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Start coding or generate with AI.
import pandas as pd
import random
from sklearn.model_selection import train_test_split
# Load the datasets
fake_df = pd.read_csv("Fake.csv")
true_df = pd.read_csv("True.csv")
# Add a label column
fake_df['label'] = 'Fake'
true_df['label'] = 'Real'
# Combine the datasets
data = pd.concat([fake_df[['title', 'text', 'label']], true_df[['title', 'text', 'label']]], ignore_index=True)
# Shuffle the data
data = data.sample(frac=1).reset_index(drop=True)
# Split into training, validation, and test sets
train, temp = train_test_split(data, test_size=0.4, random_state=42)
val, test = train_test_split(temp, test_size=0.5, random_state=42)
print(f"Training size: {len(train)}, Validation size: {len(val)}, Test size: {len(test)}")
→ Training size: 26938, Validation size: 8980, Test size: 8980
import random
def prepare(n):
    global train
    # Randomly sample n rows
    examples = []
    sampled_rows = train.sample(n) # Randomly sample n rows
    for _, row in sampled_rows.iterrows():
        examples.append(f"{row['title']} => {row['label']}")
    return "\n".join(examples)
# Example:
prepared_examples = prepare(3)
# Display the prepared examples
print("Prepared Examples:\n")
print(prepared_examples)
→ Prepared Examples:
     Trump to host Thai prime minister on October 2: White House => Real
     The U.S. Establishment vs The Rest of World => Fake
     End 'containment' of asylum-seekers on islands, aid groups tell Greek PM => Real
def generate_prompt(n, example_row):
    # Generate a prompt with n examples from the training set and one example from the validation/test
    # Prepare n examples from the training set
    few_shot_examples = prepare(n)
    # Append the example from the validation/test set
    validation_example = f"{example_row['title']} =>"
    # Combine into a single prompt
    prompt = f"{few_shot_examples}\n{validation_example}"
    return prompt
# Example
example_row = val.iloc[0] # Take the first row from the validation set
n = 3 # Number of training examples to include
prompt = generate_prompt(n, example_row)
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# Display the generated prompt
print("Generated Prompt:\n")
print(prompt)
→ Generated Prompt:
     'SLEEPY' JUSTICE GINSBURG: Excites Crowd By Saying She'd Back Abolition Of The Electoral College [Video] => Fake
      It's Not Over Yet: Jill Stein Files Federal Lawsuit Over Pennsylvania Recount => Fake
     Trump meets insurers, promises catastrophic year for Obamacare => Real
     Lockheed Martin wins $582 million U.S. defense contract: Pentagon =>
from transformers import GPT2Tokenizer, GPT2LMHeadModel
import torch
# Load GPT-2 model and tokenizer
tokenizer = GPT2Tokenizer.from_pretrained("gpt2")
model = GPT2LMHeadModel.from_pretrained("gpt2")
def gpt2_generate(prompt, max_tokens=1):
    #Generate a prediction using GPT-2 based on the given prompt.
   # Encode the prompt
   inputs = tokenizer.encode(prompt, return_tensors="pt")
   # Generate response
   outputs = model.generate(
        inputs.
        max_new_tokens=max_tokens, # Number of tokens to generate
       do_sample=True, # sampling for randomness
        temperature=0.7, # control randomness
       top_k=50, # top-k sampling for diversity
        pad_token_id=tokenizer.eos_token_id # set pad_token_id to eos_token_id
    # Decode and return the output
   return tokenizer.decode(outputs[0][inputs.size(-1):], skip_special_tokens=True).strip()
example_prompt = generate_prompt(3, val.iloc[0]) # Generate a prompt with 3 examples from training set
prediction = gpt2 generate(example prompt)
# Display the prediction
print("Generated Prediction:\n")
print(prediction)

→ Generated Prediction:
     Fake
from sklearn.metrics import classification_report
def score(n, dataset):
    #Scores GPT-2's performance on a given dataset using n examples for few-shot training.
   # Store predictions and true labels
    predictions = []
   ground_truths = []
    # Iterate through each row in the dataset
    for index, row in dataset.iterrows():
        # Generate the prompt
       prompt = generate_prompt(n, row)
        # Generate the prediction
        prediction = gpt2_generate(prompt)
        # Print the prompt and prediction
        print(f"Prompt:\n{prompt}\nPrediction: {prediction}\n")
        # Append the prediction and ground truth
        predictions.append(prediction)
        ground_truths.append(row['label'])
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# return classification metrics
   metrics = classification_report(ground_truths, predictions, output_dict=True, zero_division=0)
   return metrics
# Reduce the dataset for speed
small val = val.sample(n=10, random state=42)
# Example usage with reduced dataset
n = 3
metrics = score(n, small_val)
print(f"Validation Metrics for n = {n}:\n", metrics)
EP #17: Patrick Henningsen LIVE - 'Parallax Politics in DC' with guest Daniel Faraci => Fake
     Trump Lawyer LITERALLY Argued That Muslim Ban Shouldn't Be Based In Reality, And Twitter Is LOSING IT (VIDEO/TWEETS) => Fake
    Former House Speaker Hastert seeks sentencing delay due to poor health =>
    Prediction: Fake
     Trump's Deplorables Freak Out, Demand We Repeal Women's Right To Vote (TWEETS) => Fake
    LEBRON JAMES Brags About Being "Well-Spoken"...Proceeds To Use Vocabulary Of A 4-Yr-Old To Trash President Trump => Fake
    Tennessee lawmaker ousted after sex harassment allegations: media => Real
    EU drone rules in balance as member states hold off endorsing deal =>
    Prediction: Fake
    Prompt:
    Spain arrests four former Venezuelan officials for U.S. probe => Real
    Crowd controversy: The making of an Inauguration Day photo => Real
    Sessions confirmed as U.S. attorney general after battle with Democrats => Real
    THE JACK BLOOD SHOW: 'From May Day Riots to Globalism' with 21WIRE guest Shawn Helton =>
    Prediction: @
    Prompt:
    CNBC DEBATE HACK Proves Allegiance To Democrat Party With This Tweet Celebrating Mass Murder => Fake
    OUCH! BERNIE SANDERS Responds To Hillary's Criticism...Slams Her With Category-5 Zinger [VIDE0] => Fake
     Former KKK Leader Thanks Trump's VP For Being So Nice (AUDIO) => Fake
    Democrats link U.S. debt limit vote to Republican tax cut moves =>
    Prediction: Obama
    Prompt:
    Wheelbarrow bomb kills man pushing it in Somalia's Puntland, police say => Real
    Trump says Brexit to be 'a great thing', wants quick trade deal with UK => Real
    Elite Nazi-allied Order From Hungary Claims Trump Adviser Sebastian Gorka Is Sworn Member => Fake
    SYRIAN MUSLIM MAN WHOSE Family Perished On Trip So He Could Get Free Dental Care Has New Spokesperson Role =>
    Prediction: Real
    Prompt:
    SCOUNDREL HILLARY SUPPORTER STARTS "TrumpLeaks" Campaign...Desperate Move! => Fake
    U.S. tax debate employing more than half of Washington's lobbyists: report => Real
    [VIDEO] OUR RACIST PRESIDENT INVITES MUSLIMS To Join Blacks In Victim Pool While Celebrating Ramadan at White House => Fake
     Elizabeth Warren Calls Trump A 'Two-Bit Dictator,' Slams His RNC Speech =>
    Prediction: Real
    Prompt:
    Philippine president's Senate foes, allies vow to block budget cut for rights body => Real
    GERMANY'S DEFENSE MINISTER Refuses To Wear Hijab During Saudi Arabia Visit...Says Trump's Election Proves Political Correctness Has B
    WOW! BRITISH ACTRESS HAMMERS EU Leaders: "Every one of you who said refugees are welcome, You Are Responsible For Brussels...Europe i
     Bernie Sanders Just Received Some Excellent News About His Campaign =>
    Prediction: Fake
    Prompt:
    LT GEN JOHN KELLY'S INCREDIBLE SPEECH Just Days After His Own Son Was Killed In Action: "Brave Men On Watch All Over the World Toni
    Greg Gutfeld Scores a Priceless TV Moment...Wears CNN Box During 'The Five' [Video] => Fake
     Those Intel Reports That Nunes Ran To Trump With Reveal A BOMBSHELL That Trump Will HATE => Fake
    Tight race as South Africa's ANC prepares to elect Zuma successor =>
    Prediction: Fake
    Validation Metrics for n = 3:
     # Evaluate performance for n in a range
results = {}
range_n = range(1, 6) # n (1 to 5)
# compute validation metrics
for n in range n:
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print(f"Evaluating for n = {n}")
metrics = score(n, small_val)
results[n] = metrics

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# accuracy, precision, recall, and F1-score
    fake_metrics = metrics.get("Fake", {"precision": 0, "recall": 0, "f1-score": 0})
    real_metrics = metrics.get("Real", {"precision": 0, "recall": 0, "f1-score": 0})
    accuracy = metrics.get("accuracy", 0)
    # display summary for this value of n
    print(f"Summary for n = {n}:")
    print(f" Accuracy: {accuracy:.2f}")
    print(f" Fake News - Precision: {fake_metrics['precision']:.2f}, Recall: {fake_metrics['recall']:.2f}, F1-Score: {fake_metrics['f1-score
    print(f" Real News - Precision: {real_metrics['precision']:.2f}, Recall: {real_metrics['recall']:.2f}, F1-Score: {real_metrics['f1-score
    print("-" * 40)
# Identify the best n based on F1-score (weighted avg)
optimal\_n = max(results, \; key=lambda \; n: \; results[n]["weighted \; avg"]["f1-score"])
print(f"\nOptimal n based on validation performance: {optimal_n}")
# Evaluate on the test set with the optimal n
print(f"\nEvaluating on test set with n = {optimal_n}")
test_metrics = score(optimal_n, small_test) # Use a reduced test set for speed
# test results
test_fake_metrics = test_metrics.get("Fake", {"precision": 0, "recall": 0, "f1-score": 0})
test_real_metrics = test_metrics.get("Real", {"precision": 0, "recall": 0, "f1-score": 0})
test_accuracy = test_metrics.get("accuracy", 0)
# display summary
print("\nTest Set Summary:")
print(f" Accuracy: {test_accuracy:.2f}")
print(f" Fake News - Precision: {test_fake_metrics['precision']:.2f}, Recall: {test_fake_metrics['recall']:.2f}, F1-Score: {test_fake_metrics['precision']:.2f}
print(f" Real News - Precision: {test_real_metrics['precision']:.2f}, Recall: {test_real_metrics['recall']:.2f}, F1-Score: {test_real_metrics['precision']:.2f},
print("-" * 40)
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12/2/24, 10:25 PM h6.ipynb - Colab

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Inis Kansas Republican Inought Touting 'Protound' Hitler Quote Was A Good Idea - It Wasn't =>
Prediction: Fake

Test Set Summary:
    Accuracy: 0.50
    Fake News - Precision: 0.63, Recall: 0.63, F1-Score: 0.63
    Real News - Precision: 0.38, Recall: 0.30, F1-Score: 0.33
```

Summary for n = 1: Accuracy: 0.30 Fake News - Precision: 0.67, Recall: 0.40, F1-Score: 0.50 Real News - Precision: 1.00, Recall: 0.20, F1-Score: 0.33

Summary for n = 2: Accuracy: 0.40 Fake News - Precision: 0.50, Recall: 0.20, F1-Score: 0.29 Real News - Precision: 0.75, Recall: 0.60, F1-Score: 0.67

Summary for n = 3: Accuracy: 0.40 Fake News - Precision: 0.75, Recall: 0.60, F1-Score: 0.67 Real News - Precision: 0.33, Recall: 0.20, F1-Score: 0.25

Summary for n = 4: Accuracy: 0.40 Fake News - Precision: 0.50, Recall: 0.40, F1-Score: 0.44 Real News - Precision: 0.50, Recall: 0.40, F1-Score: 0.44

Summary for n = 5: Accuracy: 0.60 Fake News - Precision: 0.67, Recall: 0.40, F1-Score: 0.50 Real News - Precision: 0.57, Recall: 0.80, F1-Score: 0.67

Optimal n based on validation performance: 5

Evaluating on test set with n = 5

Test Set Summary: Accuracy: 0.50 Fake News - Precision: 0.63, Recall: 0.63, F1-Score: 0.63 Real News - Precision: 0.38, Recall: 0.30, F1-Score: 0.33

Part 2:

- 1. What I usually find works well is trimming the description of the problem to contain only essential requirements and break them into numbered steps. Specify the input and output and provide a sample input/output, also specify an approach, for example: "solve using Greedy/DP if ideal"
- 2. GPT often generate good stories and high quality cover letters if the prompt includes details and personal backgrounds, because many existing dataset/patterns already exists. It's not very accurate at mathematical proof because unless we are proving something that have already been proved many times before in textbooks, it's hard to follow a strict, precise logic without generating extra/creative texts that may not be suited for a proof.