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**Smart Subscription Manager:
Application for Managing User Subscriptions**

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Abbreviations

| Abbreviation | Meaning |
|--------------|-----------------------------------|
| AI | Artificial Intelligence |
| ML | Machine Learning |
| NLP | Natural Language Processing |
| LSTM | Long Short-Term Memory |
| API | Application Programming Interface |
| FinTech | Financial Technology |
| UI | User Interface |
| UX | User Experience |
| DL | Deep Learning |
| ios | iPhone Operating System |

1. System Design

1.1 Alternative Designs and Methods

Several alternative design approaches were evaluated during the system design phase, including system architecture, system decomposition, and algorithm selection. The evaluation criteria focused on scalability, maintainability, system complexity, and suitability for intelligent data processing.

A monolithic architecture was initially considered due to its simplicity; however, it was excluded because of limited scalability and maintenance difficulties. Consequently, a layered client–server architecture was adopted, as it provides better separation of concerns, supports system scalability, and facilitates integration with external services such as APIs and analytics modules.

For system decomposition, a pipeline (data-flow) model was evaluated but found unsuitable for systems requiring complex user interactions and persistent data management. Therefore, object-oriented decomposition (OOD) was selected, as it aligns with the system’s object-based structure and enhances modularity and maintainability.

Regarding algorithm selection, traditional forecasting methods such as linear regression and ARIMA were considered; however, they were rejected due to their limited ability to model complex temporal patterns. Instead, the LSTM model was selected for its effectiveness in capturing long-term dependencies in time-series data [4]. Additionally, a rule-based approach for invoice data extraction was evaluated, but an NLP-based method was chosen due to its flexibility and robustness when handling varying email formats [6].

Table 1: Comparison of Alternative Designs and Selected Approaches

| Aspect | Alternative Option | Limitations | Selected Approach | Justification |
|-------------------------|-------------------------------|--|-------------------------------------|---|
| System Architecture | Monolithic Architecture | Limited scalability, high coupling, difficult maintenance | Layered Client–Server Architecture | Better separation of concerns, scalability, and easier integration with external services |
| System Decomposition | Pipeline (Data-Flow) Model | Not suitable for complex user interactions and persistent data | Object-Oriented Decomposition (OOD) | Improved modularity and maintainability |
| Forecasting Algorithm | Linear Regression / ARIMA | Limited ability to model complex temporal patterns | LSTM | Effective modeling of long-term dependencies in time-series data |
| Invoice Data Extraction | Rule-based (Regex) Extraction | Low flexibility and sensitive to email format changes | NLP-based Extraction | Robust and flexible handling of varying email formats |

1.2 Graphical User Interface Design

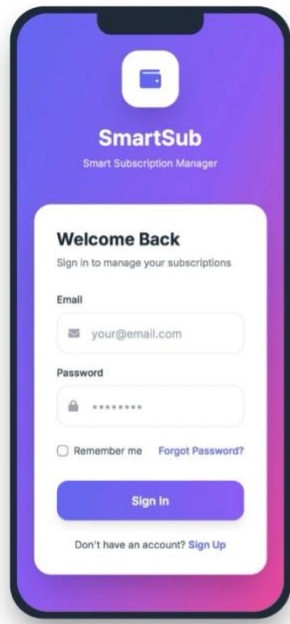


Figure 1: The login screen allows users to securely sign in to the SmartSub application.

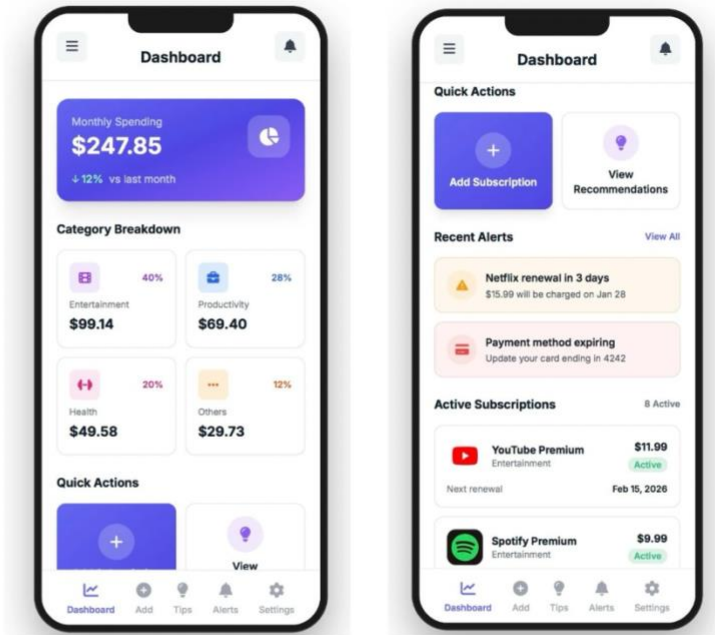


Figure 2: The dashboard screens present an overview of the user's subscription data, including monthly spending analytics, category breakdowns, quick actions, recent alerts, and a list of active subscriptions for efficient monitoring and management.

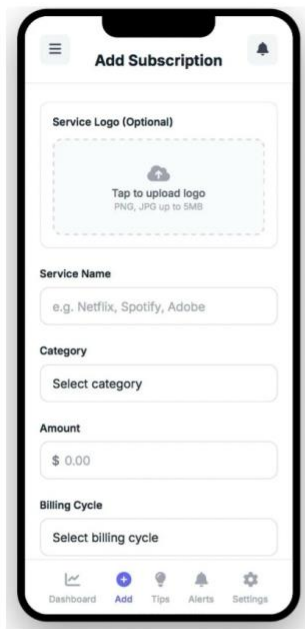


Figure 3: shows the Add Subscription screen, where users can enter subscription details such as the service name, category, amount, billing cycle, and optionally upload a service logo.

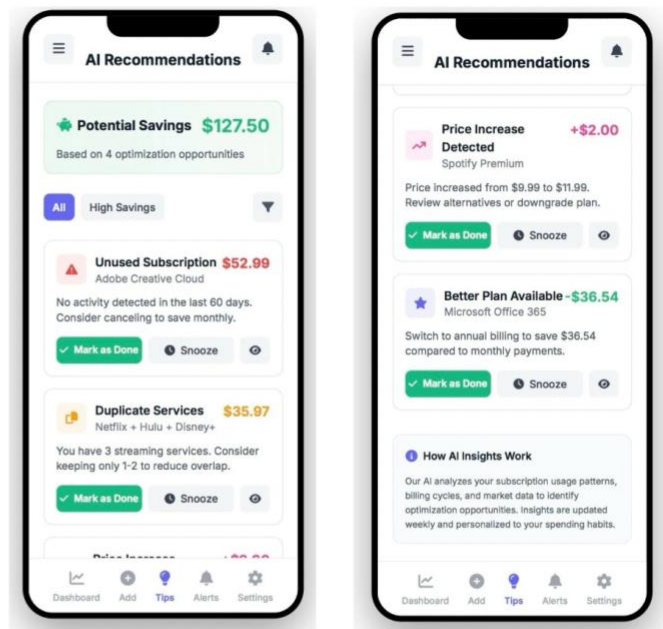


Figure 4: presents the AI Recommendations screens, highlighting price changes, unused or duplicate subscriptions, and potential savings, with quick actions to help users manage and optimize their subscriptions efficiently.

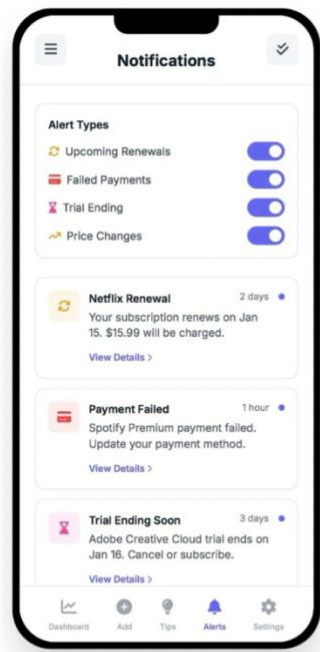


Figure 5: shows the Notifications screen, where users can manage alert preferences and view timely updates for renewals, failed payments, trials ending, and price changes.

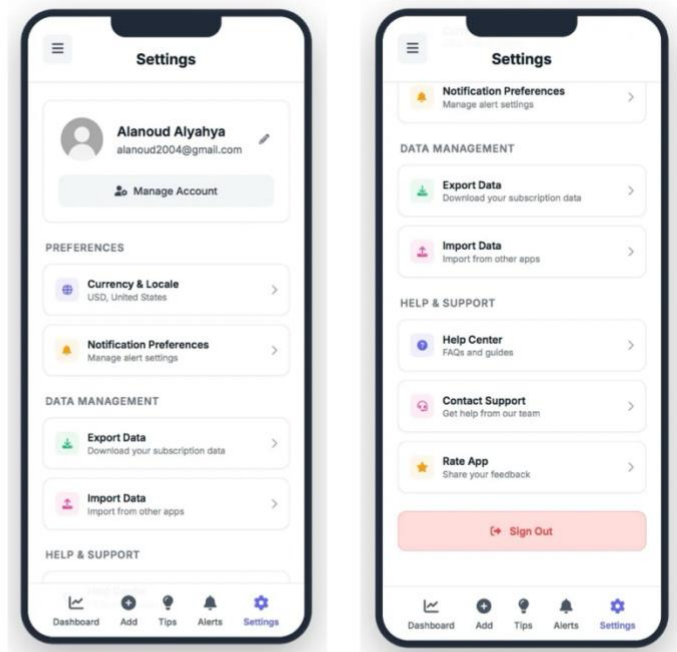


Figure 6: shows the Settings screens, where users can manage their account information, preferences, data import/export options, access help and support, and sign out of the application.

2. System Architecture

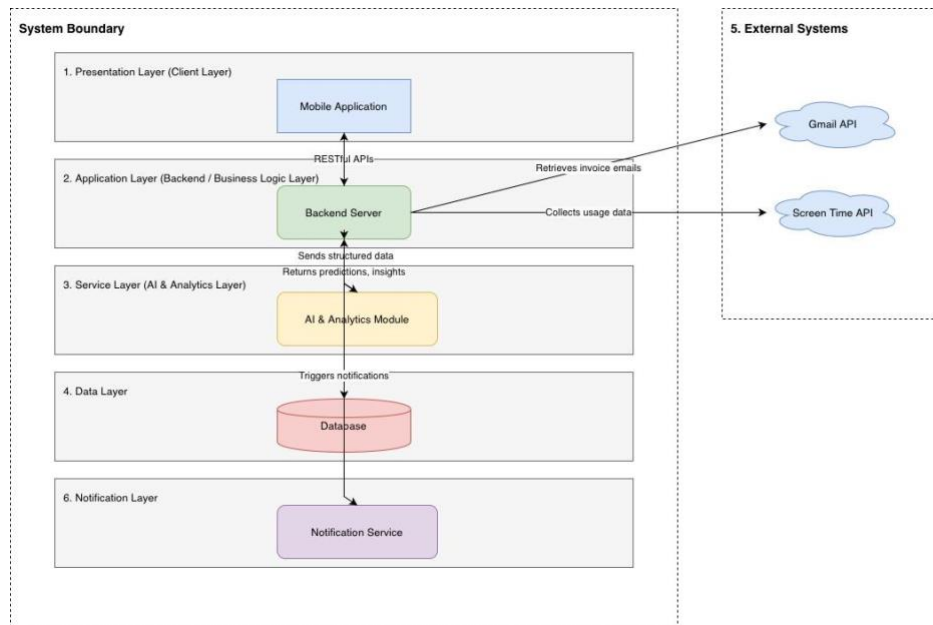


Figure 7: System architecture

The system architecture of the Smart Subscription Manager is designed using a layered architecture combined with a client–server model. This design separates system responsibilities into independent layers, improving scalability, maintainability, and system performance. The architecture consists of several interconnected subsystems that work together to collect subscription data, analyze user behavior, and provide intelligent insights and notifications.

The system is composed of the following main subsystems:

2.1 Presentation Layer (Client Layer)

The presentation layer represents the mobile application used by end users. It provides the graphical user interface through which users can register, log in, manage subscriptions, view dashboards, and receive recommendations and alerts. This layer sends user requests to the backend server through RESTful APIs and displays the processed results received from the system.

2.2 Application Layer (Backend / Business Logic Layer)

The application layer acts as the central controller of the system. It handles user requests, validates inputs, manages subscription data, and coordinates communication between internal components and external services. The backend server processes business logic such as subscription management, authentication, data synchronization, and API communication.

2.3 Service Layer (AI & Analytics Layer)

The service layer is responsible for intelligent data processing and analysis. It includes AI and analytics modules used for subscription classification, usage–cost analysis, and expense prediction. Natural Language Processing (NLP) techniques are applied to extract subscription information from invoice emails, while machine learning models analyze user behavior to generate personalized recommendations and forecasts.

2.4 Data Layer

The data layer manages persistent data storage. It stores user accounts, subscription records, invoice data, usage information, analytics results, and notification history. The database ensures data consistency, fast retrieval, and secure storage of all system information.

2.5 Notification Layer

The notification layer is responsible for delivering alerts and reminders to users. It generates renewal notifications, budget alerts, and recommendation messages based on system analysis and predefined rules.

2.6 External Systems Integration

The system integrates with external services to automate data collection. The Gmail API is used to retrieve invoice emails and detect subscription payments automatically. The Screen Time API is used to collect application usage data, allowing the system to evaluate subscription value by comparing usage with cost.

2.7 Subsystem Communication

The system follows a layered client–server communication pattern. The mobile application communicates with the backend server through REST APIs. The backend server interacts with the database for data storage, communicates with external APIs for data retrieval, and sends processed data to the AI and analytics module. The generated results are returned to the mobile application and notification service for user interaction.

2.8 Design Justification

The layered architecture was selected because it separates user interface, business logic, analytics, and data management into independent components. This improves system maintainability, allows future expansion such as adding new APIs or AI models, and reduces system complexity.

3. Database Design

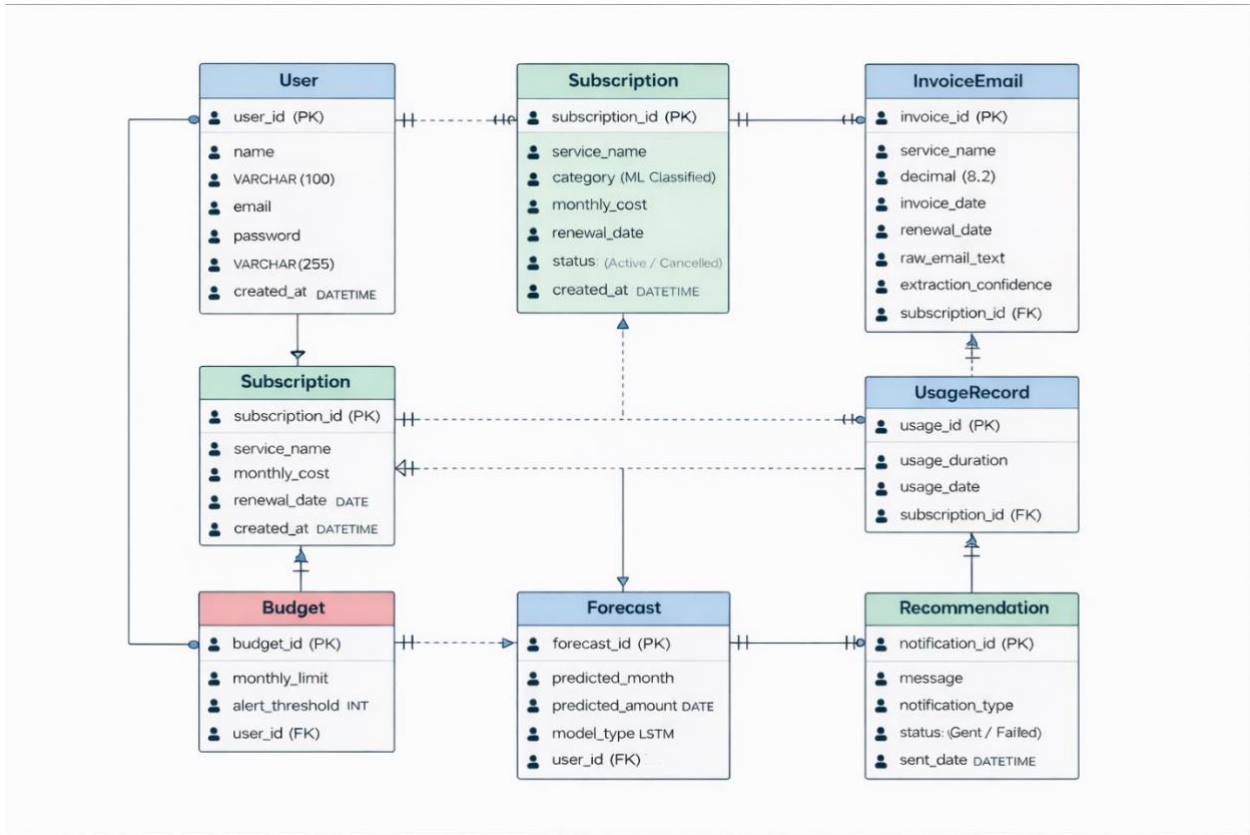


Figure 8: ER Diagram

Data Dictionary

1. User

| Field | Type | Key | Description |
|-----------------------|--------------|--------|------------------------|
| user_id | INT | PK | Unique user identifier |
| name | VARCHAR(100) | — | User full name |
| email | VARCHAR(150) | Unique | User email |
| password | VARCHAR(255) | — | Hashed password |
| account_status | VARCHAR(20) | — | Active / Suspended |
| created_at | DATETIME | — | Account creation date |

2. Subscription

| Field | Type | Key | Description |
|-----------------|--------------|-----|--------------------------|
| subscription_id | INT | PK | Unique subscription ID |
| service_name | VARCHAR(100) | — | Service name |
| category | VARCHAR(50) | — | Category (ML classified) |
| monthly_cost | DECIMAL(8,2) | — | Monthly cost |
| renewal_date | DATE | — | Renewal date |
| status | VARCHAR(20) | — | Active / Cancelled |
| created_at | DATETIME | — | Added date |
| user_id | INT | FK | Reference to User |

3. InvoiceEmail

| Field | Type | Key | Description |
|-----------------------|--------------|-----|----------------------|
| invoice_id | INT | PK | Invoice identifier |
| service_name | VARCHAR(100) | — | Service name |
| amount | DECIMAL(8,2) | — | Invoice amount |
| invoice_date | DATE | — | Invoice date |
| renewal_date | DATE | — | Renewal date |
| raw_email_text | TEXT | — | Raw email content |
| extraction_confidence | DECIMAL(4,2) | — | NLP confidence score |
| subscription_id | INT | FK | Related subscription |

4. UsageRecord

| Field | Type | Key | Description |
|-----------------|------|-----|--------------------------|
| usage_id | INT | PK | Usage record ID |
| usage_duration | INT | — | Usage duration (minutes) |
| usage_date | DATE | — | Date |
| subscription_id | INT | FK | Related subscription |

5. Budget

| Field | Type | Key | Description |
|-----------------|--------------|-----|-------------------|
| budget_id | INT | PK | Budget ID |
| monthly_limit | DECIMAL(8,2) | — | Monthly limit |
| alert_threshold | INT | — | Alert percentage |
| user_id | INT | FK | Reference to User |

6. Forecast

| Field | Type | Key | Description |
|------------------|--------------|-----|-------------------|
| forecast_id | INT | PK | Forecast ID |
| predicted_month | VARCHAR(20) | — | Forecasted month |
| predicted_amount | DECIMAL(8,2) | — | Predicted cost |
| model_type | VARCHAR(50) | — | LSTM |
| user_id | INT | FK | Reference to User |

7. Recommendation

| Field | Type | Key | Description |
|----------------------------|--------------|-----|---------------------------|
| recommendation_id | INT | PK | Recommendation ID |
| recommendation_type | VARCHAR(50) | — | Cancel / Keep / Downgrade |
| reason | TEXT | — | Recommendation reason |
| confidence_score | DECIMAL(4,2) | — | AI confidence |
| subscription_id | INT | FK | Related subscription |

8. Notification

| Field | Type | Key | Description |
|--------------------------|-------------|-----|-------------------|
| notification_id | INT | PK | Notification ID |
| message | TEXT | — | Notification text |
| notification_type | VARCHAR(50) | — | Renewal / Budget |
| status | VARCHAR(20) | — | Sent / Failed |
| sent_date | DATETIME | — | Sent time |
| user_id | INT | FK | Reference to User |

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