## Redes Neuronales Artificales

#### Instalar Theano

pip install --upgrade --no-deps git+git://github.com/Theano/Theano.git

#### Instalar Tensorflow y Keras

conda install -c conda-forge keras

## Parte 1 - Pre procesado de datos

#### Cómo importar las librerías

```
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
```

#### ▼ Importar el data set

arrav([0.0, 0.0, 619, 0, 42, 2, 0.0, 1, 1, 1, 101348.88], dtvpe=object)

#### Codificar datos categóricos

```
from sklearn.preprocessing import LabelEncoder
labelencoder_X_1 = LabelEncoder()
X[:, 1] = labelencoder_X_1.fit_transform(X[:, 1])
labelencoder_X_2 = LabelEncoder()
X[:, 2] = labelencoder_X_2.fit_transform(X[:, 2])
from sklearn.preprocessing import OneHotEncoder
from sklearn.compose import ColumnTransformer
transformer = ColumnTransformer(
    transformers=[
        ("Churn_Modelling",
                            # Un nombre de la transformación
         OneHotEncoder(categories='auto'), # La clase a la que transformar
         [1]
                        # Las columnas a transformar.
    ], remainder='passthrough'
)
X = transformer.fit_transform(X)
X = X[:, 1:]
```

Dividir el data set en conjunto de entrenamiento y conjunto de testing

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state =
```

▼ Escalado de variables

```
from sklearn.preprocessing import StandardScaler
sc_X = StandardScaler()
X_train = sc_X.fit_transform(X_train)
X_test = sc_X.transform(X_test)
```

#### ▼ Parte 2 - Construir la RNA

```
# Importar Keras y librerías adicionales
import keras
from keras.models import Sequential
from keras.layers import Dense
```

Using TensorFlow backend.

The default version of TensorFlow in Colab will soon switch to TensorFlow 2.x. We recommend you <u>upgrade</u> now or ensure your notebook will continue to use TensorFlow 1.x via the %t

```
# Inicializar la RNA
classifier = Sequential()
# Añadir las capas de entrada y primera capa oculta
classifier.add(Dense(units = 6, kernel_initializer = "uniform",
                   activation = "relu", input_dim = 11))
# Añadir la segunda capa oculta
classifier.add(Dense(units = 6, kernel_initializer = "uniform", activation = "relu"))
# Añadir la capa de salida
classifier.add(Dense(units = 1, kernel initializer = "uniform", activation = "sigmoid"))
# Compilar la RNA
classifier.compile(optimizer = "adam", loss = "binary_crossentropy", metrics = ["accuracy"
# Ajustamos la RNA al Conjunto de Entrenamiento
classifier.fit(X_train, y_train, batch_size = 10, epochs = 100)
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensc
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensor
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensor
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/optimizers.py
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tenso
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow_core/pyt
    Instructions for updating:
    Use tf.where in 2.0, which has the same broadcast rule as np.where
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tenso
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensor
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tenso
    Epoch 1/100
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tenso
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensor
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tenso
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tenso
    WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/keras/backend/tensor
    Epoch 2/100
    Epoch 3/100
    8000/8000 [======================] - 1s 126us/step - loss: 0.3945 - acc: 0
```

```
Epoch 4/100
Epoch 5/100
8000/8000 [============= ] - 1s 125us/step - loss: 0.3794 - acc: 0
Epoch 6/100
Epoch 7/100
8000/8000 [============= ] - 1s 124us/step - loss: 0.3705 - acc: 0
Epoch 8/100
8000/8000 [============== ] - 1s 129us/step - loss: 0.3671 - acc: 0
Epoch 9/100
Epoch 10/100
Epoch 11/100
Epoch 12/100
Epoch 13/100
Epoch 14/100
```

# Parte 3 - Evaluar el modelo y calcular predicciones finales

Predicción de los resultados con el Conjunto de Testing

```
y_pred = classifier.predict(X_test)
y_pred = (y_pred>0.5)
```

#### Predecir una nueva observación

Utiliza nuestro modelo de RNA para predecir si el cliente con la siguiente información abandonará el banco:

• Geografia: Francia

Puntaje de crédito: 600

Género masculino

Edad: 40 años de edad

Tenencia: 3 años.

Saldo: \$ 60000

Número de productos: 2

¿Este cliente tiene una tarjeta de crédito? Sí

¿Es este cliente un miembro activo? Sí

• Salario estimado: \$50000

Entonces, ¿deberíamos decir adiós a ese cliente?

```
new_prediction ·= · classifier.predict(sc_X.transform(np.array([[0,0,600,·1,·40,·3,·60000,·2,
print(new_prediction)
print(new_prediction ·> · 0.5)

[[0.04022561]]
[[False]]
```

## ▼ Elaborar una matriz de confusión

```
from sklearn.metrics import confusion_matrix
cm·=·confusion_matrix(y_test, ·y_pred)

(cm[0][0]+cm[1][1])/cm.sum()

0.866
```