

Final Project

Movie genre classification



Deep Learning and Neural Network

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Objective

Classify a movie genre based on plots and posters.

Methodology

- Machine Learning algorithm using both images and text
- Deep Learning algorithm using both images and text
- Embed results of the best model found in images and the best model found in text.



Using the movie plots - Text

The function TfidfVectorizer, allows summarize the text to a matrix, using the parameters: ngram, max_features and stemmer:

```
vect2 = TfidfVectorizer(ngram_range=(1, 2),max_features=15000,analyzer=split_into_stemmer)
X_dtm2_train = vect2.fit_transform(X_train)
X_dtm2_test = vect2.transform(X_test)
X_dtm2_train.shape
```

This transformation will be use to fit a random forest (Machine Learning) and also to a Neuronal Network (Deep Learning).



Using the movie posters - Image

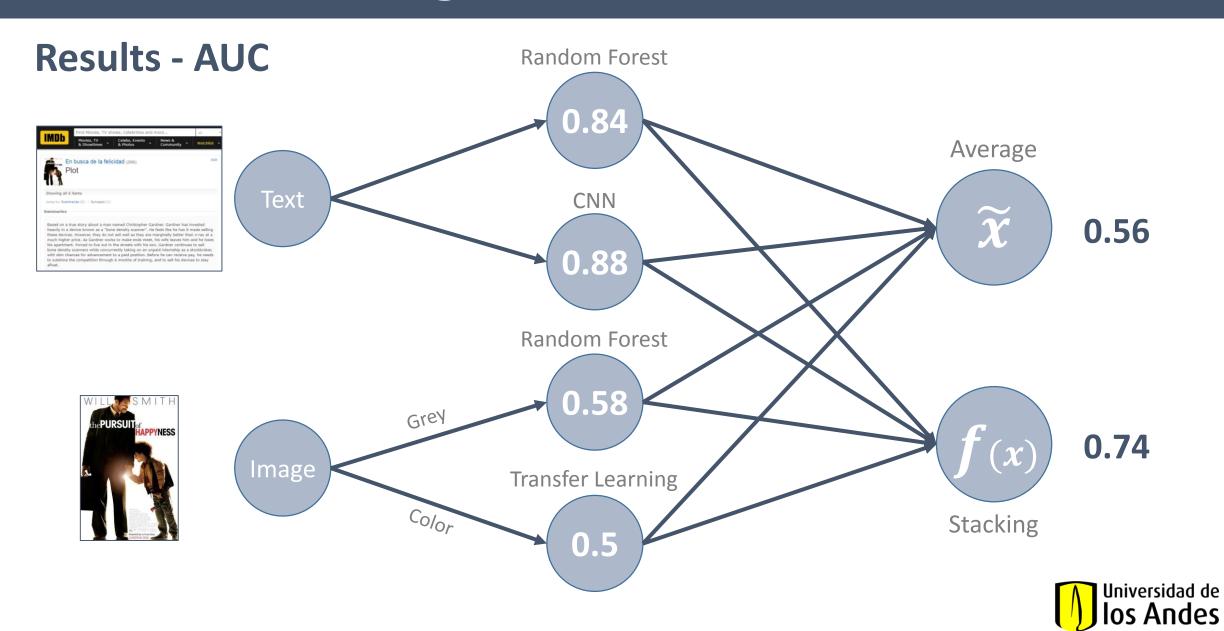
A PCA with 80 componentes was applied to the flatten grey scale and color images. For the 80 components, fit a random forest as a Machine Learning algorithm.

After split Training color images into train and validation, the pre-trained VGG16 model without de top layer was used as input of a neural network (Deep Learning).

```
from keras.applications.vgg16 import VGG16
model_vgg16 = VGG16(weights='imagenet', include_top=False)

from keras.models import Model
model1 = Model(inputs=model_vgg16.input, outputs=model_vgg16.get_layer('block5_pool').output)
```





Random Forest – Text



Random Forest – Images

```
clf.fit(X train, y train genres)
OneVsRestClassifier(estimator=RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
             max_depth=10, max_features='auto', max_leaf_nodes=None,
             min impurity decrease=0.0, min impurity split=None,
             min samples leaf=1, min samples split=2,
             min weight fraction leaf=0.0, n estimators=500, n jobs=-1,
             oob score=False, random state=42, verbose=0, warm start=False),
           n jobs=1)
y pred genres = clf.predict proba(X test)
 roc_auc_score(y_test_genres, y_pred_genres, average='macro')
0.5798856252910195
```



CNN – Text

```
max_words=15000
    Modelo4= Sequential()
    Modelo4.add(Dense(300, input shape=(max words,)))
    Modelo4.add(Activation('relu'))
    Modelo4.add(BatchNormalization())
    Modelo4.add(Dropout(0.7))
    Modelo4.add(Dense(24, activation='sigmoid'))
    Modelo4.compile(loss = 'categorical crossentropy',
                  optimizer = Adagrad(),
                  metrics = ['accuracy'])
    Modelo4.fit(X_dtm2_train, y_train, epochs=10, verbose=2)
F⇒ Epoch 1/10
     - 2s - loss: 7.0963 - acc: 0.2532
    Epoch 2/10
     - 2s - loss: 5.7879 - acc: 0.3139
y pred4=Modelo4.predict proba(X dtm2 test)
    roc_auc_score(y_test, y_pred4, average='macro')
    0.875589276062426
```



Transfer Learning – Images

model2.summary()

Layer (type)	Output	Shape	Param #
flatten_1 (Flatten)	(None,	25088)	0
dense_1 (Dense)	(None,	1024)	25691136
dropout_1 (Dropout)	(None,	1024)	0
dense_2 (Dense)	(None,	512)	524800
dropout_2 (Dropout)	(None,	512)	0
dense_3 (Dense)	(None,	24)	12312

Total params: 26,228,248
Trainable params: 26,228,248
Non-trainable params: 0

roc_auc_score(np.array(y_valCN), np.array(preds.tolist()), average='macro')

0.487433200743367

