

**UNIVERSIDAD NACIONAL DE INGENIERIA**  
**FACULTAD DE CIENCIAS**  
**CIENCIAS DE LA COMPUTACION**



**Título del Trabajo**

**Practica Calificada 5: Modelamiento Dinámico**

**Autores**

Lázaro Camasca Edson Nicks

**Curso**

Inteligencia Artificial

**Profesor**

Antonio Morán

Lima – Perú  
(2019)

# 1. Planteamiento del problema

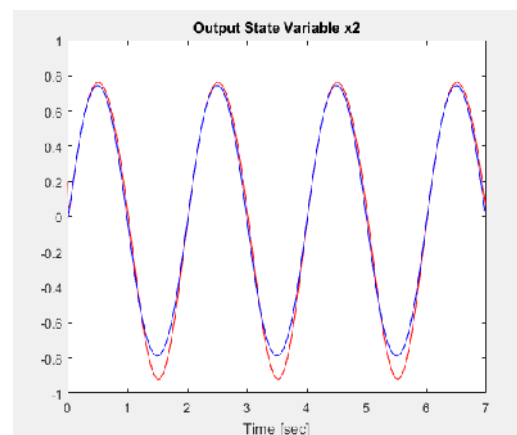
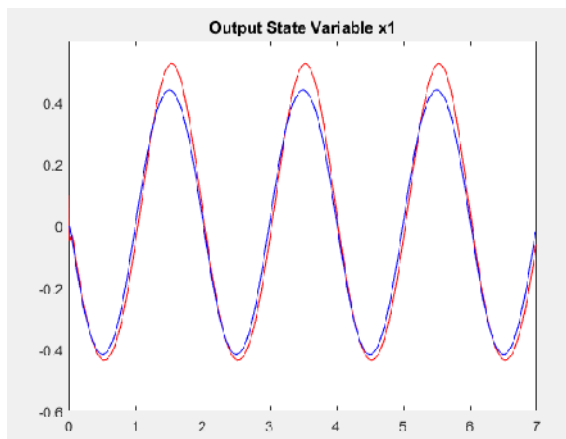
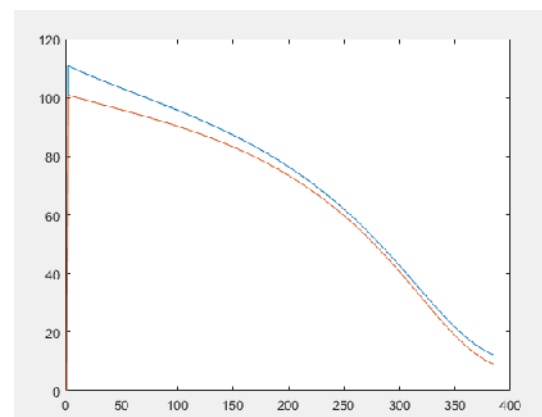
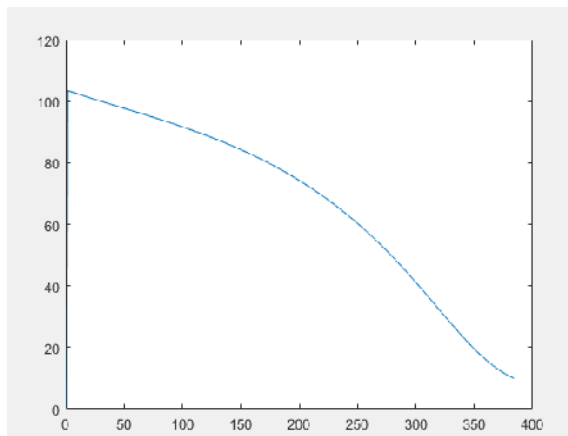
En este informe trabajaremos con un programa para el entrenamiento de una red neuronal dinámica para modelar un sistema no lineal utilizando el algoritmo DBP.

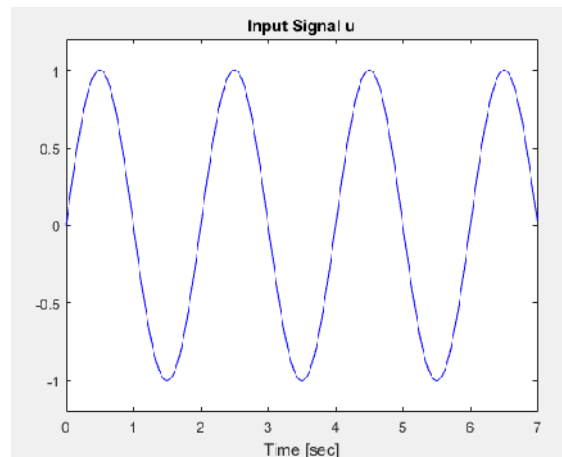
El sistema tiene 1 entrada y 2 señales de salida, variaremos los valores de centro, porcentaje de error.

## 2. Primer intento:

### 2.1. Entrenamiento\_ U1:

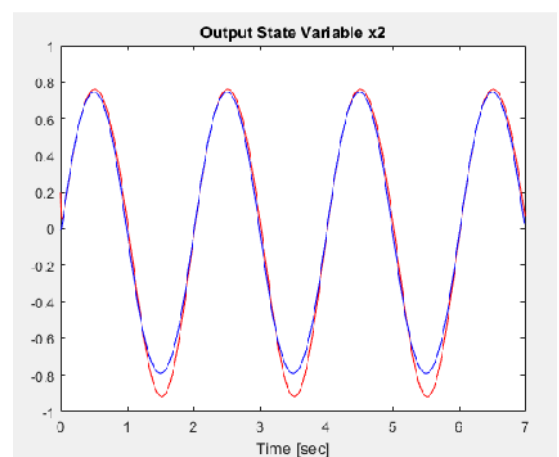
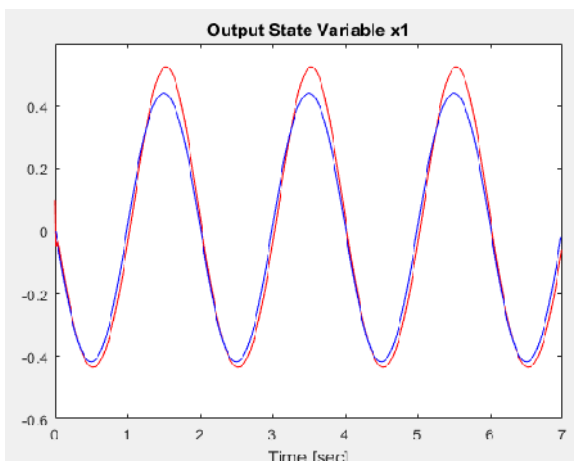
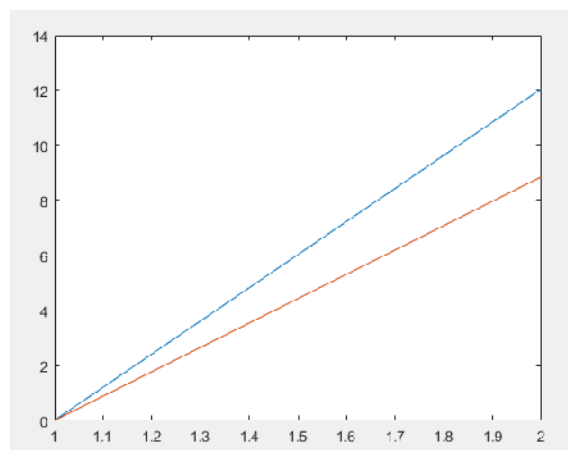
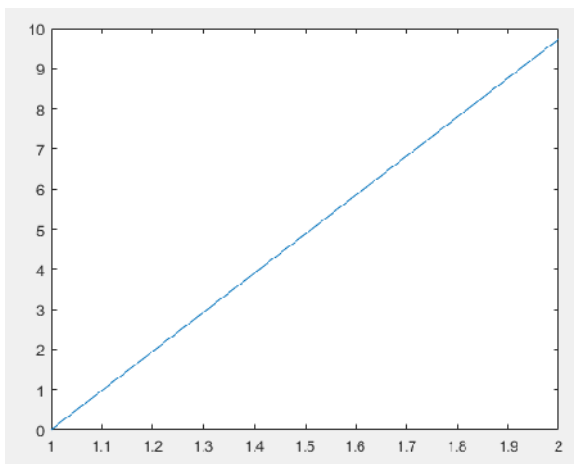
Datos iniciales	Automático
learning rate [v w]	0.02
learning rate [c: sigmoid center]	0.05
learning rate [a: sigmoid slope]	0.05
maximum value of error function (percentage %):	10
number of iteration steps	1000
erreltotal	0.0991

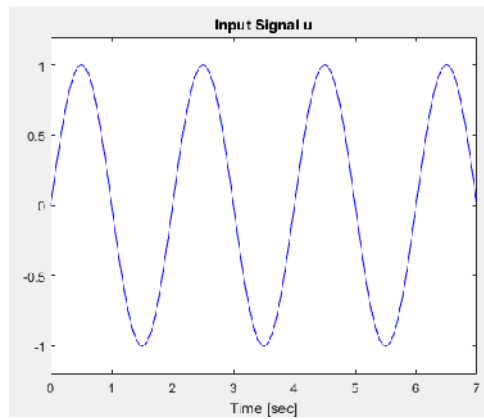




## 2.2. Validando\_ U2:

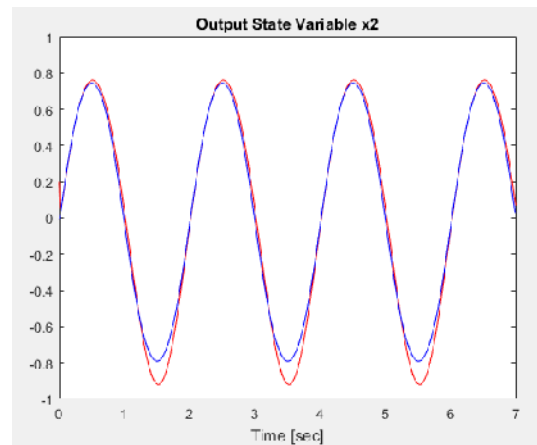
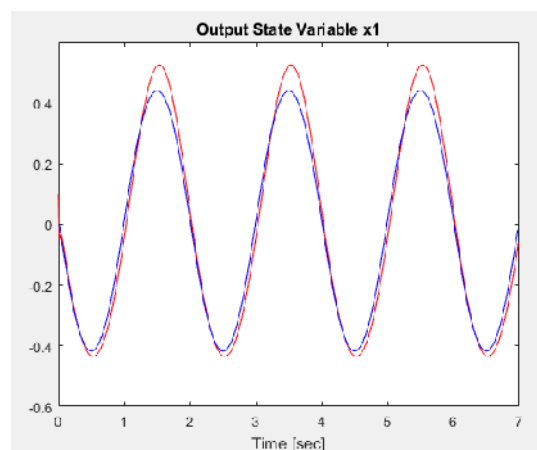
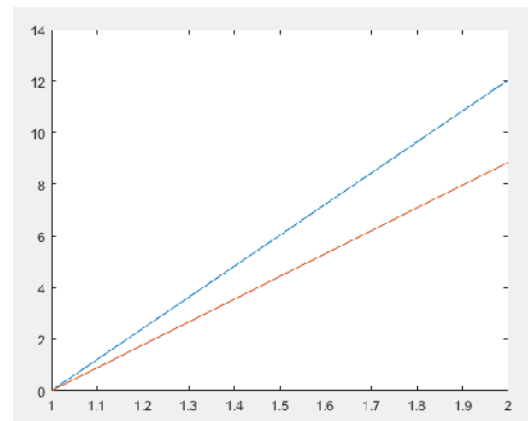
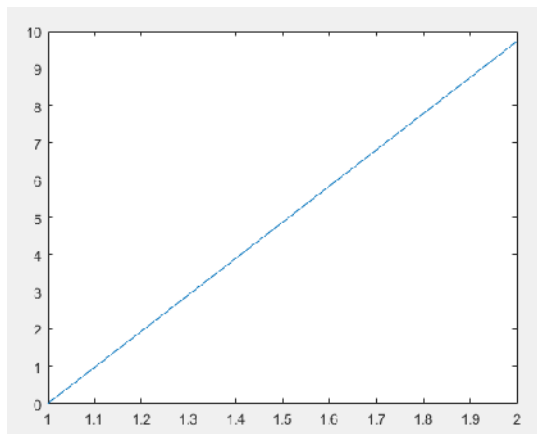
Datos iniciales	Automático
learning rate [v w]	0
learning rate [c: sigmoid center]	0
learning rate [a: sigmoid slope]	0
maximum value of error function (percentage %):	10
number of iteration steps	2
erreltotal	0.0982

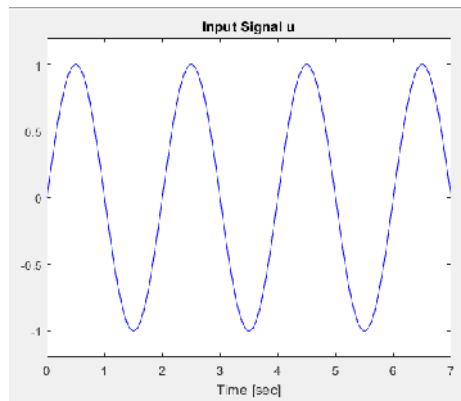




### 2.3.Validando\_ U3:

Datos iniciales	Automático
learning rate [v w]	0
learning rate [c: sigmoid center]	0
learning rate [a: sigmoid slope]	0
maximum value of error function (percentage %):	10
number of iteration steps	2
erreltotal	0.0989

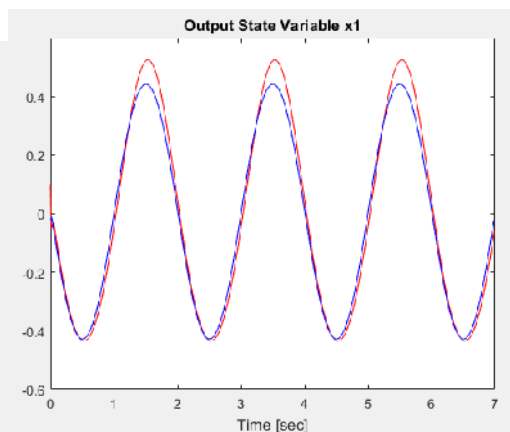
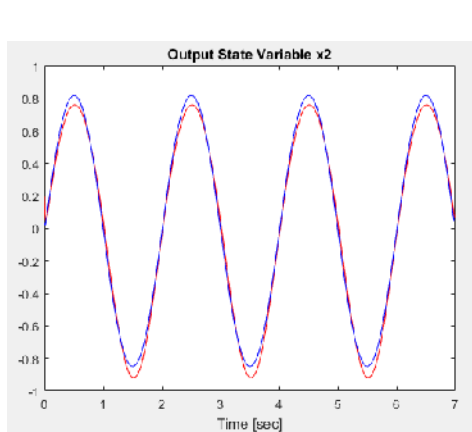
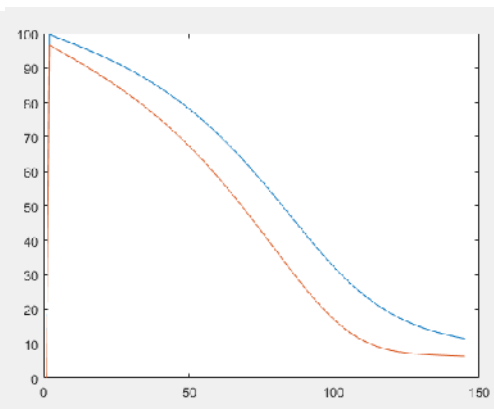
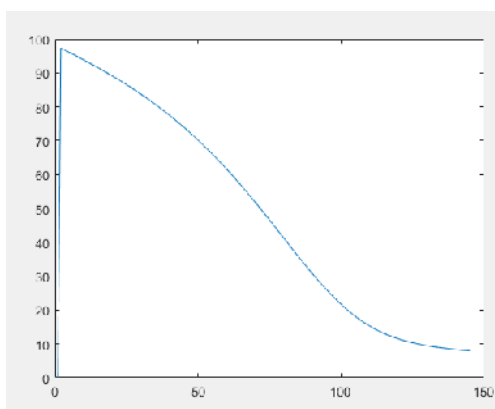


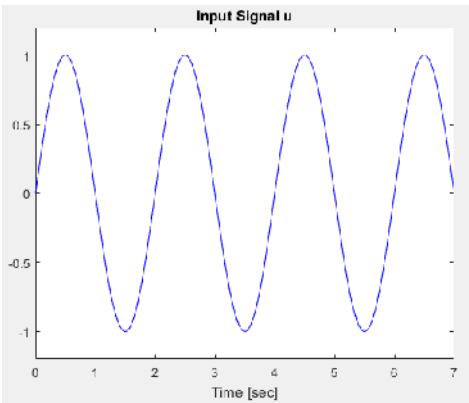


### 3. Segundo intento:

#### 3.1. Entrenamiento\_ U1:

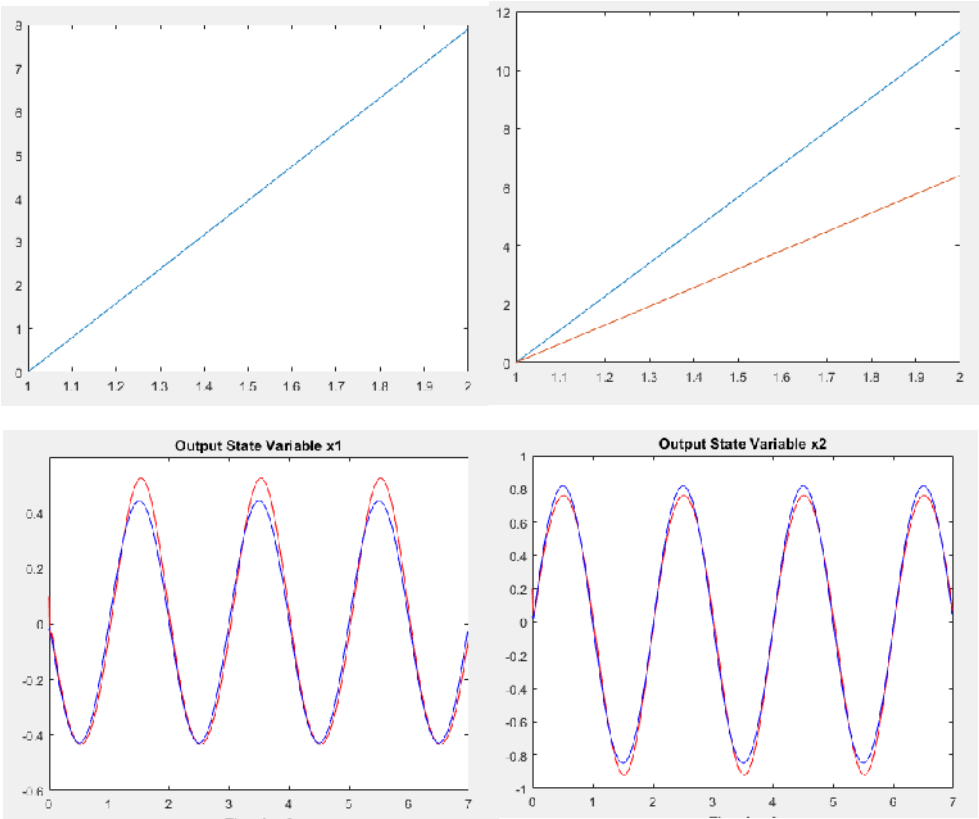
Datos iniciales	Automático
learning rate [v w]	0.08
learning rate [c: sigmoid center]	0.05
learning rate [a: sigmoid slope]	0.05
maximum value of error function (percentage %):	9
number of iteration steps	1000
erreltotal	0.0880

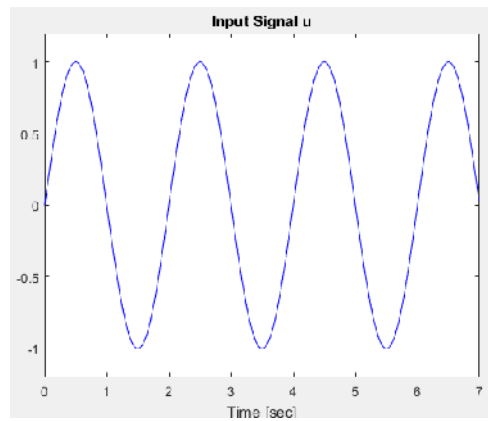




### 3.1 Validando\_ U2:

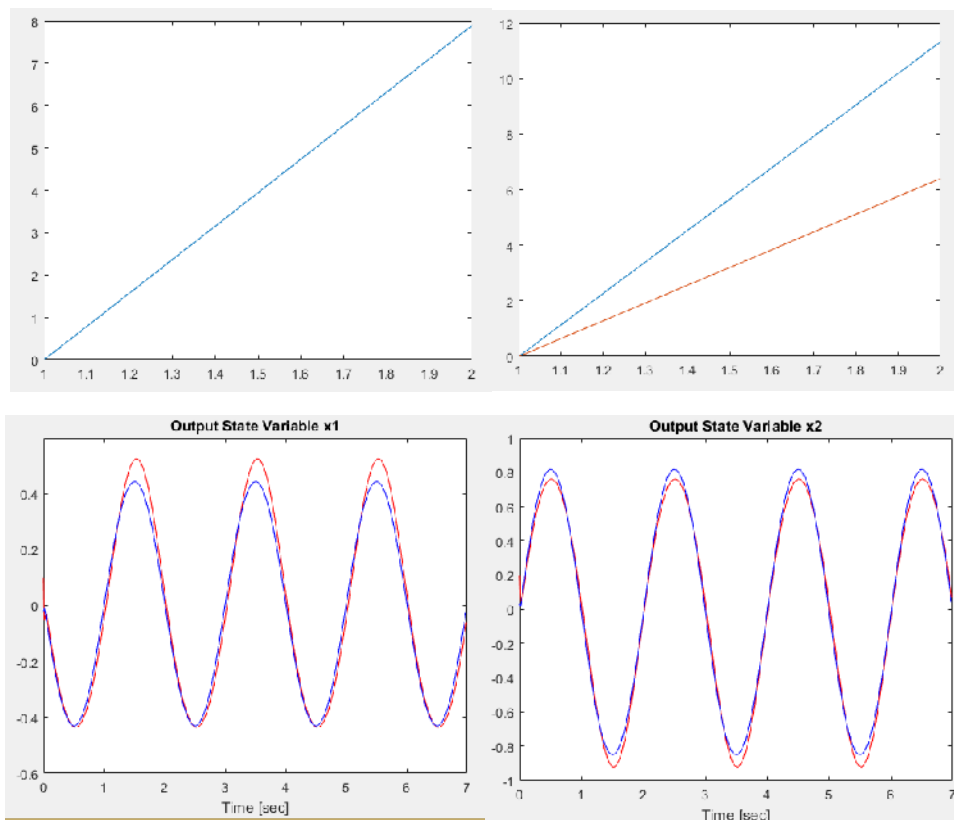
Datos iniciales	Automático
learning rate [v w]	0
learning rate [c: sigmoid center]	0
learning rate [a: sigmoid slope]	0
maximum value of error function (percentage %):	8
number of iteration steps	3
erreltotal	0.0849

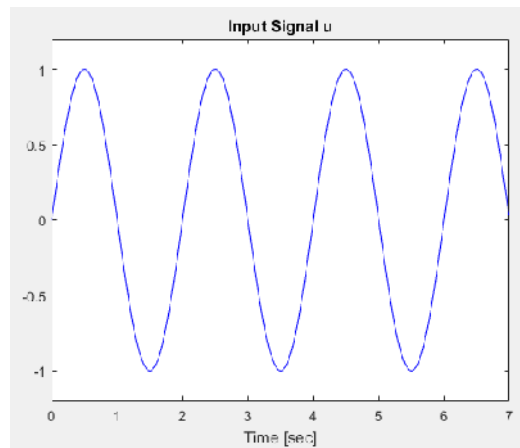




### 3.2 Validando\_ U3:

Datos iniciales	Automático
learning rate [v w]	0
learning rate [c: sigmoid center]	0
learning rate [a: sigmoid slope]	0
maximum value of error function (percentage %):	11
number of iteration steps	3
erreltotal	0.0863

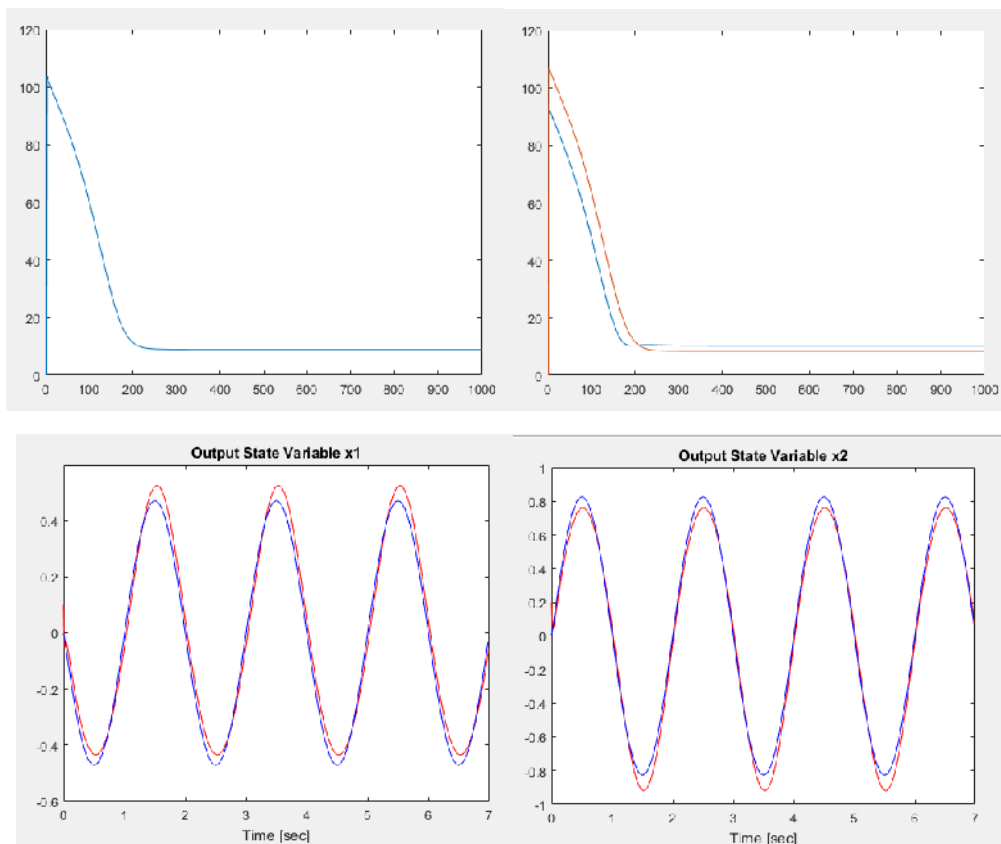




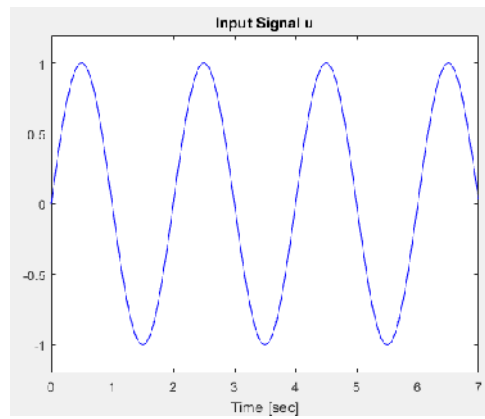
#### 4. Tercer intento:

##### 4.1. Entrenamiento\_ U1:

Datos iniciales	Automático
learning rate [v w]	0.05
learning rate [c: sigmoid center]	0.01
learning rate [a: sigmoid slope]	0.01
maximum value of error function (percentage %):	10
number of iteration steps	900
erreltotal	0.0947

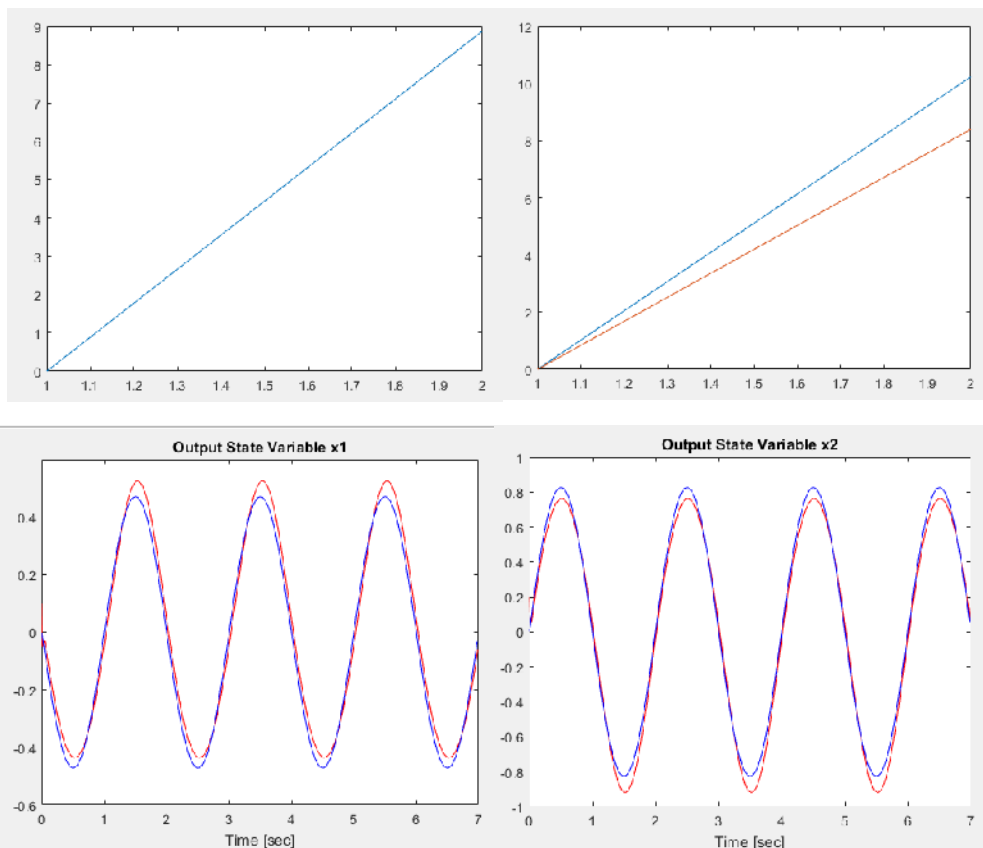


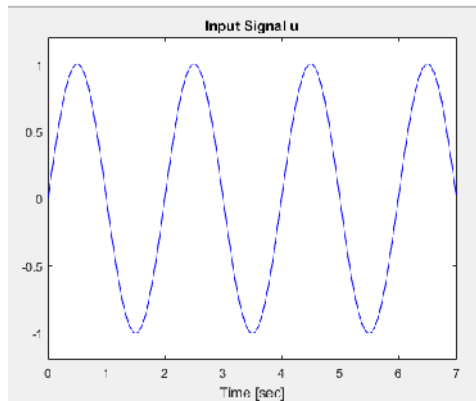




## 4.2. Validando\_U2:

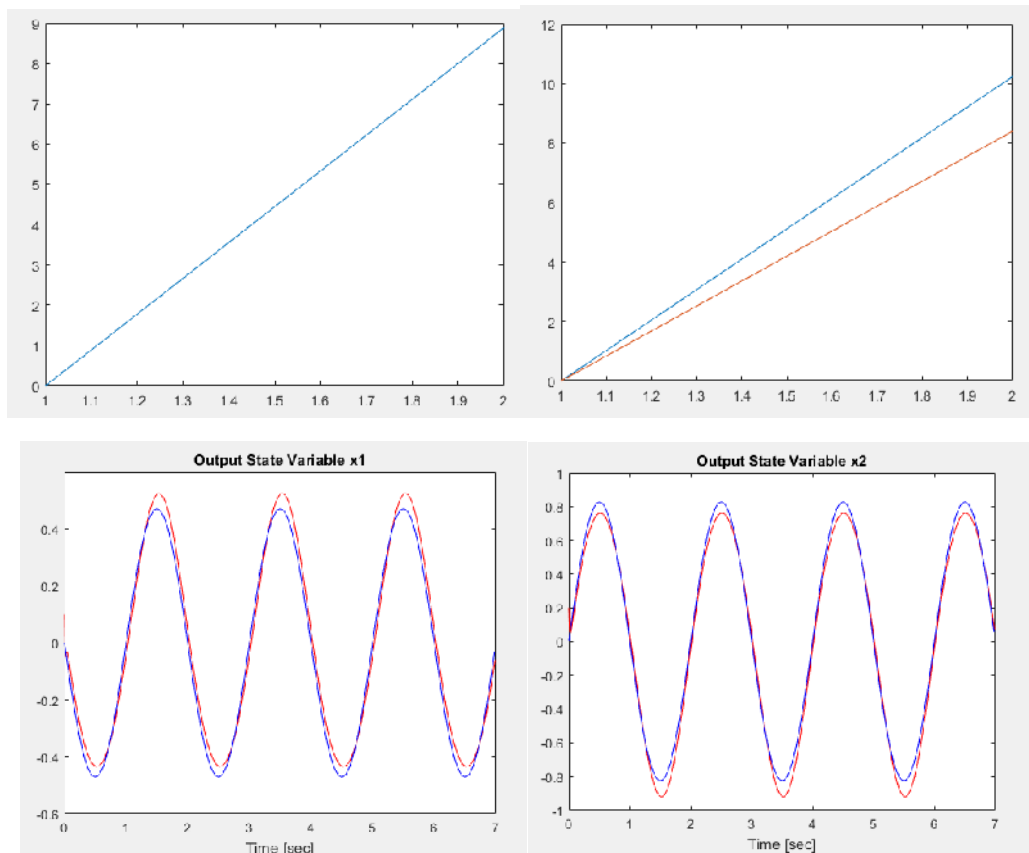
Datos iniciales	Automático
learning rate [v w]	0
learning rate [c: sigmoid center]	0
learning rate [a: sigmoid slope]	0
maximum value of error function (percentage %):	8
number of iteration steps	2
erreltotal	0.0785

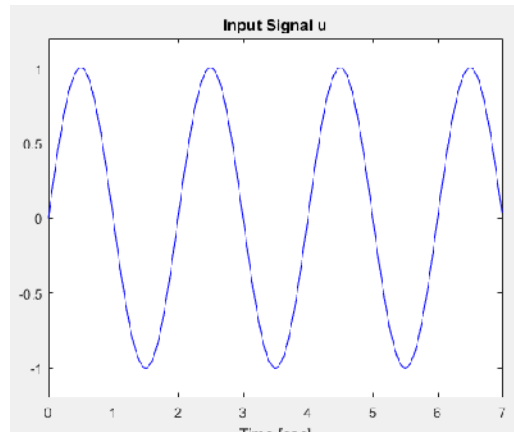




### 4.3. Validando\_U3:

Datos iniciales	Automático
learning rate [v w]	0
learning rate [c: sigmoid center]	0
learning rate [a: sigmoid slope]	0
maximum value of error function (percentage %):	8
number of iteration steps	2
erreltotal	0.0752





## CONCLUSIÓN

Como observamos cuando los valores de  $\eta$  son pequeños ( $0 < \eta < 0.05$ ), y los valores de “a” y “c” son “0” podemos notar en las gráficas que los resultados (rojo) se aproximan más a lo que queremos que salga (azul), y en caso contrario el error que obtenemos es mayor.