

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

2019/09/30

1 Dropout

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

- Referencia: <https://dl.acm.org/doi/abs/10.5555/2627435.2670313>
- 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()

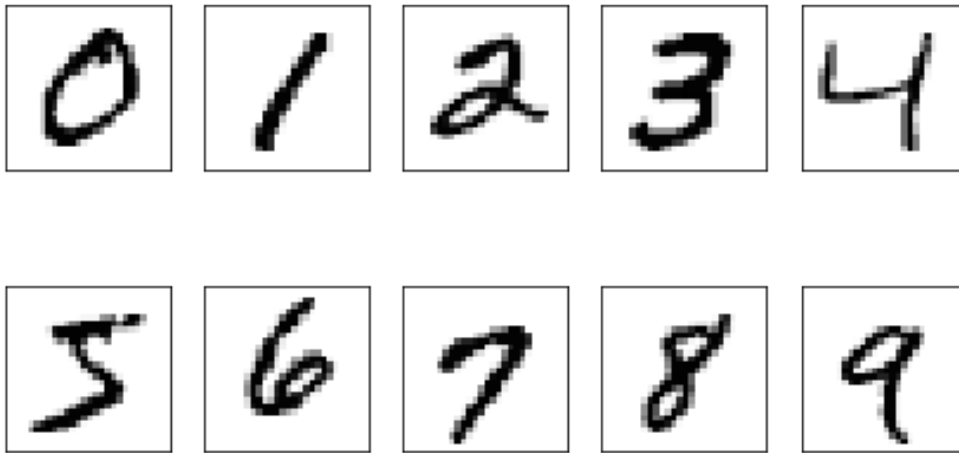
In [2]: print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)

(60000, 28, 28)
(60000,)
(10000, 28, 28)
(10000,)
```

1.1 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([])
        ax[0].set_yticks([])
        plt.tight_layout()
        plt.show()
```



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

```
In [4]: x_train = x_train.reshape(x_train.shape[0], -1)
        x_test = x_test.reshape(x_test.shape[0], -1)
```

```
print(x_train.shape) # (60000, 784)
print(y_train.shape) # (60000,)
print(x_test.shape)  # (10000, 784)
print(y_test.shape)  # (10000,)
```

```
(60000, 784)
(60000,)
(10000, 784)
(10000,)
```

```
In [5]: epoch = 50
        verbose = 0
        batch = 50
```

```
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
dense_1 (Dense)	(None, 40)	1640
dropout_1 (Dropout)	(None, 40)	0
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 accuracy:', 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"

Layer (type)	Output Shape	Param #
dense_4 (Dense)	(None, 40)	31400
dropout_3 (Dropout)	(None, 40)	0
dense_5 (Dense)	(None, 40)	1640
dropout_4 (Dropout)	(None, 40)	0
dense_6 (Dense)	(None, 40)	1640
dropout_5 (Dropout)	(None, 40)	0
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 accuracy:', test_acc)

10000/1 - 1s - loss: 2.3022 - accuracy: 0.1135

Test accuracy: 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'))
    model.add(Dropout(0.1)) # capa Dropout
    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)                0
-----
dense_9 (Dense)              (None, 40)                1640
-----
dropout_7 (Dropout)          (None, 40)                0
-----
dense_10 (Dense)             (None, 40)                1640
-----
dropout_8 (Dropout)          (None, 40)                0
-----
dense_11 (Dense)             (None, 10)                410
=====
Total params: 35,090
Trainable params: 35,090
Non-trainable params: 0
```

```

-----

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 accuracy:', test_acc)

10000/1 - 1s - loss: 0.1365 - accuracy: 0.9436

Test accuracy: 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'])

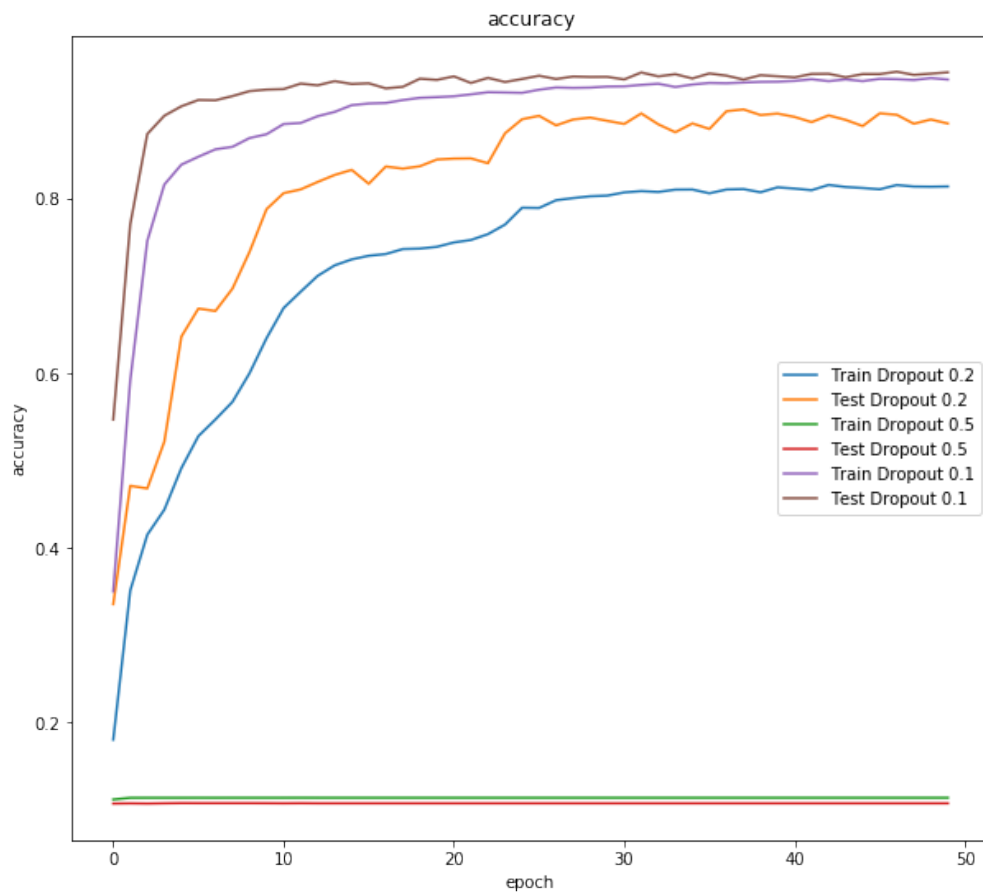
         plt.plot(history3.history['accuracy'])
         plt.plot(history3.history['val_accuracy'])

         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