ has a Laplace transform $X(s)$, then the Laplace transform of $x(t - t_0)$ is: • $\mathcal{L}\{x(t - t_0)\} = e^{-st_0}X(s)$ • Explanation: Shifting a signal in time by t_0 introduces a multiplicative exponential factor in the Laplace domain. <p>c) Convolution in Time Domain:</p> <ul style="list-style-type: none"> • Property: The convolution of two signals $x_1(t)$ and $x_2(t)$ in the time domain corresponds to the multiplication of their Laplace transforms in the frequency domain: • $\mathcal{L}\{x_1(t) * x_2(t)\} = X_1(s)X_2(s)$ • Explanation: Convolution in time becomes a simple multiplication in the Laplace domain, simplifying the analysis of systems. <p>d) Final Value Theorem:</p> <ul style="list-style-type: none"> • Property: The final value of a signal $x(t)$ as $t \rightarrow \infty$ can be determined using its Laplace transform: $\lim_{t \rightarrow \infty} x(t) = \lim_{s \rightarrow 0} sX(s)$ • Explanation: This theorem helps find the steady-state value of a system without needing the inverse Laplace transform. 	1+2+2+2 (equation- 1+explanati on-1 for each)	7	