

Ejercicio 2

January 31, 2021

1 Importamos las librerías necesarias.

```
[1]: from scipy import stats
import numpy as np
import pandas as pd
import math
import matplotlib.pyplot as plt
from random import seed
```

```
[2]: import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
from tensorflow.keras.utils import to_categorical
from sklearn.preprocessing import LabelBinarizer
```

```
[3]: from tensorflow.keras.layers import Input, Dense, Dropout
from tensorflow.keras.optimizers import SGD
from tensorflow.keras.models import Model
from tensorflow.keras.models import load_model
from tensorflow.keras.callbacks import ModelCheckpoint
```

2 Definimos los modelos 'model_et' y model_peso correspondientes a los operadores '&' y '\$'.

```
[4]: model_et = keras.Sequential([
    keras.layers.Dense(64, activation='relu', input_shape=[2]),
    keras.layers.Dense(1)
])
model_et.compile(loss='mean_squared_error', optimizer='adam', metrics=['mae', 'mse'])
model_et.summary()

model_peso = keras.Sequential([
```

```

keras.layers.Dense(64, activation='relu', input_shape=[2]),
keras.layers.Dense(1)
])
model_peso.compile(loss='mean_squared_error', optimizer='adam', metrics=['mae',
↪ 'mse'])
model_peso.summary()

```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 64)	192
dense_1 (Dense)	(None, 1)	65

Total params: 257
 Trainable params: 257
 Non-trainable params: 0

Model: "sequential_1"

Layer (type)	Output Shape	Param #
dense_2 (Dense)	(None, 64)	192
dense_3 (Dense)	(None, 1)	65

Total params: 257
 Trainable params: 257
 Non-trainable params: 0

3 Definimos nuestros datos correspondientes a los operadores ‘&’ y ‘\$’.

```

[5]: x_train = np.array([
      [0,0],
      [1,0],
      [0,1],
      [1,1]
    ])

y_train_et = np.array([1,1,1,0])

y_train_peso = np.array([0,0,0,1])

```

4 Realizamos el entrenamiento de los modelos con los datos correspondientes.

```
[6]: model_et.fit(x=x_train, y=y_train_et, batch_size=5, epochs=5000, verbose=0)
```

```
[6]: <tensorflow.python.keras.callbacks.History at 0x1ae2fee3dc8>
```

```
[7]: model_peso.fit(x=x_train, y=y_train_peso, batch_size=5, epochs=5000, verbose=0)
```

```
[7]: <tensorflow.python.keras.callbacks.History at 0x1ae311a1408>
```

5 Verificamos que el modelo funcione bien en los datos originales.

```
[8]: y_pred_et = model_et.predict(np.array([[0,0],[1,0],[0,1],[1,1]]))
y_pred_et
```

```
[8]: array([[1.0000001e+00],
          [1.0000000e+00],
          [9.9999994e-01],
          [1.4901161e-08]], dtype=float32)
```

```
[9]: y_pred_peso = model_peso.predict(np.array([[0,0],[1,0],[0,1],[1,1]]))
y_pred_peso
```

```
[9]: array([[ 0.0000000e+00],
          [-1.8626451e-09],
          [-5.5879354e-09],
          [ 1.0000001e+00]], dtype=float32)
```

6 Finalmente, realizamos las operaciones que se piden en el problema.

```
[10]: a = [1.001,0,0.001,1]
b = [0,1,0,1]
c = [0,1,1,0]

a_et_b = []
for i in range(4):
    a_et_b.append([a[i],b[i]])
a_et_b
```

```
[10]: [[1.001, 0], [0, 1], [0.001, 0], [1, 1]]
```

```
[11]: a_et_b = model_et.predict(np.array(a_et_b))  
a_et_b
```

```
[11]: array([[1.0000385e+00],  
            [9.9999994e-01],  
            [1.0001057e+00],  
            [1.4901161e-08]], dtype=float32)
```

```
[12]: a_et_b_peso_c = []  
for i in range(4):  
    a_et_b_peso_c.append([a_et_b.tolist()[i][0],c[i]])  
a_et_b_peso_c
```

```
[12]: [[1.000038504600525, 0],  
       [0.9999999403953552, 1],  
       [1.0001057386398315, 1],  
       [1.4901161193847656e-08, 0]]
```

```
[13]: a_et_b_peso_c = model_peso.predict(np.array(a_et_b_peso_c))  
a_et_b_peso_c
```

```
[13]: array([[1.16135925e-05],  
            [1.00000000e+00],  
            [1.00003064e+00],  
            [3.72529030e-09]], dtype=float32)
```

7 Observemos que el resultado es muy cercano al [0,1,1,0].