221501111_221501104_NLP PROJECT MODEL TRAINING

```
!pip install "nvidia-cublas-cu12==12.4.5.8" "nvidia-cuda-runtime-cu12==12.4.127" "nvidia-cudnn-cu12==9.1.0.70" --force-reinstall
Collecting nvidia-cublas-cu12==12.4.5.8
       Downloading nvidia_cublas_cu12-12.4.5.8-py3-none-manylinux2014_x86_64.whl.metadata (1.5 kB)
     Collecting nvidia-cuda-runtime-cu12==12.4.127
       Downloading nvidia_cuda_runtime_cu12-12.4.127-py3-none-manylinux2014_x86_64.whl.metadata (1.5 kB)
     Collecting nvidia-cudnn-cu12==9.1.0.70
       Downloading nvidia_cudnn_cu12-9.1.0.70-py3-none-manylinux2014_x86_64.whl.metadata (1.6 kB)
     Downloading nvidia_cublas_cu12-12.4.5.8-py3-none-manylinux2014_x86_64.whl (363.4 MB)
                                                      363.4/363.4 MB 4.1 MB/s eta 0:00:00
     Downloading nvidia_cuda_runtime_cu12-12.4.127-py3-none-manylinux2014_x86_64.whl (883 kB)
                                                      883.7/883.7 kB 23.2 MB/s eta 0:00:00
     Downloading nvidia_cudnn_cu12-9.1.0.70-py3-none-manylinux2014_x86_64.whl (664.8 MB)
                                                       664.8/664.8 MB 2.3 MB/s eta 0:00:00
     Installing collected packages: nvidia-cuda-runtime-cu12, nvidia-cublas-cu12, nvidia-cudnn-cu12
       Attempting uninstall: nvidia-cuda-runtime-cu12
         Found existing installation: nvidia-cuda-runtime-cu12 12.5.82
          Uninstalling nvidia-cuda-runtime-cu12-12.5.82:
            Successfully uninstalled nvidia-cuda-runtime-cu12-12.5.82
       Attempting uninstall: nvidia-cublas-cu12
          Found existing installation: nvidia-cublas-cu12 12.5.3.2
          Uninstalling nvidia-cublas-cu12-12.5.3.2:
            Successfully uninstalled nvidia-cublas-cu12-12.5.3.2
       Attempting uninstall: nvidia-cudnn-cu12
          Found existing installation: nvidia-cudnn-cu12 9.3.0.75
          Uninstalling nvidia-cudnn-cu12-9.3.0.75:
            Successfully uninstalled nvidia-cudnn-cu12-9.3.0.75
     ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the source
     torch 2.6.0+cu124 requires nvidia-cuda-cupti-cu12==12.4.127; platform_system == "Linux" and platform_machine == "x86_64", but you have n torch 2.6.0+cu124 requires nvidia-cuda-nvrtc-cu12==12.4.127; platform_system == "Linux" and platform_machine == "x86_64", but you have n
     torch 2.6.0+cu124 requires nvidia-cufft-cu12==11.2.1.3; platform_system == "Linux" and platform_machine == "x86_64", but you have nvidia
     torch 2.6.0+cu124 requires nvidia-curand-cu12==10.3.5.147; platform_system == "Linux" and platform_machine == "x86_64", but you have nvi torch 2.6.0+cu124 requires nvidia-cusolver-cu12==11.6.1.9; platform_system == "Linux" and platform_machine == "x86_64", but you have nvi
     torch 2.6.0+cu124 requires nvidia-cusparse-cu12==12.3.1.176; platform_system == "Linux" and platform_machine == "x86_64", but you have n torch 2.6.0+cu124 requires nvidia-nvjitlink-cu12==12.4.127; platform_system == "Linux" and platform_machine == "x86_64", but you have nv
     Successfully installed nvidia-cublas-cu12-12.4.5.8 nvidia-cuda-runtime-cu12-12.4.127 nvidia-cudnn-cu12-9.1.0.70
!pip install -q datasets
!pip install lime
₹
                                                     - 491.2/491.2 kB 13.0 MB/s eta 0:00:00
                                                     - 116.3/116.3 kB 5.8 MB/s eta 0:00:00
                                                     - 183.9/183.9 kB 12.8 MB/s eta 0:00:00
                                                     - 143.5/143.5 kB 12.5 MB/s eta 0:00:00
                                                     - 194.8/194.8 kB 16.5 MB/s eta 0:00:00
     ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the source
     torch 2.6.0+cu124 requires nvidia-cuda-cupti-cu12==12.4.127; platform_system == "Linux" and platform_machine == "x86_64", but you have n torch 2.6.0+cu124 requires nvidia-cuda-nvrtc-cu12==12.4.127; platform_system == "Linux" and platform_machine == "x86_64", but you have n
     torch 2.6.0+cu124 requires nvidia-cufft-cu12==11.2.1.3; platform_system == "Linux" and platform_machine == "x86_64", but you have nvidia
     torch 2.6.0+cu124 requires nvidia-curand-cu12==10.3.5.147; platform_system == "Linux" and platform_machine == "x86_64", but you have nvi torch 2.6.0+cu124 requires nvidia-cusolver-cu12==11.6.1.9; platform_system == "Linux" and platform_machine == "x86_64", but you have nvi
     torch 2.6.0+cu124 requires nvidia-cusparse-cu12==12.3.1.170; platform_system == "Linux" and platform_machine == "x86_64", but you have n
     torch 2.6.0+cu124 requires nvidia-nvjitlink-cu12==12.4.127; platform_system == "Linux" and platform_machine == "x86_64", but you have nv
     gcsfs 2025.3.2 requires fsspec==2025.3.2, but you have fsspec 2024.1\overline{2.0} which is incompatible.
     Collecting lime
       Downloading lime-0.2.0.1.tar.gz (275 kB)
                                                       - 275.7/275.7 kB 7.2 MB/s eta 0:00:00
       Preparing metadata (setup.py) ... done
     Requirement already satisfied: matplotlib in /usr/local/lib/python3.11/dist-packages (from lime) (3.10.0)
     Requirement already satisfied: numpy in /usr/local/lib/python3.11/dist-packages (from lime) (2.0.2)
     Requirement already satisfied: scipy in /usr/local/lib/python3.11/dist-packages (from lime) (1.14.1)
     Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from lime) (4.67.1)
     Requirement already satisfied: scikit-learn>=0.18 in /usr/local/lib/python3.11/dist-packages (from lime) (1.6.1)
     Requirement already satisfied: scikit-image>=0.12 in /usr/local/lib/python3.11/dist-packages (from lime) (0.25.2)
     Requirement already satisfied: networkx>=3.0 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (3.4.2)
     Requirement already satisfied: pillow>=10.1 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (11.1.0)
     Requirement already satisfied: imageio!=2.35.0,>=2.33 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (2.37.0
     Requirement already satisfied: tifffile>=2022.8.12 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (2025.3.30
     Requirement already satisfied: packaging>=21 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (24.2)
     Requirement already satisfied: lazy-loader>=0.4 in /usr/local/lib/python3.11/dist-packages (from scikit-image>=0.12->lime) (0.4)
     Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn>=0.18->lime) (1.4.2)
     Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn>=0.18->lime) (3.6.0)
     Requirement already satisfied: contourpy>=1.0.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib->lime) (1.3.1)
     Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.11/dist-packages (from matplotlib->lime) (0.12.1)
```

```
Requirement already satisfied: fonttools>=4.22.0 in /usr/local/lib/python3.11/dist-packages (from matplotlib->lime) (4.57.0)
     Requirement already satisfied: kiwisolver>=1.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib->lime) (1.4.8)
     Requirement already satisfied: pyparsing>=2.3.1 in /usr/local/lib/python3.11/dist-packages (from matplotlib->lime) (3.2.3)
     Requirement already satisfied: python-dateutil>=2.7 in /usr/local/lib/python3.11/dist-packages (from matplotlib->lime) (2.8.2)
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.7->matplotlib->lime) (1.17.0
     Building wheels for collected packages: lime
       Building wheel for lime (setup.py) ... done
       {\tt Created \ wheel \ for \ lime: filename=lime-0.2.0.1-py3-none-any.whl \ size=283834 \ sha256=263403eb25b10bc70bef09efbfb545d7e6484ad9d365fe472897} }
       Stored in directory: /root/.cache/pip/wheels/85/fa/a3/9c2d44c9f3cd77cf4e533b58900b2bf4487f2a17e8ec212a3d
     Successfully built lime
     Installing collected packages: lime
     Successfully installed lime-0.2.0.1
import os
os.environ["WANDB_DISABLED"] = "true"
import pandas as pd
import numpy as np
import torch
from sklearn.model_selection import train_test_split
from transformers import BertTokenizer, BertForSequenceClassification, Trainer, TrainingArguments
from transformers import DataCollatorWithPadding
from datasets import Dataset
from \ sklearn.metrics \ import \ accuracy\_score, \ precision\_recall\_fscore\_support
from lime.lime_text import LimeTextExplainer
import matplotlib.pyplot as plt
from google.colab import drive
drive.mount('/content/drive')
→ Mounted at /content/drive
import os
model_dir = '/content/drive/MyDrive/fake_news_models'
os.makedirs(model_dir, exist_ok=True)
# Upload your CSVs from Files -> Upload
from google.colab import files
uploaded = files.upload()
# Load datasets
fake_df = pd.read_csv('Fake.csv')
true_df = pd.read_csv('True.csv')
# Add labels: 0 - Fake, 1 - Real
fake_df['label'] = 0
true_df['label'] = 1
df = pd.concat([fake_df[['text', 'label']], true_df[['text', 'label']]], axis=0)
df = df.sample(frac=1).reset_index(drop=True)
# Train-test split
train_df, test_df = train_test_split(df, test_size=0.2, random_state=42)
 ₹
     Choose Files No file chosen
                                       Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to
     enable.
tokenizer = BertTokenizer.from_pretrained("bert-base-uncased")
def tokenize_function(example):
    return tokenizer(example["text"], truncation=True)
```

train_dataset = Dataset.from_pandas(train_df)

You will be able to reuse this secret in all of your notebooks.

Please note that authentication is recommended but still optional to access public models or datasets.

warnings.warn(

tokenizer_config.json: 100%

48.0/48.0 [00:00<00:00, 5.11kB/s]

vocab.txt: 100%

232k/232k [00:00<00:00, 1.63MB/s]

tokenizer.json: 100%

466k/466k [00:00<00:00, 3.15MB/s]

config.json: 100%

570/570 [00:00<00:00, 38.9kB/s]

Map: 100%

35918/35918 [04:29<00:00, 133.47 examples/s]

Map: 100%

8980/8980 [01:07<00:00, 132.88 examples/s]

model = BertForSequenceClassification.from_pretrained("bert-base-uncased", num_labels=2)

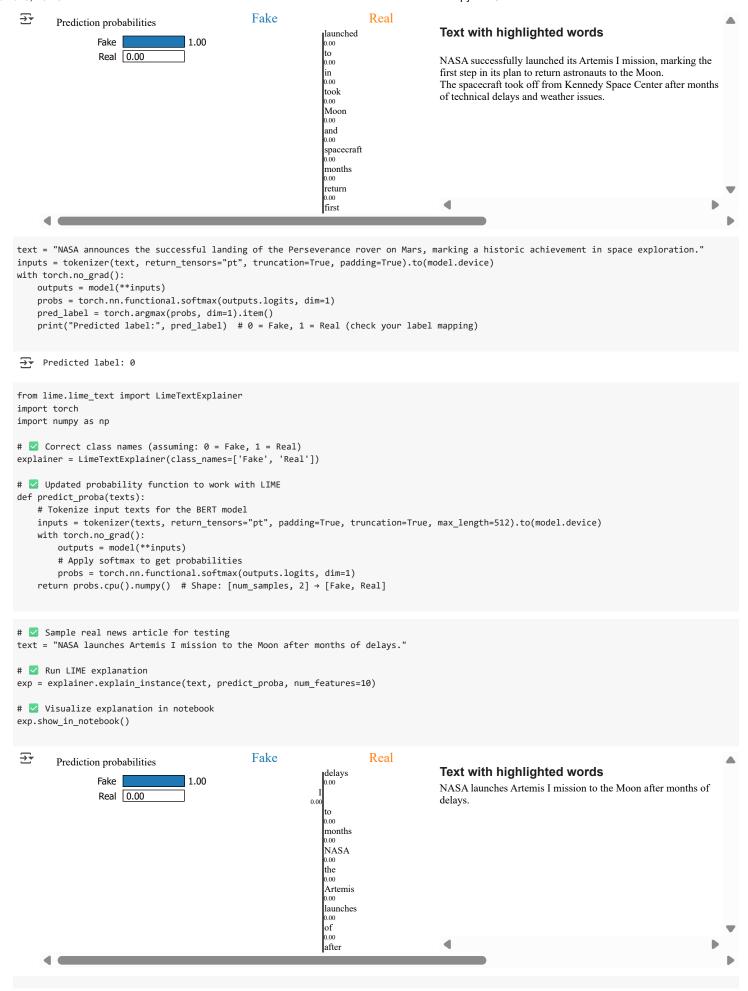
Xet Storage is enabled for this repo, but the 'hf_xet' package is not installed. Falling back to regular HTTP download. For better perfc WARNING:huggingface_hub.file_download:Xet Storage is enabled for this repo, but the 'hf_xet' package is not installed. Falling back to r model.safetensors: 100%

440M/440M [00:05<00:00, 114MB/s]

Some weights of BertForSequenceClassification were not initialized from the model checkpoint at bert-base-uncased and are newly initiali You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.

```
from sklearn.metrics import precision_recall_fscore_support, accuracy_score
import numpy as np
from transformers import TrainingArguments, Trainer, EarlyStoppingCallback
def compute metrics(eval pred):
   logits, labels = eval_pred
   predictions = np.argmax(logits, axis=-1)
   precision, recall, f1, _ = precision_recall_fscore_support(labels, predictions, average='binary')
   acc = accuracy_score(labels, predictions)
   return {
       "accuracy": acc,
       "f1": f1,
        "precision": precision,
        "recall": recall
   }
training_args = TrainingArguments(
   output_dir="./results",
   evaluation_strategy="epoch",
   learning_rate=2e-5,
   per_device_train_batch_size=16,
   per_device_eval_batch_size=16,
   num_train_epochs=3, # you can go up to 10 epochs since early stopping is enabled
   weight_decay=0.01,
   load_best_model_at_end=True,
                                     # 🔽 loads best model (based on val metric)
   metric_for_best_model="f1",
                                       # ☑ choose best by f1 score
                                      # 🔽 maximize F1
   greater_is_better=True,
   report_to="none",
                                       # ☑ disable W&B or other loggers
   save_strategy="epoch",
                                      # 🔽 save every epoch
   logging_strategy="epoch"
)
trainer = Trainer(
   model=model,
   args=training_args,
   train_dataset=train_tokenized,
   eval_dataset=test_tokenized,
   tokenizer=tokenizer,
   data_collator=data_collator,
   compute metrics=compute metrics.
   callbacks=[EarlyStoppingCallback(early_stopping_patience=2)] # ☑ stops if no improvement in 2 evals
```

```
/usr/local/lib/python3.11/dist-packages/transformers/training_args.py:1611: FutureWarning: `evaluation_strategy` is deprecated and will
       warnings.warn(
     <ipython-input-12-758431d95ecc>:33: FutureWarning: `tokenizer` is deprecated and will be removed in version 5.0.0 for `Trainer.__init__
       trainer = Trainer(
trainer.train()
<del>∑</del>₹
                                           ■ [6735/6735 3:11:56, Epoch 3/3]
      Epoch Training Loss Validation Loss Accuracy F1
                                                                Precision Recall
                  0.009000
                                                                  1.000000 0.999529
          1
                                   0.002041 0.999777 0.999764
         2
                  0.001100
                                   0.002291 0.999777 0.999764
                                                                  0.999764 0.999764
          3
                  0.000000
                                   0.002276 0.999777 0.999764
                                                                  0.999764 0.999764
     TrainOutput(global_step=6735, training_loss=0.0033688701711368277, metrics={'train_runtime': 11519.6917, 'train_samples_per_second':
model_save_path = "/content/drive/MyDrive/fake_news_models"
trainer.save_model(model_save_path)
tokenizer.save_pretrained(model_save_path)
('/content/drive/MyDrive/fake_news_models/tokenizer_config.json',
      '/content/drive/MyDrive/fake_news_models/special_tokens_map.json',
      '/content/drive/MyDrive/fake_news_models/vocab.txt',
      '/content/drive/MyDrive/fake_news_models/added_tokens.json')
trainer.evaluate()
→
                                           [562/562 04:50]
     {'eval_loss': 0.002290518721565604,
      eval_accuracy': 0.9997772828507795,
      'eval_f1': 0.9997644842204427,
      'eval_precision': 0.9997644842204427,
      'eval_recall': 0.9997644842204427,
      'eval_runtime': 291.3658,
      'eval_samples_per_second': 30.82,
      'eval_steps_per_second': 1.929,
       annch' ? al
from lime.lime_text import LimeTextExplainer
explainer = LimeTextExplainer(class_names=['Fake', 'Real'])
def predict_proba(texts):
   inputs = tokenizer(texts, return_tensors="pt", padding=True, truncation=True).to(model.device)
   with torch.no_grad():
       outputs = model(**inputs)
       probs = torch.nn.functional.softmax(outputs.logits, dim=1)
   return probs.cpu().numpy()
text =text = """
NASA successfully launched its Artemis I mission, marking the first step in its plan to return astronauts to the Moon.
The spacecraft took off from Kennedy Space Center after months of technical delays and weather issues.
exp = explainer.explain_instance(text, predict_proba, num_features=10)
exp.show_in_notebook()
```



```
def predict_proba(texts):
    inputs = tokenizer(texts, return_tensors="pt", padding=True, truncation=True, max_length=512).to(model.device)
    with torch.no_grad():
        outputs = model(**inputs)
        probs = torch.nn.functional.softmax(outputs.logits, dim=1)
    return\ probs.cpu().numpy()[:,\ [1,\ 0]]\ \ \#\ Flip\ columns:\ [Real,\ Fake]\ \rightarrow\ [Fake,\ Real]
probs = predict_proba([text])
predicted_label = np.argmax(probs[0])
print(f"Predicted Label: {'Fake' if predicted_label == 0 else 'Real'}")
print(f"Confidence: {probs[0][predicted_label]:.4f}")
→ Predicted Label: Real
     Confidence: 1,0000
# 🗹 Sample real news article for testing
text = "Scientists confirm chocolate cures COVID-19 in new breakthrough study."
# 🗹 Run LIME explanation
exp = explainer.explain_instance(text, predict_proba, num_features=10)
# 🔽 Visualize explanation in notebook
exp.show_in_notebook()
₹
                                               Fake
                                                                      Real
        Prediction probabilities
                                                                                    Text with highlighted words
                Fake 0.00
                                                                                    Scientists confirm chocolate cures COVID-19 in new
                                                              COVID
                                  1 00
                Real
                                                                                    breakthrough study.
                                                         study
                                                       confirm
                                                      chocolate
                                                         cure
                                                              breakthrough
                                                     Scientists
print(train_dataset[0]) # See the label and example
 🚁 {'text': 'If you ve been living under a rock for the past few days, you re probably unaware that Donald Trump Jr. committed treason. Aft
label_map = {0: "Fake", 1: "Real"}
class names = ['Fake', 'Real']
def predict_label(text):
    inputs = tokenizer(text, return_tensors="pt", truncation=True, padding=True).to(model.device)
    with torch.no_grad():
        outputs = model(**inputs)
        probs = torch.nn.functional.softmax(outputs.logits, dim=1)
        predicted_class = torch.argmax(probs, dim=1).item()
        label = label_map[predicted_class]
        confidence = probs[0][predicted_class].item()
    return label, confidence
from lime.lime_text import LimeTextExplainer
class_names = ['Fake', 'Real'] # index 0 = Fake, 1 = Real
explainer = LimeTextExplainer(class_names=class_names)
def predict_proba(texts):
    inputs = tokenizer(texts, return_tensors="pt", padding=True, truncation=True).to(model.device)
```

```
with torch.no_grad():
       outputs = model(**inputs)
       probs = torch.nn.functional.softmax(outputs.logits, dim=1)
   return probs.cpu().numpy()
<del>_</del>__
                                               Traceback (most recent call last)
     <ipython-input-6-85b5e70124a5> in <cell line: 0>()
     ----> 1 from lime.lime_text import LimeTextExplainer
           3 class_names = ['Fake', 'Real'] # index 0 = Fake, 1 = Real
           4 explainer = LimeTextExplainer(class_names=class_names)
     ModuleNotFoundError: No module named 'lime'
     NOTE: If your import is failing due to a missing package, you can
     manually install dependencies using either !pip or !apt.
     To view examples of installing some common dependencies, click the
     "Open Examples" button below.
     OPEN EXAMPLES
text = "NASA launches Artemis I mission to the Moon after months of delays."
label, confidence = predict_label(text)
print(f"Prediction: {label} (Confidence: {confidence:.2f})")
exp = explainer.explain_instance(text, predict_proba, num_features=10)
exp.show_in_notebook()
<del>_</del>_
     NameError
                                               Traceback (most recent call last)
     <ipython-input-1-22e62369a7cc> in <cell line: 0>()
          1 text = "NASA launches Artemis I mission to the Moon after months of delays."
     ----> 2 label, confidence = predict_label(text)
           3 print(f"Prediction: {label} (Confidence: {confidence:.2f})")
           5 # Run LIME
     NameError: name 'predict_label' is not defined
```