Multiclass Text Classification with

Logistic Regression Implemented with PyTorch and CE Loss

First, we will do some initialization.

```
In [ ]: import random
        import torch
        import numpy as np
        import pandas as pd
        from tgdm.notebook import tgdm
        # enable tqdm in pandas
        tqdm.pandas()
        # set to True to use the gpu (if there is one available)
        use_gpu = True
        # select device
        device = torch.device('cuda' if use_gpu and torch.cuda.is_available() else
        print(f'device: {device.type}')
        # random seed
        seed = 1234
        # set random seed
        if seed is not None:
            print(f'random seed: {seed}')
            random.seed(seed)
            np.random.seed(seed)
            torch.manual seed(seed)
```

device: cpu random seed: 1234

We will be using the AG's News Topic Classification Dataset. It is stored in two CSV files: train.csv and test.csv, as well as a classes.txt that stores the labels of the classes to predict.

First, we will load the training dataset using pandas and take a quick look at how the data.

```
In []: train_df = pd.read_csv('./train.csv', header=None) # leemos el archivo csv
    train_df.columns = ['class index', 'title', 'description'] # le asignamos no
    train_df # mostramos el dataframe
```

	class index	title	description
0	Class Index	Title	Description
1	3	Wall St. Bears Claw Back Into the Black (Reuters)	Reuters - Short-sellers, Wall Street's dwindli
2	3	Carlyle Looks Toward Commercial Aerospace (Reu	Reuters - Private investment firm Carlyle Grou
3	3	Oil and Economy Cloud Stocks' Outlook (Reuters)	Reuters - Soaring crude prices plus worries\ab
4	3	Iraq Halts Oil Exports from Main Southern Pipe	Reuters - Authorities have halted oil export\f
•••		•••	***
119996	1	Pakistan's Musharraf Says Won't Quit as Army C	KARACHI (Reuters) - Pakistani President Perve
119997	2	Renteria signing a top-shelf deal	Red Sox general manager Theo Epstein acknowled
119998	2	Saban not going to Dolphins yet	The Miami Dolphins will put their courtship of
119999	2	Today's NFL games	PITTSBURGH at NY GIANTS Time: 1:30 p.m. Line:
120000	2	Nets get Carter from Raptors	INDIANAPOLIS All-Star Vince

120001 rows × 3 columns

2

120000

The dataset consists of 120,000 examples, each consisting of a class index, a title, and a description. The class labels are distributed in a separated file. We will add the labels to the dataset so that we can interpret the data more easily. Note that the label indexes are one-based, so we need to subtract one to retrieve them from the list.

Nets get Carter from Raptors

```
In []: labels = open('./classes.txt').read().splitlines() # leemos las etiquetas de
    train_df = train_df.drop(0).reset_index(drop=True) # eliminamos la primera
        train_df['class index'] = train_df['class index'].astype(int) # convertimos
        classes = train_df['class index'].map(lambda i: labels[i-1]) # mapeamos los
        train_df.insert(1, 'class', classes) # insertamos la columna 'class' en la p
        train_df # mostramos el dataframe
```

Carter was trad...

Oil and Economy Cloud

Main Southern Pipe...

record, posing new...

Stocks' Outlook (Reuters)

Iraq Halts Oil Exports from

Oil prices soar to all-time

Pakistan's Musharraf Says

Renteria signing a top-shelf

Saban not going to Dolphins

Won't Quit as Army C...

Today's NFL games

Nets get Carter from

Raptors

Reuters - Soaring crude prices

Reuters - Authorities have

AFP - Tearaway world oil

KARACHI (Reuters) - Pakistani

Red Sox general manager

Theo Epstein acknowled...

The Miami Dolphins will put

PITTSBURGH at NY GIANTS

Time: 1:30 p.m. Line: ...

INDIANAPOLIS -- All-Star

Vince Carter was trad...

prices, toppling reco...

plus worries\ab...

halted oil export\f...

President Perve...

their courtship of...

Out[]:		class index	class	title	description
	0 3		Business	Wall St. Bears Claw Back Into the Black (Reuters)	Reuters - Short-sellers, Wall Street's dwindli
	1	3	Business	Carlyle Looks Toward Commercial Aerospace (Reu	Reuters - Private investment firm Carlyle Grou

120000 rows x 4 columns

2

3

4

119995

119996

119997

119998

119999

Business

Business

Business

World

Sports

Sports

Sports

Sports

1

2

2

2

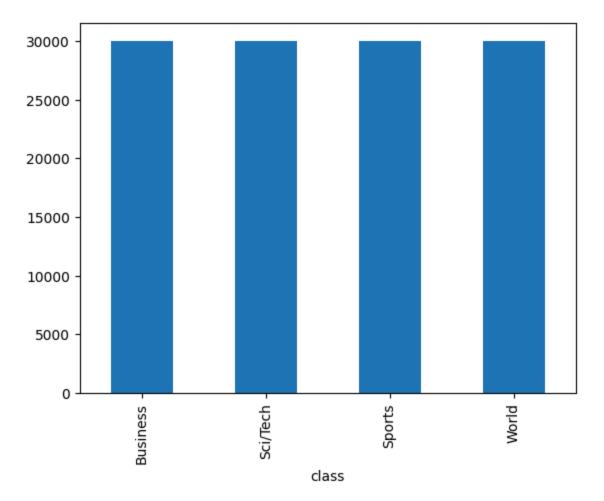
2

Let's inspect how balanced our examples are by using a bar plot.

```
In []:
       pd.value_counts(train_df['class']).plot.bar() # graficamos la distribución d
```

/var/folders/7b/g3f6kzfd1b99b918vd8tvd100000gn/T/ipykernel_25959/1245903889. py:1: FutureWarning: pandas.value_counts is deprecated and will be removed i n a future version. Use pd.Series(obj).value_counts() instead. pd.value_counts(train_df['class']).plot.bar()

Out[]: <Axes: xlabel='class'>



The classes are evenly distributed. That's great!

However, the text contains some spurious backslashes in some parts of the text. They are meant to represent newlines in the original text. An example can be seen below, between the words "dwindling" and "band".

```
In [ ]: print(train_df.loc[0, 'description']) # mostramos la descripción de la prime
```

Reuters - Short-sellers, Wall Street's dwindling\band of ultra-cynics, are seeing green again.

We will replace the backslashes with spaces on the whole column using pandas replace method.

```
In []: title = train_df['title'].str.lower() # convertimos el título a minúsculas
  descr = train_df['description'].str.lower() # convertimos la descripción a n
  text = title + " " + descr # concatenamos el título y la descripción
  train_df['text'] = text.str.replace('\\', ' ', regex=False) # reemplazamos l
  train_df # mostramos el dataframe
```

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	class index	class	title	description	text
0	3	Business	Wall St. Bears Claw Back Into the Black (Reuters)	Reuters - Short- sellers, Wall Street's dwindli	wall st. bears claw back into the black (reute
1	3	Business	Carlyle Looks Toward Commercial Aerospace (Reu	Reuters - Private investment firm Carlyle Grou	carlyle looks toward commercial aerospace (reu
2	3	Business	Oil and Economy Cloud Stocks' Outlook (Reuters)	Reuters - Soaring crude prices plus worries\ab	oil and economy cloud stocks' outlook (reuters
3	3	Business	Iraq Halts Oil Exports from Main Southern Pipe	Reuters - Authorities have halted oil export\f	iraq halts oil exports from main southern pipe
4	3	Business	Oil prices soar to all-time record, posing new	AFP - Tearaway world oil prices, toppling reco	oil prices soar to all-time record, posing new
•••			•••		•••
119995	1	World	Pakistan's Musharraf Says Won't Quit as Army C	KARACHI (Reuters) - Pakistani President Perve	pakistan's musharraf says won't quit as army c
119996	2	Sports	Renteria signing a top-shelf deal	Red Sox general manager Theo Epstein acknowled	renteria signing a top-shelf deal red sox gene
119997	2	Sports	Saban not going to Dolphins yet	The Miami Dolphins will put their courtship of	saban not going to dolphins yet the miami dolp
119998	2	Sports	Today's NFL games	PITTSBURGH at NY GIANTS Time: 1:30 p.m. Line:	today's nfl games pittsburgh at ny giants time
119999	2	Sports	Nets get Carter from Raptors	INDIANAPOLIS All-Star Vince Carter was trad	nets get carter from raptors indianapolis a

120000 rows × 5 columns

Now we will proceed to tokenize the title and description columns using NLTK's word_tokenize(). We will add a new column to our dataframe with the list of tokens.

```
In []: import nltk # importamos la librería nltk
    nltk.download('punkt_tab') # descargamos el tokenizador 'punkt', que tiene u
    [nltk_data] Downloading package punkt_tab to
    [nltk_data] /Users/hibrantapia/nltk_data...
    [nltk_data] Unzipping tokenizers/punkt_tab.zip.
```

```
Out[]: True
```

	class index	class	title	description	text	tokens
0	3	Business	Wall St. Bears Claw Back Into the Black (Reuters)	Reuters - Short- sellers, Wall Street's dwindli	wall st. bears claw back into the black (reute	[wall, st., bears, claw, back, into, the, blac
1	3	Business	Carlyle Looks Toward Commercial Aerospace (Reu	Reuters - Private investment firm Carlyle Grou	carlyle looks toward commercial aerospace (reu	[carlyle, looks, toward, commercial, aerospace
2	3	Business	Oil and Economy Cloud Stocks' Outlook (Reuters)	Reuters - Soaring crude prices plus worries\ab	oil and economy cloud stocks' outlook (reuters	[oil, and, economy, cloud, stocks, ', outlook,
3	3	Business	Iraq Halts Oil Exports from Main Southern Pipe	Reuters - Authorities have halted oil export\f	iraq halts oil exports from main southern pipe	[iraq, halts, oil, exports, from, main, southe
4	3	Business	Oil prices soar to all- time record, posing new	AFP - Tearaway world oil prices, toppling reco	oil prices soar to all- time record, posing new	[oil, prices, soar, to, all- time, record, ,, p
•••						
119995	1	World	Pakistan's Musharraf Says Won't Quit as Army C	KARACHI (Reuters) - Pakistani President Perve	pakistan's musharraf says won't quit as army c	[pakistan, 's, musharraf, says, wo, n't, quit,
119996	2	Sports	Renteria signing a top-shelf deal	Red Sox general manager Theo Epstein acknowled	renteria signing a top-shelf deal red sox gene	[renteria, signing, a, top-shelf, deal, red, s
119997	2	Sports	Saban not going to Dolphins yet	The Miami Dolphins will put their courtship of	saban not going to dolphins yet the miami dolp	[saban, not, going, to, dolphins, yet, the, mi
119998	2	Sports	Today's NFL games	PITTSBURGH at NY GIANTS Time: 1:30 p.m. Line:	today's nfl games pittsburgh at ny giants time	[today, 's, nfl, games, pittsburgh, at, ny, gi
119999	2	Sports	Nets get Carter from	INDIANAPOLIS All-Star	nets get carter from	[nets, get, carter, from,

tokens	text	description	title	class	class index
raptors, dianapoli	raptors indianapolis	Vince Carter was trad	Raptors		
	- 1		Raptors		

120000 rows × 6 columns

Now we will create a vocabulary from the training data. We will only keep the terms that repeat beyond some threshold established below.

```
In []: threshold = 10 # umbral de frecuencia de las palabras
    tokens = train_df['tokens'].explode().value_counts() # contamos la frecuenci
    tokens = tokens[tokens > threshold] # seleccionamos las palabras que superar
    id_to_token = ['[UNK]'] + tokens.index.tolist() # creamos una lista con las
    token_to_id = {w:i for i,w in enumerate(id_to_token)} # creamos un diccionar
    vocabulary_size = len(id_to_token) # obtenemos el tamaño del vocabulario
    print(f'vocabulary size: {vocabulary_size:,}') # mostramos el tamaño del voc
    vocabulary size: 19,668
In []: from collections import defaultdict # importamos la clase defaultdict de la
    def make_feature_vector(tokens, unk_id=0): # esta es la función que conviert
```

```
vector = defaultdict(int) # inicializamos un diccionario con valores ent
for t in tokens: # iteramos sobre los tokens
    i = token_to_id.get(t, unk_id) # obtenemos el índice de la palabra e
    vector[i] += 1 # incrementamos el valor del índice en el diccionario

return vector # retornamos el diccionario

train_df['features'] = train_df['tokens'].progress_map(make_feature_vector)
train df # mostramos el dataframe
```

0%| | 0/120000 [00:00<?, ?it/s]

	class index	class	title	description	text	tokens	features
0	3	Business	Wall St. Bears Claw Back Into the Black (Reuters)	Reuters - Short-sellers, Wall Street's dwindli	wall st. bears claw back into the black (reute	[wall, st., bears, claw, back, into, the, blac	{427: 2, 566: 1, 1609: 1, 15347: 1, 120: 1, 73
1	3	Business	Carlyle Looks Toward Commercial Aerospace (Reu	Reuters - Private investment firm Carlyle Grou	carlyle looks toward commercial aerospace (reu	[carlyle, looks, toward, commercial, aerospace	{16371: 2, 1077: 1, 854: 1, 1287: 1, 4243: 1,
2	3	Business	Oil and Economy Cloud Stocks' Outlook (Reuters)	Reuters - Soaring crude prices plus worries\ab	oil and economy cloud stocks' outlook (reuters	[oil, and, economy, cloud, stocks, ', outlook,	{66: 1, 9: 2, 351: 2, 4575: 1, 158: 1, 116: 1,
3	3	Business	Iraq Halts Oil Exports from Main Southern Pipe	Reuters - Authorities have halted oil export\f	iraq halts oil exports from main southern pipe	[iraq, halts, oil, exports, from, main, southe	{77: 2, 7404: 1, 66: 3, 1785: 1, 32: 2, 900: 2
4	3	Business	Oil prices soar to all- time record, posing new	AFP - Tearaway world oil prices, toppling reco	oil prices soar to all- time record, posing new	[oil, prices, soar, to, all- time, record, ,, p	{66: 2, 99: 2, 4376: 1, 4: 2, 3590: 1, 149: 1,
•••				•••			
119995	1	World	Pakistan's Musharraf Says Won't Quit as Army C	KARACHI (Reuters) - Pakistani President Perve	pakistan's musharraf says won't quit as army c	[pakistan, 's, musharraf, says, wo, n't, quit,	{383: 1, 23: 1, 1625: 2, 91: 1, 1804: 1, 285:
119996	2	Sports	Renteria signing a top-shelf deal	Red Sox general manager Theo Epstein acknowled	renteria signing a top-shelf deal red sox gene	[renteria, signing, a, top-shelf, deal, red, s	{8468: 2, 2634: 1, 5: 4, 0: 3, 127: 1, 203: 3,
119997	2	Sports	Saban not going to Dolphins yet	The Miami Dolphins will put their courtship of	saban not going to dolphins yet the miami dolp	[saban, not, going, to, dolphins, yet, the, mi	{7747: 2, 68: 1, 660: 1, 4: 2, 1440: 2, 704: 1

	class index	class	title	description	text	tokens	features
119998	2	Sports	Today's NFL games	PITTSBURGH at NY GIANTS Time: 1:30 p.m. Line:	today's nfl games pittsburgh at ny giants time	[today, 's, nfl, games, pittsburgh, at, ny, gi	{106: 1, 23: 1, 728: 1, 225: 1, 1588: 1, 22: 1
119999	2	Sports	Nets get Carter from Raptors	INDIANAPOLIS All-Star Vince Carter was trad	nets get carter from raptors indianapolis a	[nets, get, carter, from, raptors, indianapoli	{2163: 2, 226: 1, 2403: 2, 32: 1, 2994: 2, 219

120000 rows × 7 columns

```
In []: def make dense(feats): # esta función convierte el vector de características
            x = np.zeros(vocabulary_size) # inicializamos un vector de ceros
            for k,v in feats.items(): # iteramos sobre los elementos del vector de d
                x[k] = v \# asignamos el valor del elemento al vector denso
            return x # retornamos el vector denso
        X_train = np.stack(train_df['features'].progress_map(make_dense)) # converti
        y_train = train_df['class index'].to_numpy() - 1 # convertimos las etiquetas
        X_train = torch.tensor(X_train, dtype=torch.float32) # convertimos la matriz
        y_train = torch.tensor(y_train) # convertimos el arreglo de etiquetas a un t
         0%|
                      | 0/120000 [00:00<?, ?it/s]
In [ ]: from torch import nn # importamos el módulo nn de torch
        from torch import optim # importamos el módulo optim de torch
        # hyperparameters
        lr = 1.0 # learning rate
        n epochs = 5 # number of epochs
        n_examples = X_train.shape[0] # number of examples
        n feats = X train.shape[1] # number of features
        n classes = len(labels) # number of classes
        # initialize the model, loss function, optimizer, and data-loader
        model = nn.Linear(n_feats, n_classes).to(device) # initialize the model
        loss_func = nn.CrossEntropyLoss() # initialize the loss function
        optimizer = optim.SGD(model.parameters(), lr=lr) # initialize the optimizer
        # train the model
        indices = np.arange(n_examples) # create an array of indices
        for epoch in range(n_epochs): # iterate over the epochs
            np.random.shuffle(indices) # shuffle the indices
            for i in tqdm(indices, desc=f'epoch {epoch+1}'): # iterate over the exam
                # clear gradients
                model.zero_grad() # clear gradients
                # send datum to right device
```

```
x = X_train[i].unsqueeze(0).to(device) # send the feature vector to
y_true = y_train[i].unsqueeze(0).to(device) # send the label to the
# predict label scores
y_pred = model(x) # predict the label scores
# compute loss
loss = loss_func(y_pred, y_true) #
# backpropagate
loss.backward() # backpropagate the loss
# optimize model parameters
optimizer.step() # optimize the model parameters
```

```
epoch 1: 0% | | 0/120000 [00:00<?, ?it/s]

epoch 2: 0% | | 0/120000 [00:00<?, ?it/s]

epoch 3: 0% | | 0/120000 [00:00<?, ?it/s]

epoch 4: 0% | | 0/120000 [00:00<?, ?it/s]

epoch 5: 0% | | 0/120000 [00:00<?, ?it/s]
```

Next, we evaluate on the test dataset

	class index	title	description	text	tokens	features
0	Class Index	Title	Description	title description	[title, description]	{394: 1, 3701: 1}
1	3	Fears for T N pension after talks	Unions representing workers at Turner Newall	fears for t n pension after talks unions repre	[fears, for, t, n, pension, after, talks, unio	{862: 1, 11: 1, 151: 1, 5349: 1, 1471: 1, 38:
2	4	The Race is On: Second Private Team Sets Launc	SPACE.com - TORONTO, Canada A second\team o	the race is on: second private team sets launc	[the, race, is, on, :, second, private, team, 	{1: 3, 493: 1, 24: 1, 10: 1, 28: 1, 133: 2, 92
3	4	Ky. Company Wins Grant to Study Peptides (AP)	AP - A company founded by a chemistry research	ky. company wins grant to study peptides (ap) 	[ky., company, wins, grant, to, study, peptide	{11738: 1, 60: 2, 407: 1, 4362: 2, 4: 2, 514:
4	4	Prediction Unit Helps Forecast Wildfires (AP)	AP - It's barely dawn when Mike Fitzpatrick st	prediction unit helps forecast wildfires (ap) 	[prediction, unit, helps, forecast, wildfires,	{9154: 1, 612: 1, 2049: 1, 705: 1, 0: 4, 14: 1
•••			•••			
7596	1	Around the world	Ukrainian presidential candidate Viktor Yushch	around the world ukrainian presidential candid	[around, the, world, ukrainian, presidential, 	{437: 1, 1: 2, 58: 1, 4477: 1, 389: 1, 1398: 1
7597	2	Void is filled with Clement	With the supply of attractive pitching options	void is filled with clement with the supply of	[void, is, filled, with, clement, with, the, s	{9010: 1, 24: 1, 4890: 1, 19: 3, 9831: 2, 1: 5
7598	2	Martinez leaves bitter	Like Roger Clemens did almost exactly eight ye	martinez leaves bitter like roger clemens did 	[martinez, leaves, bitter, like, roger, clemen	{2131: 2, 1790: 1, 2951: 2, 317: 1, 1698: 1, 2
7599	3	5 of arthritis patients in Singapore take Bext	SINGAPORE : Doctors in the United States have	5 of arthritis patients in singapore take bext	[5, of, arthritis, patients, in, singapore, ta	{627: 1, 6: 1, 2701: 1, 2439: 1, 7: 2, 1019: 2
7600	3	EBay gets into rentals	EBay plans to buy the	ebay gets into rentals ebay	[ebay, gets, into, rentals,	{1563: 2, 537: 1, 73:

class index	title	description	text	tokens	features
		apartment and home renta	plans to buy the a	ebay, plans, to, b	1, 11111: 1, 160: 1, 4:

7601 rows × 6 columns

```
In [ ]: test_df = test_df.drop(index=0).reset_index(drop=True) # eliminamos la prime
test_df # mostramos el dataframe
```

	class index	title	description	text	tokens	features
0	3	Fears for T N pension after talks	Unions representing workers at Turner Newall	fears for t n pension after talks unions repre	[fears, for, t, n, pension, after, talks, unio	{862: 1, 11: 1, 151: 1, 5349: 1, 1471: 1, 38:
1	4	The Race is On: Second Private Team Sets Launc	SPACE.com - TORONTO, Canada A second\team o	the race is on: second private team sets launc	[the, race, is, on, :, second, private, team, 	{1: 3, 493: 1, 24: 1, 10: 1, 28: 1, 133: 2, 92
2	4	Ky. Company Wins Grant to Study Peptides (AP)	AP - A company founded by a chemistry research	ky. company wins grant to study peptides (ap) 	[ky., company, wins, grant, to, study, peptide	{11738: 1, 60: 2, 407: 1, 4362: 2, 4: 2, 514:
3	4	Prediction Unit Helps Forecast Wildfires (AP)	AP - It's barely dawn when Mike Fitzpatrick st	prediction unit helps forecast wildfires (ap) 	[prediction, unit, helps, forecast, wildfires,	{9154: 1, 612: 1, 2049: 1, 705: 1, 0: 4, 14: 1
4	4	Calif. Aims to Limit Farm- Related Smog (AP)	AP - Southern California's smog-fighting agenc	calif. aims to limit farm- related smog (ap) ap	[calif., aims, to, limit, farm- related, smog, 	{1076: 1, 1713: 1, 4: 2, 3282: 1, 0: 4, 13226:
•••		•••		•••	•••	
7595	1	Around the world	Ukrainian presidential candidate Viktor Yushch	around the world ukrainian presidential candid	[around, the, world, ukrainian, presidential, 	{437: 1, 1: 2, 58: 1, 4477: 1, 389: 1, 1398: 1
7596	2	Void is filled with Clement	With the supply of attractive pitching options	void is filled with clement with the supply of	[void, is, filled, with, clement, with, the, s	{9010: 1, 24: 1, 4890: 1, 19: 3, 9831: 2, 1: 5
7597	2	Martinez leaves bitter	Like Roger Clemens did almost exactly eight ye	martinez leaves bitter like roger clemens did 	[martinez, leaves, bitter, like, roger, clemen	{2131: 2, 1790: 1, 2951: 2, 317: 1, 1698: 1, 2
7598	3	5 of arthritis patients in Singapore take Bext	SINGAPORE: Doctors in the United States have	5 of arthritis patients in singapore take bext	[5, of, arthritis, patients, in, singapore, ta	{627: 1, 6: 1, 2701: 1, 2439: 1, 7: 2, 1019: 2

	class index	title	description	text	tokens	features
7599	3	EBay gets into rentals	EBay plans to buy the apartment and home renta	ebay gets into rentals ebay plans to buy the a	[ebay, gets, into, rentals, ebay, plans, to, b	{1563: 2, 537: 1, 73: 1, 11111: 1, 160: 1, 4:

7600 rows × 6 columns

```
In [ ]: test_df['class index'] = pd.to_numeric(test_df['class index'], errors='coerd
In [ ]: X_test = np.stack(test_df['features'].progress_map(make_dense)) # convertime
        y_test = test_df['class index'].to_numpy() - 1 # convertimos las etiquetas a
        X_test = torch.tensor(X_test, dtype=torch.float32) # convertimos la matriz d
        y_test = torch.tensor(y_test) # convertimos el arreglo de etiquetas a un ter
                      | 0/7600 [00:00<?, ?it/s]
         0%|
In []: from sklearn.metrics import classification report # importamos la función cl
        # set model to evaluation mode
        model.eval()
        # don't store gradients
        with torch.no_grad(): # no almacenar gradientes
            X test = X test.to(device) # enviar la matriz de características al disp
            y_pred = torch.argmax(model(X_test), dim=1) # predecir las etiquetas
            y_pred = y_pred.cpu().numpy() # convertir las etiquetas a un arreglo de
            print(classification_report(y_test, y_pred, target_names=labels)) # most
                     precision
                                   recall f1-score
                                                      support
              World
                          0.95
                                     0.82
                                               0.88
                                                         1900
                          0.93
                                     0.98
                                               0.95
             Sports
                                                         1900
           Business
                          0.85
                                     0.86
                                               0.85
                                                         1900
           Sci/Tech
                          0.82
                                               0.85
                                     0.89
                                                         1900
                                               0.89
                                                         7600
           accuracy
                                               0.88
                                                         7600
          macro avq
                          0.89
                                     0.89
       weighted avg
                          0.89
                                     0.89
                                               0.88
                                                         7600
```