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"import matplotlib.pyplot as plt\n",
"from sklearn.model_selection import train_test_split\n",
"from sklearn.preprocessing import LabelEncoder\n",
"from keras.models import Model\n",
"from keras.layers import LSTM, Activation, Dense, Dropout, Input, Embedding\n",
"from keras.optimizers import Adam\n",
"from keras.preprocessing.text import Tokenizer\n",
"from keras.preprocessing import sequence\n",
"from keras.utils import pad_sequences\n",
"from keras.utils import to_categorical\n",
"from keras.callbacks import EarlyStopping"
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```
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[illegible]

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0.15);\n",
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"  }\n",
" </style>\n",
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```



```

"\n",
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  "        const dataTable =\n",
  "            await google.colab.kernel.invokeFunction('convertToInteractive',\n",
  "                [key], {});\n",
  "        if (!dataTable) return;\n",
  "\n",
  "        const docLinkHtml = 'Like what you see? Visit the ' +\n",
  "            '<a target=\"_blank\" href=https://colab.research.google.com/notebooks/data_table.ipynb>data table notebook</a>'\n",
  "            + ' to learn more about interactive tables.';\n",
  "        element.innerHTML = \"\n",
  "            dataTable['output_type'] = 'display_data';\n",
  "            await google.colab.output.renderOutput(dataTable, element);\n",
  "            const docLink = document.createElement('div');\n",
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        "from wordcloud import WordCloud, STOPWORDS, ImageColorGenerator\n",
        "X = df.v2\n",
        "Y = df.v1\n",
        "le = LabelEncoder()\n",
        "Y = le.fit_transform(Y)\n",
        "Y = Y.reshape(-1,1)\n",
        "X_train,X_test,Y_train,Y_test = train_test_split(X,Y,test_size=0.25)\n",
        "max_words = 1000\n",
        "max_len = 150\n",
        "tok = Tokenizer(num_words=max_words)\n",
        "tok.fit_on_texts(X_train)\n",
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        "Epoch 4/15\n",
        "168/168 [=====] - 33s 196ms/step - loss: 0.0097 - accuracy: 0.9973 - val_loss: 0.1148 - val_accuracy: 0.9761\n",
        "Epoch 5/15\n",
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```
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  "168/168 [=====] - 30s 179ms/step - loss: 0.0070 - accuracy:  
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  "Epoch 9/15\n",  
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    "def plot_graphs1(var1, var2, string):\n",
    "    metrics[[var1, var2]].plot()\n",
    "    plt.title('Training and Validation ' + string)\n",
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}  
]  
},  
{  
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  "source": [  
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```



```
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    "test_sequences_matrix = pad_sequences(test_sequences,maxlen=max_len)"
  ],

```

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    },
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  },
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    }
  ]
},
{
```

```
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    ]
  }
]
```