

- Project Phase 1 : September 08, 2024
- Project Phase 2 : September 23, 2024
- Project Phase 3 : September 29, 2024
- + Project Phase 4 : .....UPDATING.....

Samrat Baral Algorithm and Data Structures

# Organ Matching and Donation Networks

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## Project Phase 1 Deliverable: Data Structure Design and Implementation

### Application Context

Organ donation is a crucial area in healthcare, potentially saving countless lives. The challenge is to efficiently match donors with recipients based on various criteria, including blood type, HLA matching, geographic proximity, and urgency. An effective algorithm can leverage real-time data for optimal matching, ensuring timely transplants as the demand for organs increases.

### Design

The architecture of the donation matching system needs to prioritize fast lookups and manage urgency and optimal matches based on location.

### Donor and Recipient Requirements

- **Authentication:** Users must verify their identities using medical records, adhering to HIPAA regulations and relevant laws to ensure secure access to health information.

### Potential Input Requirements

1. Blood Type
2. Human Leukocyte Antigen (HLA)
3. Geographic Proximity
4. Urgency
5. HCT levels (optional)

### Potential Data Structures

#### 1. Priority Queue (Heap)

- **Insertion/Deletion:**  $O(\log n)$
- Consider using a Fibonacci heap for frequent insertions with an amortized time of  $O(1)$ .

#### 2. Hash Table/Maps

- Maps blood types to donor/recipient lists.
- **Time Complexity:**  $O(1)$

### 3. Binary Tree (Balanced Tree or Red-Black Tree)

- For storing user information and urgency levels.
- **Time Complexity:**  $O(\log n)$

### 4. Graphs

- Use Dijkstra's algorithm for finding optimal paths based on geographic data.
- **Time Complexity:**  $O(\log n) + O(V + E \log V)$

## Key Ideas on Data Structures

### Priority Queue

```
class Recipient:
    def __init__(self, id, blood_type, urgency):
        self.id = id
        self.blood_type = blood_type
        self.urgency = urgency

    def __lt__(self, other):
        return self.urgency > other.urgency # Higher urgency gets higher priority

class UrgencyQueue:
    def __init__(self):
        self.heap = []

    def add_recipient(self, recipient):
        heapq.heappush(self.heap, recipient)

    def get_highest_priority(self):
        return heapq.heappop(self.heap) if self.heap else None
```

### Hash Table

```
class Donor:
    def __init__(self, id, blood_type, hla_type):
        self.id = id
        self.blood_type = blood_type
        self.hla_type = hla_type

class DonorHashMap:
    def __init__(self):
        self.hash_table = {blood_type: [] for blood_type in ["A", "B", "AB", "O"]}

    def add_donor(self, donor):
        self.hash_table[donor.blood_type].append(donor)

    def get_donors_by_blood_type(self, blood_type):
        return self.hash_table.get(blood_type, [])
```

## Binary Tree (AVL Tree)

```
class AVLNode:
    def __init__(self, recipient):
        self.recipient = recipient
        self.height = 1
        self.left = None
        self.right = None

class AVLTree:
    def __init__(self):
        self.root = None

    def insert(self, recipient):
        # Insertion logic goes here, including balancing the tree
        pass
```

## Architecture: Estimation and Potential Requirements

1. **Donor and Recipient Entry:** Incorporate privacy measures and smart contracts.
2. **Matching Criteria:** Implement real-time matching, considering unmatched donors.
3. **Network Security:** Leverage blockchain technology for secure data handling.
4. **Kidney Paired Exchange (KPE):** Utilize KPE algorithms to optimize organ exchanges.
5. **Compliance:** Adhere to ethical and regulatory standards.
6. **Machine Learning:** Implement predictive models for optimal donor selection.

## Python Library Implementation

- **heapq:** For priority queues.
- **sortedcontainers:** For sorted data structures.
- **bintree:** For balanced trees.
- **NetworkX:** For graph-related operations.
- **SQLAlchemy:** For database interactions.
- **scikit-learn:** For machine learning implementations.
- **NumPy** and **Pandas:** For data manipulation.
- **schedule:** For scheduling periodic tasks.

## Potential Challenges and Limitations

- **Scalability:** Addressing hash table collisions and ensuring efficient lookups.
- **Geographic Limitations:** Potentially utilizing AI models for routing.
- **Real-Time Updates:** Managing updates without excessive computational costs.
- **Healthcare Compliance:** Navigating evolving regulations.
- **Data Sources:** Automating data updates from various APIs while ensuring data integrity.

## Python API Use Case

```
import requests
import json

api_endpoint = "https://api.example.com/donors"
headers = {
    "Authorization": "Bearer YOUR_API_KEY",
    "Content-Type": "application/json"
}

def update_data():
    response = requests.get(api_endpoint, headers=headers)
    if response.status_code == 200:
        data = response.json()
        save_to_database(data)
    else:
        print(f"Failed to fetch data: {response.status_code}")

def save_to_database(data):
    # Implement database saving logic here
    pass

if __name__ == "__main__":
    update_data()
    # Schedule the update function to run periodically
```

## Demonstration of Key Operations

The provided test script demonstrates how to add users, establish product relationships, and generate recommendations. Running this script will output the top product recommendation based on the scores assigned.

```
import heapq
from collections import defaultdict

# Recipient class representing the recipient details
class Recipient:
    def __init__(self, id, blood_type, urgency):
        self.id = id
        self.blood_type = blood_type
        self.urgency = urgency

    def __lt__(self, other):
        return self.urgency > other.urgency # Higher urgency gets higher priority

# Priority queue for managing recipient urgency
class UrgencyQueue:
    def __init__(self):
        self.heap = []
```

```
def add_recipient(self, recipient):
    heapq.heappush(self.heap, recipient)

def get_highest_priority(self):
    return heapq.heappop(self.heap) if self.heap else None

# Donor class representing the donor details
class Donor:
    def __init__(self, id, blood_type, hla_type):
        self.id = id
        self.blood_type = blood_type
        self.hla_type = hla_type

# Hash table to manage donors by blood type
class DonorHashMap:
    def __init__(self):
        self.hash_table = defaultdict(list)

    def add_donor(self, donor):
        self.hash_table[donor.blood_type].append(donor)

    def get_donors_by_blood_type(self, blood_type):
        return self.hash_table.get(blood_type, [])

# Simple matching algorithm based on blood type and urgency
def match_donors_to_recipients(donor_map, urgency_queue):
    matches = []
    while urgency_queue.heap:
        recipient = urgency_queue.get_highest_priority()
        donors = donor_map.get_donors_by_blood_type(recipient.blood_type)

        if donors:
            matched_donor = donors.pop(0) # Get the first available donor
            matches.append((recipient.id, matched_donor.id))
        else:
            print(f"No suitable donor found for recipient {recipient.id} with
blood type {recipient.blood_type}.")

    return matches

# Test script demonstrating functionality
if __name__ == "__main__":
    # Initialize data structures
    donor_map = DonorHashMap()
    urgency_queue = UrgencyQueue()

    # Adding donors
    donor1 = Donor(id="donor1", blood_type="A", hla_type="HLA-A1")
    donor2 = Donor(id="donor2", blood_type="B", hla_type="HLA-B1")
    donor3 = Donor(id="donor3", blood_type="A", hla_type="HLA-A2")
    donor_map.add_donor(donor1)
    donor_map.add_donor(donor2)
    donor_map.add_donor(donor3)
```

```
# Adding recipients with varying urgency
recipient1 = Recipient(id="recipient1", blood_type="A", urgency=5)
recipient2 = Recipient(id="recipient2", blood_type="B", urgency=10)
recipient3 = Recipient(id="recipient3", blood_type="A", urgency=8)
urgency_queue.add_recipient(recipient1)
urgency_queue.add_recipient(recipient2)
urgency_queue.add_recipient(recipient3)

# Matching donors to recipients
matches = match_donors_to_recipients(donor_map, urgency_queue)
for recipient_id, donor_id in matches:
    print(f"Matched recipient {recipient_id} with donor {donor_id}.")
```

## Output

```
python3 main.py
Matched recipient recipient2 with donor donor2.
Matched recipient recipient3 with donor donor1.
Matched recipient recipient1 with donor donor3.
```

## Challenges Encountered

- **Integration of Data Structures:** Ensuring the user profiles, product relationships, and recommendations work cohesively posed initial challenges. I tackled this by defining clear interfaces for each data structure, allowing easy data flow.
- **Scoring System:** Developing a straightforward and effective scoring mechanism for recommendations required careful thought. I implemented a basic scoring method, which can be refined later.

## Solutions Implemented

- **Modular Design:** Each data structure was designed with clear responsibilities, making future expansions easier.
- **Error Handling:** Basic error handling was included to manage cases such as duplicate users or missing product relationships.

## Next Steps

1. **Enhance User Interaction Tracking:** Implement a more sophisticated method for tracking user interactions and preferences to improve recommendation accuracy.
2. **Advanced Recommendation Algorithms:** Explore collaborative filtering and content-based filtering techniques for generating more personalized recommendations.
3. **User Interface Development:** Develop a simple UI to facilitate user interaction with the recommendation system.
4. **Testing and Validation:** Conduct extensive testing to validate functionality, performance, and scalability as the application grows.

# GitHub Repository

The full codebase, including the implementation of data structures and the test script, is available in the following GitHub repository: [MSCS532\\_Project](#)

This example includes a priority queue for recipients, a hash table for donors, and a basic structure for geographical matching using a simplified approach.

## Project Phase 2 Deliverable 2: Proof of Concept Implementation

### 1. Recipient Class and Priority Queue

```
import heapq

class Recipient:
    def __init__(self, id, blood_type, urgency, location):
        self.id = id
        self.blood_type = blood_type
        self.urgency = urgency
        self.location = location #new

    def __lt__(self, other):
        return self.urgency > other.urgency # Higher urgency gets higher priority

class UrgencyQueue:
    def __init__(self):
        self.heap = []

    def add_recipient(self, recipient):
        heapq.heappush(self.heap, recipient)

    def get_highest_priority(self):
        return heapq.heappop(self.heap) if self.heap else None
```

### 2. Donor Class and Hash Table

```
class Donor:
    def __init__(self, id, blood_type, hla_type, location):
        self.id = id
        self.blood_type = blood_type
        self.hla_type = hla_type
        self.location = location

class DonorHashMap:
    def __init__(self):
        self.hash_table = {blood_type: [] for blood_type in ["A", "B", "AB", "O"]}

    def add_donor(self, donor):
        self.hash_table[donor.blood_type].append(donor)
```

```
def get_donors_by_blood_type(self, blood_type):  
    return self.hash_table.get(blood_type, [])
```

### 3. Simple Geographical Matching Function

Installation: [GeoPy](#)

```
pip install geopy
```

```
from geopy.distance import geodesic  
  
def find_best_match(recipient, donors):  
    best_match = None  
    best_distance = float('inf')  
  
    for donor in donors:  
        if donor.blood_type == recipient.blood_type: # Check blood type  
            compatibility  
                distance = geodesic(recipient.location, donor.location).miles  
                if distance < best_distance:  
                    best_distance = distance  
                    best_match = donor  
  
    return best_match
```

### 4. Proof of Concept Usage

```
if __name__ == "__main__":  
    # Create a queue for recipients  
    recipient_queue = UrgencyQueue()  
  
    # Add recipients  
    recipient1 = Recipient("rec1", "A", 5, (40.7128, -74.0060)) # New York  
    recipient2 = Recipient("rec2", "B", 10, (34.0522, -118.2437)) # Los Angeles  
    recipient_queue.add_recipient(recipient1)  
    recipient_queue.add_recipient(recipient2)  
  
    # Create a hash map for donors  
    donor_map = DonorHashMap()  
  
    # Add donors  
    donor1 = Donor("don1", "A", "HLA1", (41.8781, -87.6298)) # Chicago  
    donor2 = Donor("don2", "B", "HLA2", (34.0522, -118.2437)) # Los Angeles  
    donor_map.add_donor(donor1)  
    donor_map.add_donor(donor2)
```



```
# Process matching for the highest priority recipient
highest_priority_recipient = recipient_queue.get_highest_priority()
donors_of_same_blood_type =
donor_map.get_donors_by_blood_type(highest_priority_recipient.blood_type)
best_match = find_best_match(highest_priority_recipient,
donors_of_same_blood_type)

if best_match:
    print(f"Best match for {highest_priority_recipient.id}: Donor ID
{best_match.id} at location {best_match.location}")
else:
    print(f"No suitable donor found for {highest_priority_recipient.id}.")
```

## Explanation

- **Recipient Class:** Holds information about each recipient, including urgency and location.
- **UrgencyQueue Class:** Implements a priority queue to manage recipients based on urgency.
- **Donor Class:** Stores donor information, including blood type and location.
- **DonorHashMap Class:** Uses a hash table to organize donors by blood type.
- **find\_best\_match Function:** Compares recipients and donors based on blood type and geographic distance, returning the closest suitable donor.

## Future Enhancements

- **Integration with Databases:** Store recipient and donor data in a database for persistence.
- **Machine Learning:** Implement predictive matching algorithms based on historical data.
- **Real-time Updates:** Use webhooks or similar methods to update donor/recipient information in real time.
- **Advanced Geolocation:** Enhance the geographical matching function with more sophisticated routing algorithms.

## Phase 3: Optimization, Scaling, and Final Evaluation

1. Optimization of Data Structures Analyze performance and identify inefficiencies. Implement optimizations like caching frequently accessed user profiles.
2. Scaling for Large Datasets Modify implementations to manage larger datasets effectively, ensuring acceptable performance levels.
3. Advanced Testing and Validation Develop comprehensive test cases and perform stress testing to evaluate robustness.
4. Final Evaluation and Performance Analysis Compare the final implementation with the initial proof of concept, discussing strengths, limitations, and areas for improvement.

Here are more detailed future ideas on how to implement the future enhancements for your **Organ Matching and Donation Network**:

### 3.1 Integration with Databases

**Objective:** Store recipient and donor data for persistence, enabling easy retrieval and management.

### Implementation Steps:

- **Choose a Database:** Select a relational database (e.g., PostgreSQL, MySQL) or a NoSQL database (e.g., MongoDB) based on your data structure and querying needs.
- **Database Models:** Define models for Donor and Recipient, using an ORM like SQLAlchemy for relational databases or a library like PyMongo for NoSQL.

```
from sqlalchemy import create_engine, Column, String, Integer, Float
from sqlalchemy.ext.declarative import declarative_base
from sqlalchemy.orm import sessionmaker

Base = declarative_base()

class Donor(Base):
    __tablename__ = 'donors'
    id = Column(String, primary_key=True)
    blood_type = Column(String)
    hla_type = Column(String)
    location_lat = Column(Float)
    location_lon = Column(Float)

class Recipient(Base):
    __tablename__ = 'recipients'
    id = Column(String, primary_key=True)
    blood_type = Column(String)
    urgency = Column(Integer)
    location_lat = Column(Float)
    location_lon = Column(Float)

# Set up the database
engine = create_engine('sqlite:///organ_donation.db')
Base.metadata.create_all(engine)
Session = sessionmaker(bind=engine)
```

- **Data Access Methods:** Implement functions to add, retrieve, and update donor/recipient records.

## 3.2 Machine Learning

**Objective:** Use historical data to improve matching accuracy through predictive algorithms.

### Implementation Steps:

- **Data Collection:** Gather historical donor and recipient data, including successful matches and outcomes.
- **Feature Engineering:** Identify relevant features (e.g., blood type, urgency, location distance) that influence successful matches.

- **Model Training:**

- Use libraries like scikit-learn or TensorFlow to create and train models (e.g., logistic regression, decision trees) on your dataset.

```
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier

# Sample feature and target arrays
X = ... # Features: blood type, distance, urgency
y = ... # Target: match success

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2)

model = RandomForestClassifier()
model.fit(X_train, y_train)
```

- **Predictive Matching:** Integrate the model into your matching logic to prioritize matches based on predicted success rates.

### 3.3 Real-time Updates

**Objective:** Ensure that donor and recipient information is up-to-date without manual intervention.

**Implementation Steps:**

- **Webhooks:** Set up webhooks to listen for changes in external data sources (e.g., hospitals, health organizations).
- **API Integration:**
  - Use libraries like `requests` to fetch real-time updates from APIs.

```
import requests

def fetch_donor_data():
    response = requests.get('https://api.example.com/donors')
    if response.status_code == 200:
        data = response.json()
        # Update the database with new data
    else:
        print(f"Error fetching data: {response.status_code}")
```

- **Scheduled Tasks:** Use a task scheduler (e.g., `schedule` library, Celery) to periodically check for updates.

### 3.4 Advanced Geolocation

**Objective:** Enhance geographic matching with sophisticated algorithms.

## Implementation Steps:

- **Routing Algorithms:** Implement algorithms like Dijkstra's or A\* for more accurate pathfinding based on real traffic data.
- **Geospatial Libraries:** Use libraries like `geopy` or `Shapely` to perform complex geospatial operations.
- **Distance Calculation:**
  - Improve distance calculations by factoring in real-world conditions like traffic patterns.

```
from geopy.distance import geodesic

def calculate_distance(location1, location2):
    return geodesic(location1, location2).miles
```

Integrating blockchain with machine learning (ML) in an organ matching and donation network can provide enhanced security, data integrity, and improved decision-making through predictive analytics. Below is a structured approach to implement this integration along with relevant source references for your research.

## Implementation Overview

### 1. Objectives

- **Blockchain:** Ensure secure and transparent storage of donor and recipient data.
- **Machine Learning:** Use historical data to predict successful matches based on various features.

### 2. System Architecture

- **Blockchain Layer:** Manages donor and recipient data securely.
- **ML Model Layer:** Predicts match success and recommends optimal matches.
- **API Layer:** Interfaces between the blockchain, ML model, and frontend application.

## Implementation Steps

### Step 1: Blockchain Setup

**Smart Contract for Organ Donation** (as previously outlined).

```
contract OrganDonation {
    struct Donor {
        string id;
        string bloodType;
        string hlaType;
        address owner;
    }

    struct Recipient {
```

```

        string id;
        string bloodType;
        uint urgency;
        address owner;
    }

    mapping(string => Donor) public donors;
    mapping(string => Recipient) public recipients;

    event DonorRegistered(string id);
    event RecipientRegistered(string id);

    function registerDonor(string memory _id, string memory _bloodType, string
memory _hlaType) public {
        donors[_id] = Donor(_id, _bloodType, _hlaType, msg.sender);
        emit DonorRegistered(_id);
    }

    function registerRecipient(string memory _id, string memory _bloodType, uint
_urgency) public {
        recipients[_id] = Recipient(_id, _bloodType, _urgency, msg.sender);
        emit RecipientRegistered(_id);
    }
}

```

**Deploy the Smart Contract** using tools like Truffle or Hardhat.

## Step 2: Machine Learning Model Development

**Predictive Model Example** (using Scikit-learn):

1. **Data Preparation:** Gather historical data on donors and recipients.

```

import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score

# Load dataset
data = pd.read_csv('donor_recipient_data.csv')
X = data[['blood_type', 'urgency', 'distance']] # Features
y = data['match_success'] # Target variable

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random_state=42)

# Train model
model = RandomForestClassifier()
model.fit(X_train, y_train)

# Predictions

```

```
y_pred = model.predict(X_test)
print("Accuracy:", accuracy_score(y_test, y_pred))
```

2. **Integration with Blockchain:** After matching is predicted, save the results back to the blockchain.

```
from web3 import Web3

# Connect to Ethereum network
w3 = Web3(Web3.HTTPProvider('http://localhost:8545'))
contract = w3.eth.contract(address='YOUR_CONTRACT_ADDRESS', abi='YOUR_ABI')

def save_match_result(donor_id, recipient_id, match_success, private_key):
    tx = contract.functions.saveMatchResult(donor_id, recipient_id,
match_success).buildTransaction({
        'donor': 1,
        'level': 1,
        'urgency': w3.toWei('1', 'RED'),
        'nonce':
w3.eth.getTransactionCount(w3.eth.account.privateKeyToAccount(private_key).address
    ),
    })

    signed_tx = w3.eth.account.signTransaction(tx, private_key)
    w3.eth.sendRawTransaction(signed_tx.rawTransaction)
```

### Step 3: API Development

Develop a RESTful API (using Flask or FastAPI) to interface with both the blockchain and ML model.

```
from flask import Flask, request, jsonify

app = Flask(__name__)

@app.route('/match', methods=['POST'])
def match():
    data = request.json
    # Predict match using the ML model
    match_success = model.predict([data['blood_type'], data['urgency'],
data['distance']])

    # Save result to blockchain
    save_match_result(data['donor_id'], data['recipient_id'], match_success)

    return jsonify({'match_success': match_success.tolist()})

if __name__ == '__main__':
    app.run(debug=True)
```

Integrating smart contracts, machine learning, and optimizations into the organ matching system requires a more sophisticated approach. Here's how we can outline the implementation for these components, along with an updated code snippet to illustrate these concepts.

## 1. Smart Contract Implementation

Smart contracts can be implemented on a blockchain to ensure data integrity and automate the matching process. For this example, we'll use Solidity for the smart contract and assume you have a blockchain environment set up (like Ethereum or a test network).

### Solidity Smart Contract Example

```
// SPDX-License-Identifier: MIT
pragma solidity ^0.8.0;

contract OrganDonation {
    struct Donor {
        string id;
        string bloodType;
        string hlaType;
        string location;
        bool isAvailable;
    }

    struct Recipient {
        string id;
        string bloodType;
        uint urgency;
        string location;
    }

    mapping(string => Donor) public donors;
    mapping(string => Recipient) public recipients;

    function addDonor(string memory id, string memory bloodType, string memory hlaType, string memory location) public {
        donors[id] = Donor(id, bloodType, hlaType, location, true);
    }

    function addRecipient(string memory id, string memory bloodType, uint urgency, string memory location) public {
        recipients[id] = Recipient(id, bloodType, urgency, location);
    }

    function matchDonor(string memory recipientId) public view returns (string memory) {
        Recipient memory recipient = recipients[recipientId];
        // Add logic to find the best match based on blood type and urgency.
        // This is simplified; implement a proper search algorithm.
        for (string memory donorId in donors) {
            if (donors[donorId].isAvailable &&
```

```

compareBloodType(donors[donorId].bloodType, recipient.bloodType)) {
    return donorId; // Return matching donor ID
}
}
return "No suitable donor found";
}

function compareBloodType(string memory bloodType1, string memory bloodType2)
private pure returns (bool) {
    // Simple blood type comparison logic
    return keccak256(abi.encodePacked(bloodType1)) ==
keccak256(abi.encodePacked(bloodType2));
}
}

```

## 2. Machine Learning Integration

Machine learning can be used to predict the best donor-recipient matches based on historical data. For this, you can utilize libraries such as [scikit-learn](#) in Python.

### Example: Predictive Model for Matching

Here's an outline of how you might implement a simple predictive model:

```

import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier

# Sample data preparation
data = pd.read_csv('organ_donation_data.csv') # Assume this CSV contains
historical match data
features = data[['blood_type_donor', 'blood_type_recipient', 'urgency']]
labels = data['match_success'] # 1 if matched successfully, 0 otherwise

# Train-test split
X_train, X_test, y_train, y_test = train_test_split(features, labels,
test_size=0.2)

# Model training
model = RandomForestClassifier()
model.fit(X_train, y_train)

# Example of making predictions
def predict_match(donor_info):
    donor_df = pd.DataFrame([donor_info]) # Convert donor_info to DataFrame
    return model.predict(donor_df)

# Use the model in the matching process
recipient_info = {
    'blood_type_donor': 'A',
    'blood_type_recipient': 'A',

```



```
        'urgency': 5
    }

    match_prediction = predict_match(recipient_info)
    print(f"Match prediction: {match_prediction[0]}")
```

### 3. Optimizations

#### Code Optimizations

- **Database Indexing:** Ensure the SQLite database has indexes on commonly queried columns like `blood_type`.
- **Efficient Data Structures:** Use more efficient data structures (like a balanced binary tree) for handling recipient and donor lists, allowing for faster lookups.

#### Python Code Enhancements

Here's an updated Python code snippet that incorporates some of these changes, including a more structured approach and the machine learning integration:

```
import heapq
from geopy.distance import geodesic
import pandas as pd
from sklearn.ensemble import RandomForestClassifier
from sqlalchemy import create_engine, Column, String, Integer, Float
from sqlalchemy.ext.declarative import declarative_base
from sqlalchemy.orm import sessionmaker

# Database setup
Base = declarative_base()
engine = create_engine('sqlite:///organ_donation.db')
Session = sessionmaker(bind=engine)
session = Session()

class Recipient(Base):
    __tablename__ = 'recipients'
    id = Column(String, primary_key=True)
    blood_type = Column(String)
    urgency = Column(Integer)
    location = Column(String)

class Donor(Base):
    __tablename__ = 'donors'
    id = Column(String, primary_key=True)
    blood_type = Column(String)
    hla_type = Column(String)
    location = Column(String)

Base.metadata.create_all(engine)
```

```

class UrgencyQueue:
    def __init__(self):
        self.heap = []

    def add_recipient(self, recipient):
        heapq.heappush(self.heap, recipient)

    def get_highest_priority(self):
        return heapq.heappop(self.heap) if self.heap else None

def find_best_match(recipient, donors):
    best_match = None
    best_distance = float('inf')

    for donor in donors:
        if donor.blood_type == recipient.blood_type: # Check blood type
            # compatibility
            distance = geodesic(recipient.location, donor.location).miles
            if distance < best_distance:
                best_distance = distance
                best_match = donor

    return best_match

# Machine Learning Model Preparation
data = pd.read_csv('organ_donation_data.csv')
features = data[['blood_type_donor', 'blood_type_recipient', 'urgency']]
labels = data['match_success']

# Train the model
model = RandomForestClassifier()
model.fit(features, labels)

# Example Matching Process
def match_recipients_to_donors(recipient_queue):
    while recipient_queue.heap:
        recipient = recipient_queue.get_highest_priority()
        donors = session.query(Donor).filter(Donor.blood_type ==
recipient.blood_type).all()

        if donors:
            best_match = find_best_match(recipient, donors)
            if best_match:
                # Use machine learning model to predict match success
                prediction = model.predict([[best_match.blood_type,
recipient.blood_type, recipient.urgency]])
                if prediction[0] == 1: # If the prediction is successful
                    print(f"Matched recipient {recipient.id} with donor
{best_match.id}.")
                else:
                    print(f"No suitable donor found for recipient
{recipient.id}.")
            else:
                print(f"No donors available for recipient {recipient.id}.")

```

```
# Main Execution Flow
if __name__ == "__main__":
    recipient_queue = UrgencyQueue()

    # Add recipients and donors to the database (omitted for brevity)
    # Call match_recipients_to_donors(recipient_queue)
```

## Conclusion

Incorporating smart contracts, machine learning, and optimization techniques significantly enhances the functionality and robustness of the organ matching system. This approach not only ensures data integrity and security but also leverages predictive analytics for better matching outcomes. Future work can focus on refining these components and expanding the system's capabilities.

## Conclusion

By integrating blockchain and machine learning, you create a secure, transparent, and intelligent organ matching system. This approach not only enhances data security but also improves the efficiency of the matching process. The provided research papers will offer deeper insights into this intersection of technologies.

- **Integration with Mapping APIs:** Use services like Google Maps API to retrieve real-time data about distances and estimated travel times.

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