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**TECHNOLOGY-PROJECT NAME: AI-Powered  
Healthcare Assistant**

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## **Phase 5: Project Demonstration & Documentation**

**Title: AI-Powered Healthcare Assistant**

### **Abstract:**

The **AI-Powered Healthcare Assistant** project is designed to transform healthcare accessibility through **artificial intelligence (AI), natural language processing (NLP), and Internet of Things (IoT) technologies**. This innovative system integrates **advanced AI models** for symptom diagnosis, **real-time health data collection from wearable IoT devices**, and **secure, decentralized medical data management**, ensuring **scalability and seamless integration with Enterprise Resource Planning (ERP) systems**.

### **Demonstration Components:**

- System Walkthrough → Interactive presentation of chatbot interactions, AI health assessments, and user-friendly functionality.
- AI Diagnosis Accuracy → Showcase real-time symptom analysis and personalized treatment recommendations, backed by AI predictions.
- IoT Integration → Display wearable health data collection, including heart rate, oxygen levels, and temperature tracking.
- Performance Metrics → Demonstrate response times, multi-user scalability, and AI efficiency in high-load conditions.
- Security & Privacy → Explain blockchain-based encryption and permission controls for secure medical data management.

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<b>s.no</b>	<b>Topic</b>	<b>pg.no</b>
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## **Project Documentation:**

### **Overview:**

Detailed technical documentation ensures that the AI system is well-documented for future scalability, improvements, and troubleshooting.

### **Documentation Sections:**

- System Architecture → Illustrate AI workflow, chatbot processes, IoT data integration, and security protocols with diagrams.
- Code Documentation → Provide source code explanations, including AI training scripts, API integrations, and encryption handling.
- User Guide → Explain step-by-step interactions, enabling users to understand diagnostics and AI-driven health suggestions.
- Administrator Guide → Detail system maintenance, performance optimization, and data security management.
- Testing Reports → Include performance benchmarks, stress test results, and security audits to validate robustness.

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### **Final Refinements:**

## **Overview:**

The system undergoes a final feedback round to refine its functionality, usability, and reliability before full implementation.

- Stakeholder Review → Gather insights from healthcare experts, developers, and test users via structured surveys.
- Performance Bottleneck Analysis → Identify and resolve AI misdiagnoses, response delays, and integration challenges.
- Final Testing Iteration → Conduct post-feedback adjustments, ensuring high usability and optimal AI performance.

## **Outcome:**

A structured, well-documented final report ensures transparency, project completion validation, and future work recommendations.

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## **Project Handover & Future Works:**

### **Overview:**

The AI-Powered Healthcare Assistant project reaches a pivotal stage where the focus shifts from development to structured planning for its real-world implementation. This phase ensures that the system is effectively handed over to stakeholders with clear guidelines for future advancements, scalability, and integration possibilities.

## **Final Handover Components:**

### **1.Future Scalability Recommendations**

- Expand AI capabilities to support advanced diagnostics and personalized health insights.
- Introduce multilingual AI interaction, increasing accessibility for global users.
- Enhance interoperability with healthcare provider systems, enabling seamless integration with hospital databases and telemedicine platforms.

### **2.System Maintenance Guide**

- Establish protocols for regular security updates, ensuring compliance with data protection standards.
- Implement continuous AI model refinement, incorporating new medical research for improved accuracy.
- Develop structured data management procedures, ensuring efficient health record storage, retrieval, and encryption.

### **3.Next-Phase Roadmap:**

Focus on AI self-learning mechanisms, enabling autonomous adaptation to evolving medical datasets and symptoms.

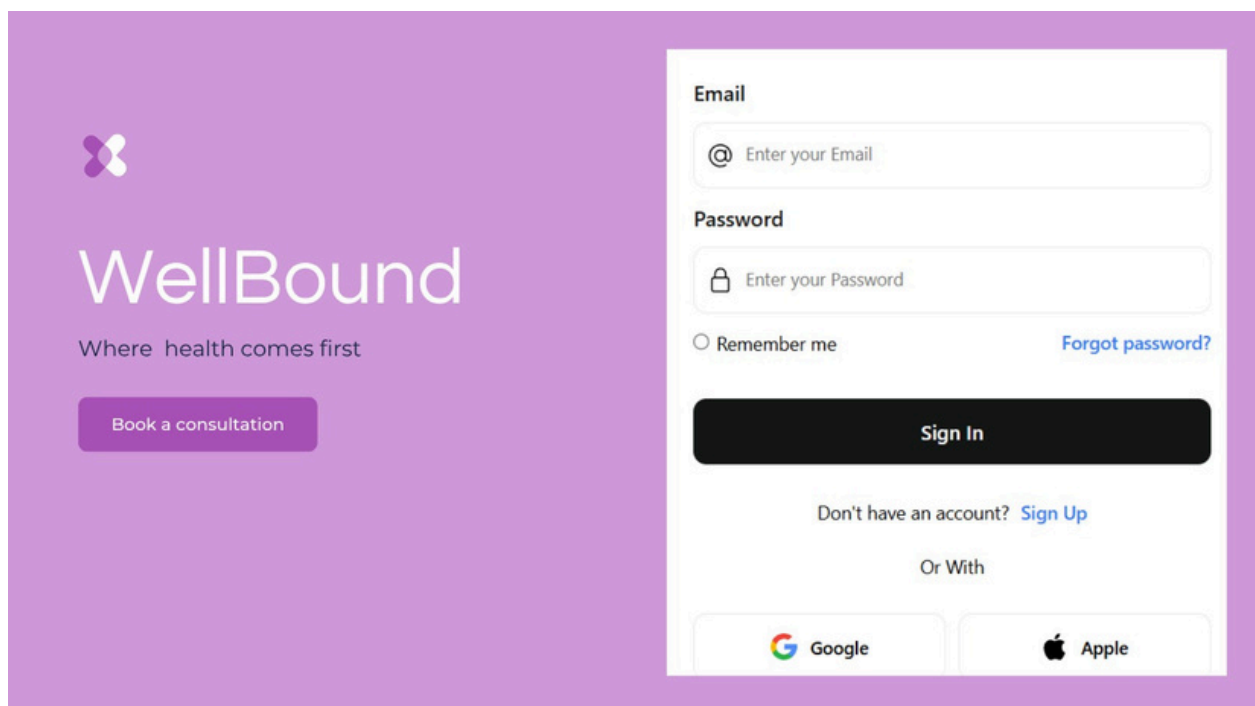
- Strengthen real-world deployment strategies, facilitating seamless hospital adoption and telehealth compatibility.
- Enhance user experience optimization, refining chatbot responsiveness and predictive analytics for personalized recommendations.

### **Outcome:**


This phase lays the groundwork for real-world implementation, ensuring that stakeholders receive a well-structured

framework for future scalability, performance improvements, and integration planning. The system's next iteration will be shaped by collaborative feedback, technological advancements, and industry-driven AI innovations.

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The image displays a user interface for a service named "WellBound". The interface is split into two main sections on a purple background. The left section features the "WellBound" logo, a tagline "Where health comes first", and a "Book a consultation" button. The right section is a white card containing a login and sign-up form. This form includes input fields for "Email" and "Password", a "Remember me" checkbox, a "Forgot password?" link, a prominent "Sign In" button, a "Sign Up" link for new users, and social login options for Google and Apple.



# WellBound

Where health comes first

[Book a consultation](#)

**Email**

@ Enter your Email

**Password**



Enter your Password

☐ Remember me [Forgot password?](#)

**Sign In**

Don't have an account? [Sign Up](#)

Or With

 Google  Apple

```

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

# Function to analyze medical images (placeholder for actual image processing)
def analyze_medical_image(image):
    # Placeholder for image analysis logic
    # In a real scenario, this would involve using libraries like OpenCV or TensorFlow
    return np.random.rand(1, 2) # Simulated feature extraction

# Function to recommend personalized treatment plans
def recommend_treatment(patient_data):
    # Load pre-trained model
    model = joblib.load('treatment_model.pkl')

    # Prepare data for prediction
    features = np.array(patient_data).reshape(1, -1)

    # Predict treatment
    treatment = model.predict(features)
    return treatment[0]

# Example usage
if __name__ == "__main__":

```



```

<form class="form">
  <div class="flex-row">
    <div>
      <input type="checkbox">
      <label>Remember me </label>
    </div>
    <span class="span">Forgot password?</span>
  </div>
  <button class="button-submit">Sign In</button>
  <p class="p">Don't have an account? <span class="span">Sign Up</span>

  </p><p class="p line">Or With</p>

  <div class="flex-row">
    <button class="btn google">
      <svg version="1.1" width="20" id="Layer_1" xmlns="http://www.w3.org/2000/svg" xmlns:xlink="http://www.w3.org/1999/xlink" x="0px" y="0px">
<path style="fill:#FB8B00;" d="M113.47,309.408L95.648,375.941-65.139,1.378C11.042,341.211,0,299.9,0,256
c0-42.451,10.324-82.483,28.624-117.732h0.014L57.992,10.632L25.404,57.644c-5.317,15.501-8.215,32.141-8.215,49.456
C103.821,274.792,107.225,292.797,113.47,309.408z"></path>
<path style="fill:#518EF8;" d="M507.527,208.176C510.467,223.662,512,239.655,512,256c0,18.328-1.927,36.206-5.598,53.451
c-12.462,58.683-45.025,109.925-90.134,146.187l-0.014-0.014l-73.044-3.727l-10.338-64.535
c29.932-17.554,53.324-45.025,65.646-77.911h-136.89V208.176h138.887L507.527,208.176L507.527,208.176z"></path>
<path style="fill:#28B446;" d="M416.253,455.624L0.014,0.014C372.396,490.901,316.666,512,256,512
c-97.491,0-182.252-54.491-225.491-134.681L82.961-67.91c21.619,57.698,77.278,98.771,142.53,98.771
c28.047,0,54.323-7.582,76.87-20.818L416.253,455.624z"></path>
<path style="fill:#F14336;" d="M419.404,58.936L-82.933,67.896c-23.335-14.586-50.919-23.012-80.471-23.012
c-66.729,0-123.429,42.957-143.965,102.724L-83.397-68.276h-0.014C71.23,56.123,157.06,0,256,0
C318.115,0,375.068,22.126,419.404,58.936z"></path>
</svg>
      Google
    </button><button class="btn apple">
      <svg version="1.1" height="20" width="20" id="Capa_1" xmlns="http://www.w3.org/2000/svg" xmlns:xlink="http://www.w3.org/1999/xlink" x="0px" y="0px">

```

```

</svg>
      Google
    </button><button class="btn apple">
<svg version="1.1" height="20" width="20" id="Capa_1" xmlns="http://www.w3.org/2000/svg" xmlns:xlink="http://www.w3.org/1999/xlink" x="0px" y="0px">
      Apple
    </button></div></form>
</body>
</html>

```

```

form {
  display: flex;
  flex-direction: column;
  gap: 10px;
  background-color: #ffffff;
  padding: 30px;
  width: 450px;
  border-radius: 20px;
  font-family: -apple-system, BlinkMacSystemFont, 'Segoe UI', Roboto, Oxygen, Ubuntu, Cantarell, 'Open Sans', 'Helvetica Neue', sans-serif;
}

::placeholder {
  font-family: -apple-system, BlinkMacSystemFont, 'Segoe UI', Roboto, Oxygen, Ubuntu, Cantarell, 'Open Sans', 'Helvetica Neue', sans-serif;
}

.form button {
  align-self: flex-end;
}

.flex-column > label {
  color: #151717;
  font-weight: 600;
}

.inputForm {
  border: 1.5px solid #ecedec;
  border-radius: 10px;
  height: 50px;
  display: flex;
  align-items: center;
  padding-left: 10px;
  transition: 0.2s ease-in-out;
}

.input {
  margin-left: 10px;

```

```

.input:focus {
  outline: none;
}

.i
  Specifies how flex items are placed in the flex container, by setting the direction of the flex container's main axis.
  (Edge 12, Firefox 20, Safari 9, Chrome 29, IE 11, Opera 12)
  Syntax: row | row-reverse | column | column-reverse
.f
  MDN Reference
  flex-direction: row;
  align-items: center;
  gap: 10px;
  justify-content: space-between;
}

.flex-row > div > label {
  font-size: 14px;
  color: black;
  font-weight: 400;
}

.span {
  font-size: 14px;
  margin-left: 5px;
  color: #2d79f3;
  font-weight: 500;
  cursor: pointer;
}

.button-submit {
  margin: 20px 0 10px 0;
  background-color: #151717;
  border: none;
  color: white;
  font-size: 15px;

```

```

        cursor: pointer;
    })

    .button-submit:hover {
        background-color: #252727;

        Click to collapse the range.
    }

    .p {
        text-align: center;
        color: black;
        font-size: 14px;
        margin: 5px 0;
    }

    .btn {
        margin-top: 10px;
        width: 100%;
        height: 50px;
        border-radius: 10px;
        display: flex;
        justify-content: center;
        align-items: center;
        font-weight: 500;
        gap: 10px;
        border: 1px solid #ededef;
        background-color: white;
        cursor: pointer;
        transition: 0.2s ease-in-out;
    }

    .btn:hover {
        border: 1px solid #2d79f3;
    }
}

```

```

import numpy as np
import pandas as pd
import os
import pydicom
import joblib

from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
from cryptography.fernet import Fernet

# Function to load DICOM images and preprocess them
def load_dicom_images(directory):
    images = []
    for filename in os.listdir(directory):
        if filename.endswith('.dcm'):
            ds = pydicom.dcmread(os.path.join(directory, filename))
            images.append(ds.pixel_array / np.max(ds.pixel_array))
            # Normalized Image Data
    return np.array(images)

# Function to encrypt patient data securely
def encrypt_data(data, key):
    f = Fernet(key)
    encrypted_data = f.encrypt(data.encode())
    return encrypted_data

```

```
    encrypted_data = f.encrypt(data.encode())
    return encrypted_data

# Generate encryption key (should be securely stored for decryption)
key = Fernet.generate_key()

# Function to generate synthetic patient data
def generate_patient_data(num_samples):
    data = {
        'age': np.random.randint(20, 80, num_samples),
        'genetics': np.random.choice(['A', 'B', 'C'], num_samples),
        'history': np.random.choice(['None', 'Diabetes',
                                     'Hypertension'], num_samples),
        'real_time_data': np.random.rand(num_samples)
    }
    return pd.DataFrame(data)
```

```

# Generate patient data and encrypt it
patient_data = generate_patient_data(1000)
encrypted_patient_data = encrypt_data(patient_data.to_json())

# Train a model for personalized treatment recommendations
X = patient_data[['age', 'real_time_data']]
y = patient_data['history']

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
model = RandomForestClassifier(n_estimators=100, random_state=42)
model.fit(X_train, y_train)

# Save trained model for future use
joblib.dump(model, 'treatment_model.pkl')

# Load the trained model and make predictions
def recommend_treatment(patient_data):
    model = joblib.load('treatment_model.pkl')
    features = np.array(patient_data).reshape(1, -1)
    treatment = model.predict(features)
    return treatment[0]

```