

HEALTHCARE AI

PROJECT DOCUMENTATION

1.INTRODUCTION

PROJECT TITLE: HEALTH AI: INTELLIGENT HEALTHCARE ASSISTANT

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2.PROJECT OVERVIEW:

An Artificial Intelligence (AI) in Health care project aims to leverage machine learning and other AI technologies to improve patient diagnosis, treatment, and management, enhance operational efficiency, and reduce healthcare costs. These projects involve large medical datasets to detect patterns, predict outcomes, automate administrative tasks, develop personalized treatment plans, accelerate drug discovery, and provide virtual health assistance. Key components include advanced algorithms for medical image analysis, predictive models for disease risk, and AI-powered tools for remote patient monitoring, all while addressing critical challenges like data privacy, bias, and regulatory frameworks.

Features

Key Features of AI in Healthcare

- **Enhanced Diagnostics and Prediction:**
 - **Medical Image Analysis:** AI algorithms can analyze medical images (X-rays, CT scans, etc.) with high accuracy, leading to earlier and more precise disease detection, such as cancer.
 - **Predictive Analytics:** AI systems analyze patient data, including genetic information and lifestyle factors, to predict potential health risks and disease outbreaks, enabling proactive and preventive care.
- **Personalized Medicine and Treatment:**
 - **Tailored Treatment Plans:** By analyzing individual patient data, AI can create personalized treatment plans that are more effective and targeted.
 - **Drug Discovery:** AI accelerates the complex process of drug development and reduces the cost of clinical trials by identifying patterns and predicting outcomes.
- **Improved Patient Care & Engagement:**
 - **Virtual Health Assistants:** AI-powered chatbots provide virtual health support, answer questions, offer health tips, and send appointment reminders.

- **Enhanced Compliance:** AI tools can track and improve patient adherence to treatment plans and promote better engagement with their health.
- **Streamlined Operations & Efficiency:**
 - **Automated Administrative Tasks:** AI automates routine tasks like scheduling, billing, and managing electronic health records, reducing administrative workload for healthcare professionals.
 - **Optimized Resource Allocation:** Predictive helps forecast patient admissions, allowing for the efficient allocation of hospital beds, staff, and equipment.
- **Research and Population Health:**
 - **Data Analysis:** AI can vast amounts of medical data to identify trends, monitor public health, and track the spread of infectious diseases.

3.ARCHITECTURE

AI in healthcare architecture integrates data analysis to design efficient, patient hospitals, leveraging tools for layout optimization, workflow, and resource management, while also encompassing the underlying hardware/software architecture like [Armv9](#) and advanced AI models such as neuro-symbolic strategies to process vast amounts of medical data for improved patient outcomes.

Architectural Design & Planning

- **Generative Design Tools:**

AI assists architects by generating design options for healthcare facilities, optimizing space utilization, operational flow, and cost-effectiveness.

- **[BIM](#) Integration:**

AI-driven Building Information Model (BIM) enhances collaborative design processes, crucial for the complexity of healthcare facilities, by optimizing layouts and improving efficiency.

- **Data-Driven Decisions:**

Architects use AI to data and guide design decisions, such as optimizing the placement of natural light to promote patient recovery and comfort.

- **Predictive Analytics:**

AI algorithms help predict patient demand and resource needs, informing the design of more efficient and adaptive healthcare spaces.

Underlying AI & System Architecture

- **Hardware Architecture:**

Architectures like Armv9 are designed to deliver enhanced AI performance and security in healthcare, enabling faster, more accurate disease detection and robust data protection.

- **Neuro-Symbolic Models:**

These models combine symbolic and sub-symbolic approaches to process large datasets, leading to innovative and optimal architectural designs for complex hospital needs.

- **Data Integration:**

A unified digital infrastructure connecting clinics, hospitals, social care services, patients, and caregivers via sensors and ambient intelligence is a long-term goal, requiring robust AI architecture for integration.

Impact on Patient Care & Operations

- **Personalized Medicine:**

AI vast medical data (genetics, lifestyle) to help create personalized treatment programs.

- **Improved Diagnosis:**

AI algorithms medical images like X-rays and CT scans to improve diagnostic accuracy and speed up disease detection.

- **Enhanced Patient Flow:**

AI optimizes patient flow and scheduling in hospitals, leading to reduced waiting times and an improved patient experience.

4.SET OF INSTRUCTIONS

1. GRADIO Framework Knowledge: GRADIO Documentation 2. IBM Granite Models (Hugging Face): IBM Granite models 3. Python Programming Proficiency: Python Documentation 4. Version Control with Git: Git Documentation 5. Google Collab's T4 GPU Knowledge: Google collab

5.RUNNING THE APPLICATION

AI's AI for Medicine Specialisation provides practical experience in applying machine learning to concrete problems in medicine, such as predicting patient survival rates, estimating treatment plan efficacy, and diagnosing diseases from 3D MRI brain scans.

6.AUTHENTICATION

To access digital health portals, patients must use AI-driven multi-factor authentication. This method ensures a good balance between increased security, ease of use and convenience. In addition, AI-driven authentication can be used to protect patient data in hospital databases and prevent data breaches.

7.USER INTERFACE

Artificial intelligence (AI) is making big moves in every industry, including healthcare. Think of chatbots' handling triage, machine-learning (ML) algorithms' spotting early signs of disease, and easy-to-use systems that help doctors make faster data-driven decisions.

8.TESTING

Testing artificial intelligence (AI) in healthcare is a crucial, multi-faceted process that verifies AI system performance, safety, and efficacy through various methods, including clinical trials, regulatory compliance, and performance evaluations across diverse KPIs. It involves [Health Technology](#)

[Assessment \(HTA\)](#), data quality checks, [bias detection](#), and [safety validation](#) to ensure responsible integration into clinical practice, ultimately ensuring patient safety and accurate results.

Key Aspects of AI Testing in Healthcare

- **Performance and [Diagnostic Accuracy](#):**

AI algorithms must be rigorously tested to ensure high diagnostic accuracy and sensitivity, especially in medical imaging, where their performance is compared against radiologists.

- **Data Quality and Integrity:**

AI systems rely heavily on high-quality data; therefore, testing must include validation of data inputs to prevent errors and bias.

- **Bias Detection:**

AI systems are susceptible to bias present in training data, necessitating testing to identify and mitigate biases that could lead to unfair or inaccurate outcomes for certain patient groups.

- **Safety and Validation:**

Ensuring the safety of AI in healthcare involves [human-AI collaboration](#), [safety validation](#), and robust protocols to prevent harm to patients.

- **Clinical Validation:**

This involves testing AI solutions in real-world clinical settings to assess their actual impact on patient care and clinical decision-making.

- **[Regulatory Compliance](#):**

AI-powered software needs to comply with stringent regulations, such as those from the FDA, which require documented processes and adherence to standards for medical devices.

- **Ethical Frameworks:**

Robust ethical and legal frameworks guide the testing process to ensure the responsible and effective integration of AI.

Key Performance Indicators (KPIs)

- **User KPIs:** Measure user engagement and the overall user experience with the AI system.
- **Diagnostic Task KPIs:** Assess the correctness and performance of the AI in diagnostic tasks.
- **Image Quality KPIs:** Evaluate image appearance characteristics relevant for AI algorithms that process images.
- **Processing Performance KPIs:** Summarize the technical capabilities and processing speeds of the AI system.
- **Diagnostic Process KPIs:** Determine the overall clinical return on investment of the AI tool.

9.SCREENSHOT

The top screenshot shows a Google Colab notebook titled 'Untitled1.ipynb'. The code in the cell is as follows:

```
plan_output = gr.Textbox(label="Personalized Treatment Plan", lines=20)

plan_btn.click(treatment_plan, inputs=[condition_input, age_input, gender_input, history_input], outputs=plan_output)

app.launch(share=True)
```

The output shows progress bars for downloading various Hugging Face models:

- tokenizer_config.json: 8.88k/? [00:00<00:00, 250kB/s]
- vocab.json: 777k/? [00:00<00:00, 9.66MB/s]
- merges.txt: 442k/? [00:00<00:00, 8.56MB/s]
- tokenizer.json: 3.48M/? [00:00<00:00, 45.1MB/s]
- added_tokens.json: 100% [87.0/87.0 [00:00<00:00, 2.92kB/s]
- special_tokens_map.json: 100% [701/701 [00:00<00:00, 49.0kB/s]
- config.json: 100% [786/786 [00:00<00:00, 53.4kB/s]
- 'torch_dtype' is deprecated! Use 'dtype' instead!
- model.safetensors.index.json: 29.8k/? [00:00<00:00, 2.41MB/s]
- Fetching 2 files: 0% [0/2 [00:00<?, ?kB/s]
- model-00001-of-00002.safetensors: 19% [960M/5.00G [00:28<02:38, 25.5MB/s]
- model-00002-of-00002.safetensors: 100% [67.1M/67.1M [00:28<00:00, 2.05MB/s]

The bottom screenshot shows the same notebook titled 'Untitled2.ipynb'. The code in the cell is as follows:

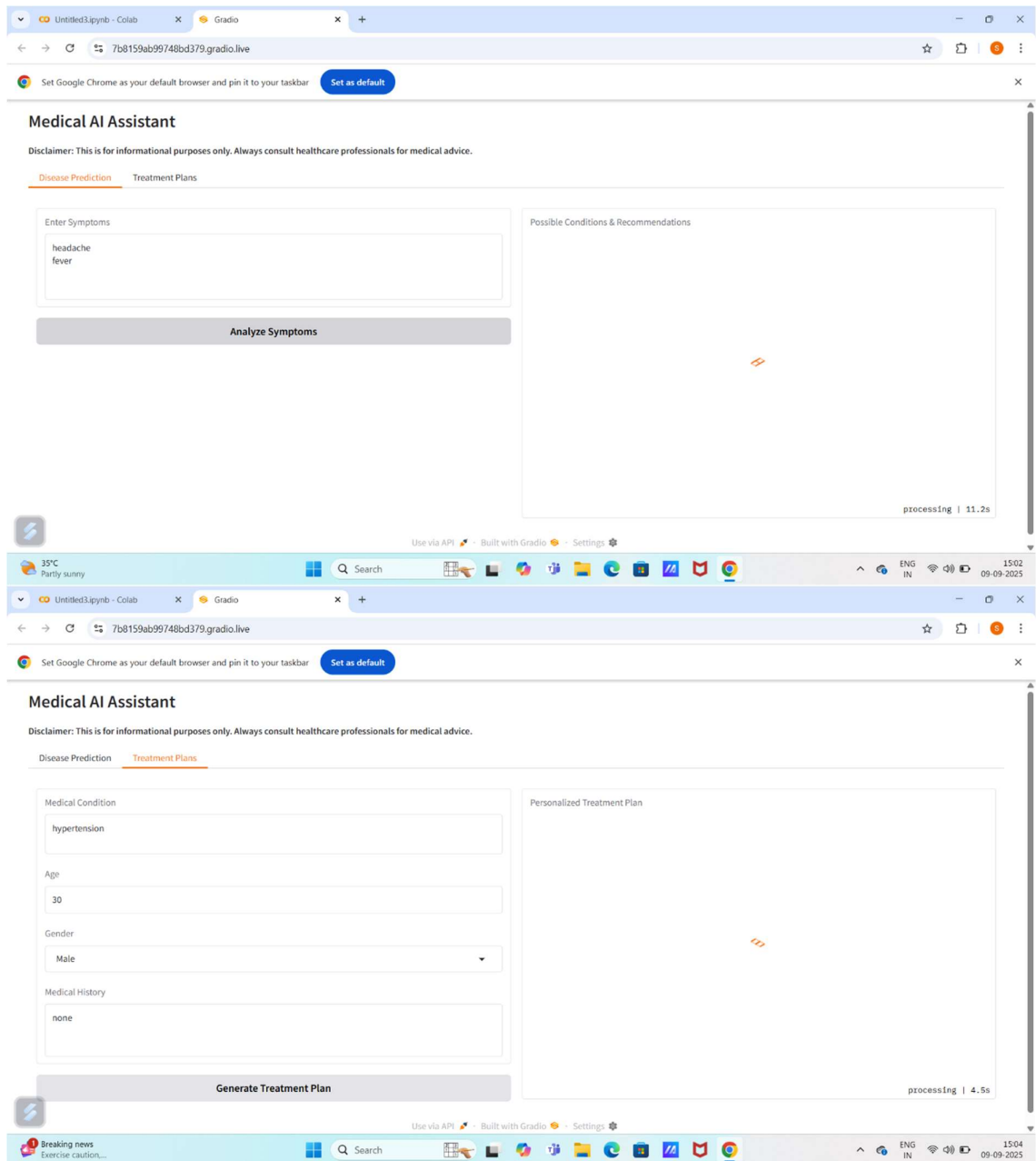
```
model-00002-of-00002.safetensors: 100% [67.1M/67.1M [00:01<00:00, 37.7MB/s]
```

The output shows progress bars for downloading various Hugging Face models:

- Loading checkpoint shards: 100% [2/2 [00:19<00:00, 8.09s/it]
- generation_config.json: 100% [137/137 [00:00<00:00, 9.47kB/s]

The notebook also displays a Gradio app interface with the following components:

- Enter Symptoms**: A text input field with the placeholder text "e.g., fever, headache, cough, fatigue...".
- Analyze Symptoms**: A button to trigger the analysis.
- Possible Conditions & Recommendations**: A list of conditions and recommendations, including:
 - 1. "Viral Infections (e.g., Flu, Common Cold):"
 - "Causes:" Influenza virus, rhinovirus, coronaviruses (including SARS-CoV-2 and MERS-CoV)
 - "Symptoms:" Fever, headache, cough, fatigue, body aches
 - "Treatment:"
 - Supportive care (hydration, rest)
 - Over-the-counter medications for fever and pain relief (acetaminophen or ibuprofen)
 - Antiviral medications may be considered for severe cases or in the early stages, especially with influenza (e.g., oseltamivir, zanamivir, or peramivir)
 - 2. "Bacterial Infections (e.g., Meningitis, Encephalitis):"
 - "Causes:" Various bacteria, such as Streptococcus pneumoniae, Neisseria meningitidis, or Haemophilus influenzae type b
 - "Symptoms:" Fever, severe headache, neck stiffness, confusion, nausea, vomiting
 - "Treatment:"
 - Empiric antibiotic therapy (broad-spectrum antibiotics before culture results)
 - Consultation with an infectious disease specialist or neurologist



10.FUTURE ENHANCEMENT

11.CODING

```
File Edit Selection View Go Run Terminal Help Search
Restricted Mode is intended for safe code browsing. Trust this window to enable all features. Manage Learn More

healthai.py X
C:\Users\Gayathri\Desktop\Health AI\CODING> healthai.py
1 import gradio as gr
2 import torch
3 from transformers import AutoTokenizer, AutoModelForCausalLM
4
5 # Load model and tokenizer
6 model_name = "ibm-granite/granite-3.2-2b-instruct"
7 tokenizer = AutoTokenizer.from_pretrained(model_name)
8 model = AutoModelForCausalLM.from_pretrained(
9     model_name,
10     torch_dtype=torch.float16 if torch.cuda.is_available() else torch.float32,
11     device_map="auto" if torch.cuda.is_available() else None
12 )
13
14 if tokenizer.pad_token is None:
15     tokenizer.pad_token = tokenizer.eos_token
16
17 def generate_response(prompt, max_length=1024):
18     inputs = tokenizer(prompt, return_tensors="pt", truncation=True, max_length=512)
19
20     if torch.cuda.is_available():
21         inputs = {k: v.to(model.device) for k, v in inputs.items()}
22
23     with torch.no_grad():
24         outputs = model.generate(
25             **inputs,
26             max_length=max_length,
27             temperature=0.7,
28             do_sample=True,
29             pad_token_id=tokenizer.eos_token_id
30         )
31
32     response = tokenizer.decode(outputs[0], skip_special_tokens=True)
33     response = response.replace(prompt, "").strip()
34     return response
35
36 def disease_prediction(symptoms):
37     prompt = f"Based on the following symptoms, provide possible medical conditions and general medication suggestions. Always emphasize the importance of consulting a doctor."
38     return generate_response(prompt, max_length=1200)
39
40 def treatment_plan(condition, age, gender, medical_history):
41     prompt = f"Generate personalized treatment suggestions for the following patient information. Include home remedies and general medication guidelines.\n\nMedical Condition: {condition}, Age: {age}, Gender: {gender}, Medical History: {medical_history}"
42     return generate_response(prompt, max_length=1200)
43
44 # Create Gradio interface
45 with gr.Blocks() as app:
46     gr.Markdown("# Medical AI Assistant")
47     gr.Markdown("**Disclaimer: This is for informational purposes only. Always consult healthcare professionals for medical advice.**")
48
49     with gr.Tabs():
50         with gr.TabItem("Disease Prediction"):
51             with gr.Row():
52                 with gr.Column():
53                     symptoms_input = gr.Textbox(
54                         label="Enter Symptoms",
55                         placeholder="e.g., fever, headache, cough, fatigue...",
56                         lines=4
57                     )
58                     predict_btn = gr.Button("Analyze Symptoms")
59
60                 with gr.Column():
61                     prediction_output = gr.Textbox(label="Possible Conditions & Recommendations", lines=20)
62
63             predict_btn.click(disease_prediction, inputs=symptoms_input, outputs=prediction_output)
64
65         with gr.TabItem("Treatment Plans"):
66             with gr.Row():
67                 with gr.Column():
68                     condition_input = gr.Textbox(
69                         label="Medical Condition",
70                         placeholder="e.g., Diabetes, Hypertension"
71                     )
72                     age_input = gr.Textbox(
73                         label="Age",
74                         placeholder="e.g., 35"
75                     )
76                     gender_input = gr.Textbox(
77                         label="Gender",
78                         placeholder="e.g., Male, Female"
79                     )
80                     medical_history_input = gr.Textbox(
81                         label="Medical History",
82                         placeholder="e.g., Allergies, Chronic Conditions"
83                     )
84
85                     treatment_btn = gr.Button("Generate Treatment Plan")
86
87             treatment_btn.click(
88                 treatment_plan,
89                 inputs=[condition_input, age_input, gender_input, medical_history_input],
90                 outputs=treatment_output
91             )
92
93     treatment_output = gr.Textbox(label="Personalized Treatment Suggestions", lines=10)
94
95 app.launch()
```

```
File Edit Selection View Go Run Terminal Help Search
Restricted Mode is intended for safe code browsing. Trust this window to enable all features. Manage Learn More

healthai.py x
C:\Users\Gayathri\Desktop>Health AI>CODING>healthai.py
45 with gr.Blocks() as app:
46     with gr.Tab():
47         with gr.Tabitem("Disease Prediction"):
63             predict_btn.click(disease_prediction, inputs=symptoms_input, outputs=prediction_output)
64
65         with gr.Tabitem("Treatment Plans"):
66             with gr.Row():
67                 with gr.Column():
68                     condition_input = gr.Textbox(
69                         label="Medical Condition",
70                         placeholder="e.g., diabetes, hypertension, migraine...",
71                         lines=2
72                     )
73                     age_input = gr.Number(label="Age", value=30)
74                     gender_input = gr.Dropdown(
75                         choices=["Male", "Female", "Other"],
76                         label="Gender",
77                         value="Male"
78                     )
79                     history_input = gr.Textbox(
80                         label="Medical History",
81                         placeholder="Previous conditions, allergies, medications or None",
82                         lines=3
83                     )
84                     plan_btn = gr.Button("Generate Treatment Plan")
85
86                 with gr.Column():
87                     plan_output = gr.Textbox(label="Personalized Treatment Plan", lines=20)
88
89             plan_btn.click(treatment_plan, inputs=[condition_input, age_input, gender_input, history_input], outputs=plan_output)
90
91 app.launch(share=True)
92
```

32°C Mostly cloudy 18:52 09-09-2025