Report

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Platform: Python(2.7) + some dependencies, libraries (NLTK toolkit, Word2Vec)

Analysing Dataset:

We are given 2601 examples which have following attributes:

- asset_id: identifier of the movie;
- title: the title of the movie;
- summary: a short text describing the movie;
- poster_movie_url: it is the url location of the movie poster;
- poster trailer url: if available, it is the url location of the movie trailer.
- Moreover, we have 17 binary attributes assign the movie to a genre. The same movie can belong to multiple genres simultaneously.

Training Data = 80% Testing Data = 20%

Classified Label Prediction : For predicting the output label a threshold was set and based on that we predict the genre label.

Evaluation Metric Used : F1 score is used as an evaluation metric which is harmonic mean of precision and recall.

Here,

Precision = True Positive / (True Positive + False Positive)
Recall = True Positive / (True Positive + False Negative)
F1 score = 2*Precision*Recall/(Precision + Recall)

Method 1: This approach is based on working with only summaries of each movie and extract features and convert it into a feature vector.

In this approach I used pretrained Glove Embeddings trained on Wikipedia Italian Corpus . Then the steps involved were:

- Tokenization of Summary text
- Removal of Stop Words
- Now, use Glove Embeddings which gives featurised representation of each word (300 dimension)which is then used as a feature vector.
- Now, mean average of these embeddings is taken according to frequency of each word occurring in the summary.

Method 2: This approach is based on working with only images of each movie poster and extract features and convert it into a feature vector.

In this approach I used pretrained VGG feature extractor(pretrained model from imagenet data) and obtained a featurised representation of each image. Each feature vector is of 8192 dimension. Then these features are fed into the neural network to train our model.

Method 3: This approach is slight modification of 2nd approach and now I extract featurised representation of both move_poster image and movie_trailer image.

In this approach I used pretrained VGG feature extractor and obtained a featurised representation of each image . Finally our feature vector is mean of feature vector_1(movie_poster image) and vector_2(movie_trailer image). This gives slightly better results as compared to 2nd approach.

Method 4: This approach is based on concatenating features(both visual and text) and then this feature vector is used to train our model.

Feature vectors obtained individually based on approach 1 and approach 4 are concatenated and then these features are fed into neural network to train our model.

Comparison of Different Approaches:

Approach	Accuracy(%)	Precision	Recall	F1 Score
Method 1	89.07	0.6191	0.6062	0.6126
Method 2	92.54	0.4034	0.4186	0.4108
Method 3	92.61	0.4540	0.4399	0.4468
Method 4	92.47	0.4319	0.4982	0.4626

Neural Network Architecture:

3 layer neural network architecture was used in which the input layer, hidden layer has "relu" activation function and at output layer "softmax" activation function was used which gives probability of occurrence of each genre.

References:

- https://medium.com/@franky07724 57962/using-ke ras-pre-trained-models-for-feature-extraction-in-ima ge-clustering-a142c6cdf5b1
- http://hlt.isti.cnr.it/wordembeddings/