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PHASE 3: IMPLEMENTATION OF PROJECT

TITLE: AI AFFORDABLE HOUSE FINDER

OBJECTIVE:

The **objective** of an Al Affordable House Finder is to leverage artificial intelligence to help users identify housing options that are **affordable**, **suitable**, and aligned with their preferences and financial situation. It aims to simplify and enhance the house-hunting process, especially for individuals and families with limited budgets.

1.AI MODEL DEVELOPMENT

Overview

The AI Affordable House Finder is a tool designed to help users search for affordable housing options based on their unique needs, preferences, and financial constraints. The goal is to leverage AI to provide intelligent recommendations, streamline the home-buying or renting process, and make housing more accessible to a wider range of people. This process involves developing models that analyze housing data, predict affordability, and personalize recommendations based on user profiles.

Implementation:

- Identify affordable housing options based on financial constraints and preferences.
- Match housing features (price, location, size, amenities) with user needs.
- Provide real-time data on the housing market and predict future trends.

Outcome:

The AI Affordable House Finder represents a highly impactful application of AI in the real estate industry. By combining predictive modeling, recommendation systems, and user personalization, it can create a smarter, more accessible, and efficient way for users to find affordable housing

options. This system can have far-reaching effects on both individuals and communities by making homeownership more achievable and ensuring that affordable housing opportunities are not missed.

2.CHATBOT DEVELOPMENT

Overview

A Chatbot for AI Affordable House Finder is an AI-powered conversational agent designed to assist users in finding affordable housing options. The chatbot leverages natural language processing (NLP) and machine learning techniques to interact with users, understand their housing needs, and provide personalized recommendations. This allows users to ask questions, share preferences, and receive real-time housing suggestions, making the home search process more efficient, accessible, and user-friendly.

Implementation

- Assist users in their housing search: Help users find affordable housing options based on their budget, preferences, and location.
- Provide personalized recommendations: Based on user input, the chatbot should be able to suggest properties that match specific criteria.
- Enable real-time, interactive communication: Users can interact with the chatbot using natural language, asking questions or refining their search as needed.

Outcome

The development of a Chatbot for AI Affordable House Finder significantly enhances the user experience by offering personalized, real-time housing recommendations through a natural conversational interface. By integrating NLP, recommendation systems, and affordability models, the chatbot streamlines the home-buying or renting process, making it more accessible and user-friendly. The outcome is a highly efficient, personalized, and scalable solution that can assist users in finding affordable housing, ultimately contributing to a more inclusive housing market.

3. DATA SECURITY IMPLEMENTATION

Overview

Data security in the **AI Affordable House Finder** focuses on securing user data throughout the entire lifecycle: from the moment it's collected (e.g., during user registration or property search) to its storage, processing, and transmission. The goal is to protect the system from unauthorized access, data breaches, and ensure compliance with data protection regulations like GDPR and CCPA. The security implementation would use a variety of encryption, authentication, and monitoring techniques to protect users' sensitive information.

Implementation

1. Data Collection and Encryption

- **Secure User Input**: When users enter personal details (e.g., income, location preferences), these should be transmitted over secure channels.
 - TLS (Transport Layer Security): Ensures encrypted communication between users and the server, preventing eavesdropping and data tampering.

Encryption of Data:

- AES (Advanced Encryption Standard): For storing sensitive data such as financial information, user preferences, and other personally identifiable information (PII). AES with a 256-bit key is considered highly secure.
- End-to-End Encryption: Ensures that data is encrypted at the point of collection (e.g., on the user's device) and only decrypted on the server, minimizing exposure during transmission.

Outcome

Implementing data security in the AI Affordable House Finder ensures that sensitive user data, including personal, financial, and housing information, is protected from unauthorized access, breaches, or theft. The implementation of encryption, secure authentication, data storage protocols, compliance with legal regulations, and ongoing monitoring helps establish trust with users while also protecting the integrity and reputation of the system. The outcome is a more secure, reliable, and user-friendly platform that not only meets security requirements but also enhances user confidence in the application.

3.TESTING AND FEEDBACK COLLECTION

Overview

Testing and feedback collection are essential components of any software development process, including AI-powered applications like the AI Affordable House Finder. Testing ensures that the system works as expected, performs efficiently, and meets security standards. Feedback collection provides insights into the user experience and helps refine the system to better meet user needs and expectations.

Implementation

1. Testing Approach

The testing of the **AI Affordable House Finder** involves multiple layers and methodologies to ensure the application works as intended.

a. Unit Testing

• **Purpose**: Ensure that individual components of the system (e.g., Al models, APIs, data encryption, and recommendation engines) function as expected.

• Implementation:

- Test Cases: Write specific test cases for each function, such as validating if a user's budget is correctly handled or if the recommendation system returns appropriate results.
- Tools: Use testing frameworks like JUnit (for Java), pytest (for Python), or Mocha (for JavaScript).

Outcome

Improved System Performance

- Bug Fixing: Through testing, issues like incorrect property recommendations, poor chatbot responses, or slow load times can be identified and fixed.
- Enhanced AI Accuracy: User feedback helps refine the recommendation system, making it more accurate in predicting properties that fit users' preferences and budget.

CHALLENGES AND SOLUTIONS:

- Challenge: Data Quality and Availability
- Challenge: User Experience and Interface Design
- Challenge: Ensuring Accurate AI Recommendations

Outcomes of Phase 3:

1.Basic Al Model: The **Al model** will be capable of understanding user inputs related to housing preferences and providing relevant property recommendations based on basic criteria.

2. Functional Chatbot Interface:

The **chatbot interface** will be fully functional and will serve as the primary method for user interaction with the AI model, allowing users to input their preferences and get personalized housing recommendations

3. Optional IoT Integration:

The **IoT integration** will be implemented if applicable, enabling the platform to gather additional data from IoT-enabled devices such as wearable health monitors or smart home devices. This will enhance the user experience by providing more personalized recommendations.

4. Data Security:

User data will be protected through **basic encryption** and **security protocols**, ensuring that sensitive information is stored securely and that privacy is maintained.

5. Initial Testing and Feedback Collection

By the end of Phase 3, **initial testing** and **feedback collection** will take place to identify issues, refine AI recommendations, and improve the platform's usability and functionality.

PROGRAM:

```
import pandas as pd
                                                                                        Affordable Houses within Budget:
\ensuremath{\text{\#}} Sample dataset: You can replace this with real CSV data
                                                                                           HouseID Location Price Bedrooms Bathrooms Score
                                                                                                 4 Village 95000
                                                                                                                                        1 0.775
    'HouseID': [1, 2, 3, 4, 5],
                                                                                                 2 Suburbs 150000
                                                                                                                                        2 0.650
    'Location': ['City Center', 'Suburbs', 'Suburbs', 'Village', 'City
                                                                                                 3 Suburbs 180000
                                                                                                                                        2 0.500
       Center'],
     'Price': [250000, 150000, 180000, 95000, 300000],
                                                                                            Code Execution Successful ===
    'Bedrooms': [2, 3, 3, 2, 4],
'Bathrooms': [1, 2, 2, 1, 3],
    'SqFt': [850, 1200, 1100, 900, 1400]
df = pd.DataFrame(data)
# Set user's budget
user_budget = 200000
# Simple AI-like logic to score houses based on affordability
def score house(row, budget):
    # The closer the price to the budget (without exceeding), the higher
       the score
        return 0
    return (budget - row['Price']) / budget + row['Bedrooms'] * 0.1 +
        row['Bathrooms'] * 0.05
```

Next Steps for Phase 4

In phase 4 team will focus on,

Improving the Al's Accuracy:

Refine the AI to make the property recommendation system more accurate and personalized by incorporating feedback from earlier phases, improving prediction models, and enhancing recommendation capabilities.

Expanding Multilingual Support

Make the platform more accessible to a global audience by offering **multilingual support** and voice-command features, allowing users to interact with the platform in their native languages and through voice-based inputs.

Scaling and Optimizing

Ensure the platform can handle a larger number of users, more complex property searches, and larger data volumes while maintaining **high performance**, **low latency**, and **system stability**.