





University of Petroleum and Energy Studies

Internship - High Level Design

On

Efficient Drug Discovery using Molecular Deep Learning

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1. Introduction

1.1 Scope of the Document

This document provides a comprehensive overview of the high-level design for the drug discovery system using molecular deep learning. It details the architecture, components, data design, interfaces, and non-functional requirements, ensuring a thorough understanding for all stakeholders involved in the project.

1.2 Intended Audience

This document is intended for:

- **Software Developers**: Responsible for implementing system components and modules.
- **Data Scientists**: Focused on developing and training machine learning models.
- **System Architects**: Tasked with designing the overall system architecture.
- **Project Managers**: Overseeing project development to ensure objectives are met.
- Quality Assurance: Ensuring the system components are tested and validated.
- **Regulatory Personnel**: Ensuring compliance with industry standards and regulations.

1.3 System Overview

The system leverages deep learning models to identify potential drug candidates and predict their toxicity. It integrates various datasets related to drug properties and toxicology, implementing both client-side and server-side validations. Key modules include data preprocessing, model training, model evaluation, and result visualization.





2. System Design

2.1 Application Design

The application is designed to be modular, scalable, and user-friendly, ensuring efficient handling of large datasets and complex computations. Key components include:

- **Frontend**: A user interface for data upload, model training, and result visualization.
- **Backend**: Handles data processing, model training, predictions, and database interactions.
- **Database**: Stores drug information, toxicology data, and model results.
- **APIs**: Facilitate communication between frontend and backend, and with external systems.

2.2 Process Flow

- 1. User Uploads Data: The user uploads datasets via the frontend.
- **2. Data Validation**: Both client-side and server-side validations are performed.
- **3. Data Preprocessing**: The backend preprocesses the data for model training.
- **4. Model Training**: The preprocessed data is used to train the deep learning models.
- **5. Predictions**: The trained models predict drug candidates and their toxicity.
- **6. Result Visualization**: The predictions are displayed to the user on the frontend.

2.3 Information Flow

- **Data Input**: Users upload datasets (e.g., drug properties, toxicology data).
- **Processing**: Backend processes data and trains models.
- **Output**: Predictions and evaluations are stored in the database and displayed on the frontend.

2.4 Components Design

• Frontend Components:





- **DataUploadComponent**: Handles data upload and validation.
- o **ModelTrainingComponent**: Manages model selection, parameter input, and training initiation.
- ResultsComponent: Displays model predictions and evaluation metrics.

• Backend Components:

- o **DataProcessingService**: Cleans and preprocesses input data.
- o **ModelTrainingService**: Manages model training, evaluation, and saving trained models.
- o **PredictionService**: Handles model inference and returns predictions.

2.5 Key Design Considerations

- **Scalability**: Ensure the system can handle increasing amounts of data and users.
- **Modularity**: Design components to be reusable and easily maintainable.
- **Performance**: Optimize processing time and response rates.
- **Security**: Implement robust security measures for data protection.
- Usability: Ensure the system is user-friendly and accessible to non-technical users.

2.6 API Catalogue

- **Data Upload API**: Allows users to upload datasets.
- Model Training API: Initiates model training and returns status.
- **Prediction API**: Provides model predictions based on input data.
- **Results API**: Retrieves and displays prediction results.
- Authentication API: Manages user authentication and authorization.





3. Data Design

3.1 Data Model

The database schema includes tables for drug information, toxicology data, and model results.

- **Drug Information Table**: Stores data on drugs including name, properties, and known targets.
 - o Fields: drug_id, drug_name, smiles, molecular_weight, targets.
- **Toxicology Data Table**: Contains toxicity information of various compounds.
 - Fields: compound_id, toxicity_level, in_vitro_results, in vivo results.
- **Model Results Table**: Stores the results of model predictions and evaluations.
 - Fields: result_id, drug_id, predicted_toxicity, prediction_date, model_version.

3.2 Data Access Mechanism

- Admin: Full access to all tables for data management and model configuration.
- User: Restricted access to view data and upload new datasets.
- **Guest**: Limited access to view public data and results.

3.3 Data Retention Policies

- **Data Retention**: Define policies for retaining and archiving data.
- **Backup**: Regular backups to prevent data loss.
- **Data Purging**: Remove outdated or irrelevant data periodically.

3.4 Data Migration

- **Migration Scripts**: Develop scripts for migrating data from legacy systems.
- Validation: Ensure data integrity during and after migration.
- **Testing**: Conduct thorough testing to validate migrated data.





4. Interfaces

The system interfaces with external APIs, model libraries, and visualization tools to ensure seamless integration and efficient operation. Detailed descriptions include:

4.1 External APIs

- **Drug and Toxicology Data APIs**: Integrate with external databases to fetch drug properties and toxicology data.
- Authentication APIs: Ensure secure user authentication and authorization processes.
- **Data Upload and Retrieval APIs**: Facilitate data upload and retrieval from external sources.

4.2 Model Libraries

- **TensorFlow/PyTorch**: Utilize these libraries for implementing and training deep learning models.
- **Scikit-learn**: Implement additional machine learning algorithms and preprocessing tools.

4.3 Visualization Tools

- Matplotlib: Generate static, animated, and interactive visualizations.
- **Seaborn**: Create informative and attractive statistical graphics.
- **Plotly/D3.js**: Develop interactive web-based visualizations for model predictions and evaluations.





5. State and Session Management

5.1 Session Management

- Flask/Django Sessions: Manage user sessions using Flask or Django frameworks.
- **Session Storage**: Securely store session data with appropriate expiration policies.
- **JWT (JSON Web Tokens)**: Implement JWT for secure session handling and token-based authentication.

5.2 User Roles and Permissions

- Admin Role: Full access to all system features and data.
- User Role: Restricted access to data upload and viewing of results.
- Guest Role: Limited access to viewing public data and results.





6. Caching

6.1 Caching Mechanism

- **Redis**: Implement Redis for efficient caching of frequently accessed data and model results.
- **In-Memory Caching**: Store frequently used data in memory to reduce database load and improve performance.

6.2 Cache Strategies

- **Time-Based Expiration**: Set time-based expiration policies for cached data to ensure freshness.
- Cache Invalidation: Implement strategies to invalidate and update the cache when underlying data changes.





7. Non-Functional Requirements

7.1 Security Aspects

- **Data Encryption**: Encrypt sensitive data in transit and at rest to protect against unauthorized access.
- Access Control: Implement role-based access control to restrict data access based on user roles.
- **Monitoring and Logging**: Continuous monitoring and logging of system activities to detect and respond to security incidents.
- **Regulatory Compliance**: Ensure compliance with data privacy regulations (e.g., GDPR) and industry standards.

7.2 Performance Aspects

- **Scalability**: Design the system to handle increased data volumes and user loads by scaling horizontally and vertically.
- **Optimization**: Optimize algorithms, database queries, and data processing to enhance performance.
- **Load Balancing**: Distribute workloads evenly across multiple servers to prevent bottlenecks and ensure high availability.
- **Response Time**: Ensure the system responds promptly to user requests, aiming for low latency and high throughput.





8. References

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