```
# Import necessary libraries
 import pandas as pd
  import matplotlib.pyplot as plt
  import seaborn as sns
 #Load the CSV file into a DataFrame
file_path = '/content/dataset.csv'
 df = pd.read_csv(file_path)
 #Examine the structure of the DataFrame
print("Dataset Shape:", df.shape)
print("\nDataset Info:")
print(df.info())
 #Check for null values in each column
print("\nNull Values in Each Column:")
print(df.isnull().sum())
 #Visualize the distribution of query lengths
 # Calculate the number of words in each query
\label{eq:df['query_length'] = df['Output'].apply(lambda x: len(str(x).split()))} $$ df['query_length'] = df['Output'].apply(lambda x: len(str(x).split())) $$ df['query_length'] = df['Query_length'] = df['Query_length'].apply(lambda x: len(str(x).split())) $$ df['query_length'] = df['query_length'].apply(lambda x: length').apply(lambda x: length'
plt.figure(figsize=(10, 6))
  sns.histplot(df['query_length'], kde=True, bins=30)
 plt.title("Distribution of Query Lengths")
plt.xlabel("Number of Words")
plt.ylabel("Frequency")
 plt.show()
```

```
→ Dataset Shape: (1867, 3)
    Dataset Info:
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 1867 entries, 0 to 1866
    Data columns (total 3 columns):
                    Non-Null Count Dtype
     # Column
    ---
        -----
        User Query 1867 non-null
                                    object
                    1867 non-null
     1
         Output
                                    object
         Annotation 1867 non-null
    dtypes: object(3)
    memory usage: 43.9+ KB
    None
    Null Values in Each Column:
    User Query
                  0
    Output
                  0
    Annotation
    dtype: int64
```

Distribution of Query Lengths 250 200 100 100 100 Number of Words

```
import re
import nltk
nltk.download('stopwords')
nltk.download('punkt')
nltk.download('wordnet')
nltk.download('punkt_tab')
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
from nltk.stem import WordNetLemmatizer
from sklearn.model_selection import train_test_split
# Define stopwords and lemmatizer
stop_words = set(stopwords.words('english'))
lemmatizer = WordNetLemmatizer()
# Function to clean text
def clean_text(text):
    # Convert text to lowercase
    text = text.lower()
    # Remove punctuation and special characters
    text = re.sub(r'[^a-z\s]', '', text)
    # Tokenize the text into words
    tokens = word_tokenize(text)
    # Remove stopwords
    tokens = [word for word in tokens if word not in stop_words]
    # Lemmatize each token
    tokens = [lemmatizer.lemmatize(word) for word in tokens]
```

```
\mbox{\tt\#} Rejoin tokens to form the cleaned text
   cleaned_text = ' '.join(tokens)
   return cleaned_text
# Apply the cleaning function to the 'User Query' column
df['cleaned_query'] = df['User Query'].apply(clean_text)
# clean the 'Output' column too
df['cleaned_output'] = df['Output'].apply(lambda x: x.lower().strip())
# Data Splitting: 80% for training and 20% for testing
train_df, test_df = train_test_split(df, test_size=0.2, random_state=42)
# Display basic information about the splits
print("Training set shape:", train_df.shape)
print("Testing set shape:", test_df.shape)
[nltk\_data] \quad \textit{Unzipping corpora/stopwords.zip.} \\
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk_data] Unzipping tokenizers/punkt.zip.
     [nltk_data] Downloading package wordnet to /root/nltk_data...
     [nltk_data] Downloading package punkt_tab to /root/nltk_data...
     [nltk_data] Unzipping tokenizers/punkt_tab.zip.
     Training set shape: (1493, 6)
     Testing set shape: (374, 6)
# Install spaCy with transformer support
!pip install -U spacy[transformers]
# Download the transformer-based English model
!python -m spacy download en_core_web_trf
```

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```
DOWNTOAUTING CUTTACEU_CTTAINSTOTTHET'S-ש.ב.ו.ד-pyz.pyp-none-any.wiit (בס גד)
     Installing collected packages: curated-tokenizers, curated-transformers, spacy-curated-transformers, en-core-web-trf
     Successfully installed curated-tokenizers-0.0.9 curated-transformers-0.1.1 en-core-web-trf-3.8.0 spacy-curated-transformers-0.3.0
     \ensuremath{\checkmark} Download and installation successful
     You can now load the package via spacy.load('en_core_web_trf')
      \Delta Restart to reload dependencies
     If you are in a Jupyter or Colab notebook, you may need to restart Python in
     order to load all the package's dependencies. You can do this by selecting the
     'Restart kernel' or 'Restart runtime' ontion.
import spacy
# Load spaCy's transformer-based English model
nlp = spacy.load("en_core_web_trf")
# Process a sample query
sample_query = df['Output'].iloc[0]
doc = nlp(sample_query)
print("Tokenization & POS Tagging:")
for token in doc:
    print(f"{token.text:12s} {token.pos_:10s} {token.tag_}")
→ Tokenization & POS Tagging:
                   ADP
                   NOUN
     beginners
                                NNS
                   PUNCT
                                VB
     start
                   VERB
     with
                    ADP
                                ΙN
     20-30
                   NUM
                                CD
                   NOUN
     minutes
                                NNS
                    ADP
                                IN
     moderate
                   ADJ
                                JJ
     cardio
                   NOUN
                                NN
     like
                   ADP
                                IN
     brisk
                   ADJ
                                JJ
     walking
                   NOUN
                                NN
                   PUNCT
     followed
                   VERB
                                VBN
                   ADP
                                IN
     by
     basic
                   ADJ
                                33
     bodyweight
                    NOUN
                                NN
     exercises
                    NOUN
                                NNS
                   ADJ
     such
                                JJ
     as
                    ADP
                                IN
     squats
                   NOUN
                                NNS
                    PUNCT
     push
                                NN
                   NOUN
                   PUNCT
                                HYPH
                    NOUN
                                NNS
     ups
                    PUNCT
                                ĆC
                   CCONT
     and
     planks
                    NOUN
                                NNS
                    PUNCT
#Chunks and dependency relation
print("\nNoun Chunks in the query:")
for chunk in doc.noun_chunks:
    print(f" - {chunk.text}")
print("\nDependency Parsing:")
    print(f"{token.text:12s} --> {token.dep_:10s} --> {token.head.text}")
<del>_</del>
     Noun Chunks in the query:
      - beginners
      - 20-30 minutes
      - moderate cardio
      - brisk walking
      - basic bodyweight exercises
      - squats
      - push-ups
      - planks
     Dependency Parsing:
     For
                   --> prep
                                  --> start
     beginners
                   --> pobj
                                  --> For
```

```
start
                  --> ROOT
                                 --> start
    with
                  --> prep
                                 --> start
    20-30
                  --> nummod
                                 --> minutes
                                 --> with
    minutes
                  --> pobj
                                 --> minutes
    of
                  --> prep
    moderate
                  --> amod
                                 --> cardio
                                 --> of
    cardio
                  --> pobj
                  --> prep
    like
                                 --> cardio
    brisk
                  --> amod
                                 --> walking
    walking
                 --> pobi
                                 --> like
                  --> punct
                                 --> minutes
    followed
                                 --> minutes
                  --> acl
                  --> agent
                                 --> followed
    by
    basic
                                 --> exercises
                  --> amod
    bodyweight
                 --> compound
                                --> exercises
    exercises
                  --> pobj
                                 --> by
    such
                  --> amod
                                 --> as
    as
                 --> prep
                                 --> exercises
    squats
                  --> pobj
                                 --> as
                  --> punct
                                 --> squats
                                 --> ups
    push
                 --> compound
                  --> punct
                                 --> ups
    ups
                  --> conj
                                 --> squats
                  --> punct
                                 --> ups
    and
                  --> cc
                                 --> ups
    planks
                  --> coni
                                 --> ups
                  --> punct
                                 --> start
print("\nNamed Entities Found in the Query:")
if doc.ents:
    for ent in doc.ents:
       print(f"{ent.text:12s} ({ent.label_})")
   print("No entities found in the sample query.")
    Named Entities Found in the Query:
    20-30 minutes (TIME)
# Install sentence-transformers
!pip install sentence-transformers
Requirement already satisfied: sentence-transformers in /usr/local/lib/python3.11/dist-packages (3.4.1)
    Requirement already satisfied: transformers<5.0.0,>=4.41.0 in /usr/local/lib/python3.11/dist-packages (from sentence-transformers) (4.49
    Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from sentence-transformers) (4.67.1)
    Requirement already satisfied: torch>=1.11.0 in /usr/local/lib/python3.11/dist-packages (from sentence-transformers) (2.6.0+cu124)
    Requirement already satisfied: scikit-learn in /usr/local/lib/python3.11/dist-packages (from sentence-transformers) (1.6.1)
    Requirement already satisfied: scipy in /usr/local/lib/python3.11/dist-packages (from sentence-transformers) (1.14.1)
    Requirement already satisfied: huggingface-hub>=0.20.0 in /usr/local/lib/python3.11/dist-packages (from sentence-transformers) (0.29.3)
    Requirement already satisfied: Pillow in /usr/local/lib/python3.11/dist-packages (from sentence-transformers) (11.1.0)
    Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.20.0->sentence-transformers)
    Requirement already satisfied: fsspec>=2023.5.0 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.20.0->sentence-trans
    Requirement already satisfied: packaging>=20.9 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.20.0->sentence-transf
    Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.20.0->sentence-transforme
    Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.20.0->sentence-transformers)
    Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.20.0->sent
    Requirement already satisfied: networkx in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-transformers) (3.4.2)
    Requirement already satisfied: jinja2 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-transformers) (3.1.6)
    Requirement already satisfied: nvidia-cuda-nvrtc-cu12==12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence
    Requirement already satisfied: nvidia-cuda-runtime-cu12==12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->senten
    Requirement already satisfied: nvidia-cuda-cupti-cu12==12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence
    Requirement already satisfied: nvidia-cudnn-cu12==9.1.0.70 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-tran
    Requirement already satisfied: nvidia-cublas-cu12==12.4.5.8 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-tra
    Requirement already satisfied: nvidia-cufft-cu12==11.2.1.3 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-tran
    Requirement already satisfied: nvidia-curand-cu12==10.3.5.147 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-t
    Requirement already satisfied: nvidia-cusolver-cu12==11.6.1.9 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-t
    Requirement already satisfied: nvidia-cusparse-cu12==12.3.1.170 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence
    Requirement already satisfied: nvidia-cusparselt-cu12==0.6.2 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-tr
    Requirement already satisfied: nvidia-nccl-cu12==2.21.5 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-transfc
    Requirement already satisfied: nvidia-nvtx-cu12==12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-trans
    Requirement already satisfied: nvidia-nvjitlink-cu12==12.4.127 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-
    Requirement already satisfied: triton==3.2.0 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-transformers) (3.2
    Requirement already satisfied: sympy==1.13.1 in /usr/local/lib/python3.11/dist-packages (from torch>=1.11.0->sentence-transformers) (1.1
    Requirement already satisfied: mpmath<1.4,>=1.1.0 in /usr/local/lib/python3.11/dist-packages (from sympy==1.13.1->torch>=1.11.0->sentence
```

Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.11/dist-packages (from transformers<5.0.0,>=4.41.0->sentence-transf Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.11/dist-packages (from transformers<5.0.0,>=4.41.0->sentence-Requirement already satisfied: tokenizers<0.22,>=0.21 in /usr/local/lib/python3.11/dist-packages (from transformers<5.0.0,>=4.41.0->sent Requirement already satisfied: safetensors>=0.4.1 in /usr/local/lib/python3.11/dist-packages (from transformers<5.0.0,>=4.41.0->sentence Requirement already satisfied: joblib>=1.2.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn->sentence-transformers) (1.4. Requirement already satisfied: threadpoolctl>=3.1.0 in /usr/local/lib/python3.11/dist-packages (from scikit-learn->sentence-transformers

--> start

--> punct

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Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.11/dist-packages (from jinja2->torch>=1.11.0->sentence-transfor Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub>=0.20 Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub>=0.20.0->sentence Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub>=0.20.0->se Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests->huggingface-hub>=0.20.0->se

```
from sentence_transformers import SentenceTransformer

# Load the Sentence-BERT model
model = SentenceTransformer('all-MiniLM-L6-v2')

# Generate an embedding for the sample query
embedding = model.encode(sample_query)
print("\nSentence-BERT Embedding for the sample query:")
print(embedding)
```

```
/usr/local/lib/python3.11/dist-packages/huggingface_hub/utils/_auth.py:94: UserWarning:
The secret `HF_TOKEN` does not exist in your Colab secrets.
To authenticate with the Hugging Face Hub, create a token in your settings tab (<a href="https://huggingface.co/settings/tokens">https://huggingface.co/settings/tokens</a>), set it as secretyou 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(
modules.json: 100%

349/349 [00:00<00:00, 5.82kB/s]

config_sentence_transformers.json: 100%

116/116 [00:00<00:00, 2.67kB/s]

README.md: 100%
```

232k/232k [00:00<00:00, 12.7MB/s]

sentence_bert_config.json: 100% 53.0/53.0 [00:00<00:00, 1.16kB/s] config.json: 100% 612/612 [00:00<00:00, 13.9kB/s]

model.safetensors: 100% 90.9M/90.9M [00:00<00:00, 161MB/s]

tokenizer_config.json: 100% 350/350 [00:00<00:00, 29.4kB/s]

tokenizer.json: 100% 466k/466k [00:00<00:00, 35.0MB/s]

special_tokens_map.json: 100% 112/112 [00:00<00:00, 8.57kB/s]

config.json: 100% 190/190 [00:00<00:00, 12.6kB/s]

vocab.txt: 100%

```
Sentence-BERT Embedding for the sample query:
[ 1.84835214e-02 -2.74094529e-02 -1.79916788e-02 3.86158749e-02
 -1.06250755e-01 -4.64252979e-02 -5.44205494e-02 -1.48432972e-02
-6.06881306e-02 -1.77154876e-02 8.29189643e-03 3.94451171e-02
 \hbox{-1.73490569e-02} \quad \hbox{3.26199532e-02} \quad \hbox{1.39670307e-03} \quad \hbox{1.48688937e-02}
  5.40559925e-02 2.02019494e-02 2.00313479e-02 2.78611761e-02
 1.95406601e-02 -7.26020522e-03 6.89539462e-02 -6.83693122e-03
 -1.27489287e-02 -4.96626785e-03 4.00171690e-02 -2.58707386e-02
 4.02860641e-02 2.22228765e-02 2.96049491e-02 -5.91720231e-02
  8.77704844e-02 3.29544991e-02 -7.88048729e-02 2.60647088e-02
  1.26677185e-01 -5.38186282e-02 -1.01688221e-01 4.09076037e-03
  2.67346203e-02 9.47994832e-03 -3.83434794e-03 6.31621201e-03
 4.16448042e-02 4.54255417e-02 4.76548485e-02 -2.63007004e-02
 4.30551320e-02 3.83962765e-02 3.57637033e-02 -3.29230428e-02
 1.00586126e-02 -1.43079972e-02 3.70694511e-02 2.65719723e-02
 -4.53203321e-02 -3.32907736e-02 8.20846204e-03 -7.44366199e-02
 1.48739768e-02 2.09486168e-02 -1.55283418e-02 7.26637489e-04
 -8.43390301e-02 -6.77855164e-02 2.23249458e-02 4.61693387e-03
 -5.07284552e-02 -1.50081078e-02 -9.43713188e-02 1.57554895e-02
 -1.23548955e-02 6.82844147e-02 -2.10242439e-03 -3.30931246e-02
 -4.75708097e-02 -1.12865362e-02 8.23356025e-03 5.62238507e-02
 -3.67332734e-02 6.23321347e-02 -1.84719581e-02 1.53332114e-01
 -1.11013157e-02 -1.45017421e-02 1.28870541e-02 3.83995362e-02
 -1.06975436e-01 9.29350965e-03 -6.47428110e-02 -1.25122108e-02
 -3.37822847e-02 -2.62645446e-02 -9.60252248e-03 1.06358575e-02
 8.15246031e-02 -1.69914644e-02 1.29946902e-01 6.10867739e-02
 -3.58257890e-02 -8.60659406e-02 5.48108108e-02 1.77901201e-02
 1.66356917e-02 9.60460603e-02 3.09957881e-02 -6.00621477e-02
 -1.63449824e-03 1.25699844e-02 5.69183454e-02 7.10281506e-02
 -2.43924186e-02 1.31562147e-02 2.24198843e-03 7.67555237e-02
-5.19442558e-02 -5.71673252e-02 4.62734792e-03 -7.18659759e-02
 -7.97729790e-02 \quad 3.90722938e-02 \quad 1.36005329e-02 \quad 2.85357315e-33
 -1.62687209e-02 4.86304164e-02 4.12776656e-02 4.61899154e-02
 -1.99135281e-02 -1.19891658e-01 7.27647636e-03 -7.69580379e-02
  6.36012405e-02 2.58996654e-02 5.37260137e-02 1.66357700e-02
 3.65951061e-02 4.26868014e-02 7.67969899e-03 -6.00061081e-02
 -1.33408131e-02 -4.48250249e-02 3.63883711e-02 -7.94426538e-03
  6.09828383e-02 -9.28344205e-02 7.08889565e-04 -2.55090352e-02
  5.43945543e-02 -3.14430743e-02 -3.93154398e-02 -2.83255074e-02
  1.46430638e-02 -1.92074291e-02 -1.74178742e-02 3.07756534e-04
 -5.67141250e-02 -5.92200318e-03 2.30634008e-02 6.63976138e-03
 1.11974232e-01 6.66338727e-02 5.85148148e-02 -3.03268954e-02
  4.63998467e-02 2.57531703e-02 2.81245876e-02 -8.65470693e-02
  1.06390655e-01 6.23061182e-03 5.60944751e-02 7.36116245e-03
 -3.14857624e-03 1.24273226e-02 1.64878666e-02 -6.17097802e-02
 4.67638373e-02 -5.22827432e-02 -1.50504513e-02 5.19814380e-02
 -1.17266709e-02 1.67018287e-02 -7.67394379e-02 4.76369411e-02
 3.50129381e-02 6.38314858e-02 -9.94396303e-03 3.89827602e-02
 -1.08212702e-01 5.39657176e-02 -2.08003428e-02 5.42124771e-02
 -1.48485927e-02 5.82120242e-03 -1.56542659e-02 3.17977965e-02
 3.56274284e-02 -9.18340459e-02 8.40264410e-02 -1.93925165e-02
 -7.62497913e-03 -6.04701638e-02 -1.08786643e-01 1.65495351e-02
 1.28396899e-02 4.03430350e-02 -5.96752726e-02 6.12376630e-03
 -7.98955280e-03 8.94102734e-03 -1.59731209e-02 3.95398587e-02
```

-5.23984879e-02 -1.05241109e-02 -6.93212673e-02 6.86114701e-03 -1.16888463e-04 5.30087203e-02 -7.45742628e-03 -2.81861615e-33

```
9.17975008e-02 -1.09253712e-02 8.62475410e-02 -1.15235476e-02
      3.08466386e-02 -3.33491676e-02 5.02797542e-03 1.02462411e-01
      8.13557282e-02 -4.89437804e-02 -2.84789130e-02 -5.76464646e-02
      -3.54441814e-03 -4.62382957e-02 3.12891565e-02 3.75519395e-02
      1.94101743e-02 3.44083123e-02 3.51605229e-02 1.89154595e-02
      1.23867497e-01 -4.57655601e-02 -1.01837680e-01 -7.40435347e-02
      -2.26007625e-02 3.22277397e-02 2.14767233e-02 9.34859551e-03
      1.89310312e-02 1.38868287e-03 5.05960314e-03 -3.45776863e-02
      6.11590706e-02 -4.96250279e-02 -1.29227161e-01 3.30411717e-02
     -8.57629701e-02 1.99739665e-01 6.56093061e-02 1.56675633e-02
      1.30372923e-02 -3.52377482e-02 -3.52281407e-02 -6.89006317e-03
     -2.76362225e-02 -1.12612091e-01 9.49432887e-03 -5.46875112e-02
     -1.13156974e-01 -9.19804648e-02 -2.01718379e-02 -6.14775270e-02
      7.44989002e-03 -5.13269864e-02 5.61711900e-02 -5.48313037e-02
     -1.86018758e-02 -1.29944170e-02 1.89890750e-02 -7.75679946e-02
     -5.87603077e-02 4.98402789e-02 -7.15531334e-02 3.50076519e-02
     -7.15085911e-03 5.13388813e-02 -5.19783869e-02 3.51954997e-02
      -4.40392084e-02 5.27255908e-02 -1.50628999e-01 6.60738274e-02
      4.64917906e-03 2.10769977e-02 -9.55304410e-03 -5.11525497e-02
      3.06484960e-02 -4.47820574e-02 -1.69125777e-02 -6.17643893e-02
      2.95035262e-02 -5.48991114e-02 -3.73484455e-02 -6.89430535e-02
      4.95837908e-03 6.94564581e-02 -4.60608229e-02 -8.08791742e-02
     -1.21860076e-02 -9.17449407e-03 1.88278891e-02 4.48427796e-02
      9.01843831e-02 6.08272217e-02 1.33690918e-02 -3.04017576e-08
     -3.36875655e-02 2.48009171e-02 -1.62181724e-02 7.97163248e-02
      8.32510740e-03 9.61110070e-02 8.70615337e-03 6.53292686e-02
     -7.57048186e-03 -3.54053043e-02 6.05215020e-02 7.29734227e-02
      1.30295560e-01 5.20841591e-03 -7.73429498e-02 4.79719136e-04
      4.98865033e-03 1.83933089e-03 -6.82095718e-03 -5.96088078e-03
      2.93115899e-02 -6.25014529e-02 6.26943773e-03 -2.71424484e-02
      2.45521311e-02 -1.18923455e-01 4.73078787e-02 -5.08301668e-02
     -1.49220822e-03 2.07754243e-02 3.25640589e-02 -1.06151709e-02
     -7.43527785e-02 8.39679316e-02 2.90378015e-02 2.65081506e-02
      8.80952831e-03 8.76302598e-04 -8.74820873e-02 3.85403447e-02
     -3.91527936e-02 1.57182571e-02 1.29646985e-02 -5.80235533e-02
     -4.70717736e-02 -3.97947915e-02 -9.89454985e-02 -4.00956720e-02
      5.12351841e-02 -1.52886603e-02 9.77758169e-02 7.12869223e-03
     -1.85055267e-02 -4.11243774e-02 -2.39809486e-03 1.00764491e-01
     -3.01155429e-02 8.01074598e-03 -1.96525943e-03 5.37771061e-02
     -1.00924753e-01 1.27574289e-02 -6.42362908e-02 2.84757763e-02]
# Encode all responses from the "cleaned_output" column into embedding vectors
responses = df['cleaned output'].tolist()
response_embeddings = model.encode(responses)
print("Encoded", len(response_embeddings), "response embeddings.")

→ Encoded 1867 response embeddings.

from sklearn.metrics.pairwise import cosine_similarity
import numpy as np
def retrieve_response(query, model, response_embeddings, responses):
   query_embedding = model.encode([query])
   # Compute cosine similarities between query embedding and all response embeddings
   cosine_scores = cosine_similarity(query_embedding, response_embeddings)
   # Get index of the best match
   best_match_idx = np.argmax(cosine_scores)
   best_response = responses[best_match_idx]
   return best_response
# Test the retrieval function with a sample query
sample_query = df['cleaned_query'].iloc[0]
print("Sample Query:", sample_query)
print("Retrieved Response:", retrieve_response(sample_query, model, response_embeddings, responses))
Sample Query: im new working suggest simple workout routine beginner
    Retrieved Response: a beginner plan should focus on low-impact exercises, proper warm-up/cool-down routines, and gradual intensity incre
    Inference function for the chatbot that returns the best matching response for a given user query.
def respond(user query):
```

```
best_response = retrieve_response(user_query, model, response_embeddings, responses)
    return best_response
# Testing the inference function with a custom query
test_query = "I need a high intensity workout for my legs."
print("Test Query:", test_query)
print("Chatbot Response:", respond(test_query))
Test Query: I need a high intensity workout for my legs.
     Chatbot Response: incorporate exercises like high knees, butt kicks, and leg swings to activate muscles and prepare your body for high-i
import numpy as np
from sklearn.metrics.pairwise import cosine_similarity
def evaluate_retrieval(test_df, model, response_embeddings, responses, k=3):
    mrr_total = 0
    recall_at_k_total = 0
    num_samples = len(test_df)
    # Loop over each test sample
    for idx, row in test_df.iterrows():
        query = row['cleaned_query']
        ground_truth = row['cleaned_output']
        # Compute the embedding for the test query
        query_embedding = model.encode([query])
        # Calculate cosine similarity between the query and all pre-computed response embeddings
        cosine_scores = cosine_similarity(query_embedding, response_embeddings)[0]
        # Rank the responses (indices) by similarity in descending order
        ranked_indices = np.argsort(cosine_scores)[::-1]
        # Find the rank position of the ground truth response
        for i, idx in enumerate(ranked_indices):
            # Simple exact string match after stripping any extra whitespace
            if responses[idx].strip() == ground_truth.strip():
                rank = i + 1 # Rank is 1-indexed
                break
        # If ground truth is not found, we consider the rank as worst-case (length + 1)
        if rank is None:
           rank = len(ranked_indices) + 1
        # Update Mean Reciprocal Rank
        mrr_total += 1 / rank
        # Update Recall@k: if the ground truth is within the top k responses
        if rank <= k:
           recall_at_k_total += 1
    mrr = mrr_total / num_samples
    recall_at_k = recall_at_k_total / num_samples
    return mrr, recall_at_k
# Evaluate the retrieval performance on the test set (with k=3 for Recall@3)
mrr, recall_at_3 = evaluate_retrieval(test_df, model, response_embeddings, responses, k=3)
print("Mean Reciprocal Rank (MRR):", mrr)
print("Recall@3:", recall_at_3)
    Mean Reciprocal Rank (MRR): 0.5070786714669944
     Recall@3: 0.5481283422459893
rmv
!pip install gradio
→ Collecting gradio
       Downloading gradio-5.23.3-py3-none-any.whl.metadata (16 kB)
```

Collecting aiofiles<24.0,>=22.0 (from gradio)

Downloading aiofiles-23.2.1-py3-none-any.whl.metadata (9.7 kB)

```
Collecting fastapi<1.0,>=0.115.2 (from gradio)
       Downloading fastapi-0.115.12-py3-none-any.whl.metadata (27 kB)
    Collecting ffmpy (from gradio)
       Downloading ffmpy-0.5.0-py3-none-any.whl.metadata (3.0 kB)
    Collecting gradio-client==1.8.0 (from gradio)
      Downloading gradio client-1.8.0-py3-none-any.whl.metadata (7.1 kB)
    Collecting groovy~=0.1 (from gradio)
       Downloading groovy-0.1.2-py3-none-any.whl.metadata (6.1 kB)
     Requirement already satisfied: httpx>=0.24.1 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.28.1)
    Requirement already satisfied: huggingface-hub>=0.28.1 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.29.3)
    Requirement already satisfied: jinja2<4.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.1.6)
    Requirement already satisfied: markupsafe<4.0,>=2.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.0.2)
    Requirement already satisfied: numpy<3.0,>=1.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.0.2)
    Requirement already satisfied: orjson~=3.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (3.10.16)
    Requirement already satisfied: packaging in /usr/local/lib/python3.11/dist-packages (from gradio) (24.2)
    Requirement already satisfied: pandas<3.0,>=1.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.2.2)
    Requirement already satisfied: pillow<12.0,>=8.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (11.1.0)
    Requirement already satisfied: pydantic<2.12,>=2.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (2.11.0)
    Collecting pydub (from gradio)
      Downloading pydub-0.25.1-py2.py3-none-any.whl.metadata (1.4 kB)
    Collecting python-multipart>=0.0.18 (from gradio)
       Downloading python_multipart-0.0.20-py3-none-any.whl.metadata (1.8 kB)
     Requirement already satisfied: pyyaml<7.0,>=5.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (6.0.2)
    Collecting ruff>=0.9.3 (from gradio)
       Downloading ruff-0.11.2-py3-none-manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (25 kB)
    Collecting safehttpx<0.2.0,>=0.1.6 (from gradio)
      Downloading safehttpx-0.1.6-py3-none-any.whl.metadata (4.2 kB)
    Collecting semantic-version~=2.0 (from gradio)
      Downloading semantic_version-2.10.0-py2.py3-none-any.whl.metadata (9.7 kB)
    Collecting starlette<1.0,>=0.40.0 (from gradio)
       Downloading starlette-0.46.1-py3-none-any.whl.metadata (6.2 kB)
    Collecting tomlkit<0.14.0,>=0.12.0 (from gradio)
       Downloading tomlkit-0.13.2-py3-none-any.whl.metadata (2.7 kB)
    Requirement already satisfied: typer<1.0,>=0.12 in /usr/local/lib/python3.11/dist-packages (from gradio) (0.15.2)
    Requirement already satisfied: typing-extensions~=4.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (4.13.0)
    Collecting uvicorn>=0.14.0 (from gradio)
       Downloading uvicorn-0.34.0-py3-none-any.whl.metadata (6.5 kB)
    Requirement already satisfied: fsspec in /usr/local/lib/python3.11/dist-packages (from gradio-client==1.8.0->gradio) (2025.3.0)
    Requirement already satisfied: websockets<16.0,>=10.0 in /usr/local/lib/python3.11/dist-packages (from gradio-client==1.8.0->gradio)
    Requirement already satisfied: idna>=2.8 in /usr/local/lib/python3.11/dist-packages (from anyio<5.0,>=3.0->gradio) (3.10)
    Requirement already satisfied: sniffio>=1.1 in /usr/local/lib/python3.11/dist-packages (from anyio<5.0,>=3.0->gradio) (1.3.1)
    Requirement already satisfied: certifi in /usr/local/lib/python3.11/dist-packages (from httpx>=0.24.1->gradio) (2025.1.31)
    Requirement already satisfied: httpcore==1.* in /usr/local/lib/python3.11/dist-packages (from httpx>=0.24.1->gradio) (1.0.7)
    Requirement already satisfied: h11<0.15,>=0.13 in /usr/local/lib/python3.11/dist-packages (from httpcore==1.*->httpx>=0.24.1->gradio)
    Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1->gradio) (3.18.0)
    Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1->gradio) (2.32.3)
    Requirement already satisfied: tqdm>=4.42.1 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub>=0.28.1->gradio) (4.67.1
    Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0->gradio) (2.8
    Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0->gradio) (2025.2)
    Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas<3.0,>=1.0->gradio) (2025.2)
    Requirement already satisfied: annotated-types>=0.6.0 in /usr/local/lib/python3.11/dist-packages (from pydantic<2.12,>=2.0->gradio) (
    Requirement already satisfied: pydantic-core==2.33.0 in /usr/local/lib/python3.11/dist-packages (from pydantic<2.12,>=2.0->gradio) (2
import gradio as gr
import pandas as pd
import datetime
# Updated knowledge base with detailed recommendations
exercise_db = pd.DataFrame({
   "exercise_name": ["Push-ups", "Squats", "Deadlifts", "Plank", "Jump Rope"],
"target_muscle": ["Chest, Triceps", "Legs, Glutes", "Back, Legs", "Core", "Cardio"],
   "difficulty_level": ["Beginner", "Beginner", "Advanced", "Beginner", "Intermediate"],
   "sets": [3, 3, 4, 3, "Timed"],
   "reps": [12, 15, 6, "Hold 30 sec", "60 sec"],
   "rest_time": ["30 sec", "30 sec", "60 sec", "N/A", "N/A"]
# Store conversation history
conversation_history = {}
# Store user queries for future improvements
query_logs = []
# Function to retrieve personalized workouts
def get_exercise_suggestions(fitness_level, equipment):
   filtered_exercises = exercise_db[
        (exercise_db["difficulty_level"] == fitness_level)
   if "None" in equipment:
     filtered exercises = filtered exercises
```

})

Requirement already satisfied: anyio<5.0,>=3.0 in /usr/local/lib/python3.11/dist-packages (from gradio) (4.9.0)

```
exercise_details = filtered_exercises.to_dict(orient="records")
   return exercise_details
# Multi-Turn Memory
def respond(user_id, user_query, mode, fitness_level, age, equipment):
   Handles normal and personalized queries, integrates multi-turn memory, and justifies responses.
   cleaned_query = user_query.strip().lower()
   # Create session for user if not exists
   if user_id not in conversation_history:
       conversation_history[user_id] = []
   if mode == "Personalized Query":
       fitness_level = fitness_level if fitness_level else "Not specified"
       age = str(age) if age else "Not specified"
       equipment_str = ", ".join(equipment) if equipment else "None"
       enriched_query = (
           final_query = enriched_query
   else:
       final query = cleaned query # Normal query without personalization
   # Retrieve response using retrieval function
   best_response = retrieve_response(final_query, model, response_embeddings, responses)
   # Store conversation history
   conversation_history[user_id].append((user_query, best_response))
   # Log queries for analysis
   query_logs.append({
       "timestamp": datetime.datetime.now().strftime("%Y-%m-%d %H:%M:%S"),
       "user_id": user_id,
       "query": user_query,
       "response": best_response
   })
   # Suggest personalized exercises
   exercise_suggestions = get_exercise_suggestions(fitness_level, equipment)
   # Justification
   justification = (
       f"Since you are a {fitness_level.lower()} level trainee and selected {', '.join(equipment) or 'no equipment'}, "
       "these workouts are suitable for your level."
   return best_response, conversation_history[user_id], exercise_suggestions, justification
# Feedback Handling Function
def collect_feedback(response, feedback):
   query_logs.append({"response": response, "feedback": feedback})
   return f"Feedback received: {feedback}"
# Toggle input fields based on mode
def toggle_inputs(mode):
visible = mode == "Personalized Query"
```