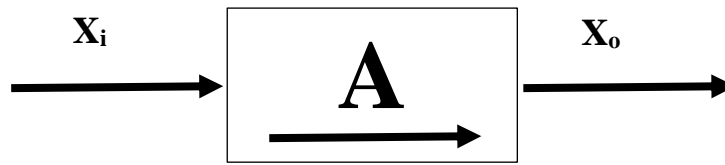


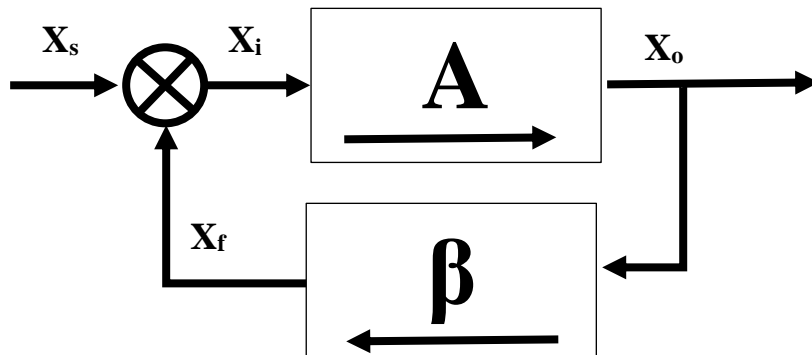
Realimentación

- Lazo abierto



$$A = \frac{X_o}{X_i}$$

- Lazo cerrado



$$X_o = AX_i$$

$$X_f = \beta X_o$$

$$X_i = X_s \pm X_f$$

Realimentación Positiva

$$X_o = AX_i$$

$$X_f = \beta X_o$$

$$X_i = X_s + X_f$$

$$\wedge X_o \uparrow X_f \uparrow \therefore X_i \uparrow$$

Entonces $X_o \uparrow \uparrow$

Realimentación Negativa

$$X_o = AX_i$$

$$X_f = \beta X_o$$

$$X_i = X_s - X_f$$

$$\wedge X_o \uparrow X_f \uparrow \text{ pero } X_i \downarrow$$

Expresión fundamental de un sistema con realimentación negativa

$$X_o = AX_i$$

$$X_o = A (X_s - X_f)$$

$$X_o = A (X_s - \beta X_o)$$

$$X_o = AX_s - \beta AX_o$$

$$X_o + \beta AX_o = AX_s$$

$$X_o (1 + \beta A) = AX_s$$

$$\frac{X_o}{X_s} = \frac{A}{1 + A \beta}$$

$$A_f = \frac{X_o}{X_s}$$

$$A_f = \frac{A}{1 + \beta A}$$

$D = 1 + \beta A$ Diferencia de retorno

$G_l = \beta A$ Ganancia de lazo

Amplificador Altamente Realimentado

$$\beta A \gg 1$$

$$\therefore \wedge A_f = \frac{A}{1 + \beta A} \text{ y } \beta A \gg 1$$

$$A_f = \frac{1}{\beta}$$

Desensibilización del sistema

Si varía A:

$$\frac{\partial A_f}{\partial A} = \frac{1 + \beta A - \beta A}{(1 + \beta A)^2} = \frac{1}{(1 + \beta A)^2}$$

$$\partial A_f = \partial \frac{A}{A} \frac{A}{(1 + \beta A)} \frac{1}{(1 + \beta A)}$$

$$\boxed{A_f}$$

$$\therefore \frac{\partial A_f}{A_f} = \frac{1}{D} \frac{\partial A}{A}$$

Si varía β :

$$\frac{\partial A_f}{\partial \beta} = \frac{-A^2}{(1 + \beta A)^2} = \frac{-A}{\underbrace{(1 + \beta A)}_{\boxed{A_f}}} \frac{A}{(1 + \beta A)}$$

$$\frac{\partial A_f}{A_f} = \frac{-\partial \beta A}{(1 + \beta A)}$$

$$\wedge \beta A \gg 1$$

$$\frac{\partial A_f}{A_f} = \frac{-\partial \beta}{\beta}$$