

```
In [1]: import pandas as pd
import numpy as np
```

```
import string
```

```
import nltk
from nltk.corpus import stopwords
nltk.download('stopwords')
```

```
from sklearn.model_selection import train_test_split
```

```
[nltk_data] Downloading package stopwords to
[nltk_data] /Users/prashanthsingaravelan/nltk_data...
[nltk_data] Package stopwords is already up-to-date!
```

```
In [2]: import tensorflow as tf
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
```

```
In [3]: from sklearn.preprocessing import LabelBinarizer
from tensorflow.keras.layers import Dense, Embedding, Input, LSTM, Dropout, GlobalMaxPool1D
from tensorflow.keras import Model
```

```
In [4]: import matplotlib.pyplot as plt
%matplotlib inline
```

```
In [5]: vocab_size = 1000
embedding_dim = 16
max_length = 120
trunc_type = 'post'
padding_type = 'post'
oov_tok = ""
training_portion = .8
```

```
In [6]: ## category is present
df_train = pd.read_csv('data-sets/BBC News Train.csv')
df_train.head()
```

```
Out[6]:
```

	ArticleId	Text	Category
0	1833	worldcom ex-boss launches defence lawyers defe...	business
1	154	german business confidence slides german busin...	business
2	1101	bbc poll indicates economic gloom citizens in ...	business
3	1976	lifestyle governs mobile choice faster bett...	tech
4	917	enron bosses in \$168m payout eighteen former e...	business

```
In [7]: ## category is not present -> We need to find
df_test = pd.read_csv('data-sets/BBC News Test.csv')
df_test.head()
```

```
Out[7]:
```

	ArticleId	Text
0	1018	qpr keeper day heads for preston queens park r...
1	1319	software watching while you work software that...
2	1138	d arcy injury adds to ireland woe gordon d arc...
3	459	india s reliance family feud heats up the ongo...
4	1020	boro suffer morrison injury blow middlesbrough...

Split between Dependent and Independent variables

```
In [8]: x = df_train['Text']
y = df_train['Category']
```

```
In [9]: print(x)
0      worldcom ex-boss launches defence lawyers defe...
1      german business confidence slides german busin...
2      bbc poll indicates economic gloom citizens in ...
3      lifestyle governs mobile choice faster bett...
4      enron bosses in $168m payout eighteen former e...
...
1485    double eviction from big brother model caprice...
1486    dj double act revamp chart show dj duo jk and ...
1487    weak dollar hits reuters revenues at media gro...
1488    apple ipod family expands market apple has exp...
1489    santy worm makes unwelcome visit thousands of ...
Name: Text, Length: 1490, dtype: object
```

```
In [10]: print(y)
0          business
1          business
2          business
3           tech
4          business
...
1485  entertainment
1486  entertainment
1487          business
1488           tech
1489           tech
Name: Category, Length: 1490, dtype: object
```

Y-Variable (Category feature) --> Tokenisation + Vectorizer

```
In [11]: lbl = LabelBinarizer()
y = lbl.fit_transform(y) ## category -> label encoding -> vector
```

```
In [12]: print(y)
[[1 0 0 0 0]
 [1 0 0 0 0]
 [1 0 0 0 0]
 ...
 [1 0 0 0 0]
 [0 0 0 0 1]
 [0 0 0 0 1]]
```

```
In [13]: print(y)
[[1 0 0 0 0]
 [1 0 0 0 0]
 [1 0 0 0 0]
 ...
 [1 0 0 0 0]
 [0 0 0 0 1]
 [0 0 0 0 1]]
```

Split between Train and Test Data

```
In [14]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2)
```

X-Variable (Category feature) --> Tokenisation + Vectorizer

```
In [15]: max_features = 1000
tokenizer = Tokenizer(num_words = max_features)
tokenizer.fit_on_texts(list(x))
```

```
In [16]: ### tokenisation + vectorization
tokenized_list_train = tokenizer.texts_to_sequences(x_train)
for i in range(4):
    print(tokenized_list_train[i])
    print("\n")
```

[47, 608, 137, 824, 47, 608, 875, 182, 1, 20, 222, 137, 3, 1, 115, 188, 468, 7, 3, 203, 824, 2, 441, 1, 110, 7, 114, 47, 566, 1, 46, 20, 39, 904, 2, 1, 4, 47, 657, 44, 1, 114, 566, 6, 1, 853, 3, 849, 4, 875, 407, 143, 3, 12, 4, 104, 49, 546, 3, 244, 70, 47, 6, 34, 41, 7, 503, 473, 193, 182, 147, 21, 148, 4, 72, 230, 1, 365, 38, 63, 6, 244, 4, 20, 149, 237, 1, 4, 4, 1, 29, 44, 1, 3, 5, 216, 139, 548, 6, 124, 4, 38, 910, 624, 5, 543, 64, 11, 14, 477, 6, 1, 72, 6, 530, 682, 13, 15, 248, 52, 508, 137, 3, 1, 3, 203, 503, 10, 122, 29, 239, 69, 164, 5, 505, 9, 5, 5, 15, 13, 50, 3, 1, 2, 1, 14, 1, 379, 110, 714, 46, 13, 29, 44, 24, 6, 861, 17, 3, 115, 138, 110, 7, 658, 1, 42, 13, 1, 407, 37, 248, 4, 5, 3, 1, 1, 25, 10, 47, 608, 731, 9, 136, 123, 2, 47, 15, 13, 578, 3, 110, 21, 193, 875, 299, 13, 339, 851, 27, 38, 607, 108, 102, 5, 829, 624, 1, 3, 1, 7, 658, 539, 5, 3, 42, 63, 158, 35, 30, 19, 32, 3, 8, 5, 910, 624, 13, 29, 44, 47, 608, 17, 32, 8, 6, 1, 3, 56, 1, 70, 2, 117, 161, 6, 1, 853, 830, 115, 299, 106, 137, 3, 1, 4, 115, 299, 125, 110, 781, 12, 1, 137, 172, 148, 875, 407, 147, 38, 346, 53, 4, 72, 6, 23, 0, 9, 974, 2, 1, 216, 49, 1, 96, 9, 248, 919, 7, 436, 104, 182, 280, 17, 5, 2, 395, 5, 6, 65, 41, 7, 853, 44, 2, 2, 4, 6, 1, 137, 169, 57, 3, 44, 895, 1, 110, 255, 21, 10, 445, 407, 44, 104, 46, 37, 1, 72, 7, 85, 911, 599, 3, 124, 44, 6, 1, 853, 1, 44, 636, 8, 8, 91, 29, 435, 44, 3, 1, 137, 6, 38, 751, 6, 3, 192, 56, 1, 58, 16, 27, 1, 13, 175, 4, 113, 923, 24, 426, 28, 20, 37, 5, 72, 137, 338, 624, 39, 5, 12, 5, 184, 142, 56, 199, 16, 8, 512, 4, 56, 93, 19, 5, 417, 151, 44, 134, 4, 2, 87, 1, 137, 172]

[73, 564, 18, 39, 6, 2, 793, 1, 292, 19, 1, 921, 42, 243, 145, 1, 208, 392, 224, 8, 31, 12, 686, 17, 84, 30, 13, 32, 36, 305, 446, 48, 40, 3, 5, 150, 61, 460, 560, 93, 19, 4, 969, 24, 30, 13, 921, 48, 189, 28, 164, 56, 245, 32, 1, 12, 1, 21, 19, 30, 1, 820, 2, 60, 1, 990, 32, 24, 30, 46, 19, 50, 3, 7, 4, 9, 589, 31, 205, 84, 292, 1, 775, 921, 122, 84, 292, 13, 15, 14, 28, 22, 30, 7, 1, 9, 10, 11, 9, 28, 146, 2, 15, 13, 107, 18, 37, 66, 4, 5, 190, 83, 2, 48, 3, 34, 60, 62, 55, 69, 872, 5, 212, 6, 10, 67, 11, 9, 5, 986, 212, 12, 556, 30, 22, 1, 775, 8, 4, 292, 66, 908, 463, 6, 5, 895, 42, 1, 292, 9, 89, 374, 22, 66, 908, 973, 46, 25, 28, 6, 78, 6, 1, 921, 29, 25, 6, 1, 107, 147, 61, 1, 578, 15, 13, 30, 9, 5, 640, 2, 1, 292, 53, 27, 84, 237, 15, 448, 427, 370, 352, 273, 1, 14, 446, 388, 1, 9, 221, 547, 2, 594, 324, 420, 332, 18, 39, 54, 292, 551, 393, 13, 15, 8, 5, 640, 3, 2, 793, 4, 56, 15, 122, 23, 793, 38, 15, 118, 1, 74, 5, 349, 3, 453, 399, 6, 1, 921, 1, 103, 16, 6, 1, 921, 19, 112, 15, 4, 5, 1, 281, 2, 933, 1, 775, 84, 292, 7, 453, 33, 33, 37, 2, 142, 78, 48, 44, 469, 1, 84, 292, 193, 427, 267, 8, 921, 122, 84, 13, 1, 292, 7, 281, 2, 94, 494, 1, 292, 73, 1, 4, 702, 29, 44, 4, 15, 254, 99, 21, 30, 7, 652, 10, 377, 2, 31, 571, 520, 46, 23, 6, 1, 179, 111, 234, 5, 230, 6, 256, 696, 3, 1, 510, 7, 6, 1, 68, 34, 41, 19, 75, 100, 494, 85, 3, 256, 68, 633, 45, 143, 31, 363, 754, 67, 2, 256]

[547, 2, 520, 1, 230, 3, 57, 67, 21, 1, 485, 1, 68, 633, 2, 112, 527, 3, 461, 6, 79, 352, 987, 18, 847, 1, 68, 216, 3, 6, 4, 589, 6, 54, 12, 1, 68, 216, 22, 106, 19, 15, 1, 230, 474, 2, 586, 7, 22, 452, 1, 696, 7, 67, 3, 9, 1, 465, 14, 79, 2, 1, 68, 216, 3, 4, 208, 68, 216, 3, 268, 7, 159, 14, 12, 968, 2, 514, 7, 216, 163, 24, 58, 28, 1, 61, 1, 243, 268, 7, 31, 567, 22, 444, 5, 159, 6, 256, 648, 46, 618, 696, 172, 633, 6, 361, 129, 7, 22, 79, 377, 2, 31, 571, 520, 46, 23, 6, 1, 179, 111, 234, 5, 230, 6, 256, 696, 3, 1, 510, 7, 6, 1, 68, 34, 41, 19, 75, 100, 494, 85, 3, 256, 68, 633, 45, 143, 31, 363, 754, 67, 2, 256]

[49, 400, 869, 1, 165, 49, 4, 18, 2, 40, 59, 32, 57, 67, 216, 309, 27, 1, 996, 8, 530, 702, 128, 125, 159, 2, 1, 04, 106, 159, 2, 12, 479, 500, 161, 3, 1, 869, 280, 27, 160, 439, 8, 372, 24, 1, 826, 951, 1, 414, 35, 770, 373, 18, 37, 420, 980, 4, 58, 264, 204, 1, 400, 869, 329, 59, 1, 906, 183, 12, 727, 5, 27, 1, 856, 22, 1, 414, 14, 80, 1, 871, 21, 106, 452, 5, 4, 5, 190, 59, 97, 1, 80, 1, 1, 414, 14, 128, 52, 377, 256, 21, 106, 1, 414, 7, 75, 9, 18, 39, 4, 308, 34, 105, 13, 3, 6, 332, 878, 486, 69, 37, 113, 980, 424, 350, 393, 191, 5, 406, 12, 1, 188, 326, 11, 14, 5, 3, 348, 264, 1, 214, 9, 619, 21, 93, 5, 486, 10, 11, 9, 3, 961, 76, 3, 35, 9, 166, 2, 587, 27, 126, 324, 1, 360, 136, 1, 49, 7, 348, 414, 595, 11, 18, 10, 11, 36, 16, 843, 60, 1, 1, 414, 18, 22, 546, 80, 1, 871, 19, 112, 19, 22, 491, 80, 1, 6, 1, 429, 91, 83, 1, 631, 114, 9, 1, 635, 418, 869, 12, 1, 50, 926, 4, 1, 40, 5, 869, 1, 165, 1, 3, 217, 6, 4, 54, 3, 1, 49, 12, 1, 70, 1, 400, 869, 9, 5, 635, 161, 3, 1, 6, 530, 1, 759, 6, 14, 333, 2, 5, 6, 192, 3, 4, 93, 19, 19, 112, 19, 3, 479, 4, 50, 499, 8, 49, 595, 14, 5, 6, 1, 869, 17, 315, 90, 2, 8, 550, 4, 70, 264, 324, 315, 7, 400, 9, 226, 2, 74, 309, 1, 996, 844, 1, 49, 7, 869, 17, 315, 14, 859, 6, 5, 30, 128, 27, 859, 1, 234, 97, 24, 1, 869, 17, 611, 14, 21, 42, 6, 40, 59, 185, 83]

```
In [17]: tokenized_list_test = tokenizer.texts_to_sequences(x_test)
for i in range(4):
    print(tokenized_list_test[i])
    print("\n")
```

[417, 9, 47, 18, 353, 12, 46, 23, 16, 1, 111, 975, 411, 5, 3, 635, 12, 417, 1, 133, 1, 559, 12, 538, 326, 275, 58, 20, 2, 1, 105, 3, 6, 1, 429, 102, 34, 20, 902, 39, 225, 267, 13, 1, 503, 768, 3, 138, 517, 22, 671, 21, 209, 125, 63, 296, 6, 1, 184, 419, 4, 18, 149, 134, 6, 93, 375, 19, 4, 1, 30, 13, 50, 37, 5, 3, 12, 1, 41, 200, 41, 7, 21, 3, 482, 106, 75, 90, 49, 75, 37, 76, 275, 75, 15, 37, 39, 2, 5, 24, 75, 69, 150, 15, 13, 704, 435, 2, 16, 6, 1, 589, 473, 307, 417, 4, 673, 307, 171, 3, 20, 3, 210, 106, 293, 4, 307, 20, 45, 39, 8, 1, 368, 6, 1, 429, 5, 635, 105, 3, 12, 417, 2, 367, 1, 63, 975, 44, 19, 5, 5, 234, 434, 1, 111, 81, 333, 8, 721, 6, 203, 18, 39, 391, 791, 333, 2, 644, 7, 3, 1, 1, 114, 332, 260, 231, 14, 477, 6, 640, 708, 4, 585, 87, 31, 243, 19, 1, 1, 417, 14, 333, 2, 87, 5, 19, 24, 14, 477, 27, 31, 6, 124]

[5, 12, 1, 18, 4, 635, 21, 6, 1, 786, 54, 65, 145, 5, 319, 346, 4, 6, 80, 1, 3, 981, 1, 876, 133, 35, 1, 169, 6, 1, 731, 18, 936, 32, 482, 359, 6, 1, 133, 84, 541, 36, 16, 340, 19, 5, 536, 3, 47, 981, 11, 13, 11, 36, 191, 34, 12, 20, 515, 29, 23, 506, 17, 1, 391, 1, 191, 32, 300, 12, 1, 1, 153, 18, 39, 6, 1, 4, 11, 58, 594, 32, 35, 36, 965, 171, 2, 441, 5, 4, 191, 32, 167, 2, 1, 76, 20, 28, 38, 21, 1, 4, 20, 1, 915, 3, 1, 226, 2, 1, 446, 935, 33, 25, 458, 19, 26, 239, 69, 277, 46, 9, 225, 24, 1, 25, 49, 13]

[4, 135, 49, 313, 4, 20, 143, 1, 855, 313, 8, 85, 4, 81, 38, 779, 34, 234, 7, 3, 640, 682, 622, 6, 37, 199, 131, 5, 396, 8, 409, 5, 6, 155, 414, 175, 794, 54, 155, 414, 4, 1, 22, 353, 1, 137, 778, 8, 85, 467, 1, 855, 49, 8, 1, 4, 184, 42, 902, 156, 12, 2, 135, 6, 70, 19, 1, 85, 651, 8, 100, 368, 19, 6, 1, 417, 262, 1, 85, 417, 396, 8, 409, 5, 778, 831, 521, 6, 155, 414, 62, 8, 622, 8, 86, 77, 4, 77, 4, 303, 49, 57, 86, 222, 698, 13, 31, 396, 6, 12, 579, 15, 45, 1, 81, 211, 62, 8, 8, 353, 5, 430, 17, 32, 461, 81, 62, 257, 140, 211, 3, 1, 41, 221, 14, 4, 13, 3, 8, 100, 211, 4, 621, 307, 26, 128, 2, 62, 13, 2, 1, 41, 200, 62, 25, 5, 60, 26, 219, 190, 1, 62, 25, 4, 190, 1, 62, 25, 64, 26, 219, 26, 23, 277, 10, 26, 328, 456, 248, 171, 4, 25, 71, 2, 16, 450, 1, 2, 94, 1, 959, 7, 778, 187, 18, 2, 5, 348, 27, 5, 8, 1, 89, 827, 8, 1, 959, 7, 137, 44, 384, 34, 145, 2, 540, 3, 1, 3, 4, 1, 9, 333, 2, 167, 12, 492, 169, 334, 97, 1, 998]

[509, 2, 305, 953, 389, 4, 48, 36, 132, 38, 205, 418, 8, 494, 953, 14, 860, 4, 174, 74, 186, 653, 150, 4, 48, 6, 50, 28, 786, 103, 16, 302, 2, 12, 38, 953, 4, 913, 6, 38, 205, 29, 45, 509, 5, 8, 218, 6, 425, 990, 2, 94, 438, 380, 698, 24, 1, 74, 659, 8, 386, 223, 573, 953, 51, 25, 106, 48, 953, 6, 130, 4, 653, 1, 105, 58, 22, 575, 7, 12, 953, 8, 6, 130, 9, 21, 5, 250, 35, 690, 162, 13, 34, 9, 1, 3, 1, 362, 11, 7, 52, 353, 399, 321, 27, 1, 362, 4, 11, 2, 4, 326, 10, 33, 23, 220, 62, 87, 196, 24, 33, 25, 28, 146, 2, 87, 82, 8, 62, 113, 40, 1, 74, 18, 199, 446, 2, 305, 48, 217, 77, 29, 55, 319, 8, 38, 218, 24, 117, 43, 3, 1, 18, 39, 653, 150, 1, 47, 186, 36, 87, 4, 412, 2, 6, 76, 446, 2, 144, 1, 29, 45, 164, 2, 108, 43, 5, 47, 190, 119, 360, 180, 786, 48, 132, 154, 217, 9, 6, 13, 8, 38, 953, 4, 220, 82, 132, 2, 943, 10, 418, 1, 23, 16, 12, 108, 391, 24, 30, 13, 1, 186, 58, 6, 76, 893, 93, 19, 43, 4, 15, 54, 386, 494, 953, 6, 130, 35, 9, 12, 300, 6, 428, 4, 251, 326, 11, 14, 449, 27, 1, 446, 74, 828, 13, 44, 380, 698, 12, 1, 3, 22, 1, 74, 4, 471, 323, 3, 1, 828, 3, 3, 218, 13, 1, 186, 58, 4, 112, 89, 8, 3, 48, 24, 796, 953, 14, 174, 4, 48, 44, 89, 936, 683, 6, 24, 28, 6, 42, 211, 253, 765, 13, 23, 28, 153, 60, 51, 25, 84, 218, 8, 48, 2, 27, 1, 463, 150, 48, 46, 319, 8, 91, 83, 223, 573, 953, 56, 163, 103, 16, 386, 953, 8, 1, 3, 38, 984, 427, 267, 13, 40, 59, 101, 223, 573, 953, 37, 39, 340, 149, 53, 641, 83, 3, 431, 3, 4, 74, 33, 24, 5, 279, 28, 5, 13, 30, 1, 570, 150, 29, 36, 803, 386, 494, 953, 22, 5, 47, 546, 207, 486, 12, 61, 101, 690, 267, 714, 13, 127, 4, 48, 2, 607, 6, 38, 205, 9, 5, 612, 33, 462, 24, 1, 9, 10, 55, 812, 73, 60, 1, 225, 410, 4, 9, 53, 9, 28, 613]

Padding

```
In [18]: maxlen = 100
x_train = pad_sequences(tokenized_list_train, maxlen = maxlen)
for i in range(4):
    print(x_train[i])
    print("\n")
```

```
[ 41   7 853   44   22   4    6    1 137 169   57   3  44 895    1 110 255   21
 10 445 407   44 104  46   37    1  72   7  85 911 599    3 124  44    6    1
853   1  44 636    8   8  91 29 435  44    3   1 137   6  38 751    6    3
192  56    1  58 16 27 113 175   4 113 923 24 426  28  20  37    5  72
137 338 624  39    5 12   5 184 142  56 199 16   8 512 456  93  19    5
417 151  44 134    4    2  87    1 137 172]
```

```
[399    6    1 921    1 103   16    6    1 921   19 112   15   45    1 281    2 933
   1 775   84 292    7 453   33   33   37    2 142   78   48  44 469    1  84 292
193 427 267    8 921 122   84 13    1 292   7 281    2  94 494    1 292  73
   1   4 702   29  44   4 15 254  99  21  30   7 652  10  32  36    6    1
 11  55 241   24 11  55  28  94  48 15 13  76  48  37  38  22 556    5
   3 627 256    1    8    9 111 686 210 530]
```

```
[   7   67    3  91 465   14   79    2    1  68 216    3    4 208   68 216    3 268
   7 159   14 12 968    2 514   7 216 163   24  58  28    1  61    1 243 268
   7  31 567 22 444    5 159    6 256 648  46 618 696 172 633    6 361 129
   7  22  79 377    2 31 571 520  46  23    6    1 179 111 234    5 230    6
256 696    3    1 510   7    6    1  68  34  41  19  75 100 494   85    3 256
 68 633   45 143   31 363 754   67    2 256]
```

```
[   3 217    6    4  54    3    1  49 12    1  70    1 400 869    9    5 635 161
   3    1    6 530    1 759    6 14 333    2    5    6 192    3    4  93 19 19
112 19    3 479    4  50 499    8  49 595 14    5    6    1 869 17 315 902
   8 550    4  70 264 324 315    7 400    9 226    2  74 309    1 996 844    1
 49   7 869 17 315 14 859    6 530 128 27 859    1 234  97  24    1 869
17 611 14  21  42    6  40  59 185  83]
```

```
In [19]: x_test = pad_sequences(tokenized_list_test, maxlen = maxlen)
for i in range(4):
    print(x_test[i])
    print("\n")
```

```
[ 15  13 704 435   2  16   6   1 589 473 307 417   4 673 307 171   3  20
   3 210 106 293   4 307  20  45  39   8   1 368   6   1 429   5 635 105
   3  12 417   2 367   1  63 975 44  19   5   5 234 434   1 111  81 333
   8 721   6 203  18  39 391 791 333   2 644   7   3   1   1 114 332 260
 231  14 477   6 640 708   4 585  87  31 243  19   1   1 417  14 333   2
   87   5  19  24  14 477  27  31   6 124]
```

```
[  6   1 731  18 936  32 482 359   6   1 133  84 541  36  16 340  19   5
536   3  47 981  11  13  11  36 191  34  12  20 515  29  23 506  17   1
391   1 191  32 300  12   1   1 153  18  39   6   1   4  11  58 594  32
 35  36 965 171   2 441   5   4 191  32 167   2   1  76  20  28  38  21
   1   4   20   1 915   3   1 226   2   1 446 935  33  25 458  19  26 239
 69 277  46   9 225  24   1  25  49  13]
```

```
[  3   1  41 221  14 413   3   8 100 211   4 621 307  26 128   2  62  13
   2   1  41 200  62  25   5  60  26 219 190   1  62  25   4 190   1  62
  25  64  26 219  26  23 277  10  26 328 456 248 171   4  25  71   2  16
450   1   2  94   1 959   7 778 187  18   2   5 348  27   5   8   1  89
827   8   1 959   7 137  44 384  34 145   2 540   3   1   3   4   1   9
333   2 167  12 492 169 334  97   1 998]
```

```
[150  48  46 319   8  91  83 223 573 953  56 163 103  16 386 953   8   1
   3  38 984 427 267  13  40  59 101 223 573 953  37  39 340 149  53 641
  83   3 431   3   4  74  33 245 279  28   5  13  30   1 570 150  29  36
803 386 494 953  22   5  47 546 207 486  12  61 101 690 267 714  13 127
   4  48   2 607   6  38 205   9   5 612  33 462  24   1   9  10  55 812
  73  60   1 225 410   4 953   9  28 613]
```

Model Creation

```
In [20]: embed_size = 128
def create_model():
    input = Input(shape=(maxlen, ))
    x = Embedding(max_features, embed_size)(input)
    x = LSTM(64, return_sequences=True)(x)
    x = GlobalMaxPool1D()(x)

    x = Dense(50, activation="relu")(x)
    x = Dropout(0.3)(x)
    x = Dense(5, activation="softmax")(x)

    return Model(inputs=input, outputs=x)
```

```
In [21]: model = create_model()
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
```

```
In [22]: model.summary()
```

Model: "model"

Layer (type)	Output Shape	Param #
=====		
input_1 (InputLayer)	[(None, 100)]	0
embedding (Embedding)	(None, 100, 128)	128000
lstm (LSTM)	(None, 100, 64)	49408
global_max_pooling1d (GlobalMaxPooling1D)	(None, 64)	0
dense (Dense)	(None, 50)	3250
dropout (Dropout)	(None, 50)	0
dense_1 (Dense)	(None, 5)	255
=====		
Total params: 180,913		
Trainable params: 180,913		
Non-trainable params: 0		

Fitting the Data into the model

```
In [23]: hist = model.fit(x_test, y_train, batch_size=32, epochs=12, validation_split=0.1)
```

```
hist
```

Epoch 1/12

2022-11-14 22:12:49.125593: W tensorflow/core/platform/profile_utils/cpu_utils.cc:128] Failed to get CPU frequency: 0 Hz

9/9 [=====] - 1s 66ms/step - loss: 1.6098 - accuracy: 0.1978 - val_loss: 1.6145 - val_accuracy: 0.2000

Epoch 2/12

9/9 [=====] - 0s 43ms/step - loss: 1.5946 - accuracy: 0.3060 - val_loss: 1.6179 - val_accuracy: 0.2000

Epoch 3/12

9/9 [=====] - 0s 42ms/step - loss: 1.5845 - accuracy: 0.3470 - val_loss: 1.6232 - val_accuracy: 0.1333

Epoch 4/12

9/9 [=====] - 0s 43ms/step - loss: 1.5757 - accuracy: 0.3470 - val_loss: 1.6239 - val_accuracy: 0.2333

Epoch 5/12

9/9 [=====] - 0s 43ms/step - loss: 1.5593 - accuracy: 0.4104 - val_loss: 1.6282 - val_accuracy: 0.2333

Epoch 6/12

9/9 [=====] - 0s 42ms/step - loss: 1.5344 - accuracy: 0.4328 - val_loss: 1.6306 - val_accuracy: 0.2333

Epoch 7/12

9/9 [=====] - 0s 41ms/step - loss: 1.5055 - accuracy: 0.5149 - val_loss: 1.6614 - val_accuracy: 0.2000

Epoch 8/12

9/9 [=====] - 0s 43ms/step - loss: 1.4750 - accuracy: 0.4142 - val_loss: 1.6385 - val_accuracy: 0.2000

Epoch 9/12

9/9 [=====] - 0s 43ms/step - loss: 1.4270 - accuracy: 0.4590 - val_loss: 1.7030 - val_accuracy: 0.2333

Epoch 10/12

9/9 [=====] - 0s 43ms/step - loss: 1.3225 - accuracy: 0.5746 - val_loss: 1.6615 - val_accuracy: 0.2000

Epoch 11/12

9/9 [=====] - 0s 43ms/step - loss: 1.2072 - accuracy: 0.5933 - val_loss: 1.7216 - val_accuracy: 0.2333

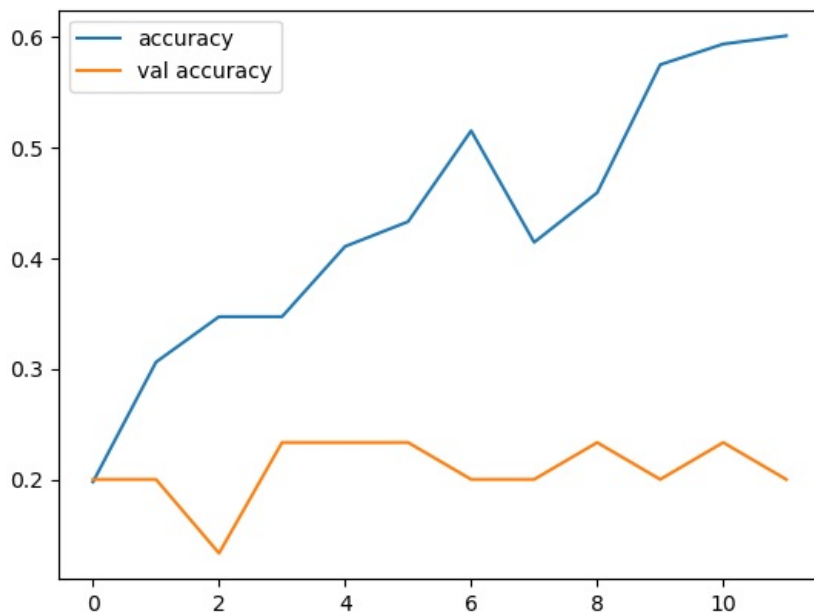
Epoch 12/12

9/9 [=====] - 0s 45ms/step - loss: 1.0568 - accuracy: 0.6007 - val_loss: 2.4154 - val_accuracy: 0.2000

Out[23]: <keras.callbacks.History at 0x17ca9bca0>

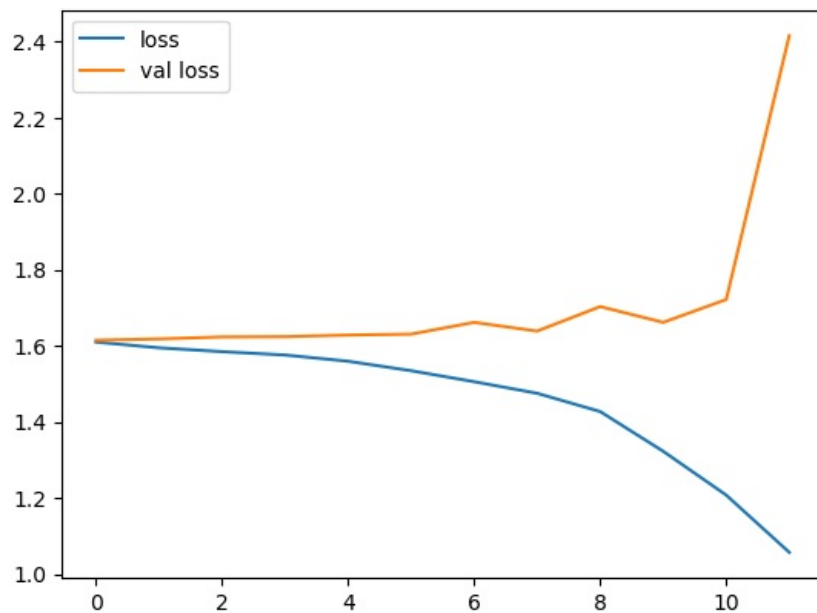
Accuracy Comparison

```
In [24]: plt.plot(hist.history['accuracy'], label='accuracy')
plt.plot(hist.history['val_accuracy'], label='val accuracy')
plt.legend()
plt.show()
```



Loss function comparison

```
In [25]: plt.plot(hist.history['loss'], label='loss')
plt.plot(hist.history['val_loss'], label='val loss')
plt.legend()
plt.show()
```



Predicting the model with the test data

```
In [26]: y_pred = model.predict(x_test, batch_size=32, verbose=1)
print(y_pred.shape)
```

```
10/10 [=====] - 0s 12ms/step
(298, 5)
```

```
In [27]: df_probs_all = pd.DataFrame(y_pred, columns=['business', 'entertainment', 'politics', 'sport', 'tech'])
df_probs_all.head()
```

```
Out[27]:
```

	business	entertainment	politics	sport	tech
0	0.112278	0.563073	0.129609	0.093734	0.101306
1	0.262064	0.286884	0.224246	0.081935	0.144870
2	0.006315	0.110508	0.004065	0.838584	0.040527
3	0.947978	0.018360	0.021687	0.000209	0.011766
4	0.037807	0.457525	0.028691	0.387221	0.088757

Cateogry with all the probabilities

```
In [28]: def top_5_predictions(df, N):
    cols = df.columns[:-1].tolist()
    a = df[cols].to_numpy().argsort()[:, :-N-1:-1]
    c = np.array(cols)[a]
    d = df[cols].to_numpy()[np.arange(a.shape[0])[:, None], a]
    df1 = pd.DataFrame(c).rename(columns=lambda x : f'max_{x+1}_col')
    predicted_genres = df1["max_1_col"] + ' ' + df1["max_2_col"] + ' ' + df1["max_3_col"] + ' ' + df1["max_4_col"]
    return predicted_genres
```

```
In [29]: pred_gen = top_5_predictions(df_probs_all, 5)
pred_gen
```

```
Out[29]:
```

0	entertainment politics business sport
1	entertainment business politics sport
2	sport entertainment business politics
3	business politics entertainment sport
4	entertainment sport business politics
...	...
293	entertainment sport business politics
294	sport entertainment business politics
295	business entertainment politics sport
296	business entertainment politics sport
297	business entertainment politics sport

Length: 298, dtype: object

```
In [30]: final_results = pd.DataFrame(columns=['text', 'keywords'])
```

```
In [31]: for index, txt in enumerate(x_test):
    final_results = final_results.append({'text': txt, 'keywords': pred_gen[index]}, ignore_index=True)
```

```
/var/folders/gq/nsqxf83n1813yysq2l8vvtxc0000gn/T/ipykernel_3632/3026006548.py:2: FutureWarning: The frame.append method is deprecated and will be removed from pandas in a future version. Use pandas.concat instead.
    final_results = final_results.append({'text': txt, 'keywords': pred_gen[index]}, ignore_index=True)
/var/folders/gq/nsqxf83n1813yysq2l8vvtxc0000gn/T/ipykernel_3632/3026006548.py:2: FutureWarning: The frame.append
```


[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

```
In [32]: final_results
```

Out[32]:

		text	keywords
0	[15, 13, 704, 435, 2, 16, 6, 1, 589, 473, 307,...	entertainment politics business sport	
1	[6, 1, 731, 18, 936, 32, 482, 359, 6, 1, 133, ...	entertainment business politics sport	
2	[3, 1, 41, 221, 14, 413, 3, 8, 100, 211, 4, 62...	sport entertainment business politics	
3	[150, 48, 46, 319, 8, 91, 83, 223, 573, 953, 5...	business politics entertainment sport	
4	[35, 11, 73, 1, 529, 137, 91, 3, 252, 1, 70, 6...	entertainment sport business politics	
...
293	[2, 16, 259, 4, 97, 10, 45, 18, 5, 619, 105, 3...	entertainment sport business politics	
294	[5, 6, 975, 7, 24, 15, 1, 6, 4, 53, 107, 7, 6,...	sport entertainment business politics	
295	[631, 14, 508, 27, 4, 5, 248, 857, 19, 15, 64,...	business entertainment politics sport	
296	[5, 6, 343, 33, 6, 1, 79, 26, 474, 585, 44, 12...	business entertainment politics sport	
297	[466, 1, 182, 13, 11, 37, 1, 149, 4, 36, 621, ...	business entertainment politics sport	

298 rows × 2 columns

```
In [33]: test_data = [""
The US says China will face harsh "consequences" if it aids Russia in its invasion of Ukraine, according to US
Unnamed officials reportedly told multiple US news outlets that Russia asked China to provide military assistance
The Chinese embassy in Washington said it was not aware of this request.
The warning comes ahead of a meeting in Rome on Monday between top US and China officials.
Since the start of the crisis Beijing has expressed strong rhetorical support for long-time ally Moscow, but is
However local media outlets citing US officials, say that Russia has in recent days asked China specifically for
In a CNN interview, US National Security Adviser Jake Sullivan said they were "communicating directly, privately"
"We will not allow that to go forward and allow there to be a lifeline to Russia from these economic sanctions.
He added that while the US believed China was aware that Russian leader Vladimir Putin was "planning something"
"Because it's very possible that [Mr] Putin lied to them the same way that he lied to Europeans and others," Mr
Mr Sullivan is due to meet Yang Jiechi, a member of China's top decision-making body the Politburo and the head
Reuters news agency quoted a US official as saying that during the meeting Mr Sullivan will spell out the consequences
Liu Pengyu, a spokesman for China's embassy in Washington DC, told US media that he had not heard of Russia's request
China so far has refrained from condemning Russia for the invasion, and has said Moscow's "legitimate security
Chinese state media and government officials, besides echoing Russia's official line that it is a "special military
But Beijing at the same time has expressed "unwavering support" for Ukraine's sovereignty. It has also called for
""]
```

```
In [34]: tokenizer = Tokenizer(num_words=max_features)
tokenizer.fit_on_texts(test_data)
tokenized_list_real = tokenizer.texts_to_sequences(test_data)
```

```
In [35]: x_real = pad_sequences(tokenized_list_real, maxlen=maxlen)
real_preds = model.predict(x_real)

1/1 [=====] - 0s 11ms/step
```

```
In [36]: real_preds = pd.DataFrame(real_preds,columns=['business', 'entertainment', 'politics', 'sport', 'tech'])
real_preds
```

Out[36]:

	business	entertainment	politics	sport	tech
0	0.150024	0.283391	0.123433	0.243319	0.199833