PG6: Use a pre-trained Hugging Face model to analyze sentiment in text. Assume a real-world application, Load the sentiment analysis pipeline. Analyze the sentiment by giving sentences to input.

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Soln:
%pip install --upgrade --quiet huggingface_hub
%pip install --upgrade langchain
from transformers import pipeline
# Load the sentiment analysis pipeline
sentiment_analyzer = pipeline("sentiment-analysis")
# Example sentences for analysis
sentences = [
  "The product quality is amazing! I'm very satisfied.",
  "I had a terrible experience with customer service.",
  "The delivery was quick, but the packaging was damaged.",
  "Absolutely love this! Best purchase I've made.",
  "Not worth the money, very disappointed."
1
# Analyze sentiment for each sentence
results = sentiment_analyzer(sentences)
# Print the results
for sentence, result in zip(sentences, results):
  print(f"Sentence: {sentence}\nSentiment: {result['label']}, Confidence:
{result['score']:.2f}\n")
Output:
Sentence: The product quality is amazing! I'm very satisfied.
Sentiment: POSITIVE, Confidence: 1.00
Sentence: I had a terrible experience with customer service.
Sentiment: NEGATIVE, Confidence: 1.00
Sentence: The delivery was quick, but the packaging was damaged.
Sentiment: NEGATIVE, Confidence: 1.00
```

```
Sentence: Absolutely love this! Best purchase I've made.
Sentiment: POSITIVE, Confidence: 1.00
Sentence: Not worth the money, very disappointed.
Sentiment: NEGATIVE, Confidence: 1.00
Results
Output:
[{'label': 'POSITIVE', 'score': 0.9998825788497925},
 {'label': 'NEGATIVE', 'score': 0.9993104934692383},
 {'label': 'NEGATIVE', 'score': 0.9997345805168152},
 {'label': 'POSITIVE', 'score': 0.9998751878738403},
 {'label': 'NEGATIVE', 'score': 0.9998034834861755}]
!pip install langchain-huggingface
Approach 2: Using API calls
from langchain_huggingface import HuggingFaceEndpoint
# get a token: https://huggingface.co/docs/api-inference/quicktour#get-your-api-token
from getpass import getpass
HUGGINGFACEHUB_API_TOKEN = getpass()
import os
os.environ["HUGGINGFACEHUB_API_TOKEN"] = HUGGINGFACEHUB_API_TOKEN
from langchain.chains import LLMChain
from langchain_core.prompts import PromptTemplate
text = ["The product quality is amazing! I'm very satisfied.",
  "I had a terrible experience with customer service.",
  "The delivery was quick, but the packaging was damaged.",
  "Absolutely love this! Best purchase I've made.",
  "Not worth the money, very disappointed."]
template = """Perform the sentiment analysis for the following: {text}.
Answer: Following is the sentiment for the given text:"""
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prompt = PromptTemplate.from_template(template)

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repo_id = "meta-llama/Llama-3.2-3B-Instruct" #"mistralai/Mistral-7B-Instruct-v0.2"
llm = HuggingFaceEndpoint(
    repo_id=repo_id,
    max_length=256,
    temperature=0.5,
    huggingfacehub_api_token=HUGGINGFACEHUB_API_TOKEN,
)
llm_chain = prompt | llm
print(llm_chain.invoke({"text": text}))
Output:
    ["Positive", "Negative", "Negative", "Negative"]
```

Explanation: The sentiment analysis is done by using a pre-trained sentiment analysis model. The model is trained on a large dataset of text and is able to identify the sentiment of a given text. Here is how the sentiment is analyzed for each text:

- 1. "The product quality is amazing! I'm very satisfied." The text contains positive words like 'amazing' and 'satisfied', hence the sentiment is positive.
- 2. "I had a terrible experience with customer service." The text contains negative words like 'terrible', hence the s entiment is negative.
- 3. "The delivery was quick, but the packaging was damaged." The text contains both positive ('quick') and negative ('damaged') words, hence the sentiment is neutral.
- 4. "Absolutely love this! Best purchase I've made." The text contains positive words like 'love' and 'best', hence the sentiment is positive.
- 5. "Not worth the money, very disappointed." The text contains negative words like 'not worth' and 'disappointed', he nce the sentiment is negative.