

TP3

Perceptrón Simple y Multicapa

Grupo 2

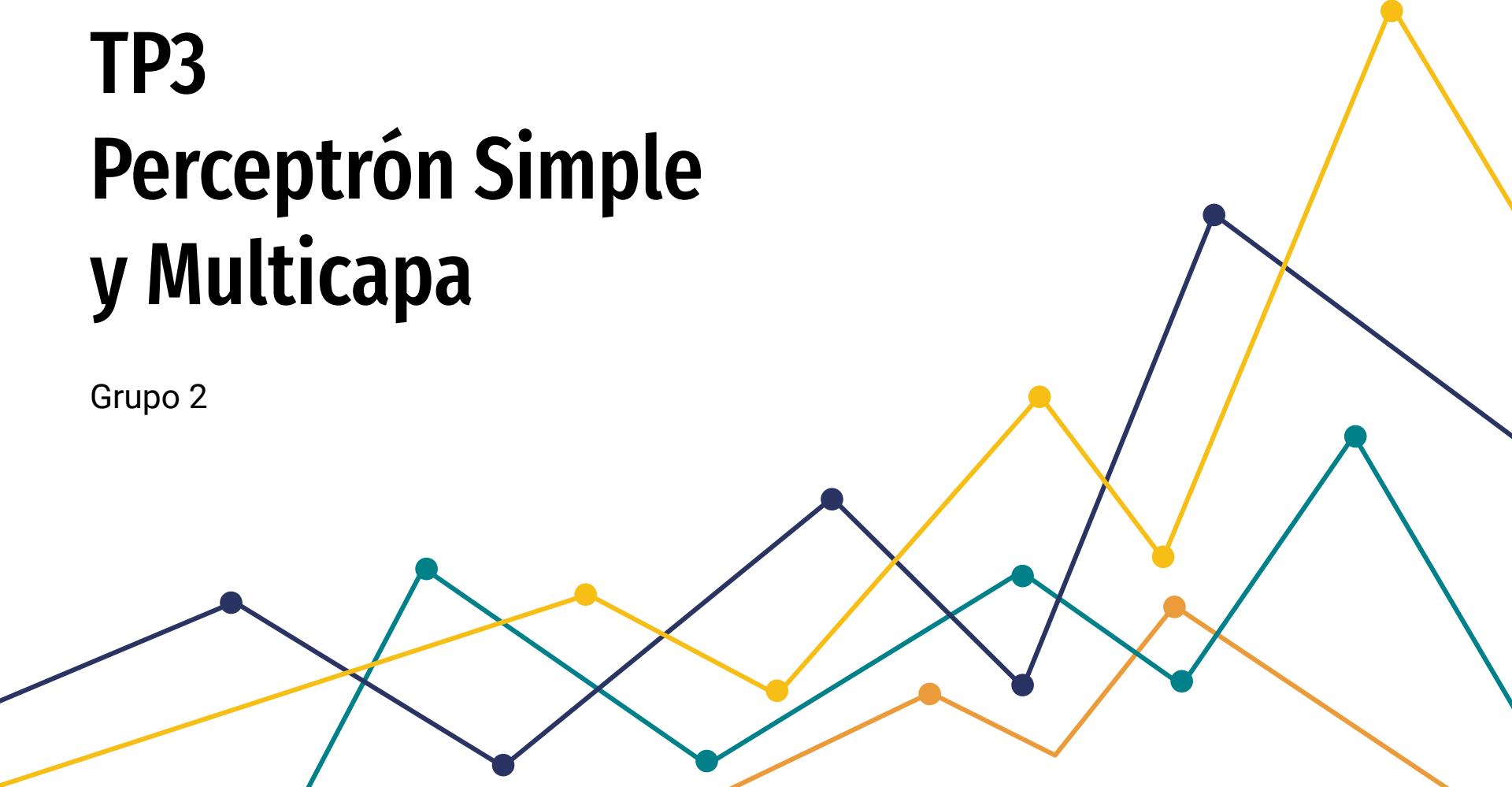


Tabla de contenidos

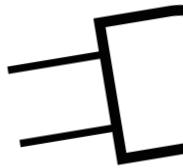
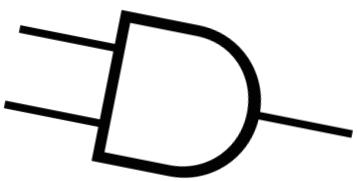
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- 4 Ejercicio 3: Perceptrón multicapa
- 5 Ejercicio 3: Pares
- 6 Reconocimiento de números

Ejercicio 1

? Preguntas:

- ¿Qué puede decir acerca de los problemas que puede resolver el perceptrón simple escalón en relación a los problemas planteados en la consigna?

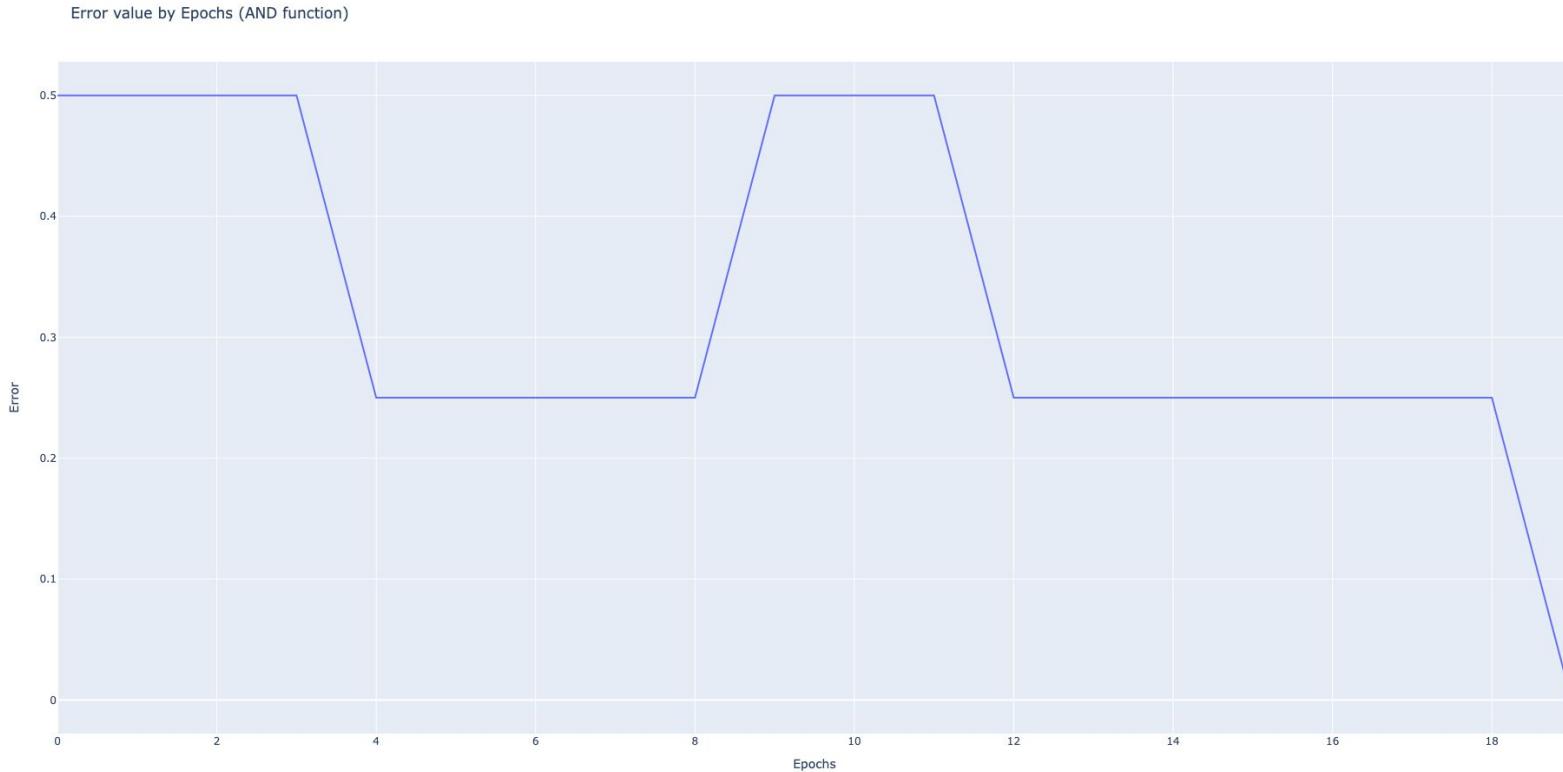
AND



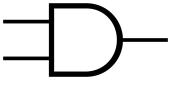
AND



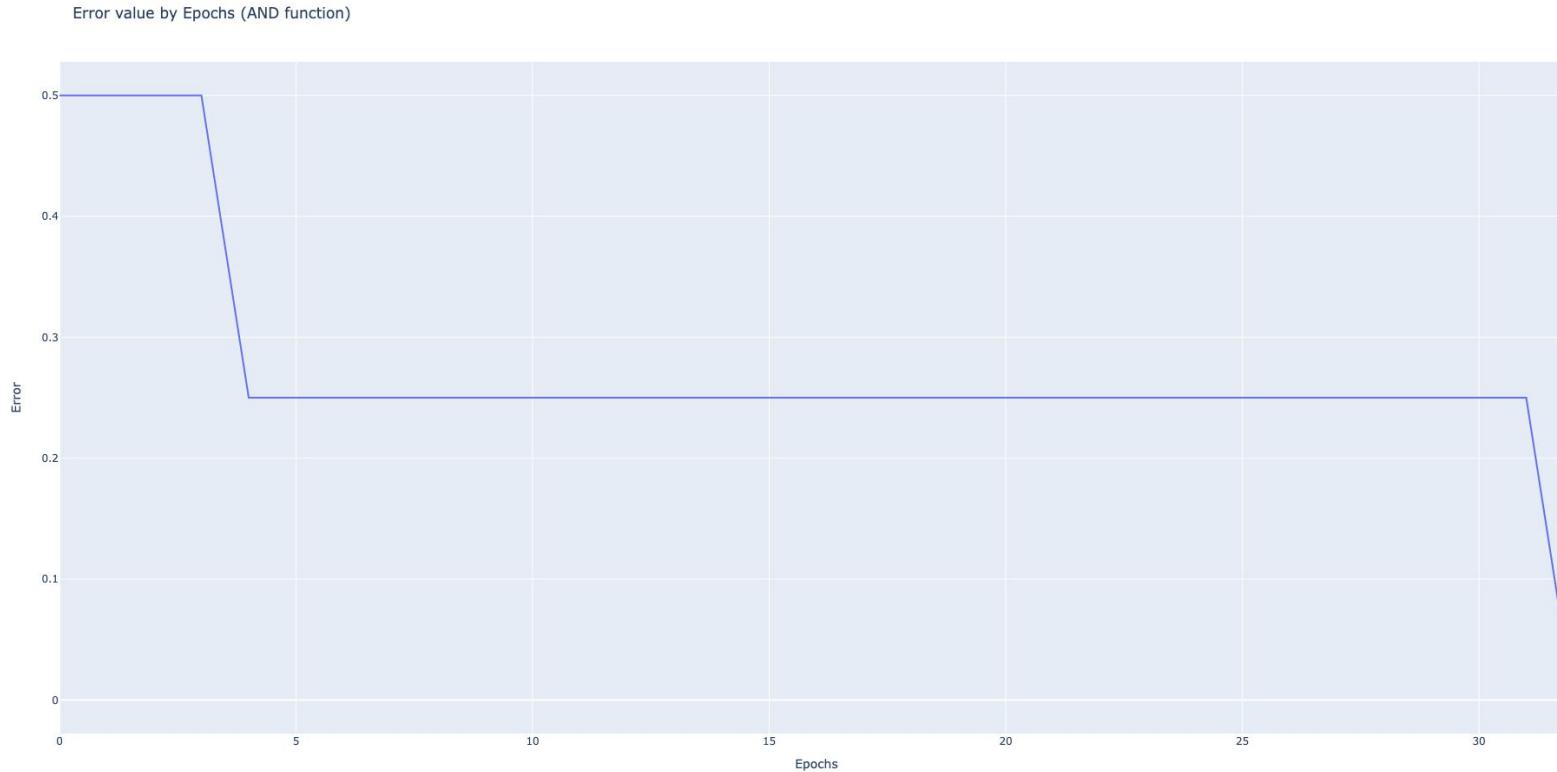
Ejercicio 1



AND



Ejercicio 1



AND



Ejercicio 1

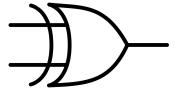




XOR

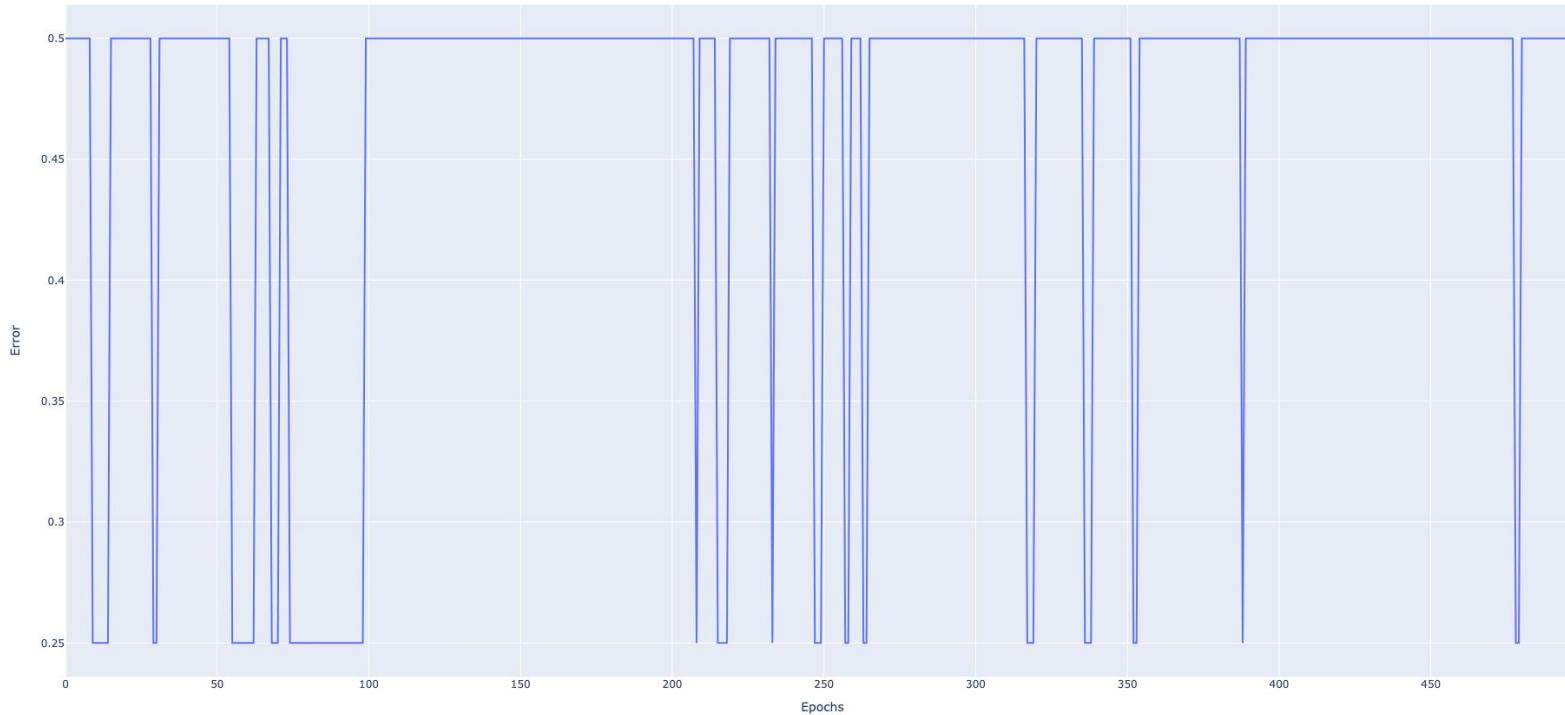


XOR

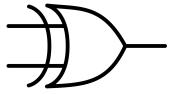


Ejercicio 1

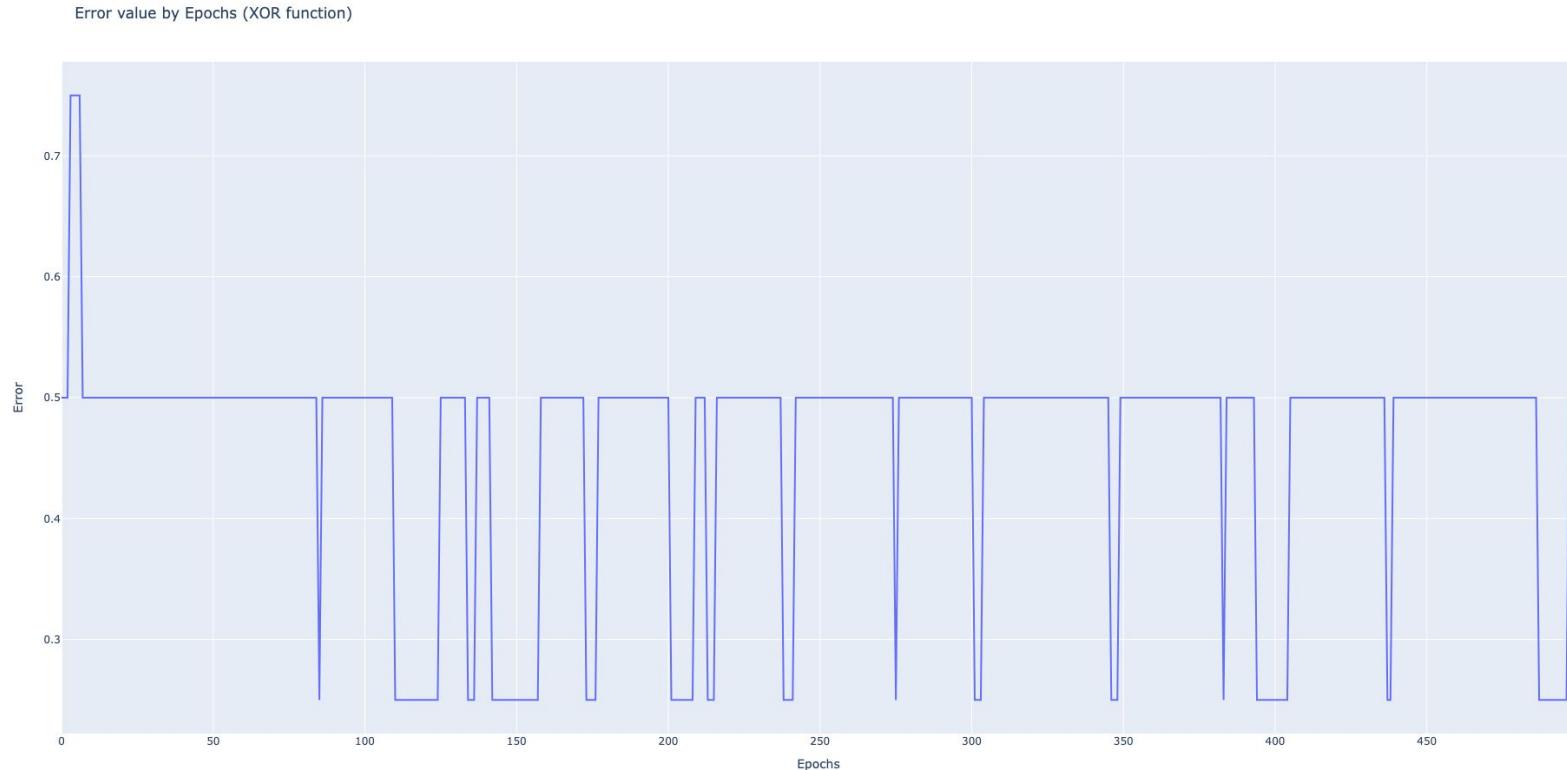
Error value by Epochs (XOR function)



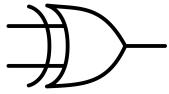
XOR



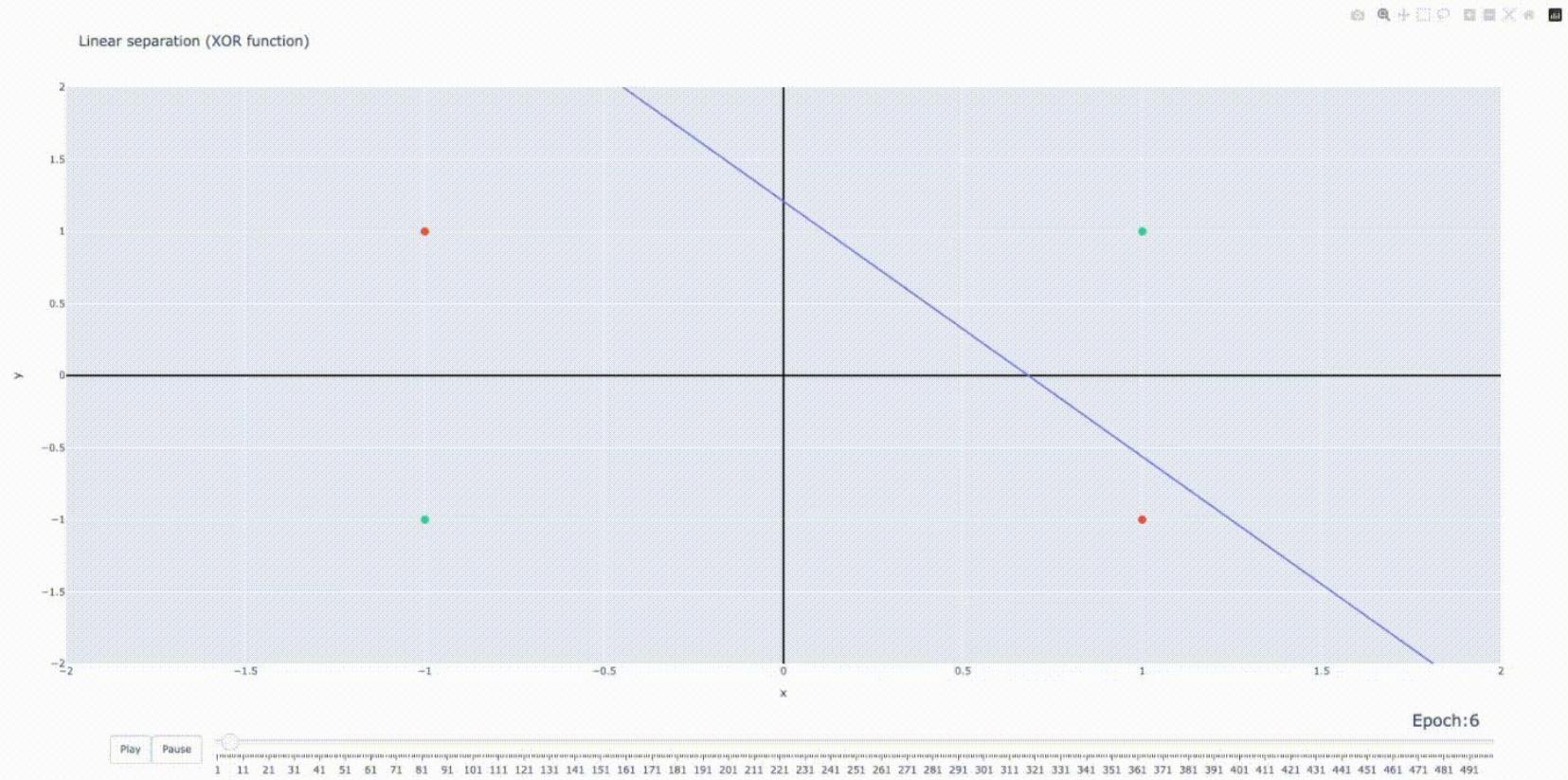
Ejercicio 1



XOR



Ejercicio 1



Ejercicio 2

? Preguntas:

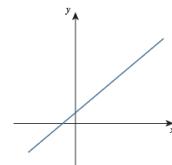
- Evaluar la capacidad de cada uno de los perceptrones para aprender la función cuyas muestras están en los archivos.
- Evaluar la capacidad de generalización del perceptrón simple no lineal utilizando, de los datos provistos, un subconjunto para entrenar y otro para testear.
- ¿Cómo elegirían el mejor conjunto de entrenamiento? ¿Qué efecto tiene la elección en la capacidad de generalización del perceptrón?

Perceptrones Lineales: Comparación de error entre entrenamiento y prueba

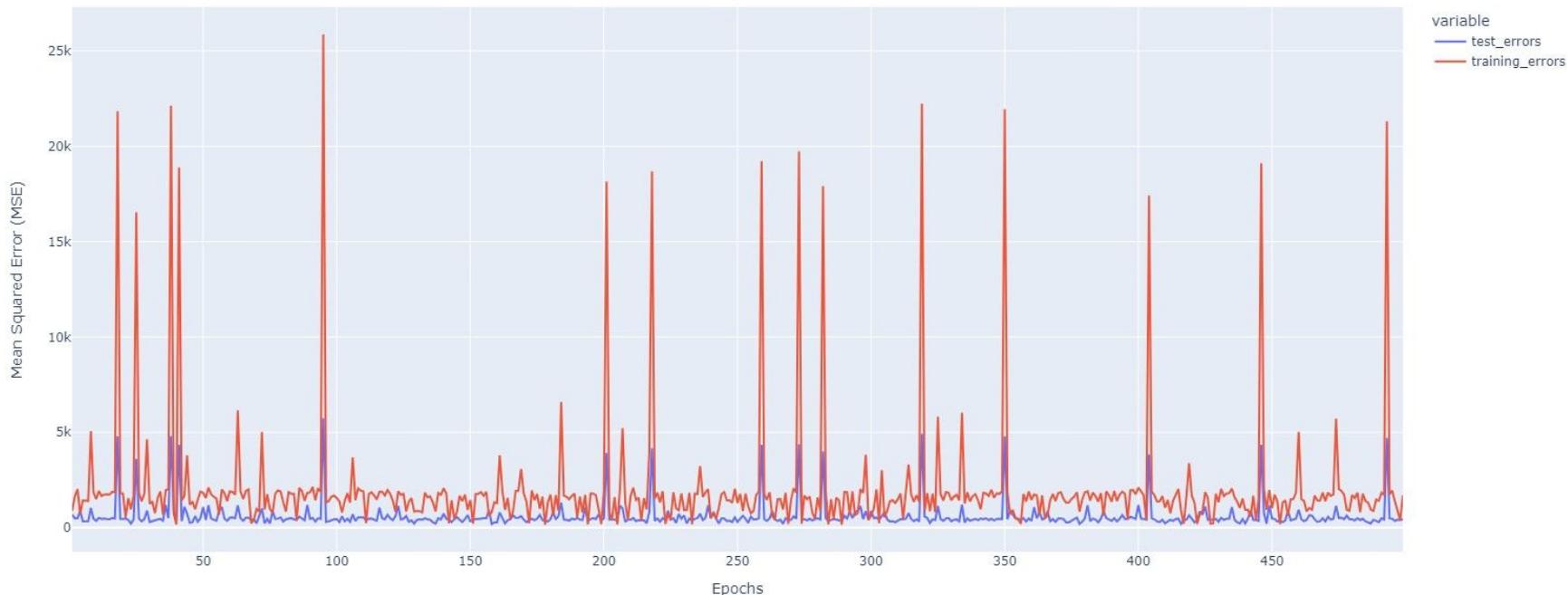
Perceptrón Lineal

Perceptrones Simples: Comparación de errores

```
dataset: TP3-ej2-conjunto.csv
activation_function = IDENTITY
learning_rate = 0.1
beta = 1
split_ratio = 0.8
```



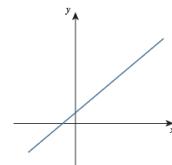
Testing and Training Mean Squared Error Comparison



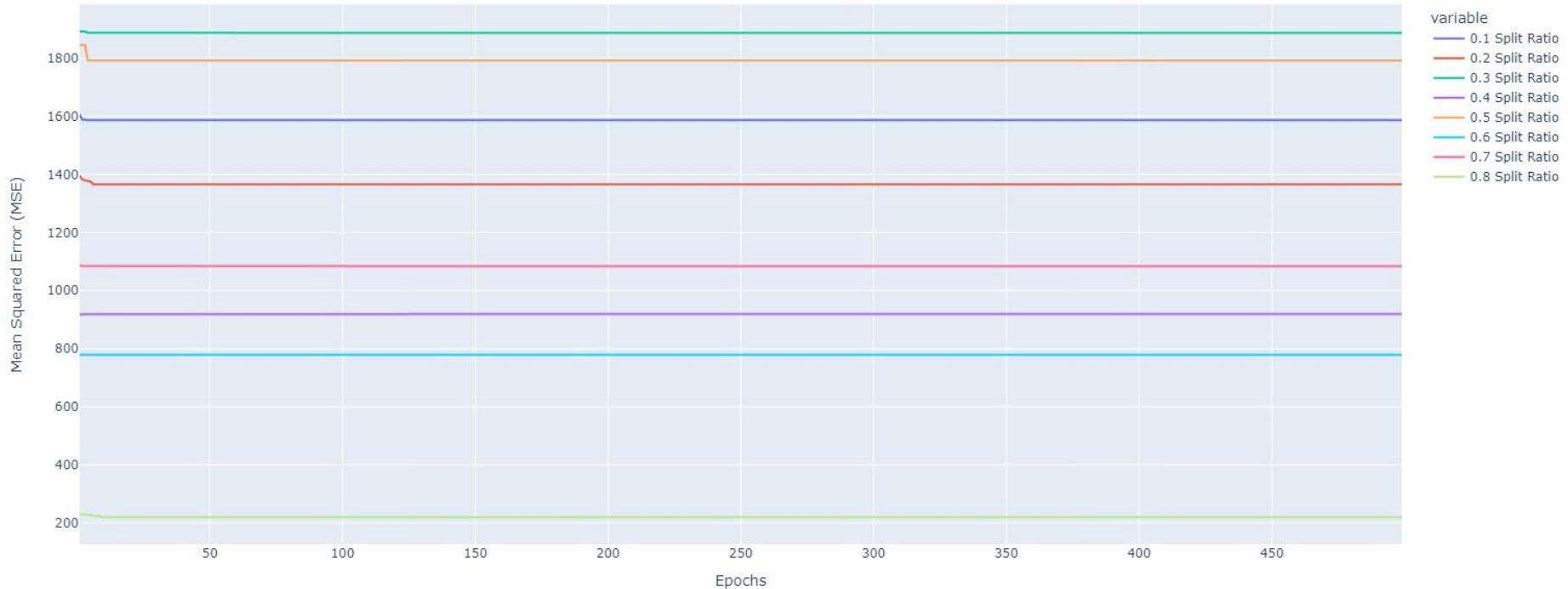
Perceptrón Lineal

Perceptrones Simples: Comparación con distintas proporciones

dataset: TP3-ej2-conjunto.csv
activation_function = IDENTITY
learning_rate = 0.1
beta = 1



Testing Mean Squared Error with Different Split Ratio Comparison

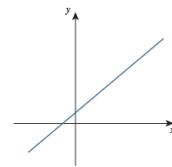


Perceptrones No Lineales: Comparación de error entre entrenamiento y prueba

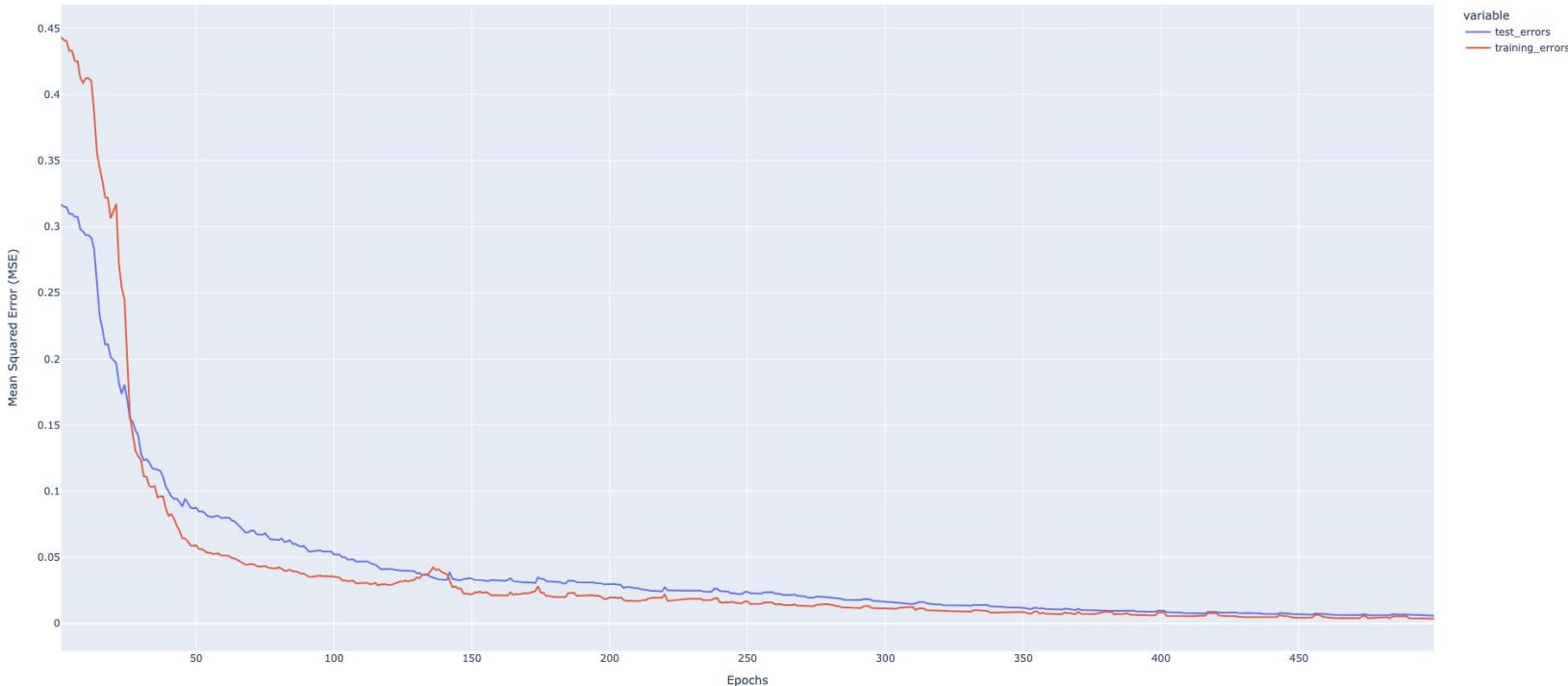
Perceptrón No Lineal

Perceptrones Simples: Comparación de errores

```
dataset: TP3-ej2-conjunto.csv
activation_function = SIGMOID
learning_rate = 0.1
beta = 1
split_ratio = 0.5
normalized (0, 1)
```



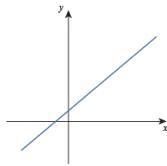
Testing and Training Mean Squared Error Comparison



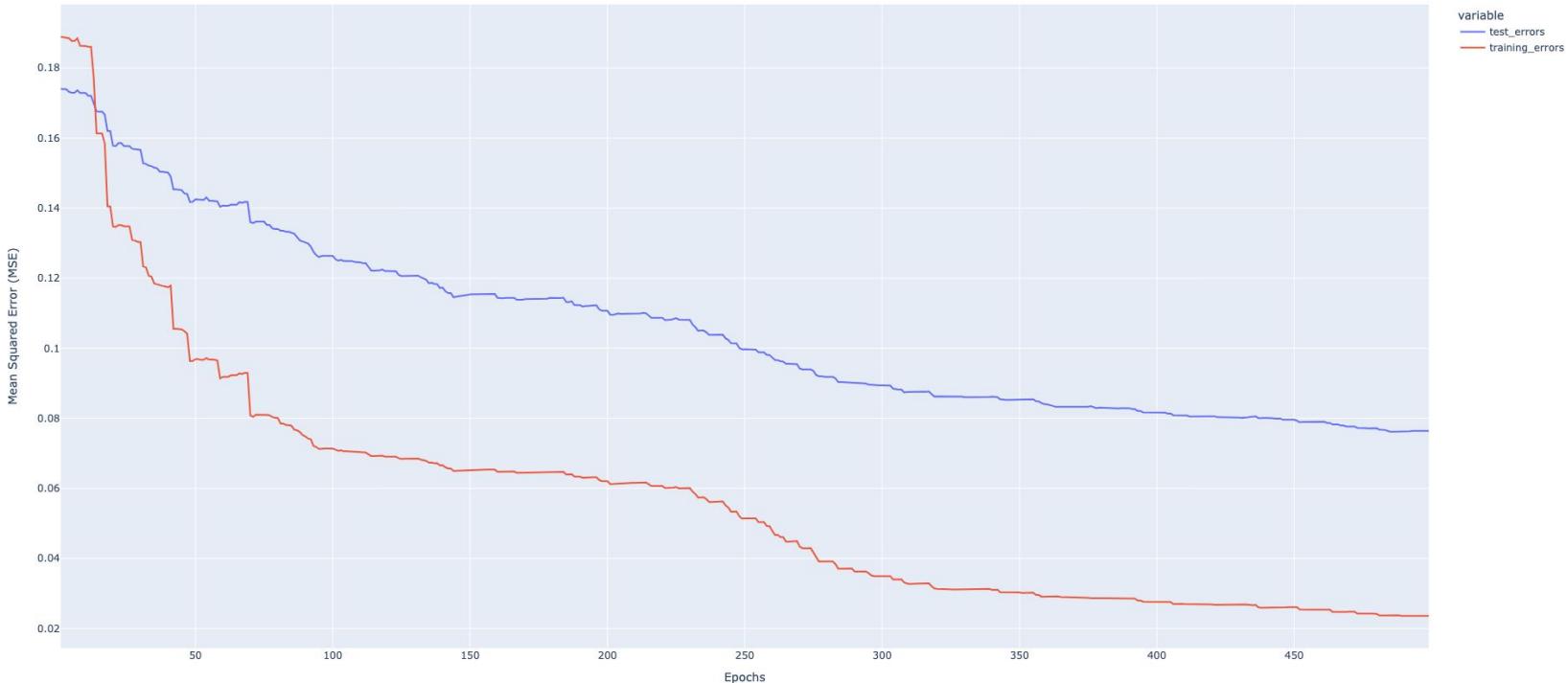
Perceptrón No Lineal

Perceptrones Simples: Comparación de errores

```
dataset: TP3-ej2-conjunto.csv
activation_function = SIGMOID
learning_rate = 0.1
beta = 1
split_ratio = 0.5
normalized (0, 1)
```



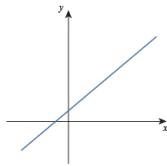
Testing and Training Mean Squared Error Comparison



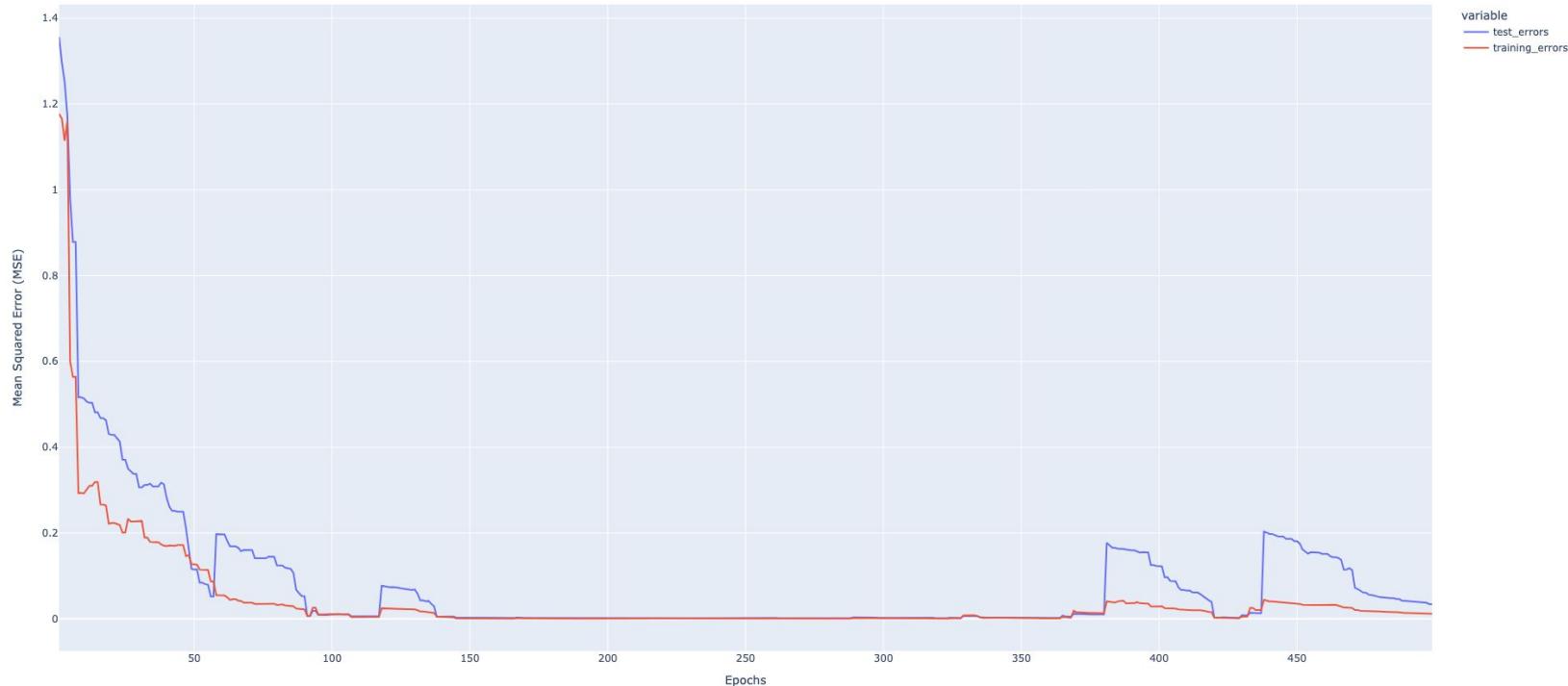
Perceptrón No Lineal

Perceptrones Simples: Comparación de errores

```
dataset: TP3-ej2-conjunto.csv
activation_function = TAN_H
learning_rate = 0.1
beta = 1
split_ratio = 0.5
normalized (-1, 1)
```



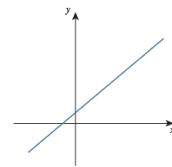
Testing and Training Mean Squared Error Comparison



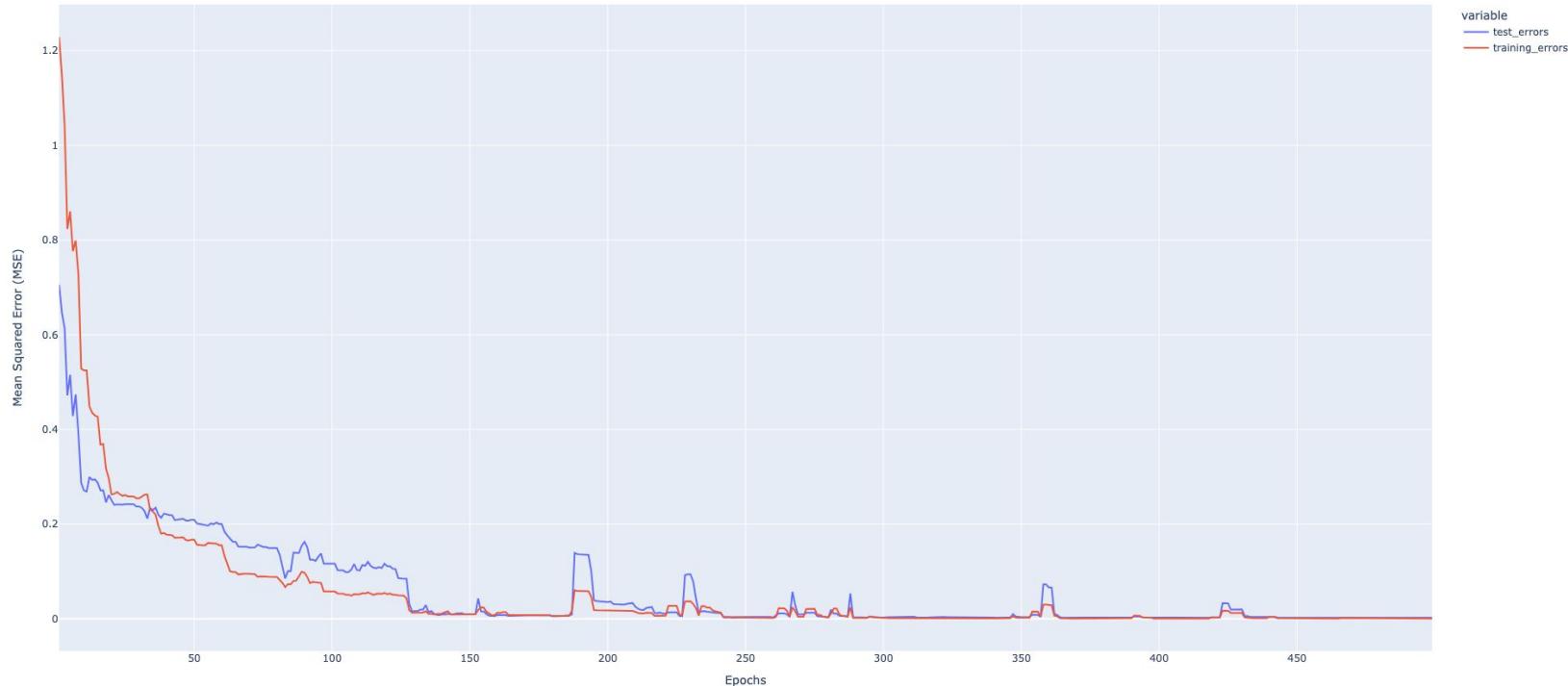
Perceptrón No Lineal

Perceptrones Simples: Comparación de errores

```
dataset: TP3-ej2-conjunto.csv
activation_function = TAN_H
learning_rate = 0.1
beta = 1
split_ratio = 0.5
normalized (-1, 1)
```



Testing and Training Mean Squared Error Comparison

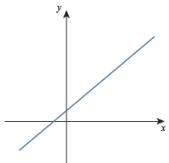


Perceptrones No Lineales: Comparación con distintas proporciones de partición

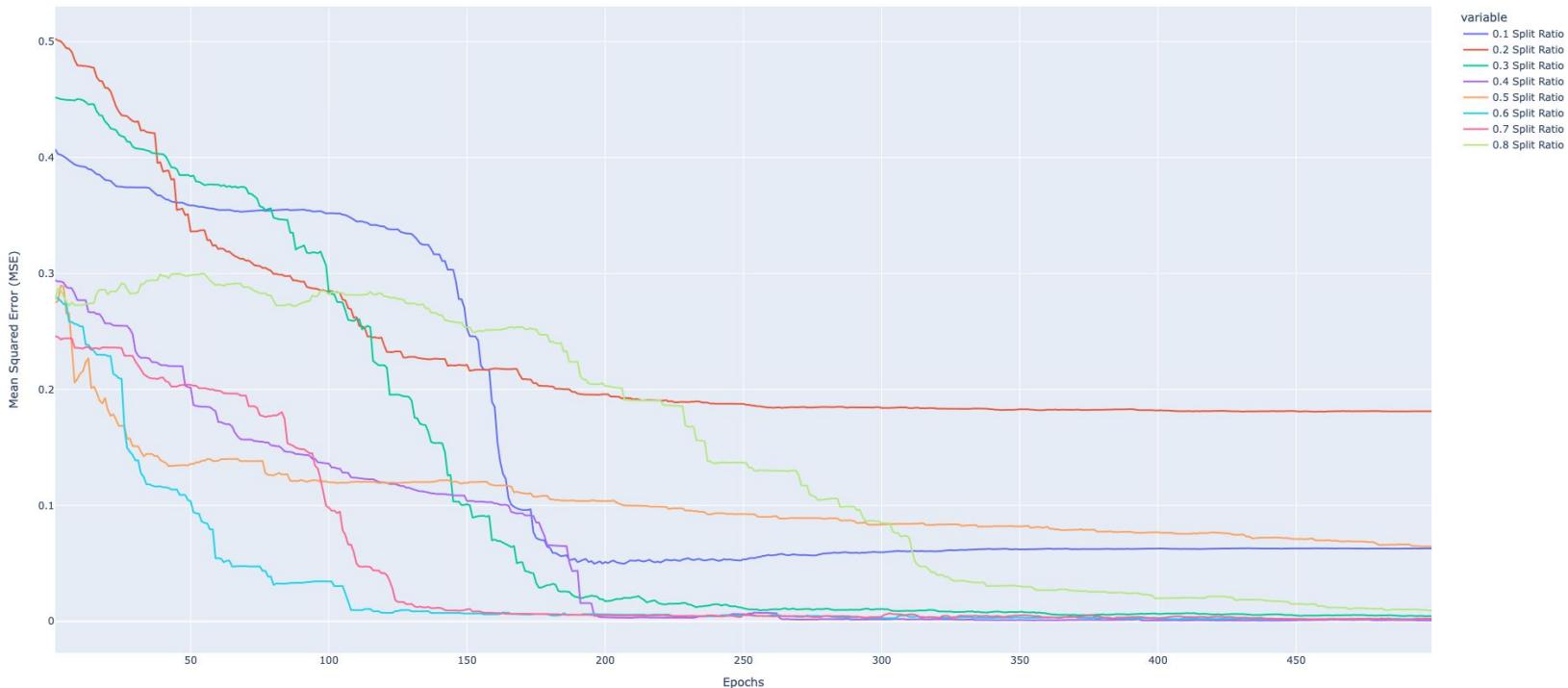
Perceptrón No Lineal

Perceptrones Simples: Comparación con distintas proporciones

```
dataset: TP3-ej2-conjunto.csv
activation_function = SIGMOID
learning_rate = 0.1
beta = 1
normalized (0, 1)
```



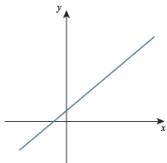
Testing Mean Squared Error with Different Split Ratio Comparison



Perceptrón No Lineal

Perceptrones Simples: Comparación con distintas proporciones

```
dataset: TP3-ej2-conjunto.csv
activation_function = SIGMOID
learning_rate = 0.1
beta = 1
normalized (0, 1)
```



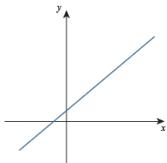
Testing Mean Squared Error with Different Split Ratio Comparison



Perceptrón No Lineal

Perceptrones Simples: Comparación con distintas proporciones

```
dataset: TP3-ej2-conjunto.csv
activation_function = SIGMOID
learning_rate = 0.1
beta = 1
normalized (0, 1)
```



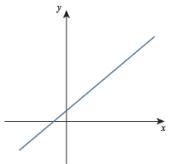
Testing Mean Squared Error with Different Split Ratio Comparison



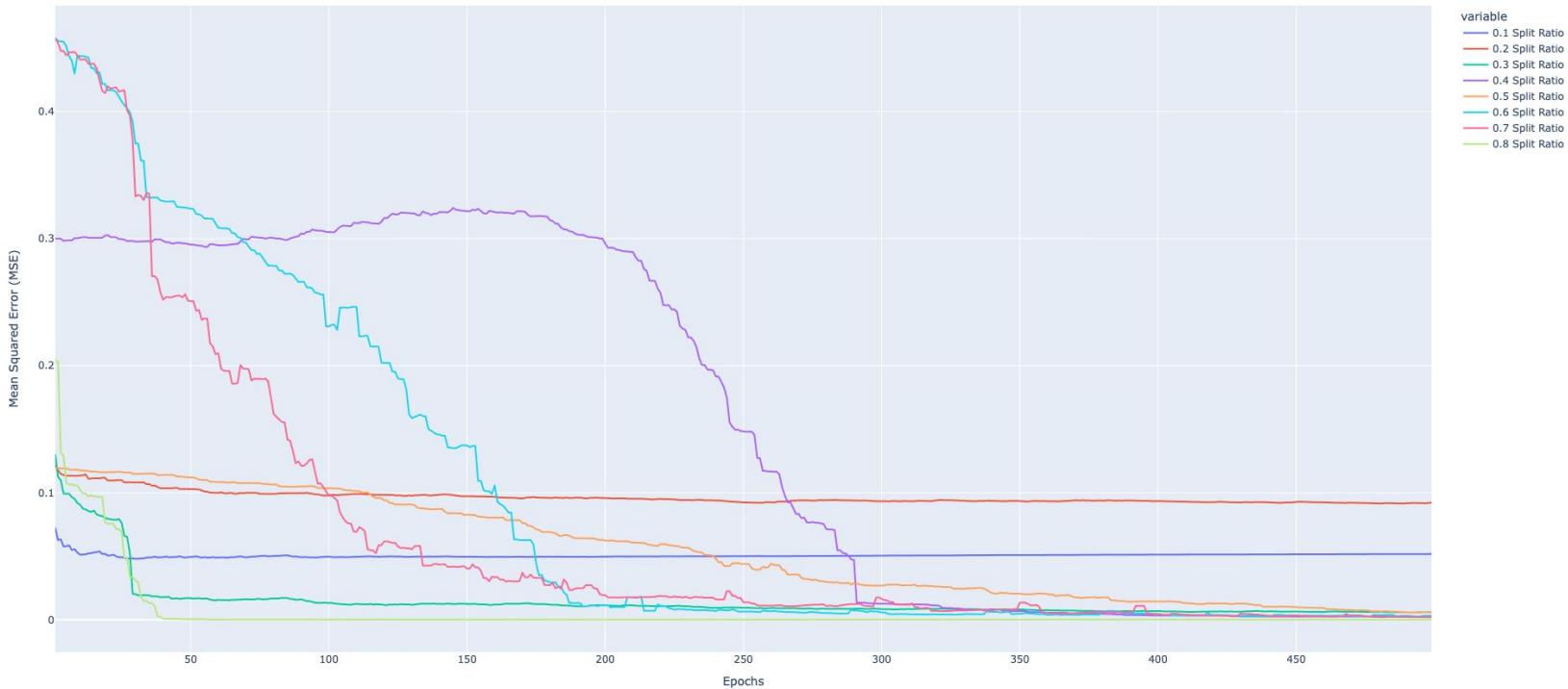
Perceptrón No Lineal

Perceptrones Simples: Comparación con distintas proporciones

```
dataset: TP3-ej2-conjunto.csv
activation_function = SIGMOID
learning_rate = 0.1
beta = 1
normalized (0, 1)
```



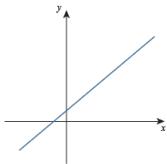
Testing Mean Squared Error with Different Split Ratio Comparison



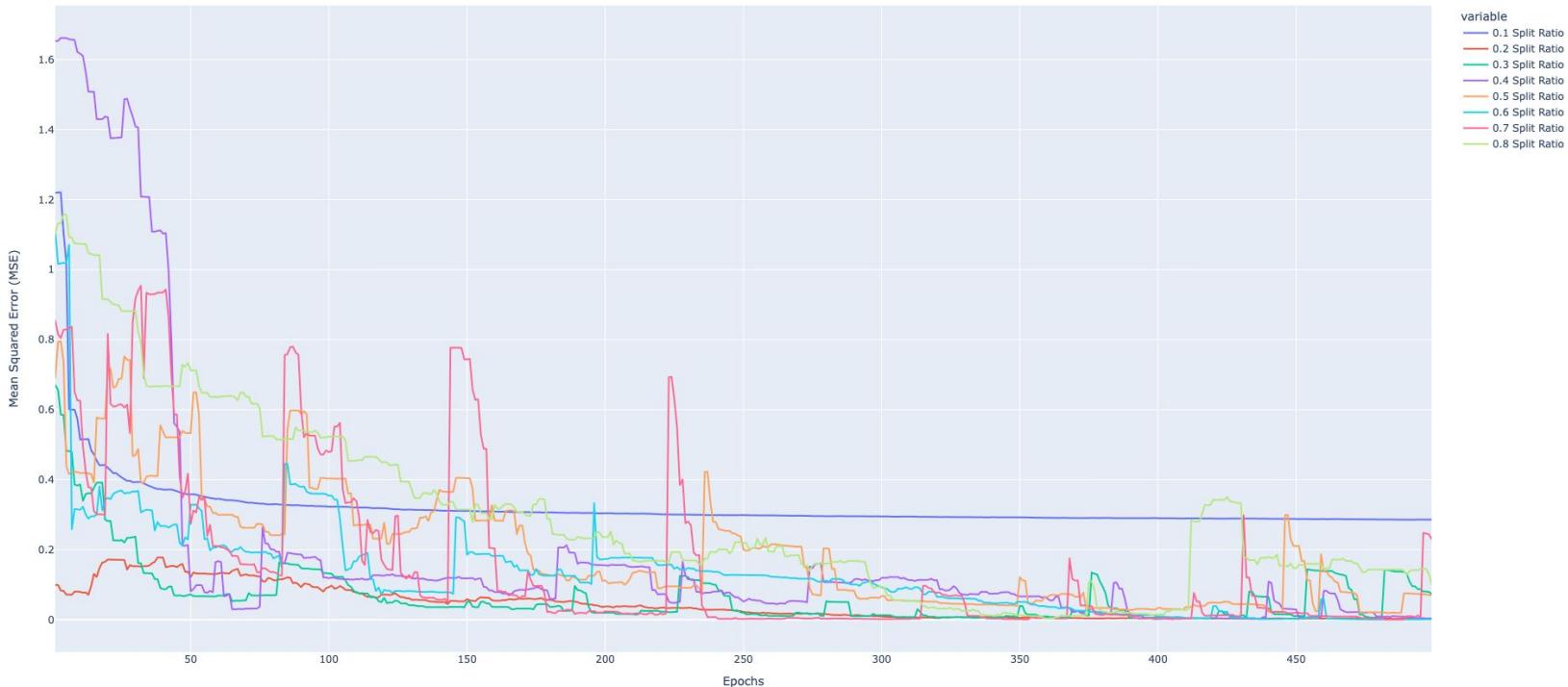
Perceptrón No Lineal

Perceptrones Simples: Comparación con distintas proporciones

```
dataset: TP3-ej2-conjunto.csv
activation_function = TAN_H
learning_rate = 0.1
beta = 1
normalized (-1, 1)
```



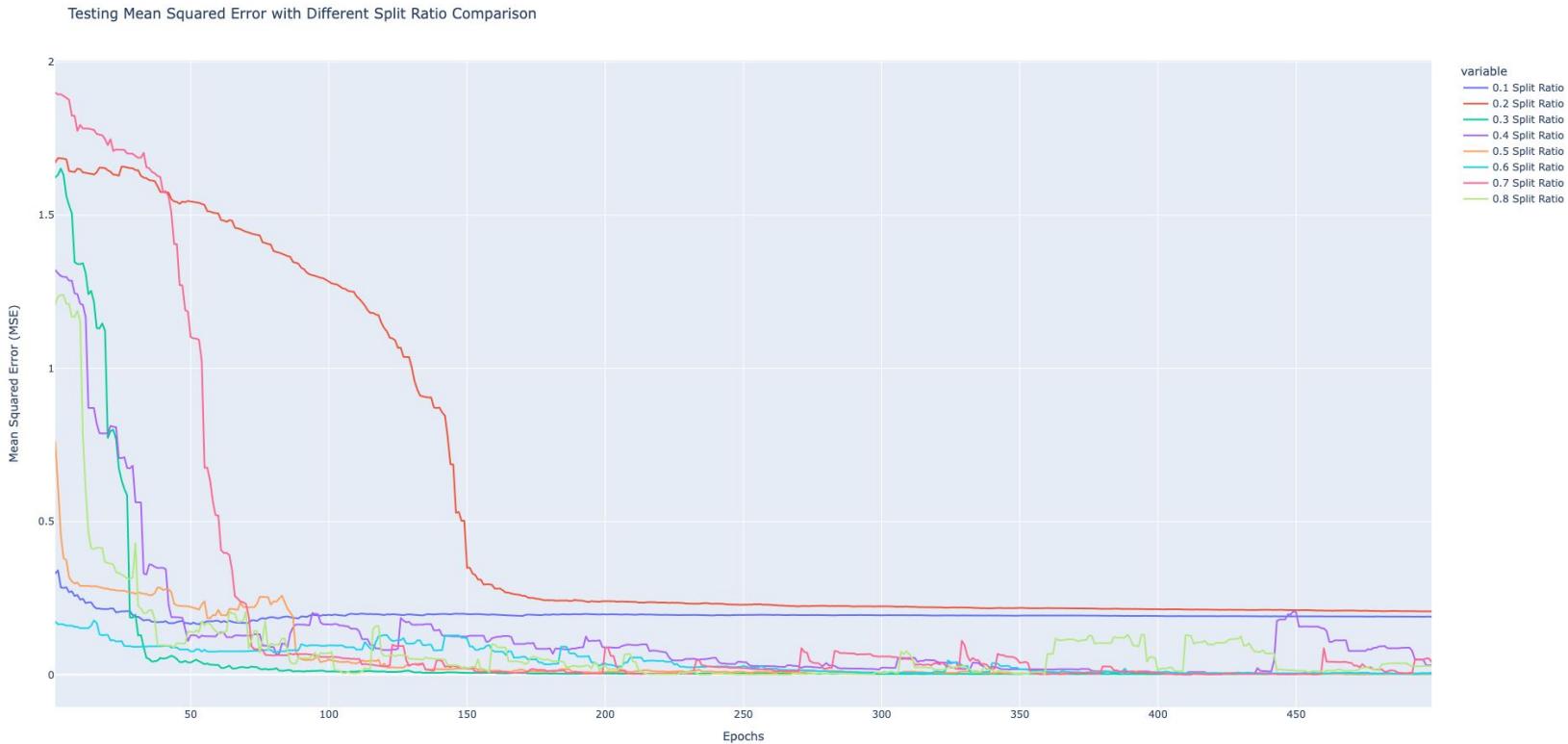
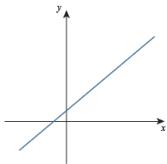
Testing Mean Squared Error with Different Split Ratio Comparison



Perceptrón No Lineal

Perceptrones Simples: Comparación con distintas proporciones

```
dataset: TP3-ej2-conjunto.csv
activation_function = TAN_H
learning_rate = 0.1
beta = 1
normalized (-1, 1)
```



Ejercicio 3

? Preguntas:

- ¿Qué problemas puede resolver el perceptrón multicapa que no puede resolver el perceptrón simple?
- ¿Cómo influye el ruido del conjunto de entrenamiento en los resultados finales?

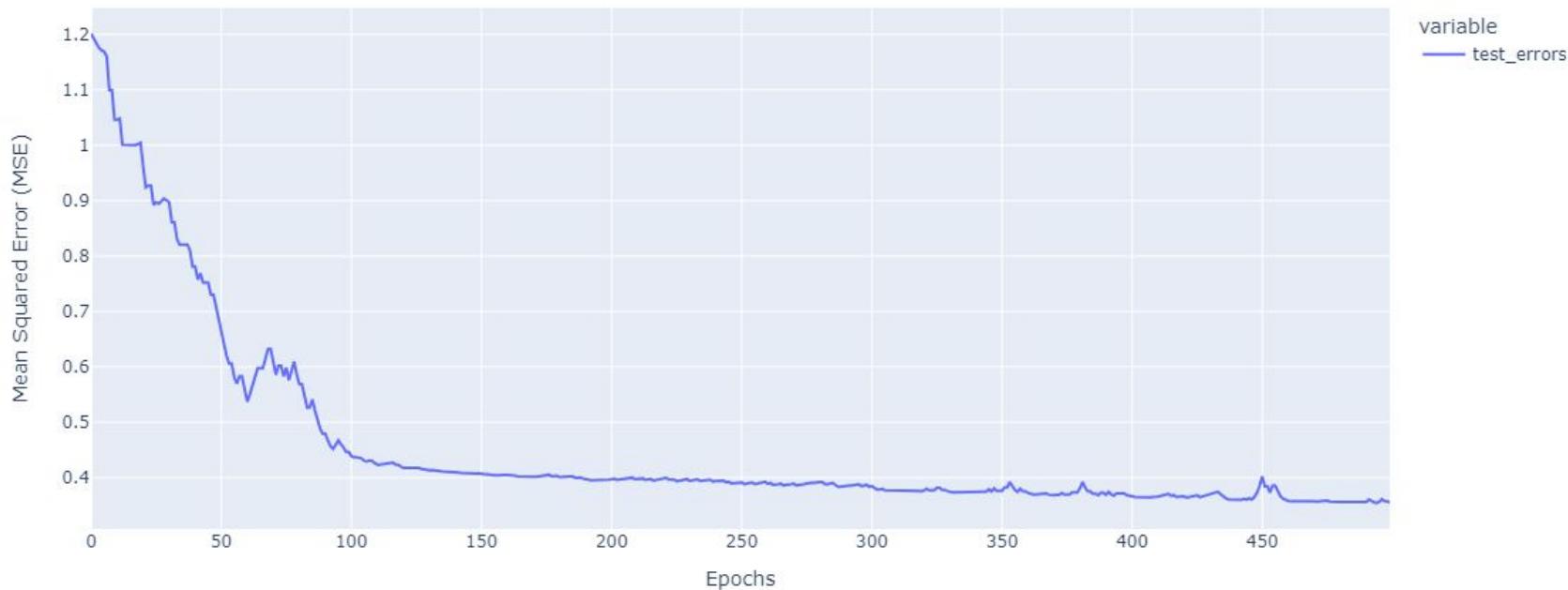
Perceptrón Multicapa: Ejercicio 1

Perceptrón Multicapa: XOR

Ejercicio 1

Testing and Training Mean Squared Error Comparison

```
dataset: XOR
activation_function = TAN_H
learning_rate = 0.1
beta = 1
NArchitecture = [2,2,2,1]
```



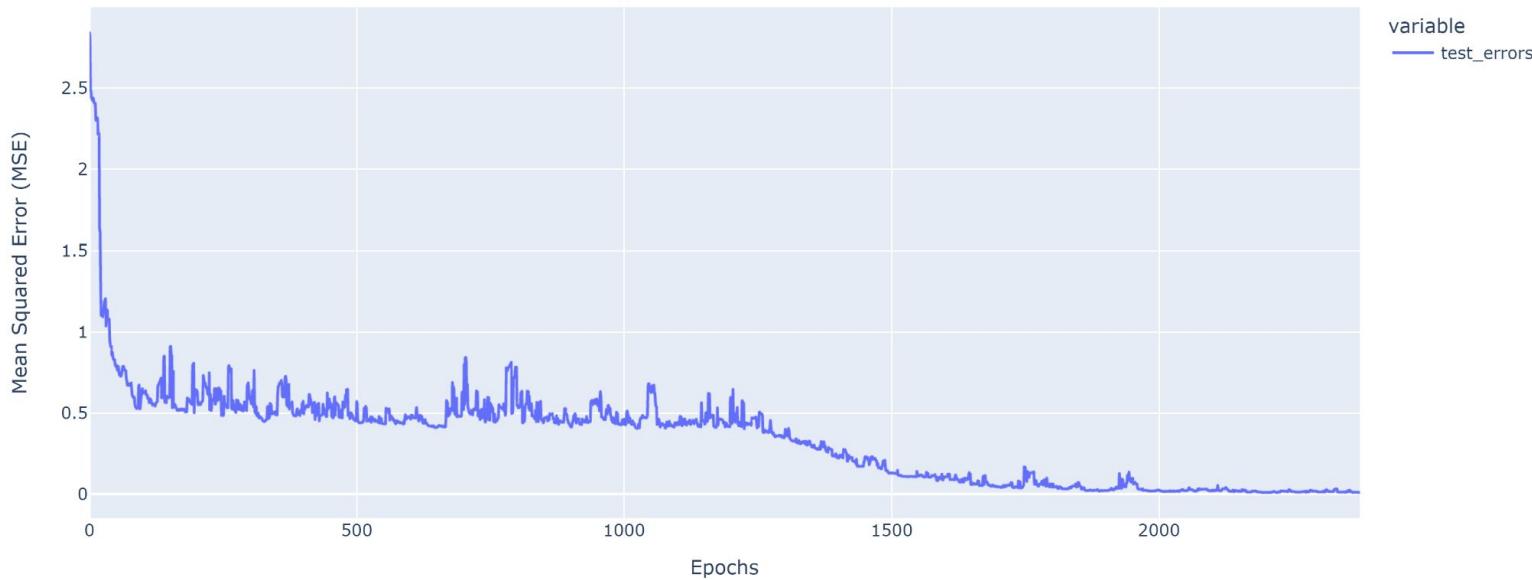
Perceptrón Multicapa: Pares

Perceptrón Multicapa: Pares

Ejercicio 2

Testing Mean Squared Error Comparison

```
dataset: TP3-ej3-digitos.txt
activation_function = TAN_H
learning_rate = 0.1
beta = 1
NArchitecture = [35,2,2,1]
```

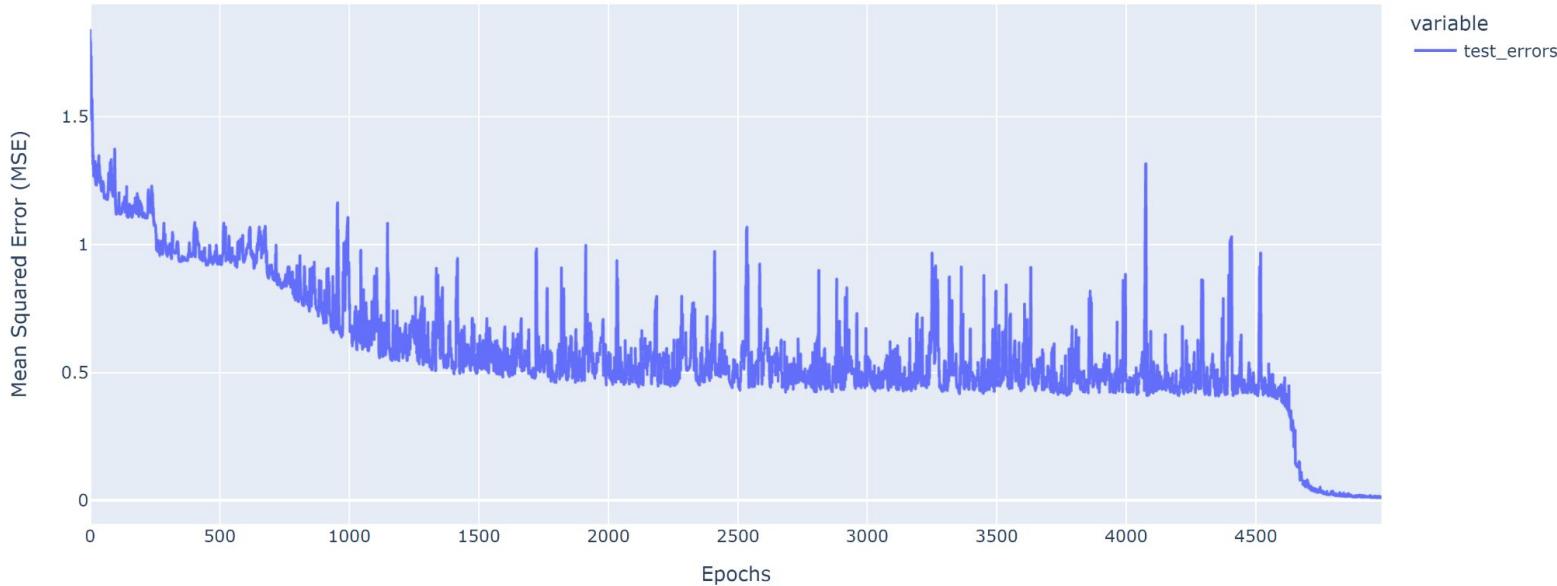


Perceptrón Multicapa: Pares (múltiples capas)

Ejercicio 2

Testing Mean Squared Error Comparison

```
dataset: TP3-ej3-digitos.txt
activation_function = TAN_H
learning_rate = 0.1
beta = 1
NArchitecture = [35,2,2,2,2,2,2,2,1]
```

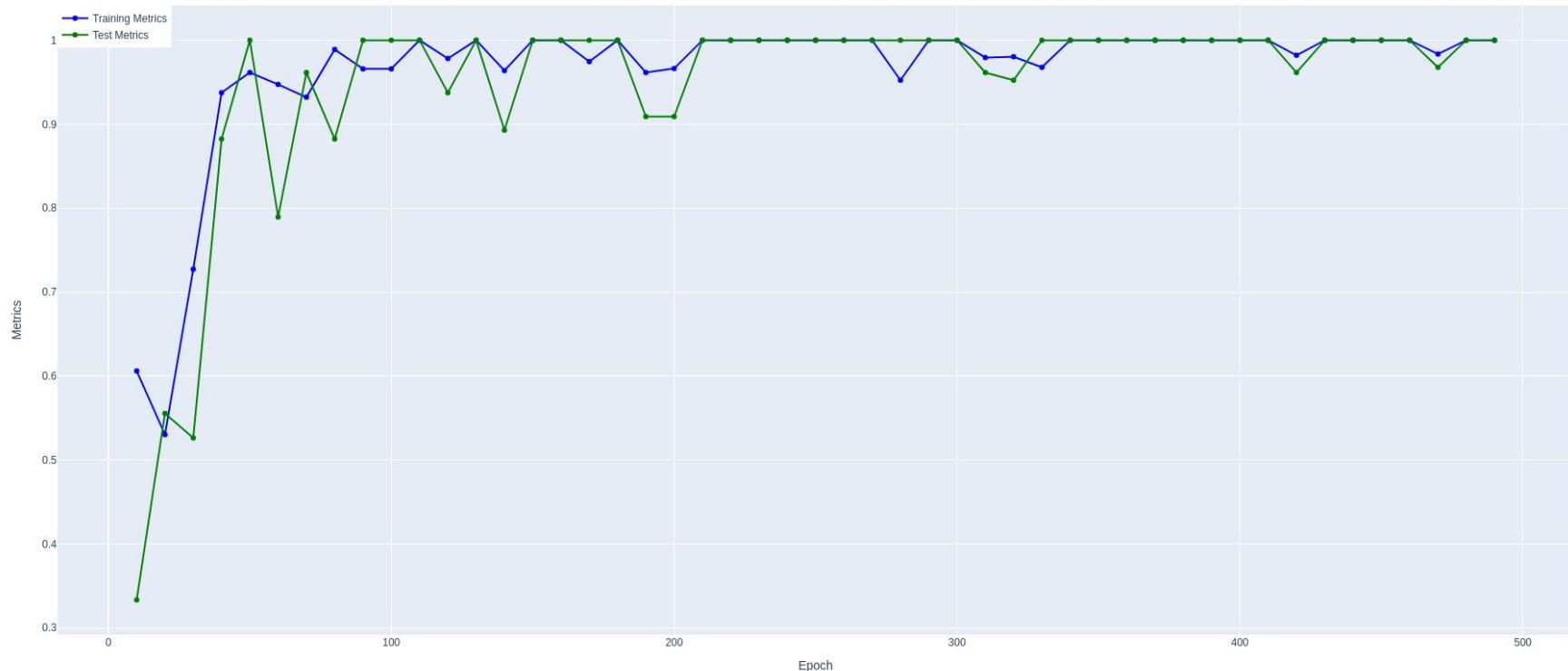


Perceptrón Multicapa: Reconocimiento de números

Perceptrón Multicapa: Reconocimiento de números

Accuracy

Training and Test Metrics Over Epochs (Accuracy)



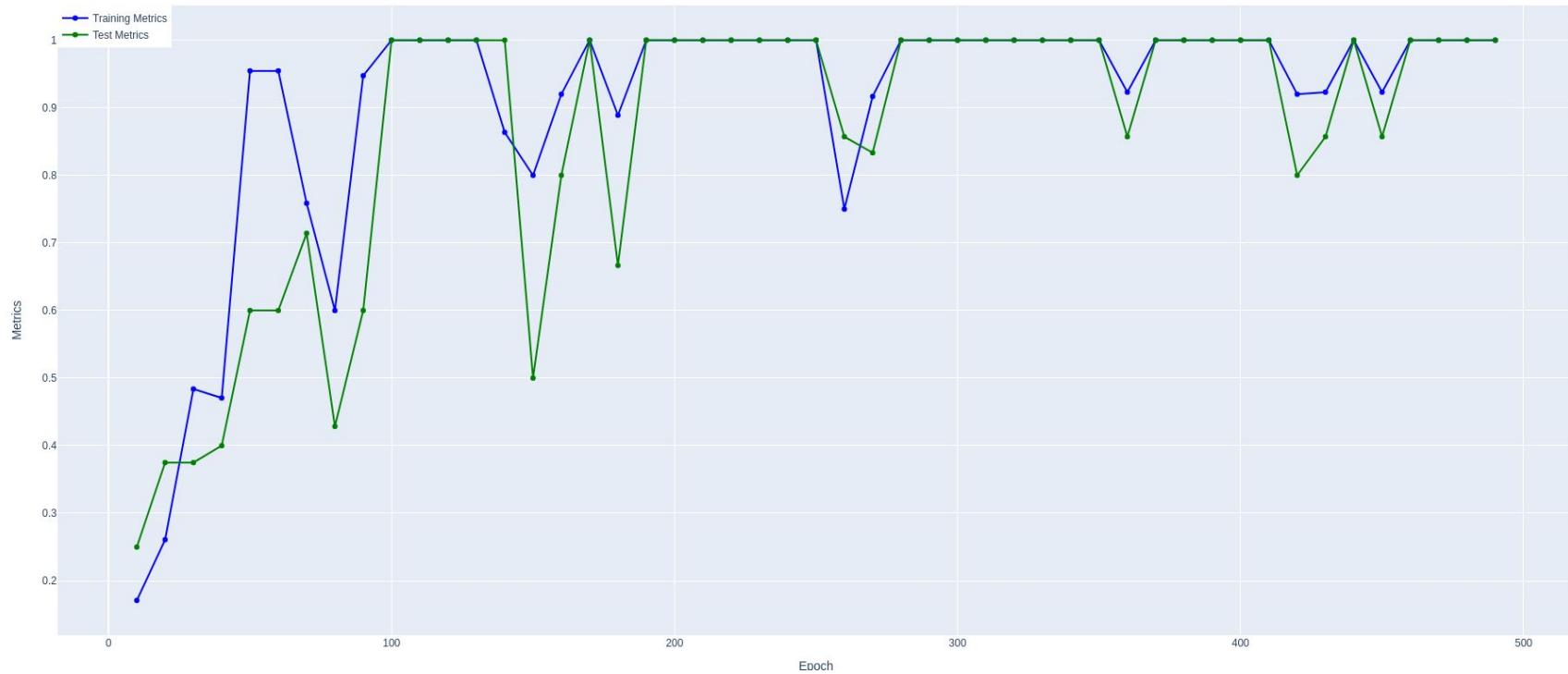
dataset: TP3-ej3-digits.txt
activation_function = TAN_H
learning_rate = 0.2
beta = 1
noise = 0.2
normalized (-1, 1)
architecture = [35, 10, 10]

Perceptrón Multicapa: Reconocimiento de números

F1

```
dataset: TP3-ej3-digits.txt
activation_function = TAN_H
learning_rate = 0.2
beta = 1
noise = 0.2
normalized (-1, 1)
architecture = [35, 10, 10]
```

Training and Test Metrics Over Epochs (F1)

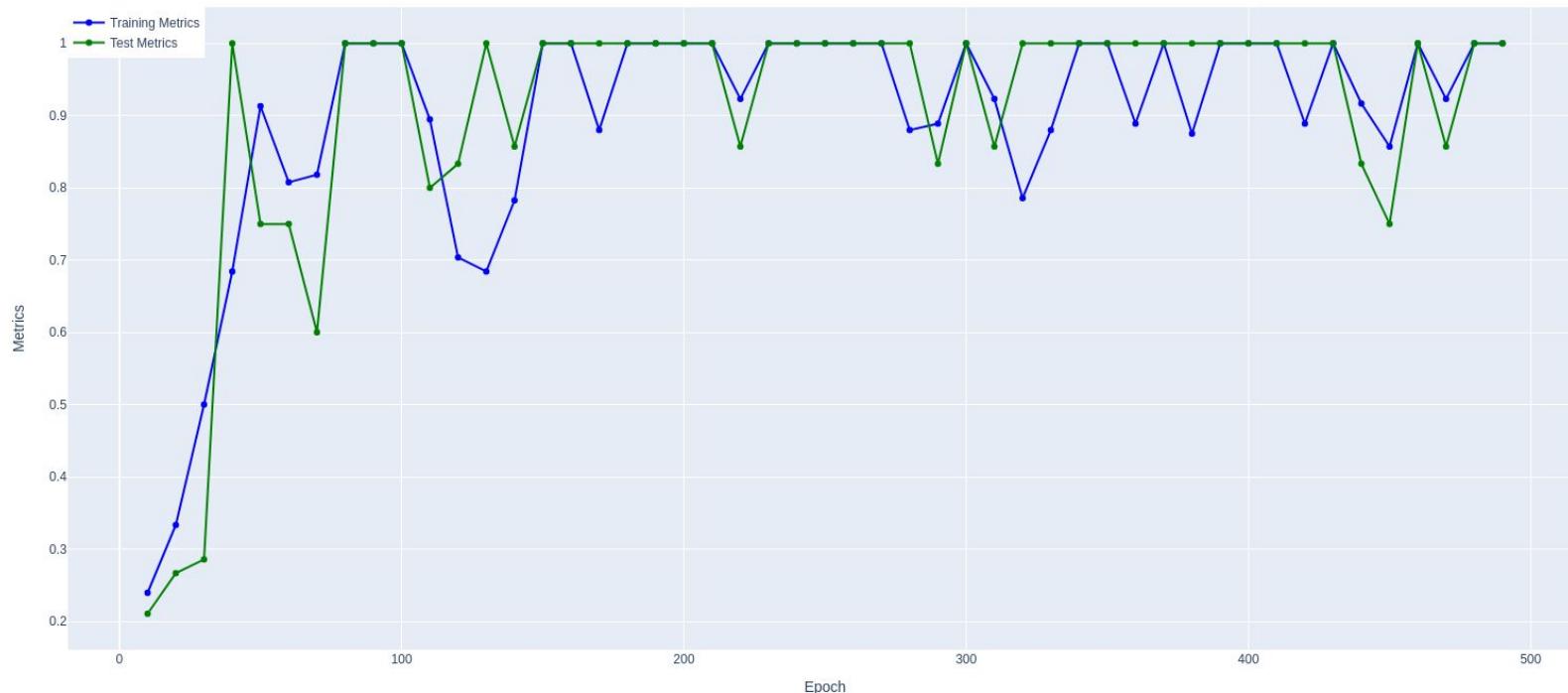


Perceptrón Multicapa: Reconocimiento de números

Precision

```
dataset: TP3-ej3-digits.txt
activation_function = TAN_H
learning_rate = 0.2
beta = 1
noise = 0.2
normalized (-1, 1)
architecture = [35, 10, 10]
```

Training and Test Metrics Over Epochs (Precision)



Perceptrón Multicapa: Reconocimiento de números

Recall

```
dataset: TP3-ej3-digits.txt
activation_function = TAN_H
learning_rate = 0.2
beta = 1
noise = 0.2
normalized (-1, 1)
architecture = [35, 10, 10]
```

Training and Test Metrics Over Epochs (Recall)

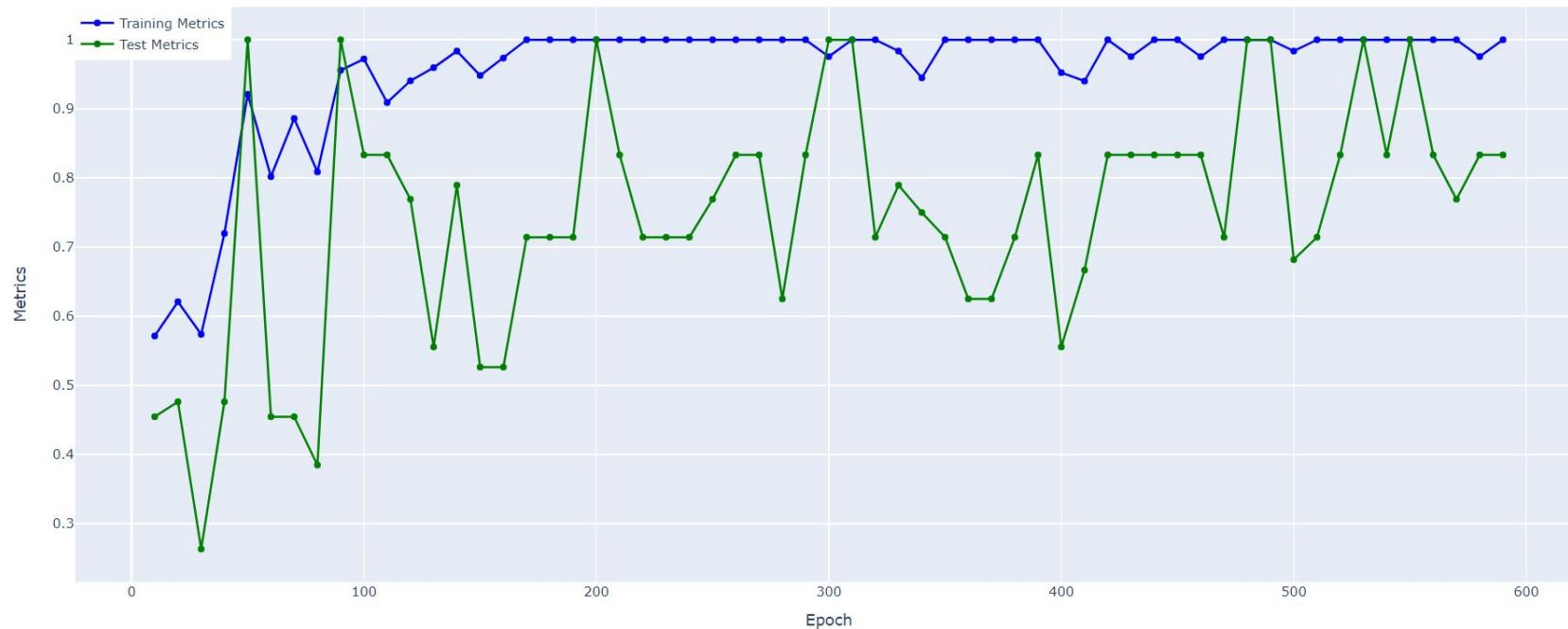


Perceptrón Multicapa: Reconocimiento de números

Accuracy

```
dataset: TP3-ej3-digits.txt
activation_function = TAN_H
learning_rate = 0.2
beta = 1
noise = 0.85
normalized (-1, 1)
architecture = [35, 10, 10]
```

Training and Test Metrics Over Epochs (Accuracy)



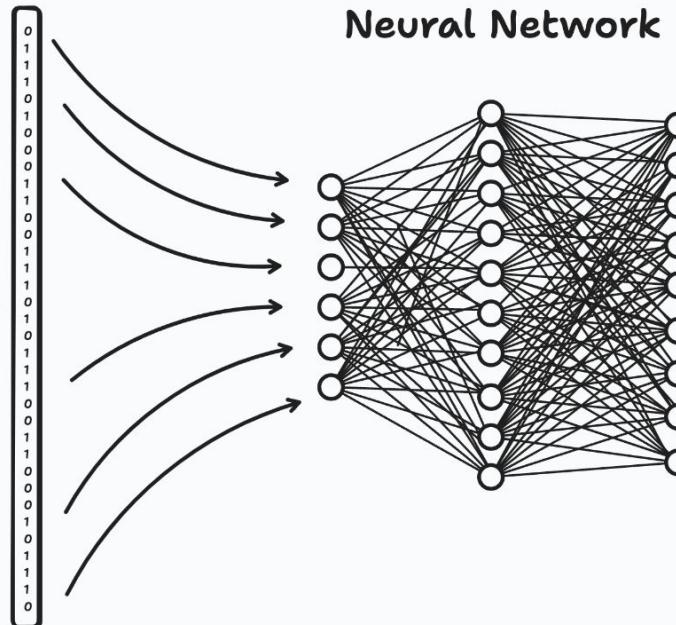
One more thing...



Number dataset

0	1	1	1	0
1	0	0	0	1
1	0	0	1	1
1	0	1	0	1
1	1	0	0	1
1	0	0	0	1
0	1	1	1	0

→ Flatten →



Neural Network

...Training

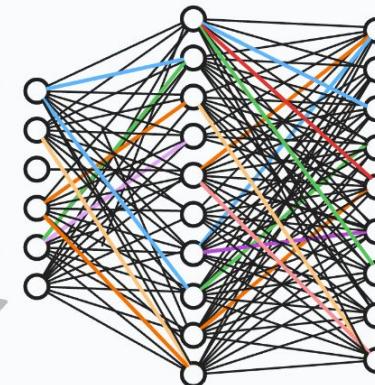
Number dataset

0	1	1	1	0
1	0	0	0	1
1	0	0	1	1
1	0	1	0	1
1	1	0	0	1
1	0	0	0	1
0	1	1	1	0

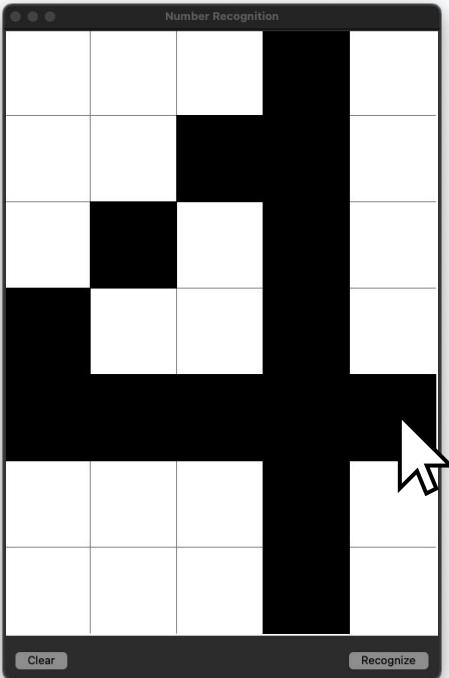
→ Flatten →



Neural Network

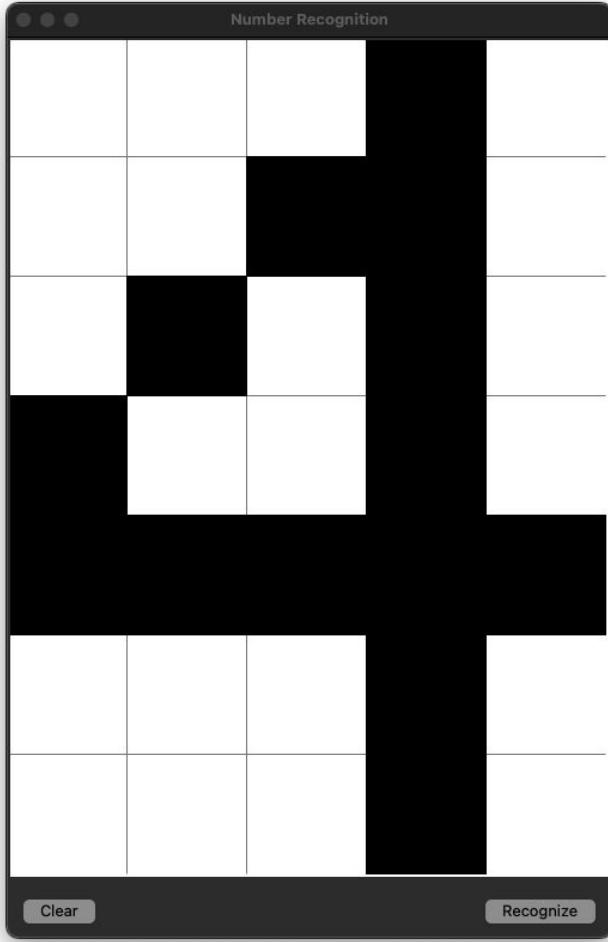


Trained 



Neural Network

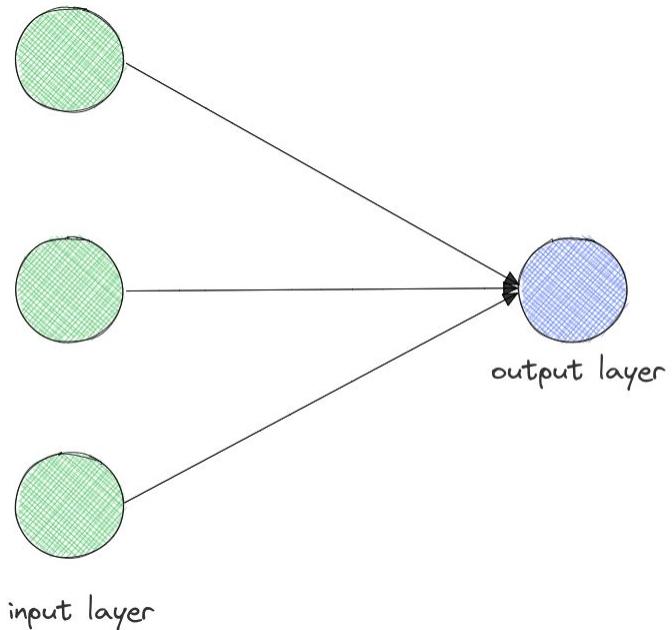


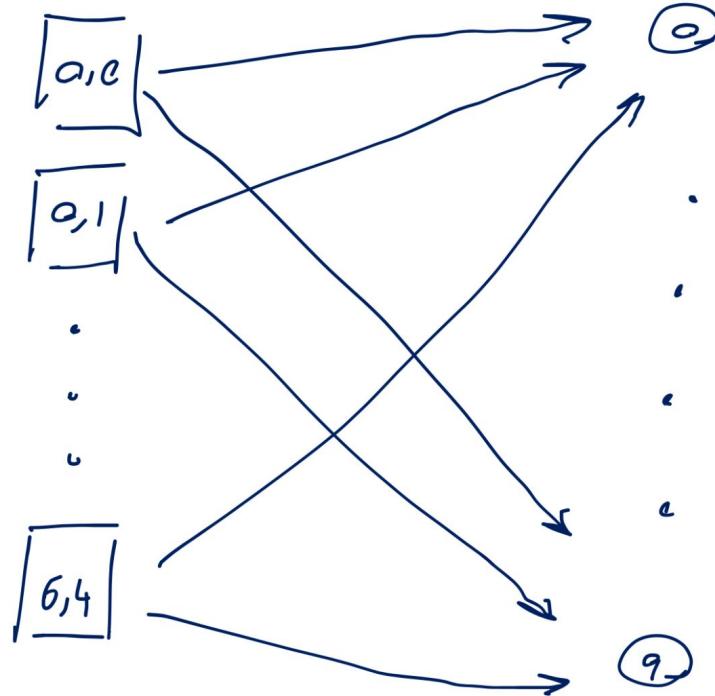
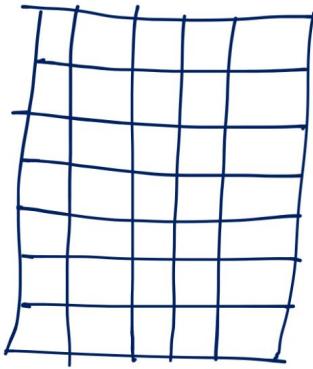


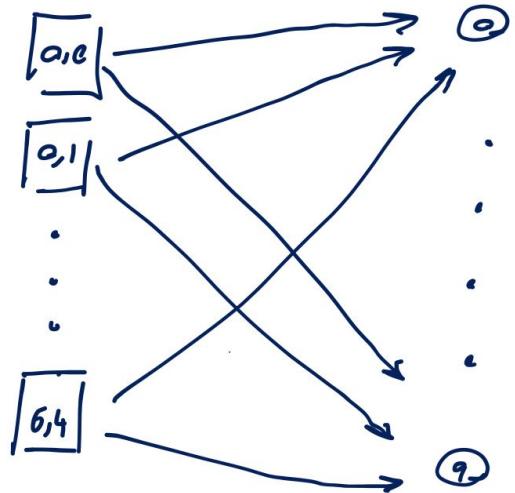
pipenv run python3 cli.py

```
~/Projects/sia/tp3 git:(main) ±16
pipenv run python3 cli.py
[0 1 1 1 0 1 0 0 0 1 1 0 0 1 1 1 0 1 0 1 1 1 0 0 1 1 0 0
 0 1 0 1 1 1 0]
Match: 0
Candidate: 3
Candidate: 6
[0 1 1 1 0 1 0 0 0 1 0 0 0 0 1 0 0 1 1 0 0 0 0 0 1 1 0 0
 0 1 0 1 1 1 0]
Match: 3
Candidate: 8
Candidate: 2
[0 1 1 1 0 1 0 0 0 1 1 0 0 0 1 0 1 1 1 1 0 0 0 0 1 0 0 0
 1 0 0 1 1 0 0]
Match: 9
Candidate: 6
Candidate: 3
[1 1 1 1 1 0 0 0 0 1 1 1 1 0 0 0 0 0 1 0 0 0 0 1 1 0 0
 0 1 0 1 1 1 0]
Match: 5
Candidate: 3
Candidate: 6
[0 1 1 1 0 1 0 0 0 1 0 0 0 0 1 0 0 1 1 0 0 0 0 0 1 1 0 0
 0 1 0 1 1 1 0]
Match: 3
Candidate: 8
Candidate: 2
[0 1 1 1 0 1 0 0 0 1 1 0 0 0 1 0 1 1 1 1 0 0 0 0 1 0 0 0
 1 0 0 1 1 0 0]
Match: 9
Candidate: 6
Candidate: 3
[0 0 0 1 0 0 0 1 1 0 0 1 0 1 0 1 0 0 1 0 1 1 1 1 0 0 0 0
 1 0 0 0 0 1 0]
Match: 4
Candidate: 8
Candidate: 1
```

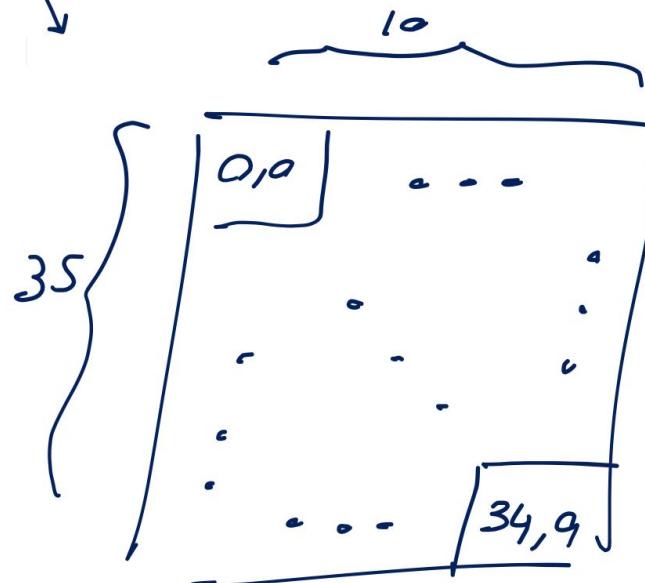
Red neuronal sin hidden layers





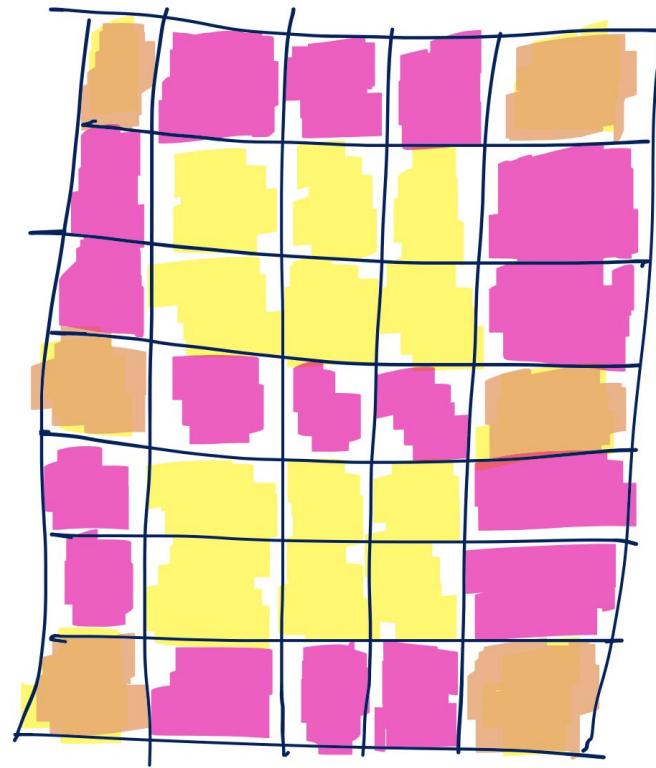


3500 weights
350 per neurone

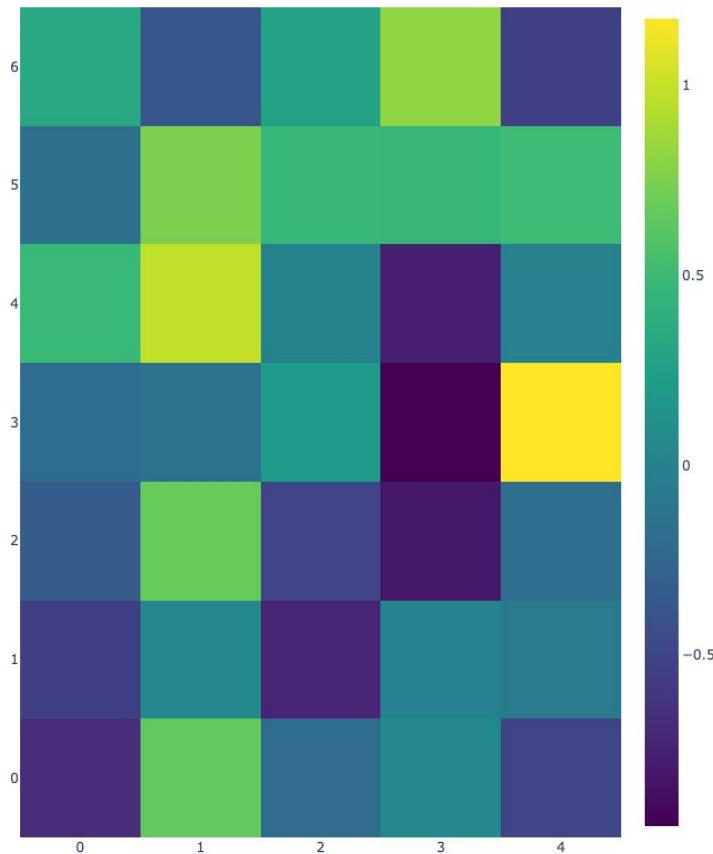


Y si dibujamos los pesos relacionados con cada neurona?

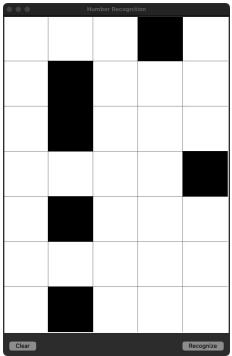
Expectativa



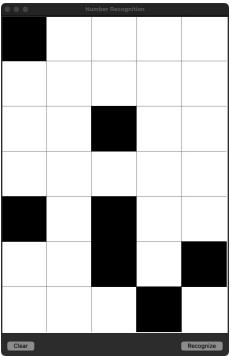
Realidad



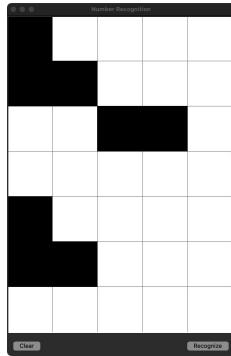
0



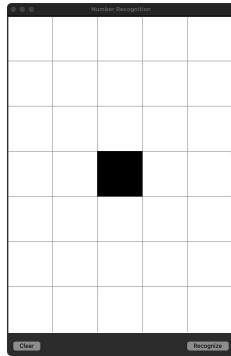
1



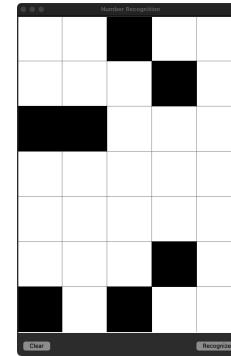
2



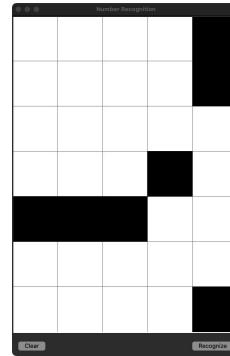
3



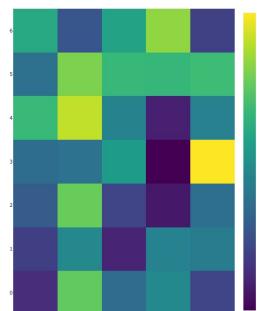
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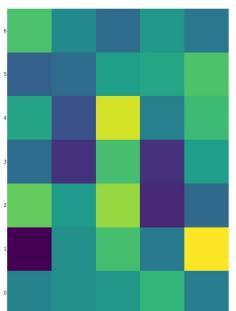
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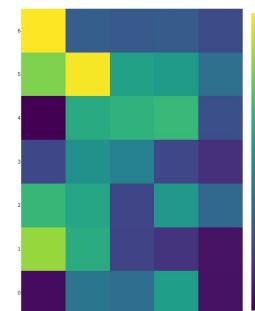
Heatmap Example Digit 0



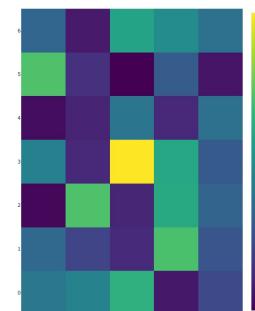
Heatmap Example Digit 1



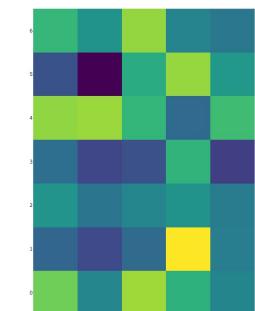
Heatmap Example Digit 2



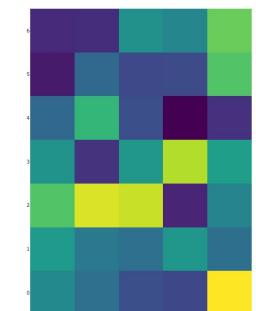
Heatmap Example Digit 3



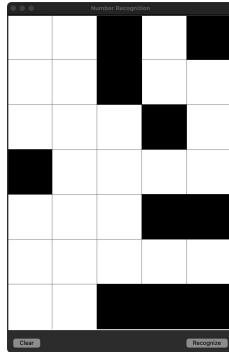
Heatmap Example Digit 4



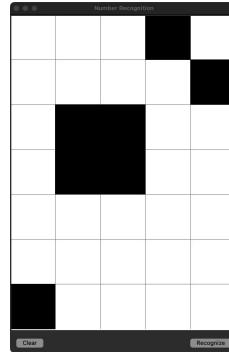
Heatmap Example Digit 5



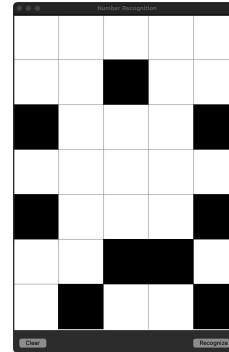
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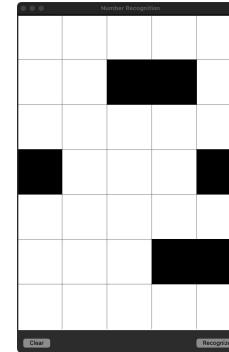
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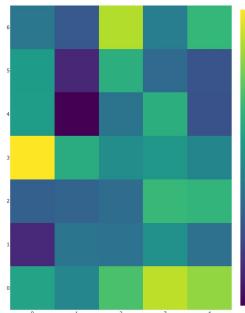
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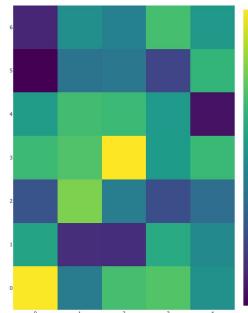
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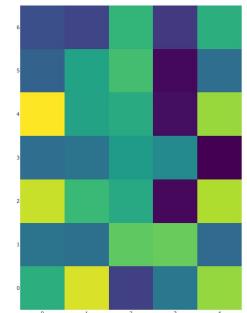
Heatmap Example Digit 6



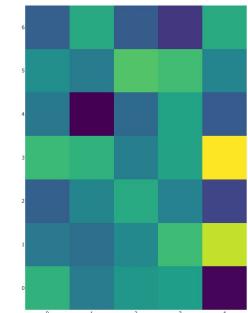
Heatmap Example Digit 7



Heatmap Example Digit 8



Heatmap Example Digit 9



¡Muchas gracias!

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