ReadMe.md

Health and Fitness Tracking App

Overview

The Health and Fitness Tracking App is a digital platform designed to assist users in monitoring and understanding their health and fitness data. The app aims to empower individuals to make informed decisions about their lifestyle and wellness routines by leveraging a robust database and user-friendly interface.

Objectives

- **Personalized Health Monitoring: ** Enable users to track specific health metrics tailored to their needs.
- **Workout Logging:** Users can log various workout details, including type,
 duration, and intensity.
- **Nutritional Tracking: ** The app comprehensively monitors users' nutritional intake, focusing on calorie tracking and macro breakdowns.
- **Sleep Pattern Analysis:** Users can record and analyze their sleep duration and quality to better understand their rest cycles.
- **Mental Wellness:** Track mental health activities, stress levels, and general mood
 to support overall well-being.
- **Health Recommendations:** Provide personalized health recommendations to users based on their logged data.

Target Audience

This app's primary audience includes health-conscious individuals looking to improve or maintain their fitness and well-being through data-driven insights.

Data Stored

- `users`: Stores personal and authentication details of each user along with links to their health and fitness records.
- body_measurements: Keeps track of user body metrics like height and weight over time.
- `workout_types`: Defines different types of workouts that users can log.
- `workouts`: Records details of each workout session, including type, duration, intensity, and calories burned.
- `meal types`: Categorizes types of meals, such as breakfast, lunch, dinner, etc.
- `nutrition`: Logs nutritional intake details, including calorie count and macronutrient breakdown for user meals.

- `sleep`: Captures data on user sleep patterns, including duration and quality of sleep.
- `health_metrics`: Tracks various health metrics such as heart rate and blood pressure for users.
- `mood_types`: Lists types of moods that users can record for mental health tracking.
- `mental_health`: Logs mental health data, including stress levels, mood types, and mindfulness activities.
- `health_recommendations`: Contains personalized health and fitness recommendations for each user.

Database Schema Design

The database schema provides a comprehensive overview of users' health and fitness data while ensuring data integrity and efficient querying. Primary keys ensure the uniqueness of records, while foreign keys enforce relationships between different data entities. Indexes on frequently searched fields like `username,` `email,` and `recorded_date` across various tables optimize query performance. Data validation is also applied to ensure data input is within reasonable expected range.

Database Design Considerations

Primary and Foreign Key Usage

- **Primary Keys** are unique identifiers for each record within a table. In this database, they ensure that each user, workout, nutrition log, etc., has a unique identifier, such as `id,` which is crucial for indexing and managing relationships between tables.
- **Foreign Keys** establish a link between related data across different tables. For example, `user_id` in the `workouts` table references the `id` in the `users` table, enabling the database to maintain consistent and coherent data about which workouts belong to which users.

Normalization

Normalization is organizing data in a database to reduce redundancy and improve data integrity. This database is normalized to the Third Normal Form (3NF), building upon the requirements of the First (1NF) and Second (2NF) Normal Forms.

First Normal Form (1NF)

- **Atomicity**: Each field contains indivisible atomic values with no repeating groups or arrays.
- **Unique Identifiers**: A primary key in each table ensures each record is unique and identifiable.
- **Data Consistency**: Using consistent data types and formats across similar fields
 prevents data conflicts.

By satisfying 1NF, the database ensures a solid foundation where each table represents a single entity or concept, and each row/column intersection contains a single value.

Second Normal Form (2NF)

- **Elimination of Partial Dependencies**: By ensuring that all non-key attributes are fully functionally dependent on the primary key, not just part of it, the database eliminates partial dependencies.

This step requires the presence of a primary key, and in the case of composite primary keys, it ensures that no attribute is dependent on only a part of the key. Thus, all the fields in a table relate directly to the primary key, enhancing data retrieval speed and consistency.

Third Normal Form (3NF)

- **Removal of Transitive Dependencies**: Non-key attributes are not allowed to depend on other non-key attributes, which means all attributes are directly dependent on the primary key.

By reaching 3NF, the database avoids transitive dependencies, which not only keeps data redundancy to a minimum but also protects data integrity by preventing anomalies that can occur during data operations (inserts, updates, or deletions).

Importance of 3NF

Maintaining 3NF is crucial for several reasons:

- **Data Integrity**: There's a single source of truth for each piece of information, which prevents data duplication and inconsistency.
- **Maintenance**: Simplifies maintenance tasks because updates, inserts, or deletes need to happen in only one place.
- **Performance**: Improves query performance due to reduced data duplication and a
 more efficient structure.
- **Flexibility**: Makes the database more adaptable to changes, as adjustments for evolving business requirements can be made without extensive redesign.

- **Clarity**: Provides more apparent relationships between entities, which can be crucial for developers and analysts when understanding the database structure.

The careful design aligned with 1NF, 2NF, and 3NF principles is essential for a robust, reliable, and efficient database system like the Health and Fitness Tracking App.

Query Optimization with Indices

Indices in this health and fitness tracking application are used to speed up data retrieval from the database. They are especially valuable when querying large datasets, as they prevent the need for full table scans. For instance, the `users` table has indices on `username` and `email,` which are likely to be used often to look up user information, thus accelerating search operations. Similarly, tables like `body_measurements,` `workouts,` `nutrition,` `sleep,` `health_metrics,` and `mental_health` have indices on user-related foreign keys and dates. Queries often filter or sort based on dates and user IDs. Indices on these columns mean the database can quickly locate and retrieve the relevant records without scanning the table. This leads to faster response times for end-users and reduced load on the database server, which is critical for providing a smooth user experience.

Transactions and ACID Concepts in Data Insertion

The insert_data file illustrates database transactions, a key part of the ACID properties (Atomicity, Consistency, Isolation, Durability) essential for maintaining database integrity.

Atomicity

Atomicity is demonstrated through transactions when adding new users and related records like body measurements and health metrics. The code ensures that all operations within a transaction block are treated as a single unit, meaning that they either succeed or are not applied. This is achieved by wrapping user creation and related data insertions within a `try` block, flushing to obtain user IDs for related records without committing, and then committing all at once if all insertions are successful.

Consistency

Consistency is maintained by enforcing data integrity constraints and relationships.

If a transaction fails, a `rollback` is initiated to undo any changes made during the transaction, ensuring the database remains consistent.

Isolation

Isolation is implicitly managed by SQLAlchemy's session and transaction management system. Each transaction is isolated from others, ensuring that the operations within a transaction are completed without interference from other concurrent transactions.

Durability

Durability is guaranteed by committing the transactions to the database. Once the transaction is executed, the changes are permanent, even in the event of a system failure, ensuring the reliability of the data.

The use of transactions in this code shows a commitment to the ACID properties, ensuring that the database operations are reliable, consistent, and correct.

Database Query Intentions

The SQL queries in the code aim to extract specific insights from a health and fitness tracking application's database:

Total Calories Burned Per User

This query calculates the sum of calories each user burns through their workout sessions, giving an overview of their total energy expenditure.

Latest Body Measurements Per User

This query retrieves each user's most recent height and weight measurements, providing up-to-date information on their physical dimensions.

Average Sleep Duration

This query determines the average sleep duration for each user over the past month, offering insight into their sleep patterns.

Top 5 Frequent Workout Types

This query identifies the five most common workout types across all users, highlighting the most popular fitness activities within the app's community.

Last Nutrition Log Entry

This query finds the details of each user's last recorded nutrition log, showing their most recent dietary intake.

Average Heart Rate Per User

This query computes the average heart rate for each user, which can indicate cardiovascular fitness or stress.

Latest Mental Health Log

```
This query fetches each user's latest recorded mental health log, revealing their most
recent stress levels and mindfulness activity durations.
## File Overview
### `create.py`
 **Purpose**: Establishes the structure of the database.
 **Functionality**: Contains class definitions for each table using SQLAlchemy ORM,
sets up the database engine, and executes commands to create the database with the
defined schema.
### `insert data.py`
 **Purpose**: Populates the database with data.
 **Functionality**: Utilizes the Faker library to generate and insert randomized but
realistic data into the tables for users, body measurements, workouts, meal types,
nutrition logs, sleep records, health metrics, and mental health logs.
### `query_data.py`
 **Purpose**: Retrieves and displays information from the database.
 **Functionality**: Contains various SQL queries executed via SQLAlchemy ORM that
pull and aggregate data from the database, such as total calories burned per user,
latest body measurements, average sleep duration, frequent workout types, last
nutrition log, and latest mental health logs. Outputs the results of these queries to
the console.
## Setting Up the Development Environment
Instructions for setting up the Python virtual environment and installing
dependencies.
python3 -m venv venv
source venv/bin/activate
pip3 install -r requirements.txt
python create.py
python insert_data.py
python query data.py
## Testing
```

Create.py:

```
from sqlalchemy import (
from sqlalchemy.ext.declarative import declarative base
from sqlalchemy.orm import relationship, sessionmaker
import datetime
engine = create engine('sqlite:///health system.db')
Base = declarative base()
class User(Base):
  id = Column(Integer, primary key=True, autoincrement=True)
  email = Column(String, nullable=False, unique=True)
  date of birth = Column(DateTime, nullable=False)
   workouts = relationship("Workout", back populates="user")
  measurements = relationship("BodyMeasurement", back populates="user")
  nutrition logs = relationship("Nutrition", back populates="user")
   sleep records = relationship("Sleep", back populates="user")
  mental health logs = relationship("MentalHealth", back populates="user")
  health recommendations = relationship("HealthRecommendation",
back populates="user")
   table args = (
```

```
class BodyMeasurement(Base):
  id = Column(Integer, primary key=True, autoincrement=True)
  user id = Column(Integer, ForeignKey('users.id'), nullable=False)
  height = Column(Float)
  weight = Column(Float)
  recorded date = Column(DateTime, default=datetime.datetime.utcnow)
   table args = (Index('ix body measurements recorded date', 'recorded date'),)
  user = relationship("User", back populates="measurements")
class WorkoutType(Base):
  id = Column(Integer, primary key=True, autoincrement=True)
  type = Column(String, nullable=False, unique=True)
class Workout(Base):
  user id = Column(Integer, ForeignKey('users.id'), nullable=False)
  workout type id = Column(Integer, ForeignKey('workout types.id'), nullable=False)
  duration = Column(Integer, nullable=False)
  intensity level = Column(Integer)
```

```
table args = (
  user = relationship("User", back populates="workouts")
  workout type = relationship("WorkoutType", back populates="workouts")
class Nutrition(Base):
  meal type id = Column(Integer, ForeignKey('meal types.id'), nullable=False)
  calories = Column(Integer, nullable=False)
  carbs = Column(Float)
  fats = Column(Float)
  date = Column(DateTime, default=datetime.datetime.utcnow)
  user = relationship("User", back populates="nutrition logs")
class Sleep(Base):
```

```
id = Column(Integer, primary key=True, autoincrement=True)
  user id = Column(Integer, ForeignKey('users.id'), nullable=False)
  duration = Column(Float, nullable=False)
  date = Column(DateTime, default=datetime.datetime.utcnow)
  table args = (
class HealthMetric(Base):
  user id = Column(Integer, ForeignKey('users.id'), nullable=False)
  blood pressure = Column(String)
  recorded date = Column(DateTime, default=datetime.datetime.utcnow)
```

```
user = relationship("User", back populates="health metrics")
class MoodType(Base):
  id = Column(Integer, primary key=True, autoincrement=True)
class MentalHealth(Base):
```

```
mood_type_id = Column(Integer, ForeignKey('mood_types.id'), nullable=False)
  activity description = Column(String)
  user = relationship("User", back populates="mental health logs")
  is active = Column(Boolean, default=True)
  user = relationship("User", back_populates="health_recommendations")
def get new session(database url):
```

```
Returns:
    Session: A SQLAlchemy session object.

"""

engine = create_engine(database_url)
Session = sessionmaker(bind=engine)
return Session()

def create_tables():
    """
    Create all tables in the database according to the defined classes and relationships.
    """
Base.metadata.create_all(engine)

create_tables()
```

Insert data.py

```
populate_users(session)
session.commit() # Users need to exist before related records can be added
users = session.query(User).all()
def add body measurements(session, users, measurement range=(1, 3)):
       for in range(random.randint(*measurement range)):
              user id=user.id,
               height=random.uniform(1.5, 2.0),
               weight=random.uniform(50.0, 100.0),
               recorded_date=fake.past_date()
           session.add(measurement)
add body measurements(session, users)
workout types = ['Running', 'Swimming', 'Cycling', 'Yoga', 'Weight Training']
for workout in workout types:
  workout type = WorkoutType(type=workout)
session.commit() # WorkoutType IDs are needed for Workout records
def populate_workouts(session, users, workout_type_ids, workout_range=(5, 20)):
       for in range(random.randint(*workout range)):
               workout_type_id=random.choice(workout_type_ids),
               duration=random.randint(20, 120),
               date=fake.past date()
           session.add(workout)
workout_type_ids = [wt.id for wt in session.query(WorkoutType).all()]
```

```
populate_workouts(session, users, workout_type_ids)
meal types = ['Breakfast', 'Lunch', 'Dinner', 'Snack']
for meal in meal types:
  meal type = MealType(type=meal)
  session.add(meal type)
session.commit()
meal type ids = [mt.id for mt in session.query(MealType).all()]
for user in users:
          user id=user.id,
           meal type id=random.choice(meal type ids),
          calories=random.randint(100, 900),
       session.add(nutrition)
           quality=random.randint(1, 10),
           date=fake.past_date()
```

```
blood_pressure=f"{random.randint(100, 140)}/{random.randint(60, 90)}",
           recorded date=fake.past date()
moods = ['Happy', 'Sad', 'Angry', 'Excited', 'Stressed']
for mood in moods:
  mood type = MoodType(mood=mood)
  session.add(mood type)
session.commit()
mood type ids = [mt.id for mt in session.query(MoodType).all()]
          user id=user.id,
          mood type id=random.choice(mood type ids),
           activity description=fake.text(max nb chars=200),
           recorded_date=fake.past_date()
       session.add(mental health log)
session.commit()
session.close() # Always close the session when done
session = Session()
try:
lead
```

```
username=fake.user name(),
      password hash=fake.sha256(raw output=False),
the user's profile.
user-related
  new measurement = BodyMeasurement(
      user id=new user.id,
  new health metric = HealthMetric(
      user id=new user.id,
```

```
except SQLAlchemyError as e:
  session.rollback()
free resources.
  session.close()
{\tt def} populate health recommendations(session, users, recommendation range=(1, 5)):
                   user id=user id,
                   is_active=fake.boolean(chance_of_getting_true=75),
```

Query data.py

```
# Import necessary libraries and modules from SQLAlchemy and datetime package.

from sqlalchemy import func, and_
from datetime import datetime, timedelta

# Import classes for each table in the database and the engine object from create.py.

from create import (
    User, BodyMeasurement, WorkoutType, Workout, MealType,
    Nutrition, Sleep, HealthMetric, MentalHealth, engine
)

# Import sessionmaker for creating a session with the database.

from sqlalchemy.orm import sessionmaker

# Set up a new session factory bound to the engine.

Session = sessionmaker(bind=engine)

session = Session()  # Instantiate a session.

# --- Calories Burned Query ---
# Query to find the total number of calories burned per user.

# Joins User and Workout tables, groups by username, and sums the calories burned.

total_calories_per_user = session.query(
    User.username,
    func.sum(Workout.calories_burned).label('total_calories_burned')
```

```
.join(User.workouts).group_by(User.username)
print("Total Calories Burned Per User:")
for username, total calories in total calories per user:
print("-" * 35)
latest body measurement subquery = session.query(
   func.max(BodyMeasurement.recorded date).label('latest date')
).group by(BodyMeasurement.user id).subquery('latest measurements')
latest body measurement = session.query(
  latest body measurement subquery,
       BodyMeasurement.user id == latest body measurement subquery.c.user id,
       BodyMeasurement.recorded_date == latest_body_measurement_subquery.c.latest_date
 .all()
print("Users and Their Latest Body Measurements:")
for username, height, weight in latest body measurement:
  print(f"{username:20} | Height: {height:10} cm | Weight: {weight:10} kg")
print("-" * 60)
average sleep duration = session.query(
  func.avg(Sleep.duration).label('average sleep duration')
  Sleep.date.between(datetime.now() - timedelta(days=30), datetime.now())
.group by(User.username)
```

```
print("Average Sleep Duration Over the Last Month:")
for username, average duration in average sleep duration:
  print(f"{username:20} | {average duration:10.2f} hours")
print("-" * 35)
top workout types = session.query(
  WorkoutType.type,
  func.count(Workout.id).label('frequency')
).join(WorkoutType.workouts).group by(WorkoutType.type).order by(
   func.count(Workout.id).desc()
).limit(5)
print("Top 5 Most Frequent Workout Types:")
for workout type, frequency in top workout types:
print("-" * 35)
last nutrition date subquery = session.query(
).group by(Nutrition.user id).subquery('last nutrition log')
nutrition on last logged day = session.query(
  Nutrition.calories,
).join(User.nutrition logs).join(
  last nutrition date subquery,
      Nutrition.user_id == last_nutrition_date_subquery.c.user_id,
      Nutrition.date == last nutrition date subquery.c.last date
.all()
print("Users' Nutrition Intake on Their Last Logged Day:")
```

```
for username, calories, protein, carbs, fats in nutrition_on_last_logged_day:
Carbs: {carbs:5}g | Fats: {fats:5}g")
print("-" * 85)
average heart rate = session.query(
  func.avg(HealthMetric.heart rate).label('average heart rate')
).join(User.health metrics).group by(User.username)
print("Average Heart Rate Per User:")
for username, avg_hr in average_heart_rate:
  print(f"{username:20} | {avg hr:10.2f} bpm")
print("-" * 35)
latest mental health_subquery = session.query(
  func.max(MentalHealth.recorded date).label('latest date')
).group by(MentalHealth.user id).subquery('latest mental health')
latest mental health = session.query(
).join(User.mental health logs).join(
  latest mental health subquery,
       MentalHealth.user id == latest mental health subquery.c.user id,
       MentalHealth.recorded_date == latest_mental_health_subquery.c.latest_date
print("Users and Their Latest Mental Health Logs:")
for username, stress level, mindfulness duration in latest mental health:
{mindfulness duration:10} min")
print("-" * 85)
```

Close the session after you're done with the queries to release resources session.close()