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

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Perceptrón Multicapa Imperative/Training loop

- Objetivo: Programar un MLP en notación orientada a objetos con Tensorflow 2.0
- Los modelos son más personalizables para tareas como investigación

https://blog.tensorflow.org/2019/01/what-are-symbolic-and-imperative-apis.html

```
In [1]: import matplotlib.pyplot as plt
        from sklearn.datasets import load_breast_cancer
        from sklearn.model_selection import train_test_split
        import tensorflow as tf
        import numpy as np
In [2]: data = load_breast_cancer()
In [3]: X_data = data.data
        y_data = data.target
In [4]: x_train, x_test, y_train, y_test = train_test_split(X_data, y_data, test_size = 0.33, rar
In [5]: print(x_train.shape)
       print(x_test.shape)
        print(y_train.shape)
        print(y_test.shape)
(381, 30)
(188, 30)
(381,)
(188,)
In [6]: batch_size = 32
        train_ds = tf.data.Dataset.from_tensor_slices(
            (x_train, y_train)).shuffle(10000).batch(batch_size)
```

In [7]: from tensorflow.keras.layers import Dense
 from tensorflow.keras import Model

Crear modelo

• Se crea una clase, las capas se definen en el constructor y el método call indica el flujo

```
In [8]: class MLP(Model):
         def init (self):
             super(MLP, self).__init__()
             self.d1 = Dense(30, activation='sigmoid', name='input')
             self.d2 = Dense(20, activation='sigmoid', name='hidden')
             self.d3 = Dense(1, activation='sigmoid', name='output')
         def call(self, x): # método call que pasa 'x' por capa
             x = self.d1(x)
             x = self.d2(x)
             return self.d3(x)
In [9]: model = MLP()
      model.build(input_shape=(None, 30))
      model.summary()
Model: "mlp"
Layer (type)
            Output Shape Param #
______
             multiple
input (Dense)
hidden (Dense) multiple
                                            620
output (Dense) multiple 21
______
Total params: 1,571
Trainable params: 1,571
Non-trainable params: 0
  • Optimizador y función de costo
In [10]: loss_fn = tf.keras.losses.BinaryCrossentropy()
       optimizer = tf.keras.optimizers.SGD()

    Métricas

In [11]: train_loss = tf.keras.metrics.BinaryCrossentropy(name='train_loss')
       train_accuracy = tf.keras.metrics.BinaryAccuracy(name='train_accuracy')
  · Listas para plot
In [12]: hist_loss = []
       hist_acc = []
```

Entrenamiento

- Se hace un ciclo por épocas en donde se itera por cada época sobre cada par de datos y etiquetas de entrenamiento
- Nota: El entrenamiento con 'tf.function decorator' tiene un mejor desempeño al compilarse en grafo. Para simplificar el ejemplo, también se omite el entrenamiento en conjunto de prueba

```
In [13]: EPOCH = 20
         for epoch in range(EPOCH):
             #entrenamiento
             for data, target in train_ds:
                 with tf.GradientTape() as tape:
                     predictions = model(data)# predicciones
                     loss = loss_fn(target, predictions) # target y predicciones para obtener acc
                 gradients = tape.gradient(loss, model.trainable_variables) # gradiente sobre var
                 optimizer.apply_gradients(zip(gradients, model.trainable_variables))
                 # se guardan metricas
                 train_loss(target, predictions)
                 train_accuracy(target, predictions)
             template = 'Epoch \{\}/\{\} \setminus n - loss: \{\} - accuracy: \{\}'
             print(template.format(epoch+1, EPOCH,
                                  train_loss.result(), train_accuracy.result()))
             # lista para plot
             hist_loss.append(train_loss.result())
             hist_acc.append(train_accuracy.result())
             # reinicia las metricas para la siguiente epoca
             train_loss.reset_states()
             train_accuracy.reset_states()
WARNING: tensorflow: Layer mlp is casting an input tensor from dtype float64 to the layer's dtype of
If you intended to run this layer in float32, you can safely ignore this warning. If in doubt, th
To change all layers to have dtype float64 by default, call `tf.keras.backend.set_floatx('float64
Epoch 1/20
 - loss: 0.6844579577445984 - accuracy: 0.6170976758003235
 - loss: 0.6588442921638489 - accuracy: 0.7556573748588562
Epoch 3/20
 - loss: 0.6512103080749512 - accuracy: 0.7567349076271057
Epoch 4/20
```

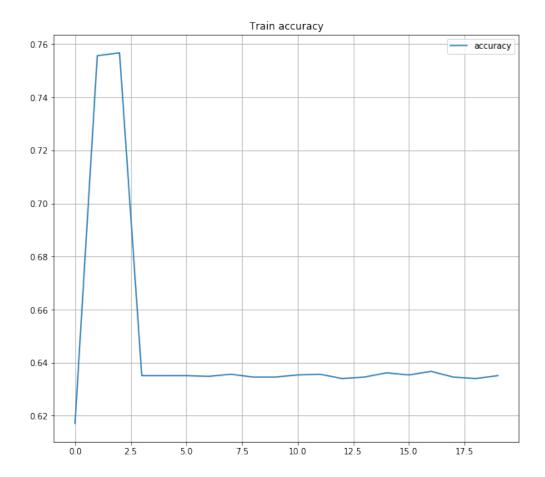
- loss: 0.6424896121025085 - accuracy: 0.6350574493408203

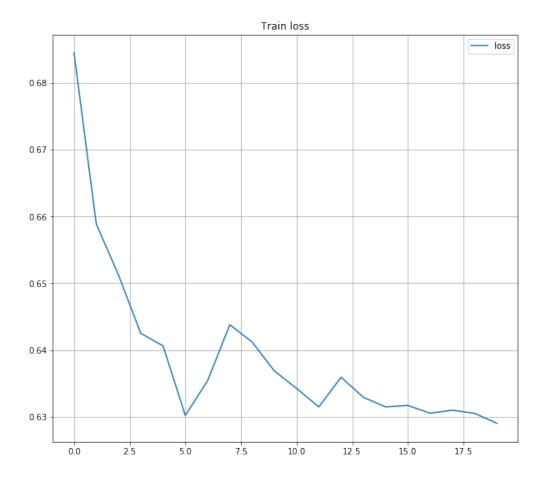
- loss: 0.6406248211860657 - accuracy: 0.6350574493408203

Epoch 5/20

Epoch 6/20

```
- loss: 0.6302149295806885 - accuracy: 0.6350574493408203
Epoch 7/20
 - loss: 0.6354005932807922 - accuracy: 0.6347880959510803
Epoch 8/20
 - loss: 0.6437819600105286 - accuracy: 0.6355962753295898
Epoch 9/20
 - loss: 0.6412071585655212 - accuracy: 0.6345186829566956
Epoch 10/20
 - loss: 0.6368744969367981 - accuracy: 0.6345186829566956
Epoch 11/20
- loss: 0.6342592239379883 - accuracy: 0.6353268623352051
Epoch 12/20
- loss: 0.6314993500709534 - accuracy: 0.6355962753295898
Epoch 13/20
- loss: 0.6359333395957947 - accuracy: 0.633979856967926
Epoch 14/20
- loss: 0.6329274773597717 - accuracy: 0.6345186829566956
- loss: 0.6314908862113953 - accuracy: 0.6361350417137146
Epoch 16/20
 - loss: 0.6317090392112732 - accuracy: 0.6353268623352051
Epoch 17/20
 - loss: 0.6305532455444336 - accuracy: 0.6366738677024841
Epoch 18/20
 - loss: 0.6310041546821594 - accuracy: 0.6345186829566956
Epoch 19/20
- loss: 0.6305307149887085 - accuracy: 0.633979856967926
Epoch 20/20
- loss: 0.6290386319160461 - accuracy: 0.6350574493408203
In [14]: plt.figure(figsize=(10,9))
         plt.plot(np.arange(len(hist_acc)), hist_acc)
         plt.title('Train accuracy')
         plt.legend(['accuracy'])
         plt.grid()
```





- Mejorar el modelo Implementar con otro dataset