## Análisis de Datos y Aprendizaje Máquina con Tensorflow 2.0: Perceptrón Multicapa

## 2019/09/30

## **Dropout**

Objetivo: Conocer Dropout, sus parámetros y como afecta los entrenamientos de una red neuronal para clasificación de imágenes

- Dropout ignora algunas neuronas durante el entrenamiento, de forma que se puede decir que se entrenan sub-ensambles de redes
- La probabilidad de que cada neurona se desconecte se indica por 'tasa de dropout'
- Dropout ha sido empleado tambien para RNN con el método de Zoneout

```
In [1]: import matplotlib.pyplot as plt
        import tensorflow as tf
        from tensorflow import keras
        from tensorflow.keras.models import Sequential
        from tensorflow.keras.layers import Dropout
        from tensorflow.keras.layers import Dense
        from tensorflow.keras import backend as K
        K.clear_session()
        mnist = keras.datasets.mnist
        (x_train, y_train), (x_test, y_test) = mnist.load_data()
In [2]: print(x_train.shape)
       print(y_train.shape)
       print(x test.shape)
       print(y_test.shape)
(60000, 28, 28)
(60000,)
(10000, 28, 28)
(10000,)
```

## Leer Dataset

```
In [3]: fig, ax = plt.subplots(nrows=2, ncols=5, sharex=True, sharey=True,)
    ax = ax.flatten()
    for i in range(10):
        img = x_train[y_train == i][0].reshape(28, 28)
        ax[i].imshow(img, cmap='Greys', interpolation='nearest')

ax[0].set_xticks([])
    plt.tight_layout()
    plt.show()
```

 $\bullet\,$  Se modifica la forma de los datos de 2-d (n, 28, 28) a 1-d (n, 784)

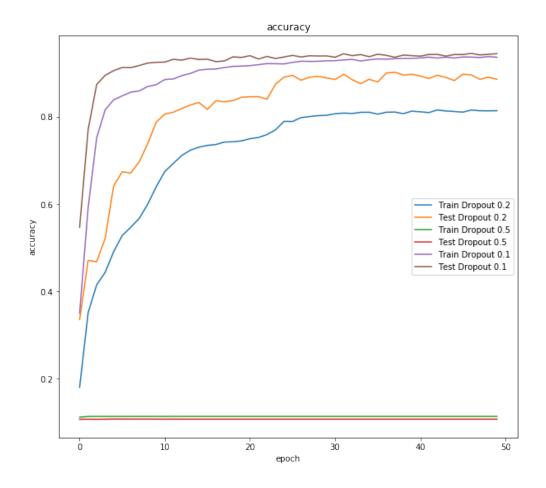
```
In [6]: def make_model():
        model = Sequential()
        model.add(Dense(40, input_shape = (784, ), activation = 'relu'))
        model.add(Dropout(0.2))
                                           # capa Dropout
        model.add(Dense(40, activation = 'relu'))
        model.add(Dropout(0.2))
                                           # capa Dropout
        model.add(Dense(40, activation = 'relu'))
        model.add(Dropout(0.2))
                                           # capa Dropout
        model.add(Dense(10, activation = 'softmax'))
        model.compile(optimizer='adam', loss='sparse_categorical_crossentropy',
                   metrics=['accuracy'])
        return model
In [7]: model = make_model()
      model.summary()
Model: "sequential"
Layer (type) Output Shape Param #
______
dense (Dense)
                     (None, 40)
                                        31400
dropout (Dropout) (None, 40) 0
                    (None, 40)
dense_1 (Dense)
                                       1640
_____
dropout_1 (Dropout) (None, 40)
dense_2 (Dense) (None, 40)
                                       1640
dropout_2 (Dropout) (None, 40) 0
dense_3 (Dense) (None, 10) 410
_____
Total params: 35,090
Trainable params: 35,090
Non-trainable params: 0
______
In [8]: history1 = model.fit(x_train, y_train, batch_size = batch, validation_split = 0.3,
                    epochs = epoch, verbose = verbose)
In [9]: test_loss, test_acc = model.evaluate(x_test, y_test, verbose=2)
      print('\nTest acccuracy:', test_acc)
```

```
10000/1 - 1s - loss: 0.3316 - accuracy: 0.8815
Test acccuracy: 0.8815
In [10]: def make_model():
          model = Sequential()
          model.add(Dense(40, input_shape = (784, ), activation = 'relu'))
          model.add(Dropout(0.5))
                                                # capa Dropout
          model.add(Dense(40, activation = 'relu'))
          model.add(Dropout(0.5))
                                               # capa Dropout
          model.add(Dense(40, activation = 'relu'))
          model.add(Dropout(0.5))
                                               # capa Dropout
          model.add(Dense(10, activation = 'softmax'))
          model.compile(optimizer='adam', loss='sparse_categorical_crossentropy',
                     metrics=['accuracy'])
          return model
In [11]: model = make_model()
       model.summary()
Model: "sequential_1"
 ayer (type) Output Shape Param #
Layer (type)
______
dense_4 (Dense)
                      (None, 40)
                                           31400
dropout_3 (Dropout) (None, 40) 0
dense_5 (Dense) (None, 40)
                                          1640
dropout_4 (Dropout) (None, 40)
dense_6 (Dense) (None, 40)
                                          1640
dropout_5 (Dropout) (None, 40)
dense_7 (Dense) (None, 10) 410
_____
Total params: 35,090
Trainable params: 35,090
Non-trainable params: 0
In [12]: history2 = model.fit(x_train, y_train, batch_size = batch, validation_split = 0.3,
                       epochs = epoch, verbose = verbose)
```

```
In [13]: test_loss, test_acc = model.evaluate(x_test, y_test, verbose=2)
       print('\nTest acccuracy:', test_acc)
10000/1 - 1s - loss: 2.3022 - accuracy: 0.1135
Test acccuracy: 0.1135
In [14]: def make_model():
          model = Sequential()
          model.add(Dense(40, input_shape = (784, ), activation = 'relu'))
          model.add(Dropout(0.1))
                                                # capa Dropout
          model.add(Dense(40, activation = 'relu'))
                                                # capa Dropout
          model.add(Dropout(0.1))
          model.add(Dense(40, activation = 'relu'))
          model.add(Dropout(0.1))
                                                # capa Dropout
          model.add(Dense(10, activation = 'softmax'))
          model.compile(optimizer='adam', loss='sparse_categorical_crossentropy',
                     metrics=['accuracy'])
          return model
In [15]: model = make model()
       model.summary()
Model: "sequential_2"
Layer (type)
                    Output Shape Param #
______
dense_8 (Dense)
                 (None, 40)
                                           31400
dropout_6 (Dropout) (None, 40)
dense_9 (Dense) (None, 40) 1640
dropout_7 (Dropout) (None, 40)
dense_10 (Dense) (None, 40) 1640
dropout_8 (Dropout) (None, 40)
dense_11 (Dense) (None, 10)
                                           410
_____
Total params: 35,090
Trainable params: 35,090
Non-trainable params: 0
```

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```
In [16]: history3 = model.fit(x_train, y_train, batch_size = batch, validation_split = 0.3,
                             epochs = epoch, verbose = verbose)
In [17]: test_loss, test_acc = model.evaluate(x_test, y_test, verbose=2)
         print('\nTest acccuracy:', test_acc)
10000/1 - 1s - loss: 0.1365 - accuracy: 0.9436
Test acccuracy: 0.9436
In [18]: #plot
         plt.figure(figsize=(10,9))
         plt.plot(history1.history['accuracy'])
         plt.plot(history1.history['val_accuracy'])
         plt.plot(history2.history['accuracy'])
         plt.plot(history2.history['val_accuracy'])
         plt.plot(history3.history['accuracy'])
         plt.plot(history3.history['val_accuracy'])
         plt.legend(['Train Dropout 0.2', 'Test Dropout 0.2',
                     'Train Dropout 0.5', 'Test Dropout 0.5',
                     'Train Dropout 0.1', 'Test Dropout 0.1'])
         plt.title('accuracy')
         plt.ylabel('accuracy')
         plt.xlabel('epoch')
         plt.show()
```



- Experimentar los demás parámetros de dropout
- Describir el comportamiento de Dropout con diferentes funciones de costo, capas y optimizadores.
- Comparar y combinar con otros métodos de regularización