



PulseMind
AI-driven Behavioral assessment
and Intervention for ADHD

Presented By:
24-25J-261

PulseMind: AI-Driven Behavioural Assessment and Intervention for ADHD



SUPERVISOR DETAILS



Supervisor
Dr. Sanvitha
Kasthuriarachchi



Co-Supervisor
Ms. Mihiri
Samaraweera

TEAM MEMBERS



Dharmasena U.D.S.V.
IT21288326



Manamperi R.S.
IT21290992



Dilshani H.T.D.P.
IT21379574



Halliyadda H.U.M.S.
IT21380532



INTRODUCTION

ADHD (Attention-Deficit / Hyperactivity Disorder) is a neuro developmental disorder.[1]

Characterized by symptoms of
Inattention

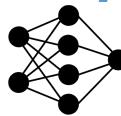
Hyperactivity
Impulsivity



ADHD can affect children's academic performance, social skills, and emotional well-being. It may lead to long-term challenges in adulthood, such as low self-esteem and career difficulties[2]

OBJECTIVES

ADHD diagnosis and provides personalized intervention strategies for managing ADHD symptoms in children



Assessing ADHD Symptoms



Enhance focus and organizational skills



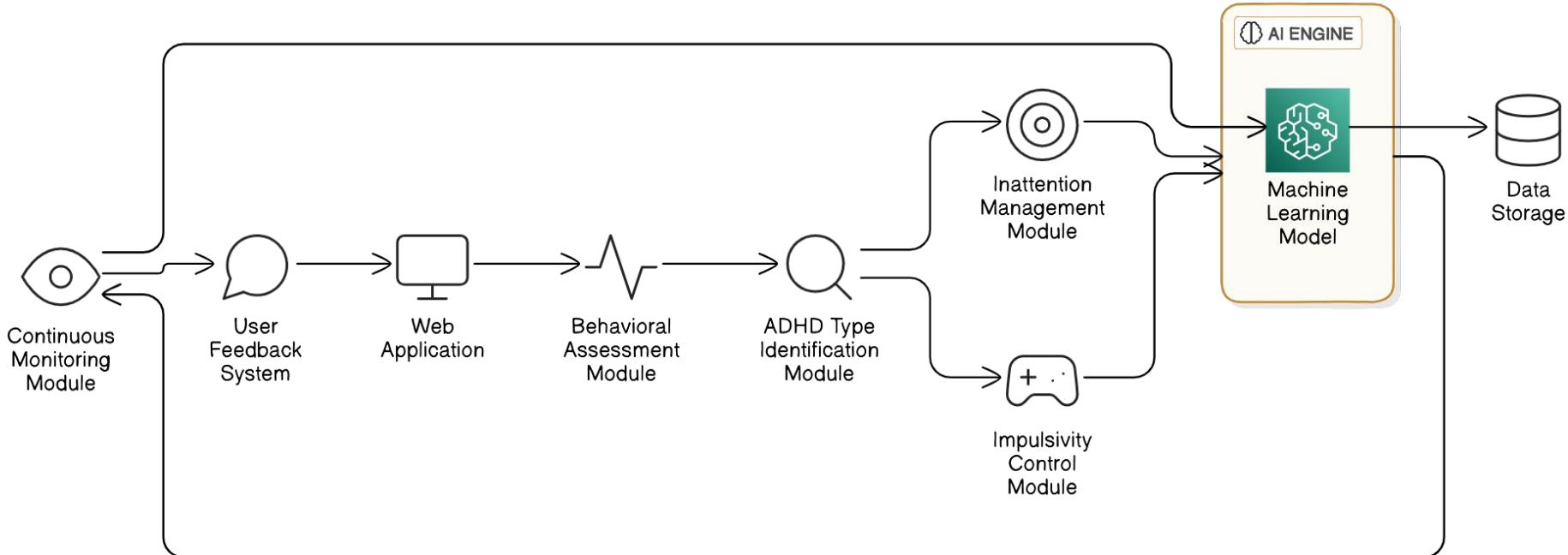
Enhancing Impulse Control



Create Adaptive Learning with Predictions

OVERALL SYSTEM DIAGRAM

ADHD Management Web Application Architecture



COMMERCIALIZATION PLAN

Basic Plan

Free

- ✓ ADHD symptom assessment through interactive questionnaires
- ✓ Basic inattention and hyperactivity tracking
- ✓ Limited progress reports available within the app

Suitable for individual users (parents or teachers) who need a basic assessment tool

Premium Plan

\$10 /month

- ✓ Full ADHD assessment with symptom identification and subtype classification
- ✓ AI-driven personalized interventions and cognitive training exercises
- ✓ Advanced progress tracking with weekly and monthly reports
- ✓ 24/7 customer support

Ideal for families and schools seeking a comprehensive ADHD support tool

Group Plan

\$150 /month

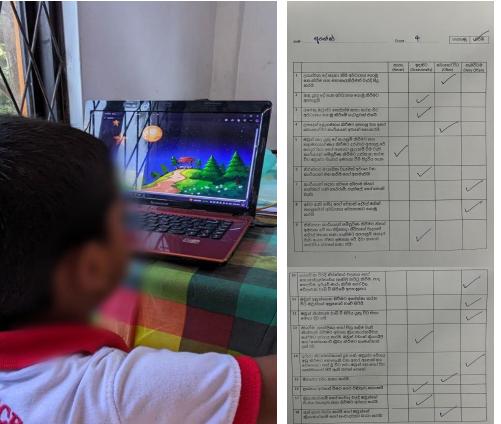
- ✓ All features from the Premium Plan
- ✓ Multi-user access for up to 25 students
- ✓ Ability to manage and monitor multiple classrooms or groups within a single dashboard

Best suited for schools, clinics needing to support multiple children

RISK MITIGATION

Risk	Severity	Mitigation Measure
Privacy and ethical concerns when collecting and storing sensitive data about children	High	Completed the ethical clearance form, obtained permission from the education division and ensured anonymization of data to protect privacy
Some parents hesitated to give consent for their children to participate	Medium	Implement robust data anonymization and encryption protocols
Target users (parents, teachers, children) find the interface or activities difficult to use or unengaging	Medium	Conduct user testing with parents, teachers, and children during development
The system does not adequately address the cultural or linguistic context of Sri Lankan users.	High	Involve local psychologists, educators, and linguists in system design and content development.

DATA COLLECTION



REFERENCES

- [1] J. J. S. Kooij, "ADHD: a Neurodevelopmental Disorder," *European Psychiatry*, vol. 30, Suppl. 1, p. 45, 2015, doi: 10.1016/S0924-9338(15)30036-5.
- [2] T. E. Wilens and T. J. Spencer, "Understanding attention-deficit/hyperactivity disorder from childhood to adulthood," *Postgrad. Med.*, vol. 122, no. 5, pp. 97-109, Sep. 2010, doi: 10.3810/pgm.2010.09.2206.

AI-Driven Gamified ADHD Symptom Assessment for Children Aligned with DSM-5



IT21288326 | Dharmasena U.D.S.V.
Specialization: Information technology

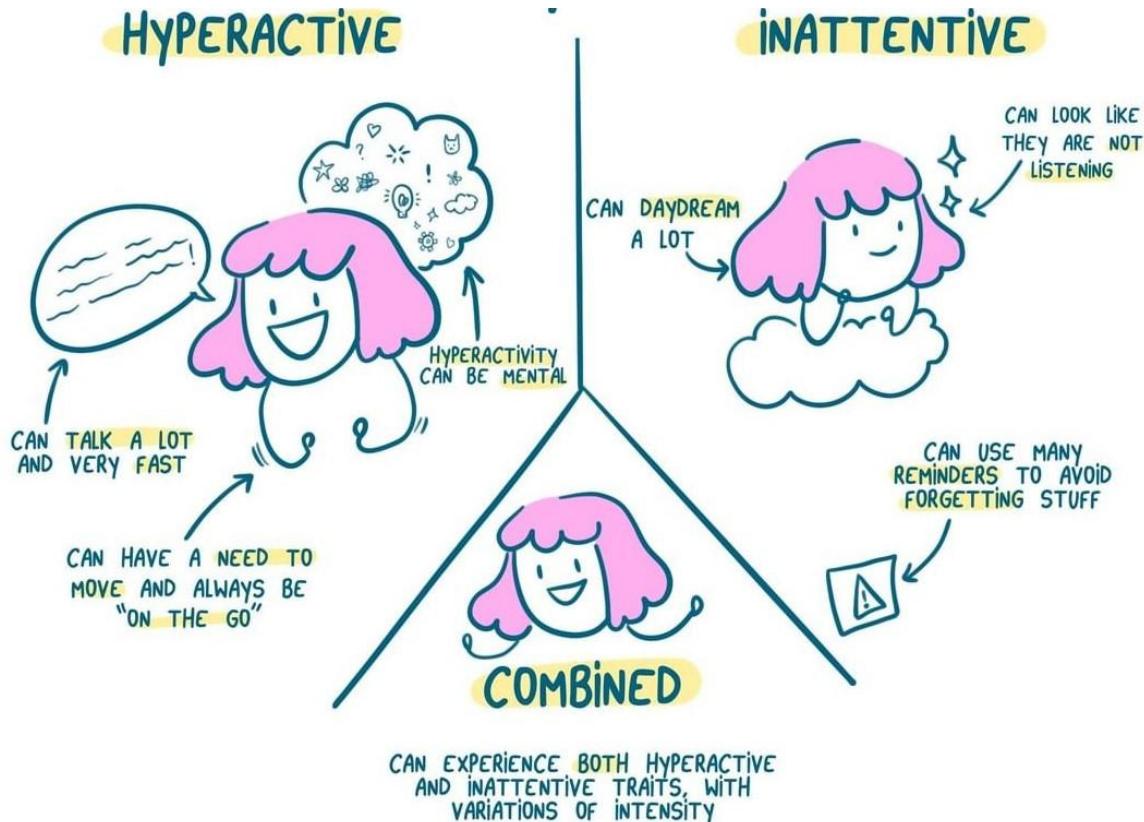
INTRODUCTION

Early diagnosis helps children with ADHD learn coping skills and get the right support, which can improve their success their overall quality of life.

ADHD subtypes

- Predominantly inattentive ADHD
- Hyperactive-impulsive ADHD
- Combined ADHD





Tools for assessing ADHD symptoms.

- Vanderbilt ADHD Diagnostic Parent Rating Scale
- Conners Rating Scale
- ADHD Self-Report Scale(for Adults)

RESEARCH GAP

Component	[3]	[4]	[5]	[6]	Proposed System
Use of digital tools for ADHD assessment	✓	✓	✗	✓	✓
Automated model training	✗	✗	✗	✗	✓
Tailored interventions for ADHD symptoms	✓	✗	✓	✓	✓
Accessibility and affordability of interventions	✗	✗	✓	✗	✓
Symptom assessment based on DSM-5	✓	✓	✓	✓	✓

[3] Jan. 2018, "Evaluating Digital ADHD Assessment Tools for Children."

[4] Apr. 2019, "Understanding Cultural Factors in ADHD Diagnosis: A Global Perspective"

[5] Jul. 2020, "Evaluating Accessibility of ADHD Intervention Programs for Diverse Populations"

[6] May. 2021, "Interactive Applications for DSM-5 Based ADHD Symptom Assessment"

RESEARCH PROBLEM

How can we Assess ADHD
Symptoms in Children age 5-10,
through an AI-Driven Gamified
Focus & Impulse Control
Module Aligned with DSM-5
Criteria ?



OBJECTIVES

Assess ADHD
Symptoms in Children
Through an AI-Driven
Gamified Focus &
Impulse Control
Module Aligned with
DSM-5 Criteria



Assess cognitive functions through
interactive activities.



Collect DSM-5 symptoms based on
gamified behavioral performance.



Integrate behavioral and questionnaire
data to classify ADHD symptoms.

PP1 to PP2 PROGRESS

Re-trained and dumped (save) a machine learning model to detect ADHD sub type

Enabled a function to process the real-time data and update the model correspondingly

Labeled the assessment criteria results to the corresponding sub type

Model prediction is given by analysing the sub type label that is taken from the assessment criteria & the results of the game (reaction time, score, no of missing attempts..etc.)

Created the dashboard to check the model health

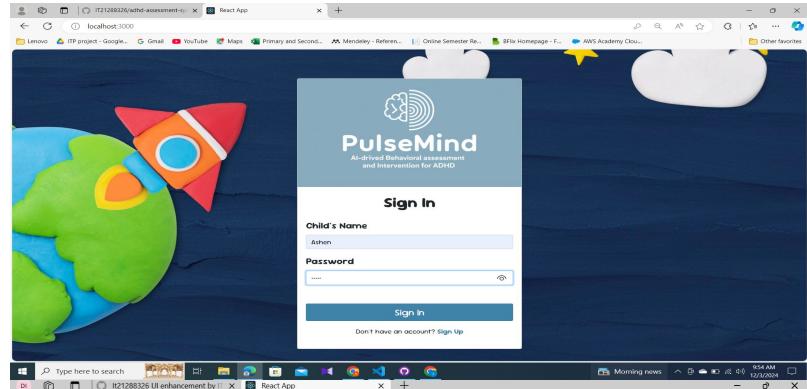
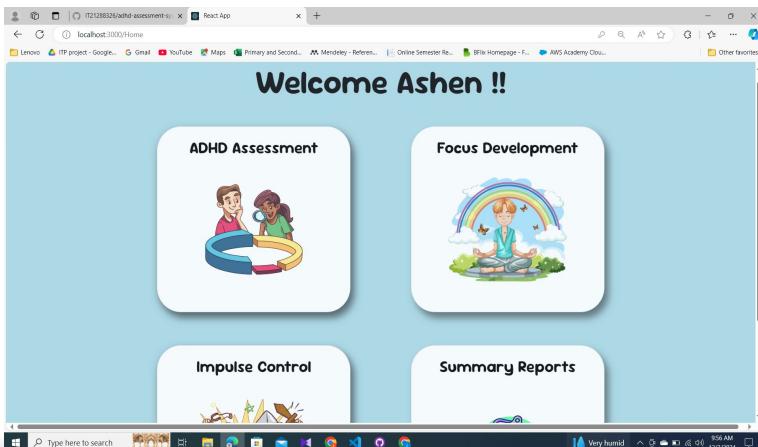
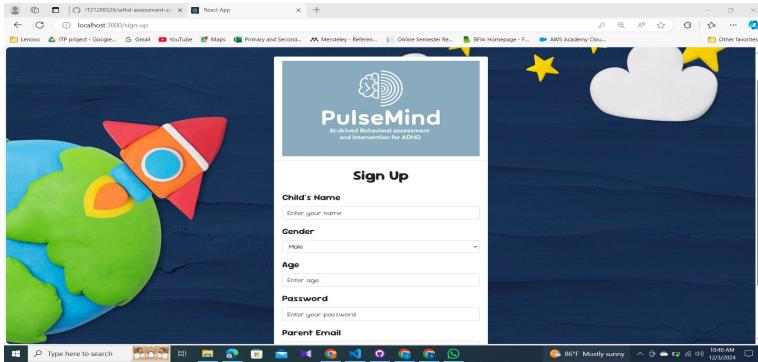
Created dashboards to show game result, child progress, insights and model prediction

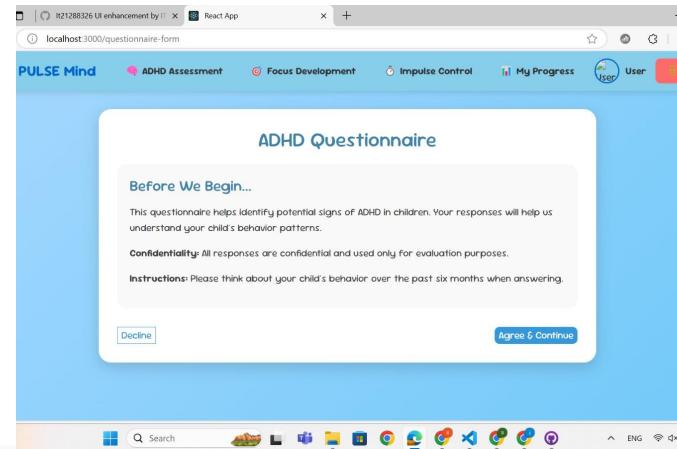
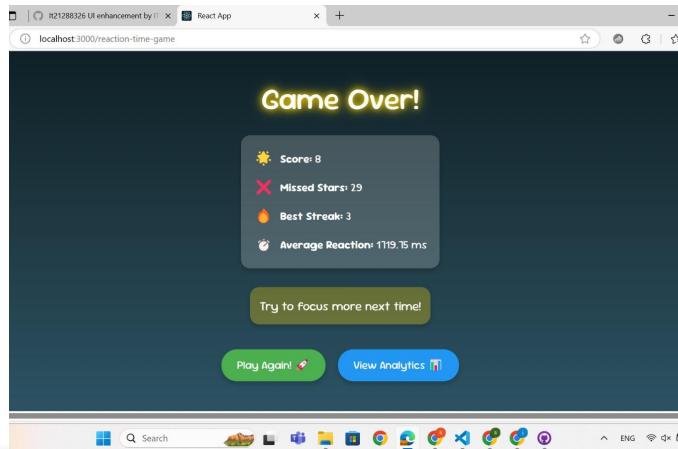
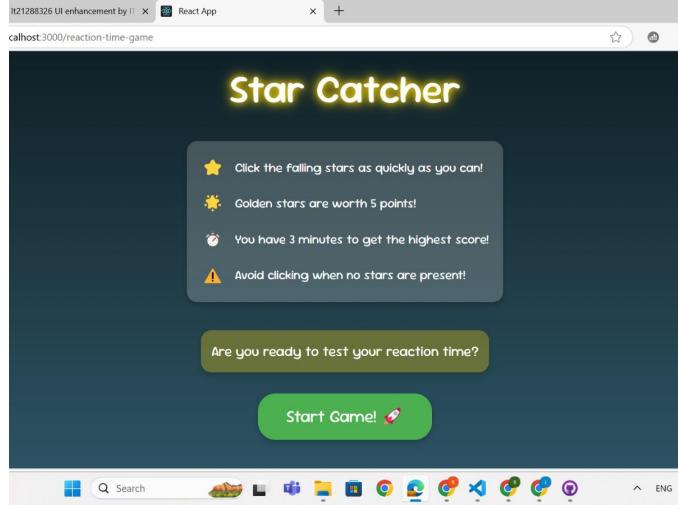
Create quiz results and generate report based on venderbilit ADHD assesment critiria

Create interfaces and save data in the database



PROOF OF COMPLETION





Question 18 of 55

Interrupts or bothers others when they are talking or playing games.

Never Occasionally
Often Very Often

Previous Next

You're doing great! Keep going.

Scale	Score	Clinical Cutoff	Percentile
Inattentive Subtype	8	Below	30
Hyperactive/Impulsive Subtype	10	Below	31
Combined Subtype	18	Below	33
Oppositional Defiant Disorder	3	Below	13
Conduct Disorder	23	Above	55
Anxiety/Depression	11	Above	52

Percentile Comparison

Category	Percentile
Inattentive	30%
Hyperactive	31%
Combined	33%

Note: These potential comorbidities are screening indicators only and require further comprehensive evaluation for diagnosis.

Recommendations

- Immediate evaluation for Conduct Disorder by a mental health professional
- Evaluation for anxiety and/or depression

Important Notice

This report is generated based on the Vanderbilt ADHD Diagnostic Parent Rating Scale (VADPRS). Results should be interpreted by qualified healthcare professionals as part of a comprehensive evaluation. This screening tool alone is not sufficient for diagnosis.

[Return to Dashboard](#) [Print Report](#)

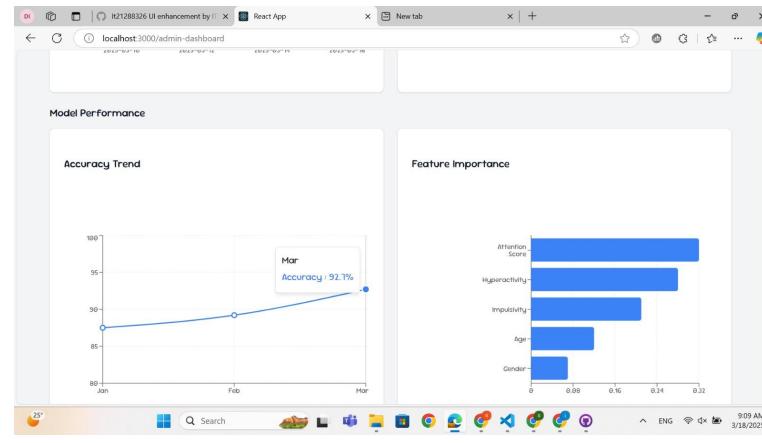
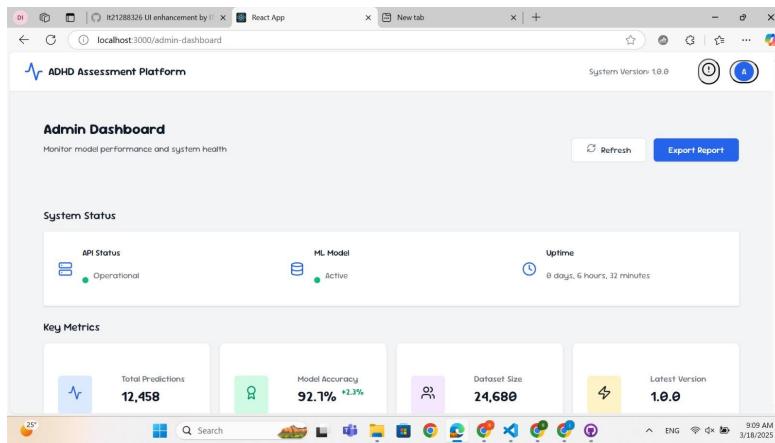
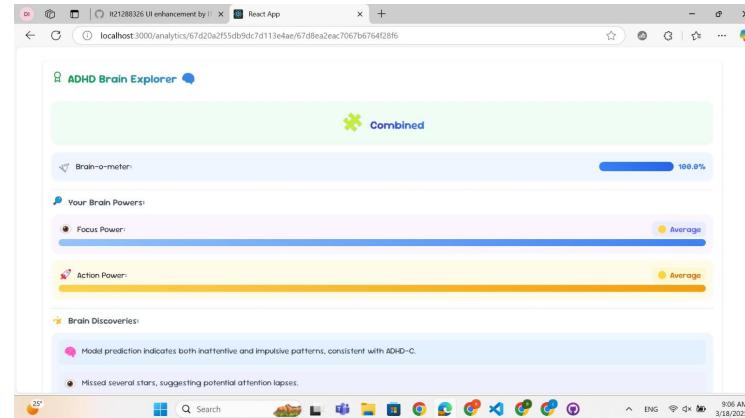
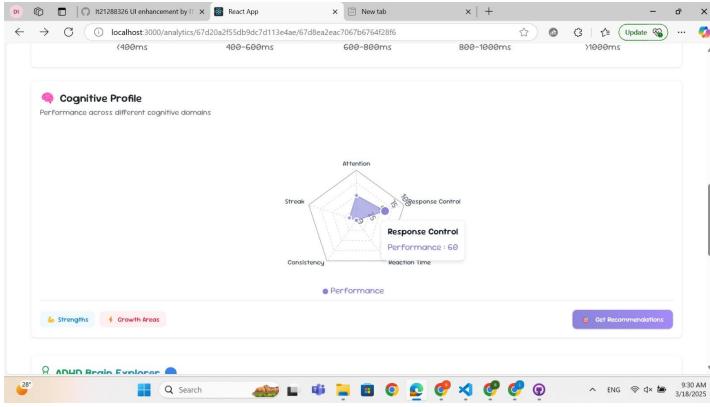
Game Performance Analytics

Detailed analysis of your cognitive performance patterns

Attention Score
59.0 Average ~

Impulsivity Score
49.0 Average ~

Combined Score
49.0 Average ~



8097 Party sunny 12/24/2024



```
frontend > ac > component > R21288326 > QuestionnaireForm > [M] QuestionnaireForm
1
2
3 import React, { useState, useEffect } from "react";
4 import axios from "axios";
5 import { useLocation } from "react-router-dom";
6 import { useStateContext } from "../../context";
7 import { QuestionnaireForm } from "./QuestionnaireForm";
8 const QuestionnaireForm = () => {
9   const location = useLocation();
10  const [questions, setQuestions] = useState([]); // Store questions fetched from the backend
11  const [responses, setResponses] = useState([]); // Store user responses
12  const [childId, setChildId] = useState(location.state?.childId || null); // Get childId from location state
13  const [showDisclaimer, setShowDisclaimer] = useState(true); // Set initial state for disclaimer
14  const [showQuestionnaire, setShowQuestionnaire] = useState(false); // Set initial state for questionnaire
15  const navigate = useNavigate(); // Initialize navigate function
16
17  const handleBackline = () => {
18    navigate(-1); // Navigate to the previous page
19  };
20
21 // Fetch questions from the backend
22 useEffect(() => {
23   const fetchQuestions = async () => {
24     try {
25       const response = await axios.get(`http://localhost:8880/api/questionnaire/questions`, {
26         withCredentials: true, // Ensure requests include credentials
27       });
28       setQuestions(response.data);
29     } catch (error) {
30       console.error(error);
31     }
32   };
33   fetchQuestions();
34 });
35
36 return (
37   <div>
38     <h1>Reaction Time Game</h1>
39     <h2>Questionnaire Form</h2>
40     <div>
41       <button onClick={handleBackline}>Back</button>
42       <button onClick={() => setShowQuestionnaire(!showQuestionnaire)}>Next</button>
43     </div>
44     <div>
45       {showDisclaimer ? <DisclaimerForm /> : <QuestionnaireForm />}
46     </div>
47   </div>
48 );
49
50 </QuestionnaireForm>
```

The screenshot shows the MongoDB Cloud interface. On the left, the sidebar has sections for Project 0, Database (selected), and Services. The main area shows the 'questionnaireresponses' collection with a storage size of 34KB and logical data size of 2.54KB. A search bar at the top right contains the URL 'cloud.mongodb.com/v2/6739f5d555d5e1f113acd42/metrics/replicaSet/6739f8e084a6f4e3489ed6/explore/test/questionnaireresponses/_find'. The interface includes tabs for Find, Indexes, Schema Anti-Patterns, Aggregation, and Search Indexes. A button for 'INSERT DOCUMENT' is visible. Below these are 'Generate queries from natural language in Compose!' and a query builder with a 'Type a query: { Field: 'value' }' input field. The bottom section displays the 'QUERY RESULTS: 1-11 OF 11' with one document listed:

```
_id: ObjectId("6749f3c15247b03417408084")
__v: 0
responses: [
  {
    __v: 0,
    responses: [
      {
        __v: 0,
        combinedScore: 7
      }
    ],
    combinedScore: 14
  }
]
subtype: "Combined"
createdAt: ISODate("2024-11-29T05:11:09.780+00:00")
--V
```

4:45 PM 89° Partly sunny 12/4/2024

The screenshot shows the MongoDB Cloud interface. On the left sidebar, under the 'Project 0' section, 'DATABASE' is selected, revealing the 'test' database which contains the 'gametrics' collection. The main panel displays the 'test' database's metrics: STORAGE SIZE: 62KB, LOGICAL DATA SIZE: 272KB, TOTAL DOCUMENTS: 82, and INDEXES TOTAL SIZE: 36KB. Below this, there are tabs for 'Find', 'Indexes', 'Schema Anti-Patterns', 'Aggregation', and 'Search Indexes'. A search bar at the top says 'Search Namespaces'. A query builder is shown with the following JSON:

```
{ "_id": { $objectID: "4f74d31232bb92f46b2c856e" }, "children": [ "test", "gametrics" ], "questionnaireResponses": [ { "reactionTime": { "reactionTime": 1000, "reactionTimeError": 100, "reactionTimeMin": 700, "reactionTimeMax": 1300, "reactionTimeUnit": "ms" }, "reactionTimes": [ 22 ] }, { "averageReactionTime": 4587, "count": 22, "minReactionTime": 1000, "maxReactionTime": 1300, "percentileClicks": 10, "percentileStars": 15, "score": 4.5, "standardDeviation": 1000, "standardError": 100, "standardScore": 4.5, "standardUpperBound": 5.5, "standardLowerBound": 3.5, "totalClicks": 22, "totalScore": 99.5 } ] }
```

The results pane shows '1-20 of many results'.

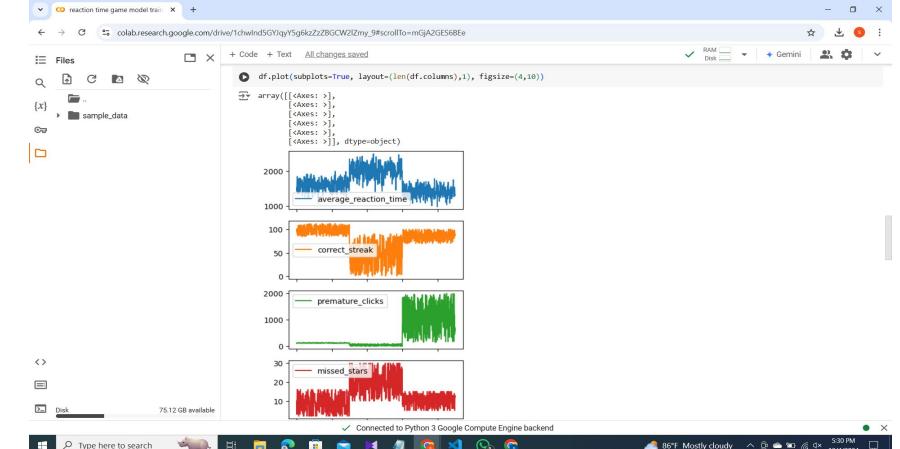
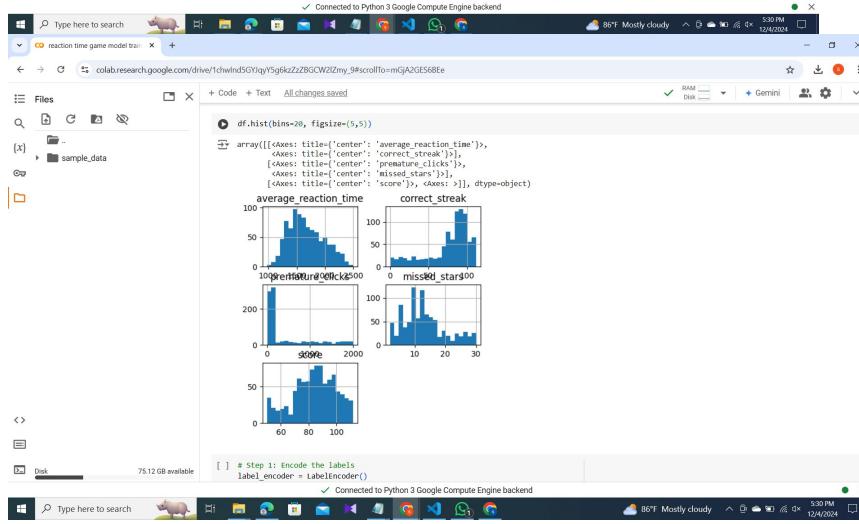
reaction time game model train

```
[ ] # Import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.svm import SVC
from sklearn.metrics import accuracy_score, classification_report
import joblib

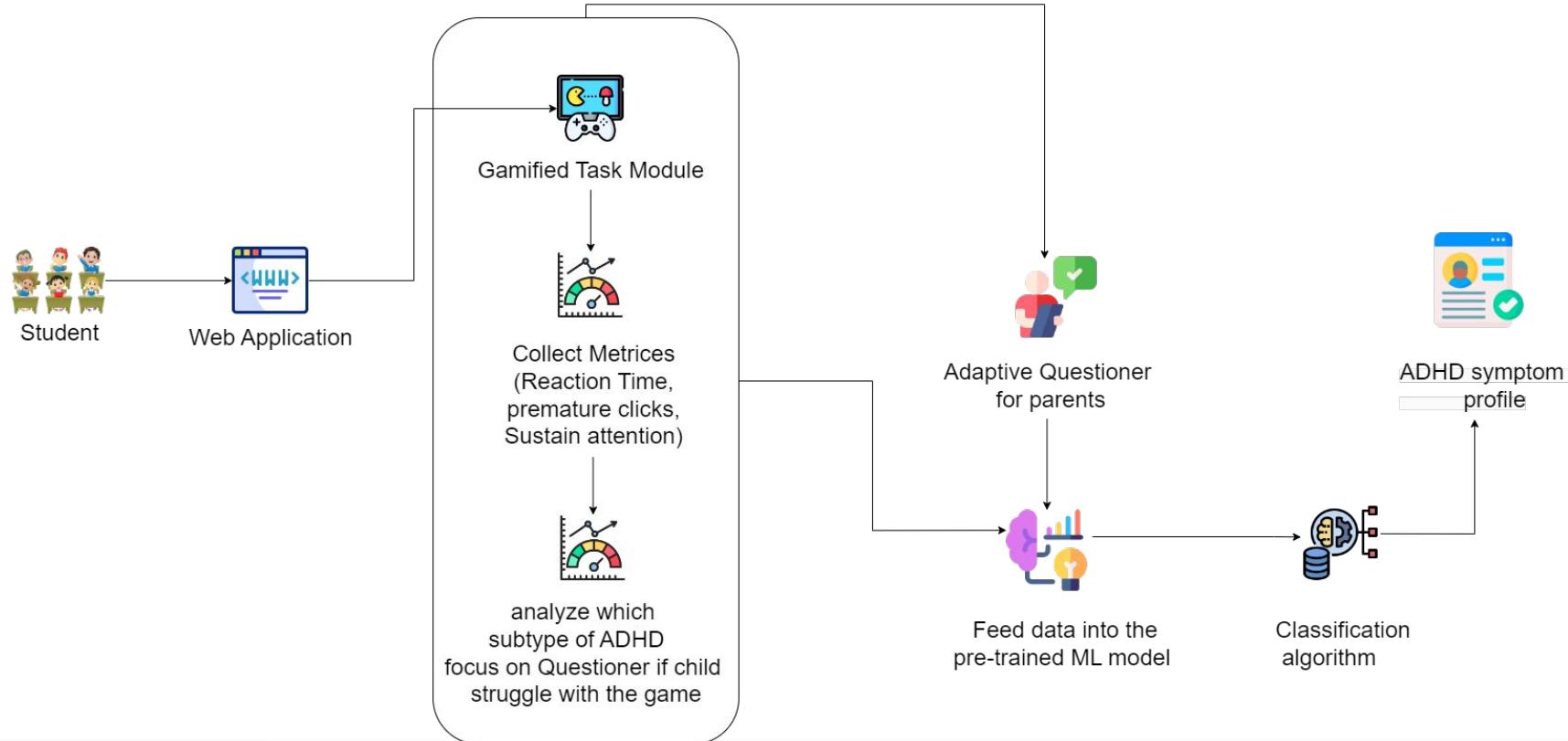
[ ] # Load the dataset
data_path = "/content/adhd_dataset.csv"
df = pd.read_csv(data_path)

[ ] df.head(10)
```

	average_reaction_time	correct_streak	premature_clicks	missed_stars	score	label
0	1528.573381	95	113	10	91	Normal
1	1502.739612	100	116	3	99	Normal
2	1417.351373	101	112	3	103	Normal
3	1794.103257	93	129	12	104	Normal
4	1846.537748	89	126	16	97	Normal
5	1715.479048	97	123	3	92	Normal
6	1546.972819	97	137	5	87	Normal
7	1655.932410	66	140	16	87	Normal
8	1616.580324	94	121	8	93	Normal
9	1350.794768	107	114	15	90	Normal



SYSTEM DIAGRAM



TECHNOLOGIES

Python



TensorFlow



mongoDB



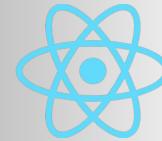
JWT



Phaser



React

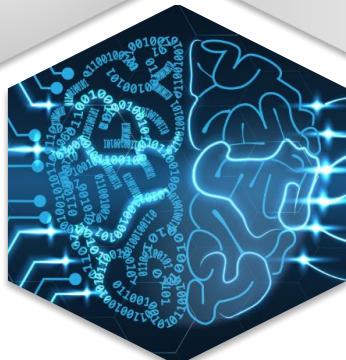


Pandas



KEY PILLARS

UI/UX



Machine learning



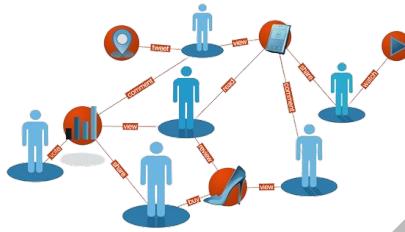
Data Collection and Management



Game Development

COMPONENT SPECIFIC REQUIREMENTS

Functional Requirement



Data Collection and Processing



Gamified Behavioral Task

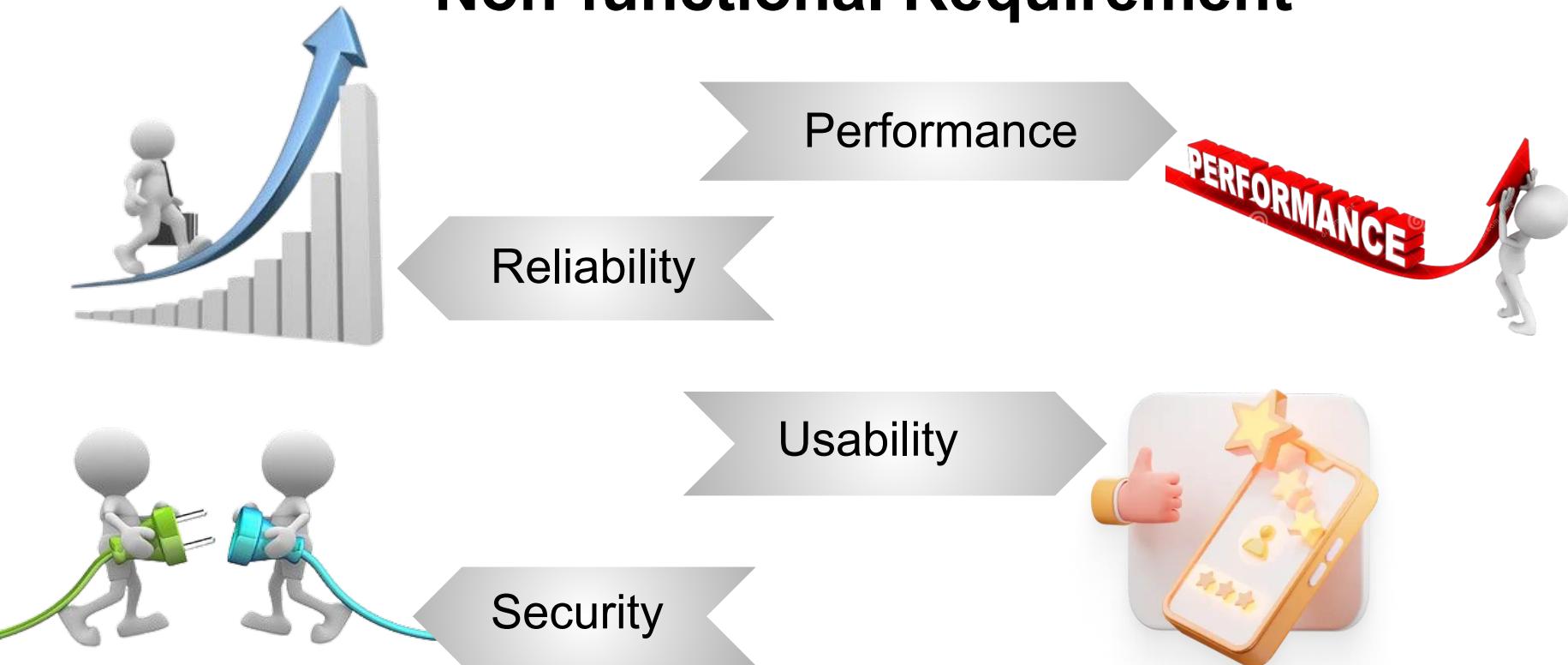


Adaptive Questionnaire

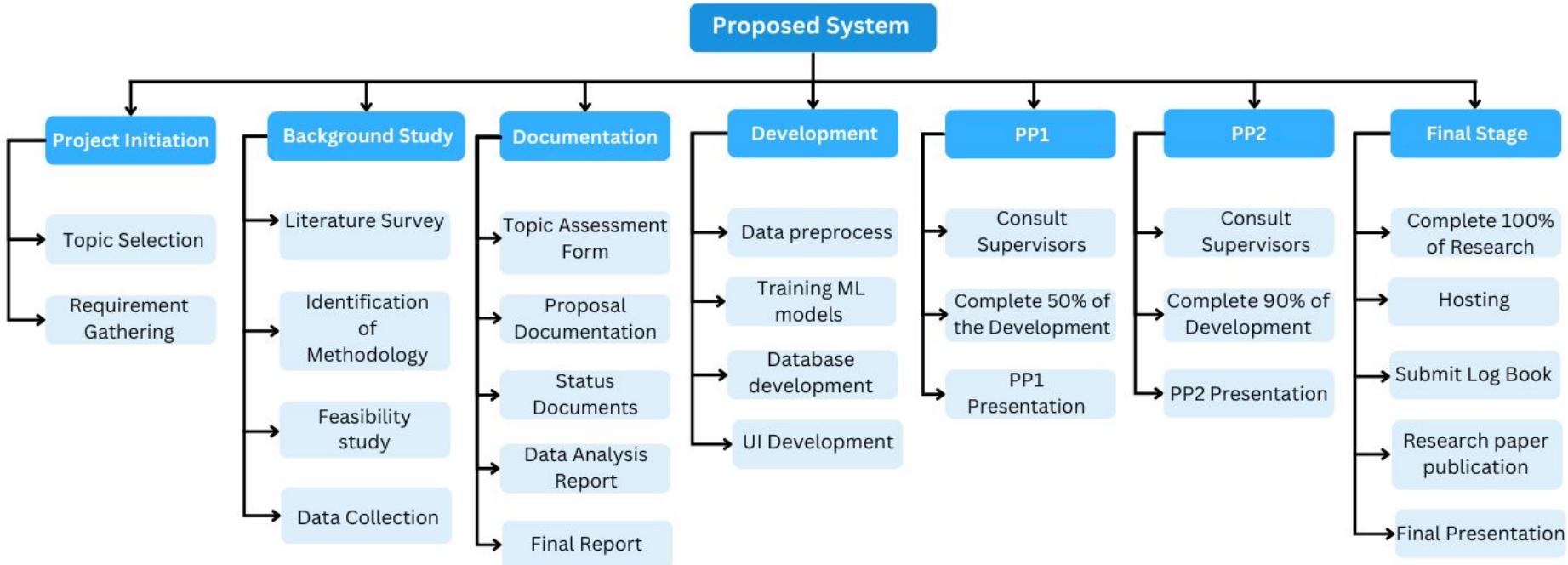
User Interface

COMPONENT SPECIFIC REQUIREMENTS

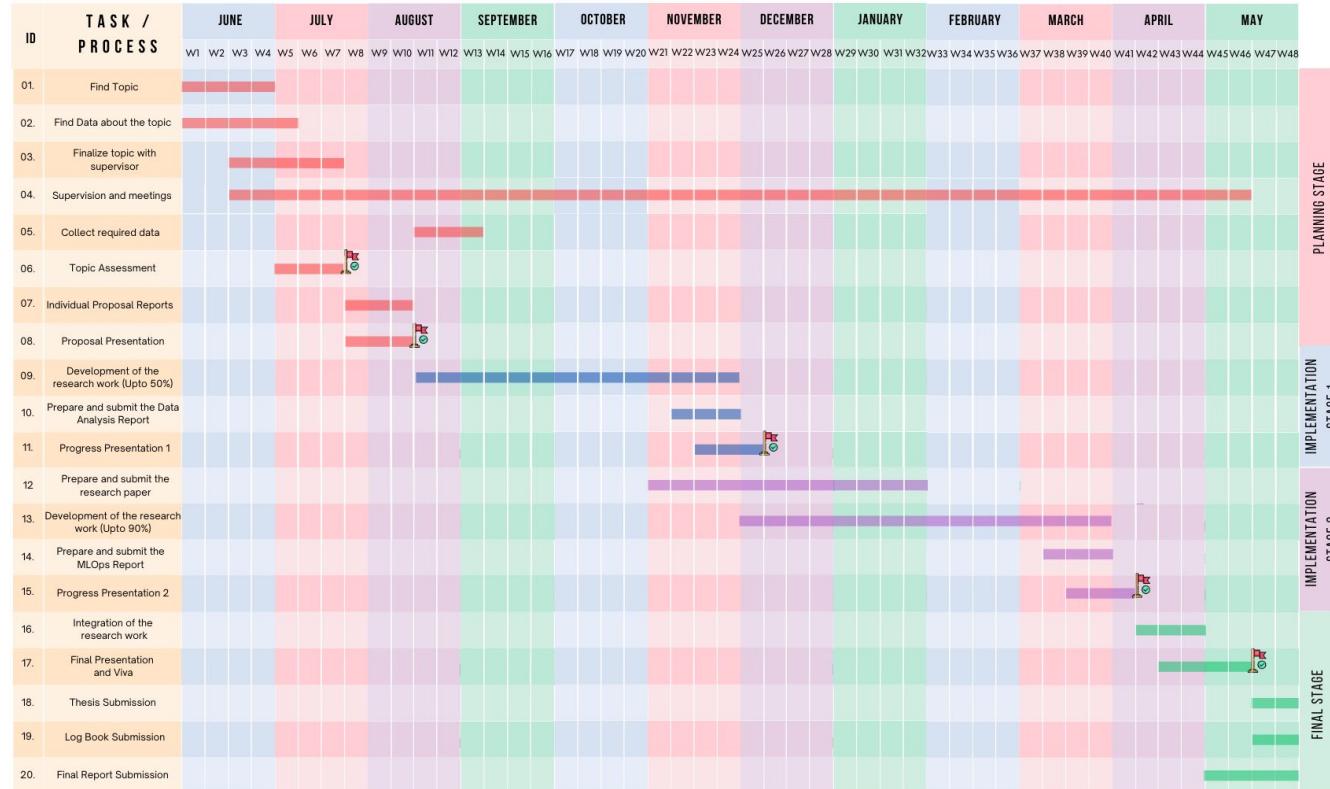
Non-functional Requirement



WORK BREAKDOWN CHART



GANTT CHART



REFERENCES

- [3]A. Smith, B. Johnson, and C. Lee, "Evaluating Digital ADHD Assessment Tools for Children: A Review," *IEEE Access*, vol. 6, pp. 12345-12353, Jan. 2018. doi: 10.1109/ACCESS.2018.2812345.
- [4]K. Chen and M. Patel, "Understanding Cultural Factors in ADHD Diagnosis: A Global Perspective," *IEEE Transactions on Psychology*, vol. 12, no. 2, pp. 345-356, Apr. 2019. doi: 10.1109/TP.2019.1234567.
- [5]L. Kumar and R. Thompson, "Evaluating Accessibility of ADHD Intervention Programs for Diverse Populations," *IEEE Journal of Public Health*, vol. 17, no. 4, pp. 567-579, Jul. 2020. doi: 10.1109/JPH.2020.2345678.
- [6]M. Rogers, S. Chang, and D. Patel, "Interactive Applications for DSM-5 Based ADHD Symptom Assessment," *IEEE Software*, vol. 22, no. 3, pp. 78-85, May 2021. doi: 10.1109/IS.2021.1234567.

Creates specific tools and activities to help children with predominantly inattentive ADHD improve focus and stay organized.



IT21290992 | Manamperi R.S.
Specialization: Information technology

INTRODUCTION

Inattention is a core characteristic of ADHD that manifests as difficulty sustaining focus, organizing tasks, and following through on instructions.

Individuals may struggle with maintaining attention in both academic and social settings.



How to Identify ?

- Difficulty paying attention to details, leading to careless mistakes
- Frequent forgetfulness in daily activities
- Challenges in organizing tasks and activities
- Avoidance of tasks requiring sustained mental effort
- Easily distracted by extraneous stimuli



Solution

- Games to Help Improve Focus
- Emotion-Based Activity Suggestion
- Personalized Game Flow
- Task Organization Tools
- Progress Tracking and Rewards
- Adaptive Learning Environment

Helping Children
Grow Step by
Step Towards
Overcoming Their
Challenges

RESEARCH GAP

Component	[7]	[8]	[9]	[10]	Proposed System
Gamified Learning Approaches	✓	✗	✗	✗	✓
Emotional and Behavioral Analysis Tools	✗	✗	✗	✗	✓
Adaptive Task Management Tools	✓	✗	✓	✗	✓
Tailored Activities for Focus Enhancement	✓	✓	✓	✓	✓
Support for Self-Regulation Skills	✗	✗	✓	✗	✓
Personalized Game Flow	✗	✗	✗	✗	✓

[7] April 2022, " Mobile Application: A Serious Game Based in Gamification for Learning Mathematics in High School Students."

[8] May 2024, " Decreased impulsiveness and MEG normalization after AI-digital therapy in ADHD children: a RCT"

[9] December 2021, "Information and Communication Technologies Learning Methodologies for Children with ADHD"

[10] October 2024, "Provision of digital health interventions for young people with ADHD in primary care: findings from a survey and scoping review"

RESEARCH PROBLEM

How can we use machine learning and real-time emotion detection to improve focus and organization in children with predominantly inattentive ADHD, enhancing their learning and reducing behavioral issues?



OBJECTIVES

To create tools and activities that enhance focus, organizational skills, and attention span in children with predominantly inattentive ADHD, using adaptive learning and technology-assisted methods.



Create engaging and **personalized activities** that cater to the unique needs of children with ADHD, focusing on improving their attention and concentration.



Implement **gamified elements** into learning methodologies to enhance motivation and engagement among children with ADHD.



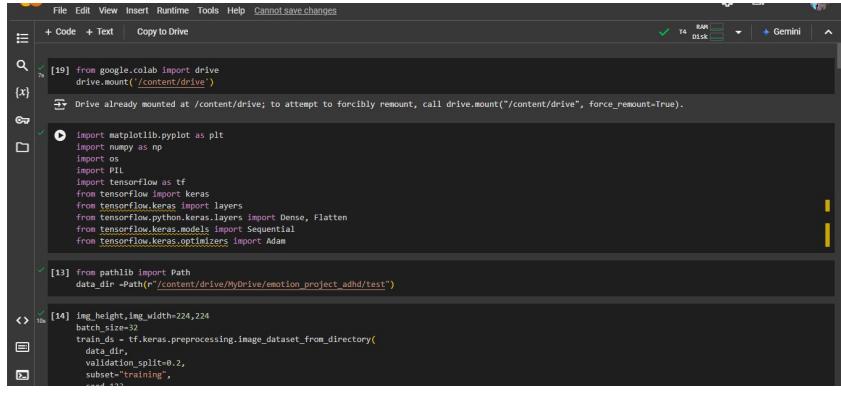
Utilize technology to analyze **emotional and behavioral responses**, enabling personalized interventions that adapt to the child's emotional state.

PP1 to PP2 PROGRESS

- Design user interfaces.
- Real time emotion capture and suggest activities
- Suggest activities to improve emotion.
- Developed Story based games
- Personalized game flow
- Detailed Reports (daily and full progress)
- Suggest activities according to the performance



PROOF OF COMPLETION



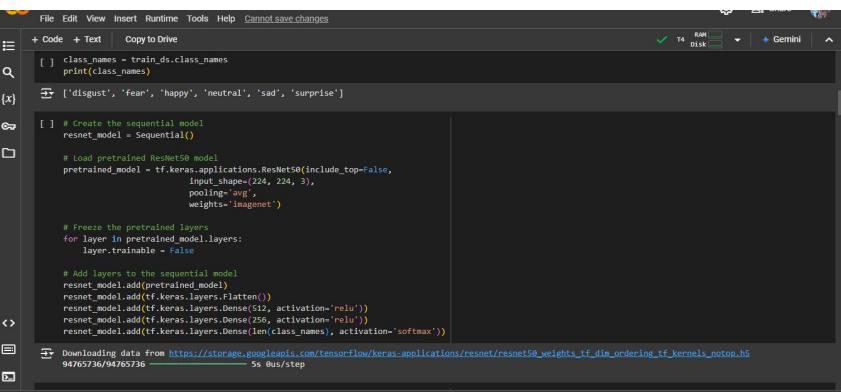
```
[19]: from google.colab import drive
drive.mount('/content/drive')

# Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

[x]
  ✓ import matplotlib.pyplot as plt
  ✓ import numpy as np
  ✓ import os
  ✓ import PIL
  ✓ import tensorflow as tf
  ✓ from tensorflow import keras
  ✓ from tensorflow.keras import layers
  ✓ from tensorflow.python.keras.layers import Dense, Flatten
  ✓ from tensorflow.keras.models import Sequential
  ✓ from tensorflow.keras.optimizers import Adam

[13]: from pathlib import Path
data_dir = Path('/content/drive/MyDrive/emotion project_adhd/test')

[x]
  [14]: img_height, img_width=224,224
batch_size=32
train_ds = tf.keras.preprocessing.image_dataset_from_directory(
    data_dir,
    validation_split=0.2,
    subset='training',
    image_size=(img_height, img_width),
    batch_size=batch_size)
```



```
[ ]
  ✓ class_names = train_ds.class_names
  print(class_names)

[x]
  [15]: ['disgust', 'fear', 'happy', 'neutral', 'sad', 'surprise']

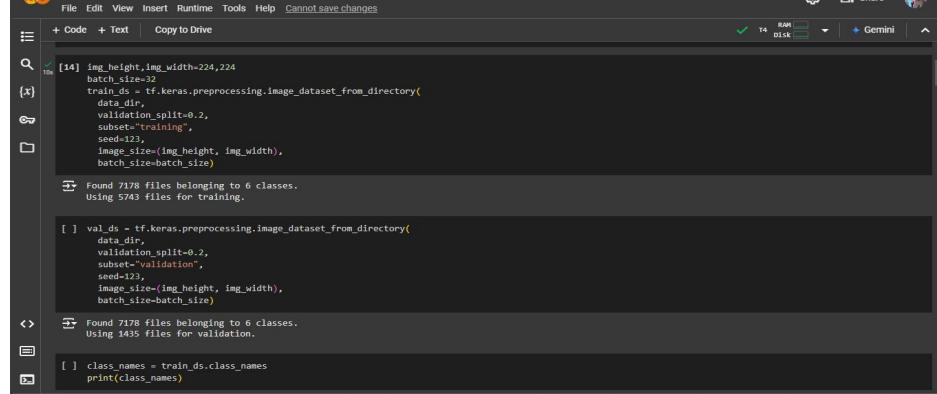
[x]
  [ ] # Create the sequential model
resnet_model = Sequential()

# Load pretrained ResNet50 model
pretrained_model = tf.keras.applications.ResNet50(include_top=False,
    input_shape=(224, 224, 3),
    pooling='avg',
    weights='imagenet')

# Freeze the pretrained layers
for layer in pretrained_model.layers:
    layer.trainable = False

# Add layers to the sequential model
resnet_model.add(pretrained_model)
resnet_model.add(tf.keras.layers.Flatten())
resnet_model.add(tf.keras.layers.Dense(512, activation='relu'))
resnet_model.add(tf.keras.layers.Dense(26, activation='relu'))
resnet_model.add(tf.keras.layers.Dense(len(class_names), activation='softmax'))
```

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/resnet50_weights_tf_dim_ordering_tf_kernels_notop.h5 94765736/94765736 - 55 MB/step



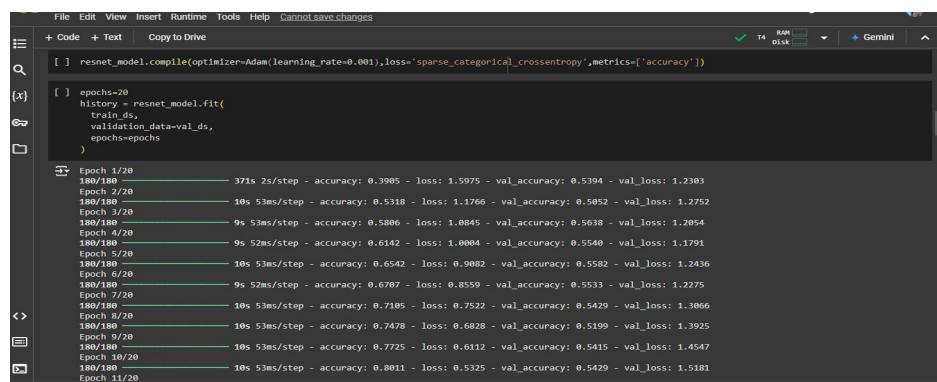
```
[14]: img_height, img_width=224,224
batch_size=32
train_ds = tf.keras.preprocessing.image_dataset_from_directory(
    data_dir,
    validation_split=0.2,
    subset='training',
    seed=123,
    image_size=(img_height, img_width),
    batch_size=batch_size)

# Found 7178 files belonging to 6 classes.
Using 5743 files for training.

[ ]
  val_ds = tf.keras.preprocessing.image_dataset_from_directory(
    data_dir,
    validation_split=0.2,
    subset='validation',
    seed=123,
    image_size=(img_height, img_width),
    batch_size=batch_size)

# Found 7178 files belonging to 6 classes.
Using 1435 files for validation.

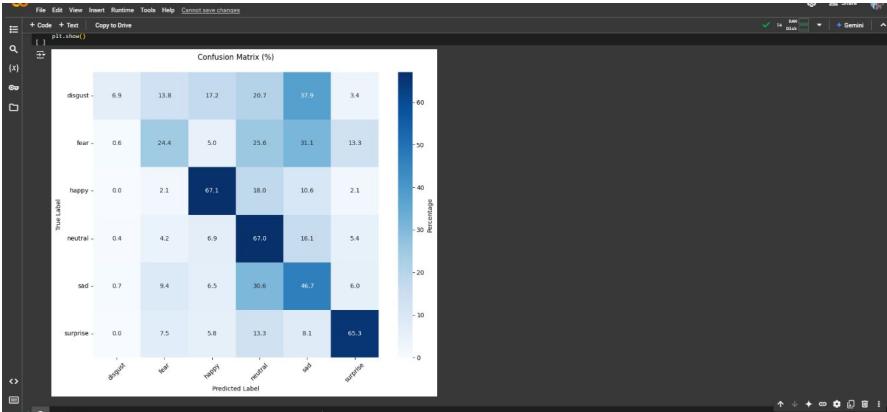
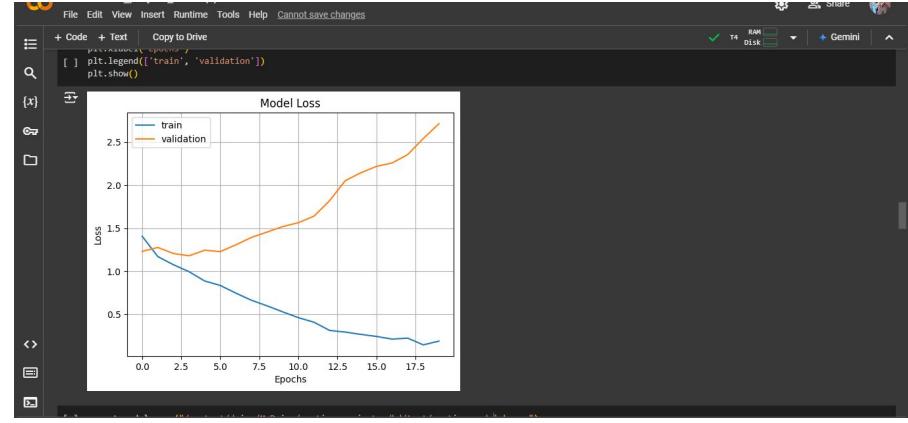
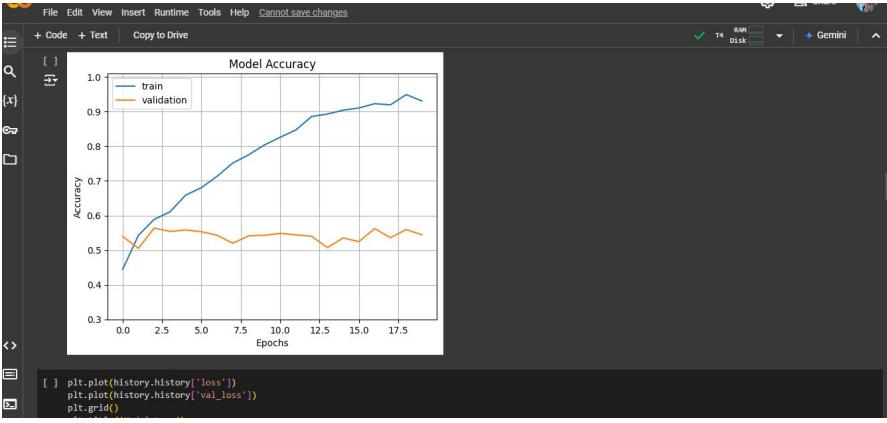
[ ]
  class_names = train_ds.class_names
  print(class_names)
```



```
[ ]
  resnet_model.compile(optimizer='Adam(learning_rate=0.001)', loss='sparse_categorical_crossentropy', metrics=['accuracy'])

[x]
  [16]: epochs=20
history = resnet_model.fit(
    train_ds,
    validation_data=val_ds,
    epochs=epochs
)

Epoch 1/20
180/180 - 371s 2s/step - accuracy: 0.3905 - loss: 1.5975 - val_accuracy: 0.5394 - val_loss: 1.2303
Epoch 2/20
180/180 - 10s 53ms/step - accuracy: 0.5318 - loss: 1.1766 - val_accuracy: 0.5952 - val_loss: 1.2752
Epoch 3/20
180/180 - 9s 53ms/step - accuracy: 0.5806 - loss: 1.0845 - val_accuracy: 0.5638 - val_loss: 1.2054
Epoch 4/20
180/180 - 9s 52ms/step - accuracy: 0.6142 - loss: 1.0004 - val_accuracy: 0.5540 - val_loss: 1.1791
Epoch 5/20
180/180 - 10s 53ms/step - accuracy: 0.6542 - loss: 0.9882 - val_accuracy: 0.5582 - val_loss: 1.2436
Epoch 6/20
180/180 - 9s 52ms/step - accuracy: 0.6707 - loss: 0.8559 - val_accuracy: 0.5533 - val_loss: 1.2275
Epoch 7/20
180/180 - 10s 53ms/step - accuracy: 0.7085 - loss: 0.7522 - val_accuracy: 0.5429 - val_loss: 1.3066
Epoch 8/20
180/180 - 10s 53ms/step - accuracy: 0.7478 - loss: 0.6828 - val_accuracy: 0.5199 - val_loss: 1.3925
Epoch 9/20
180/180 - 10s 53ms/step - accuracy: 0.7725 - loss: 0.6112 - val_accuracy: 0.5415 - val_loss: 1.4547
Epoch 10/20
180/180 - 10s 53ms/step - accuracy: 0.8011 - loss: 0.5325 - val_accuracy: 0.5429 - val_loss: 1.5181
```



Predicted Label

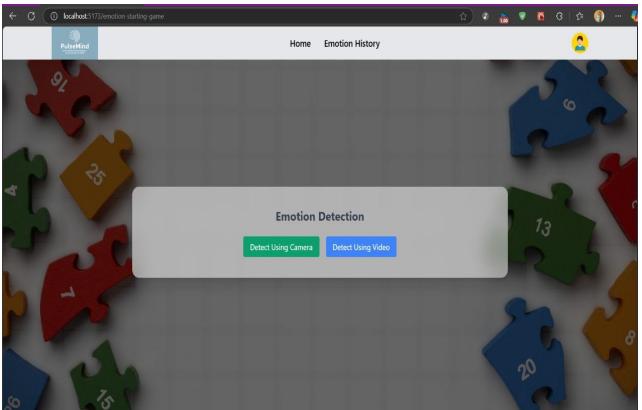
```
[ ] import cv2

def process_image(image_path):
    image=cv2.imread(image_path)
    img_height,img_width= 224,224
    image_resized = cv2.resize(image, (img_height,img_width))
    image=np.expand_dims(image_resized, axis=0)

    pred=model.predict(image)
    class_names = ['disgust', 'fear', 'happy', 'neutral', 'sad', 'surprise']
    output_class=class_names[np.argmax(pred)]
    probability=float(str(round(max(pred[0]),6)))
    return {"class":output_class , "probability":probability}

print(process_image(r"/content/drive/MyDrive/Dataset/archive (2)/test/sad/PrivateTest_10650656.jpg"))

1/1 -- 0s 23ms/step
{'class': 'sad', 'probability': 0.997832}
```



localhost:1711/suggestion-free/3/0003c91fca/30fb6

PulseMind

Home Emotion History

Emotion: Fear (Anxious, Nervous, Stressed)

Goal: Reduce anxiety and bring the child to a relaxed state

Bubble Breathing
5-10 mins
Blow soap bubbles slowly to regulate breathing and calm down.
[How to Perform](#)

Squeeze a Soft Toy
2-3 mins
Help release frustration through controlled squeezing.
[How to Perform](#)

localhost:1711/suggestion-free/3/0003c91fca/30fb6

PulseMind

Home Emotion History

Emotion: Fear (Anxious, Nervous, Stressed)

Goal: Reduce anxiety and bring the child to a relaxed state

Bubble Breathing

- Give the child a bottle of soap bubbles.
- Ask them to blow slowly, making big, gentle bubbles.
- Tell them to watch the bubbles float and pop.
- Repeat until they feel calmer.

[Close](#) [Complete](#)

localhost:1711/daily-progress/0/0003c91fca/30fb6

PulseMind

Home Emotion History

Daily Progress Dashboard

Game Details
Game Name: N/A
Plan Type: N/A
Session Started: 3/17/2025, 11:48:04 PM
Defaulting to initial emotion data as game details are incomplete.

You have not started the game yet.

Emotion Change Over Time

100
75
50
25

Initial Emotion

0:00 / 0:02

fear: 48.28%
anx: 51.72%

localhost:1711/daily-progress/0/0003c91fca/30fb6

PulseMind

Home Emotion History

User Progress Dashboard

Game Details
Game Name: Story 2: Luna and the Secret Garden
Plan Type: 7 Days

[Send Report To Doctor](#)

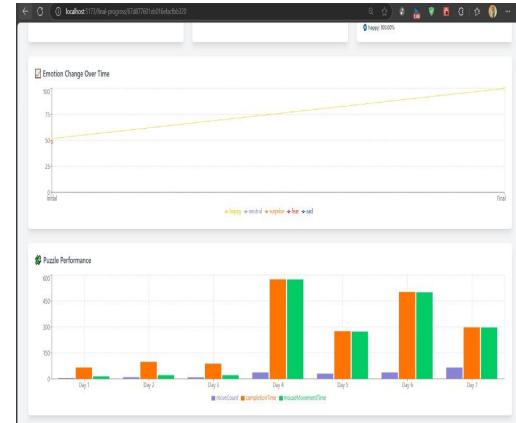
Initial Emotion

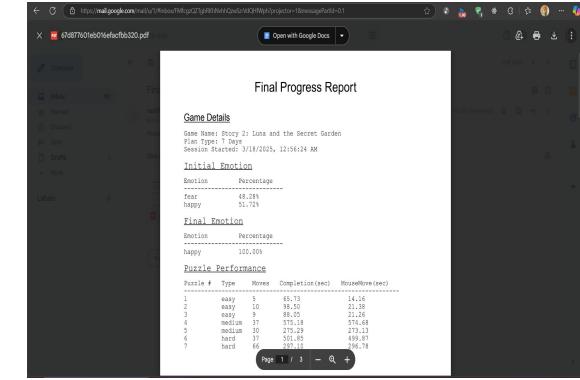
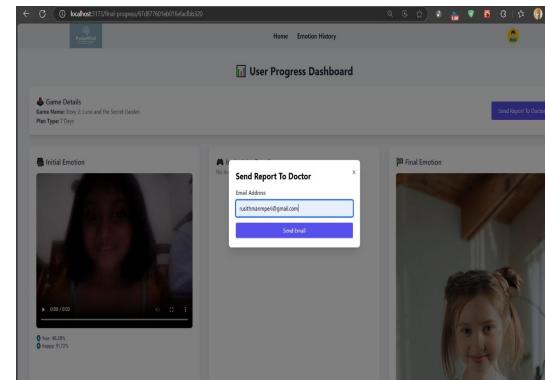
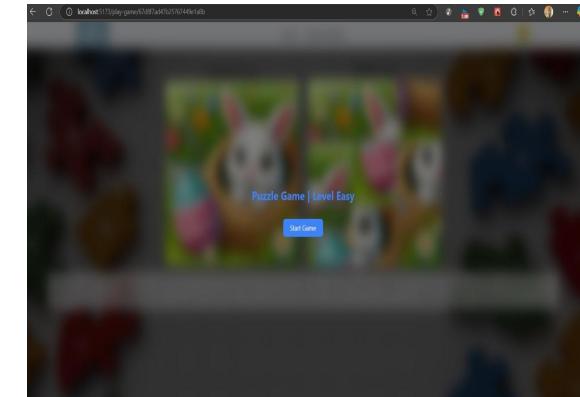
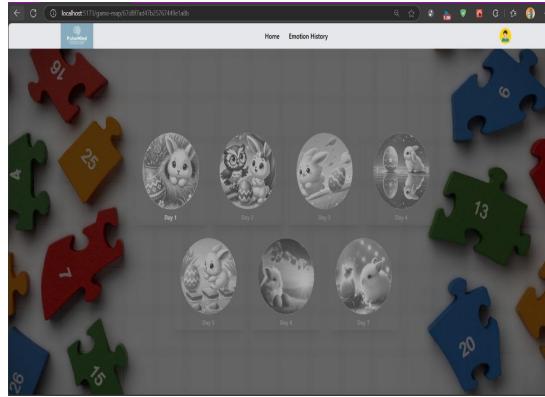
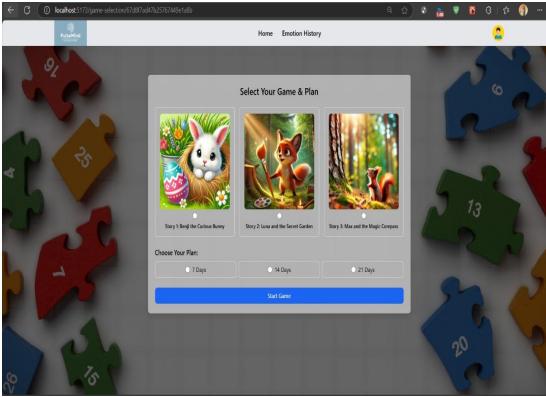
0:00 / 0:02

fear: 48.28%
anx: 51.72%

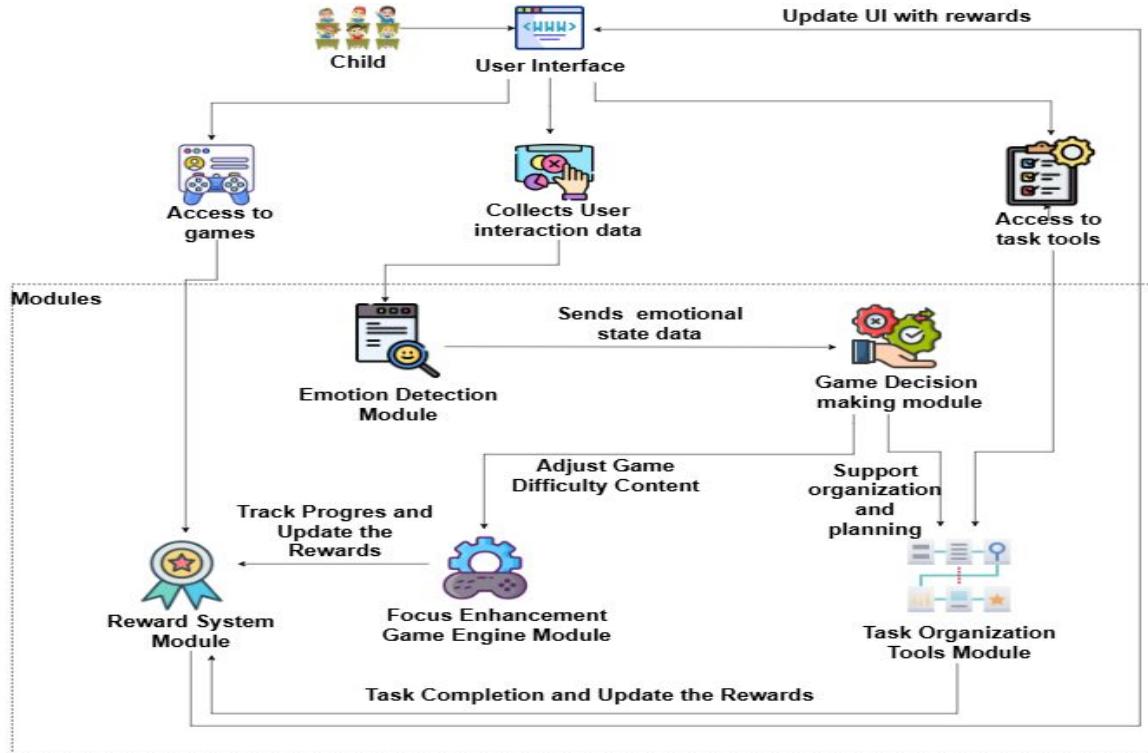
In-Activity Emotions
No in-game emotions recorded.

Final Emotion

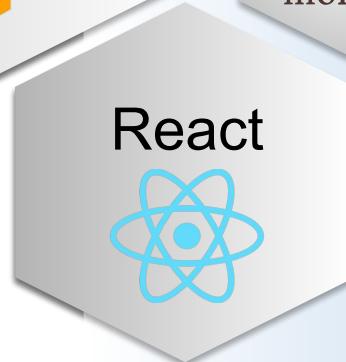
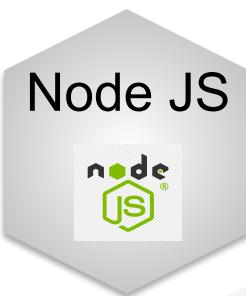




SYSTEM DIAGRAM



TECHNOLOGIES



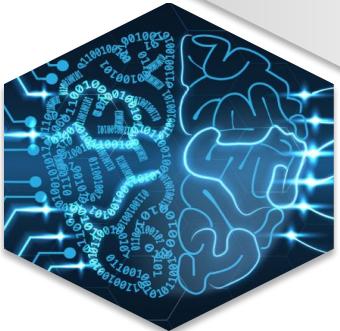
KEY PILLARS

Machine
learning



Games
Development

Image
Processing



COMPONENT SPECIFIC REQUIREMENTS

Functional Requirement



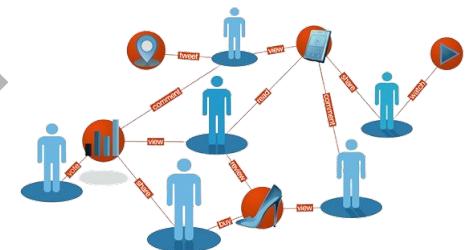
Identifying a child's behaviors through facial expressions



Performance and emotional analysis



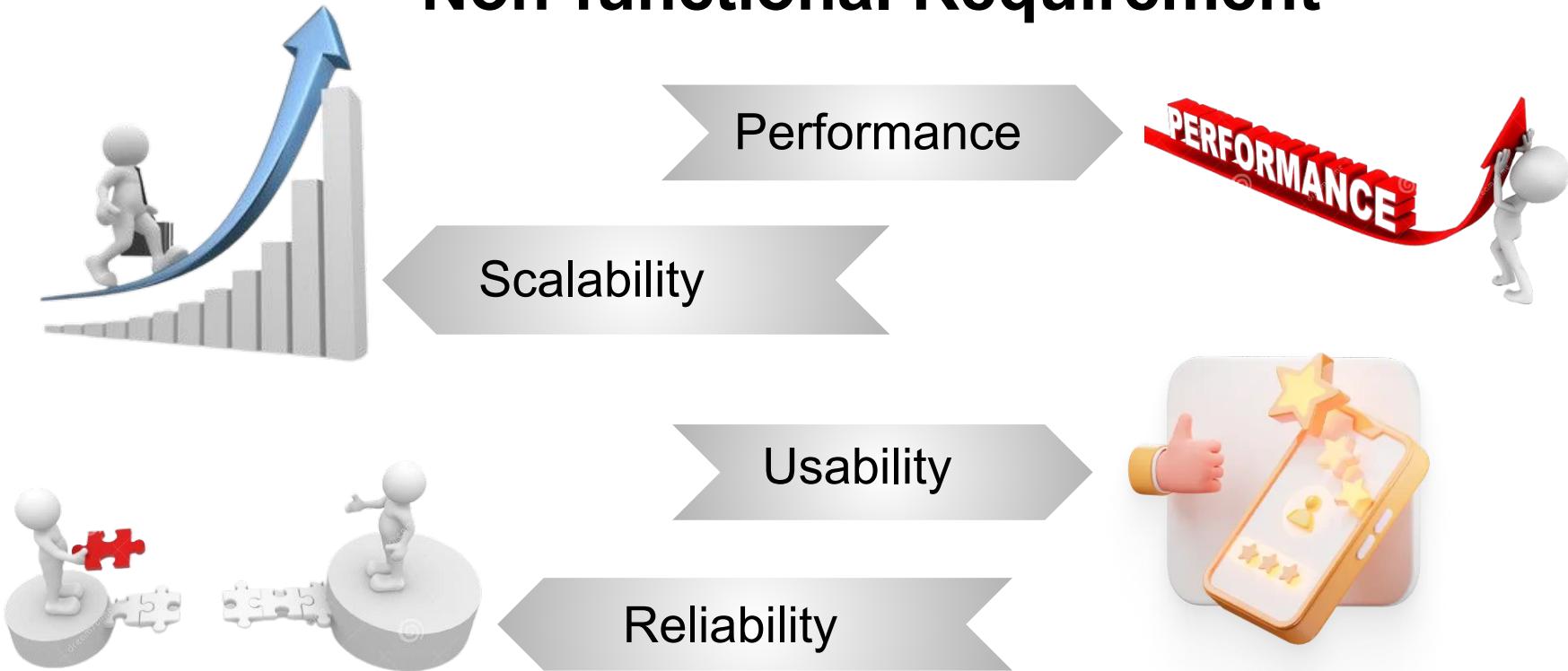
Image Preprocessing



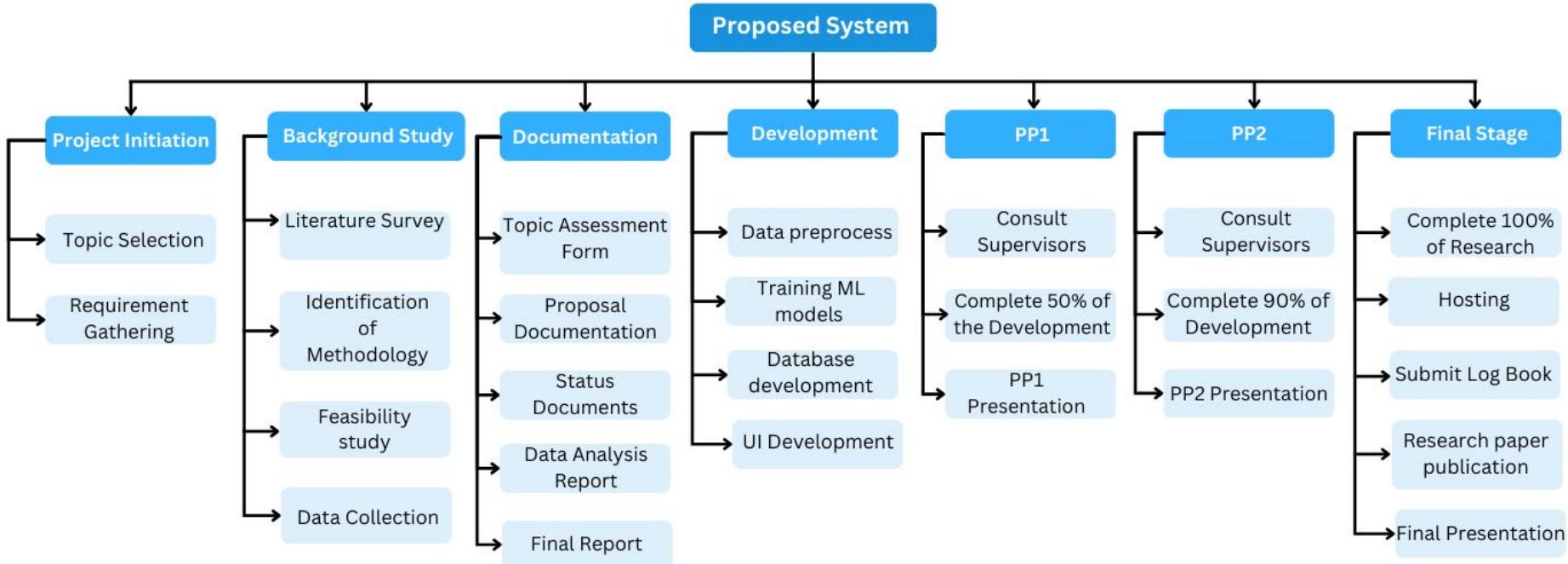
Personalized Game Flow

COMPONENT SPECIFIC REQUIREMENTS

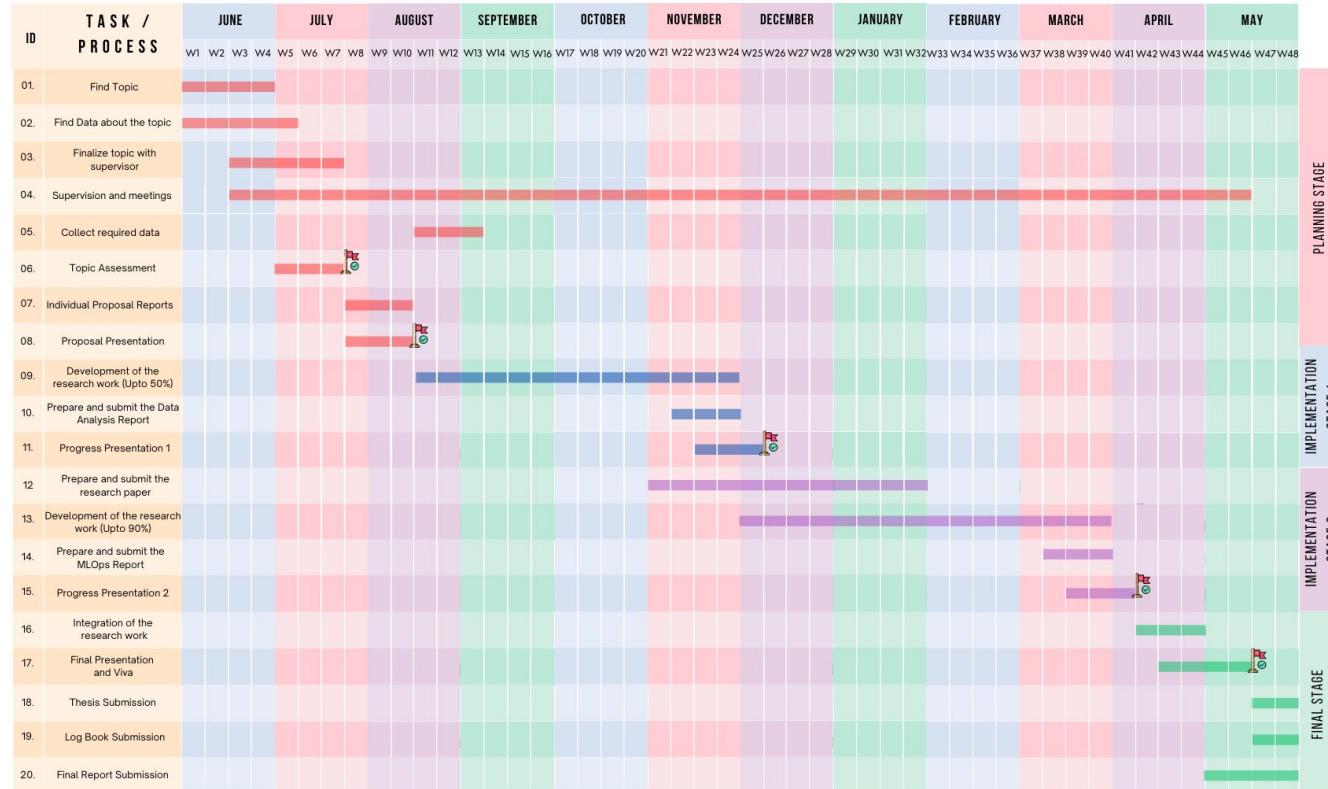
Non-functional Requirement



WORK BREAKDOWN CHART



GANTT CHART



REFERENCES

- [7]W. Ortiz, D. Castillo and L. Wong, "Mobile Application: A Serious Game Based in Gamification for Learning Mathematics in High School Students," 2022 31st Conference of Open Innovations Association (FRUCT), Helsinki, Finland, 2022, pp. 220-228, doi: 10.23919/FRUCT54823.2022.9770917. keywords: {Technological innovation;Avatars;User experience;Mobile handsets;Mathematics;Serious games;Mobile applications},
- [8]Danylyna Shpakivska Bilan, Irene Alice Chicchi Giglioli, Pablo Cuesta et al. Decreased impulsiveness and MEG normalization after AI- digital therapy in ADHD children: a RCT, 02 May 2024, PREPRINT (Version 1) available at Research Square [<https://doi.org/10.21203/rs.3.rs-4329802/v1>]
- [9]Andrade, Luis Javier Serpa, Roberto Agustin Garcia Velez, and Graciela Del Serpa Andrade. "Information and Communication Technologies Learning Methodologies for Children with Adhd." *AHFE 2022 Conference on Applied Human Factors and Ergonomics International*. 2022.
- [10]Gudka, R., Becker, K., Newlove-Delgado, T. et al. Provision of digital health interventions for young people with ADHD in primary care: findings from a survey and scoping review. *BMC Digit Health* 2, 71 (2024).
<https://doi.org/10.1186/s44247-024-00129-1>

Enhancing Impulse Control in ADHD Students Through Structured Timetables



**IT21379574 | Dilshani H.T.D.P.
Specialization: Information technology**

INTRODUCTION

The system for children with hyperactive-impulsive ADHD offers personalized timetables and physical activities, empowering caregivers with tailored solutions for effective symptom management and improved impulse control.



RESEARCH GAP

Component	[11]	[12]	[13]	Proposed System
Current Approaches	✓	✗	✗	✓
Limited Integration of Activities	✓	✗	✗	✓
Lack of Personalization	✗	✓	✓	✓
Engagement and Enjoyment	✓	✓	✗	✓
Need for Structured Timetables	✗	✗	✗	✓

[11] August 2024-The effectiveness of parent training programs for children with ADHD aged 6-11 years: A systematic

[12] August 2021-Characterizing neuroanatomic heterogeneity in people with and without ADHD based on subcortical brain volumes

[13] June 2019:Development and Standardization of an Impulse Control Scale for Adolescents

RESEARCH PROBLEM

How can personalized, structured activities, including physical exercises and impulsivity level predictions, help children with hyperactive-impulsive ADHD improve impulse control and manage their energy effectively through tailored timetables?



OBJECTIVES

The project aims to create a personalized app that aids children with ADHD by offering structured activities like physical exercises and mindfulness techniques.



Develop five pre-designed schedules with activities to address different impulsive behaviors in children with ADHD.



Allow teachers/caregivers to customize and modify timetables, ensuring the activities are engaging and fit the specific needs of each child.



This will provide a personalized, structured routine for each child, helping them manage their energy, improve impulse control, and develop self-regulation through engaging, enjoyable activities.

PP1 to PP2 PROGRESS

- Implement frontend and backend
- Collect data and generate dataset
- Train a model to identify children's impulsivity level
- Implement the recommended physical activities
- Implement the questioner dashboard
- Improve the model and make customized time tables
- Develop the progress level and recommendations.



PROOF OF COMPLETION

Step 1 of 6

Basic Details

Name: Savi

Age: 4

Gender: Female

ADHD Subtype: Select Subtype

Step 2 of 6

A1 Questions

What do you do when you feel angry?
Select an answer

What do you do when you got stuck and need help with homework?
Select an answer

How do you feel when someone ask about you?
Select an answer

Previous Next

ADHD Timetable
Loading timetable...

Recommendations

Calm Sounds Meditation Music Therapy

Start Over

Menu

Questionnaire Dashboard



ome

Questionnaire

History

Predictions Monitoring

Week 1 Progress

0% Completed

Week 2 Progress

0% Completed

Week 3 Progress

0% Completed

- Unknown Answer
 - Unknown Answer
 - Calm
 - Help them pick it up
 - Raise your hand and wait to be called on
 - Get angry and yell at them
 - Unknown Answer
 - Complain loudly to everyone
 - Unknown Answer
 - Focuses and attempts to complete it.
 - Tries again and learns from the mistake.
 - Unknown Answer
 - Distracts others or avoids participation.
 - Take it to a teacher or an adult.
 - Unknown Answer
 - Ask a friend for help.
 - Get upset and demand they replace it.
 - Tell an adult about it.

The screenshot displays a developer's environment with several open tabs and windows. The browser tabs include 'Welcome', 'app.py', 'model_selection.py', 'model_training.py', and 'ADHD_dataset.csv'. The code editor shows 'ActivityForms.js' with React code for a form component. The terminal shows pip upgrade logs. The status bar at the bottom indicates the file is 'master' and shows system information like battery level, signal strength, and date/time.

The screenshot shows a Jupyter Notebook interface with several open files in the left sidebar:

- Untitled (Workspace)
- ADHD Questioner
- model
- A3_Q5_encoder.pkl
- A4_Q1_encoder.pkl
- A4_Q2_encoder.pkl
- A4_Q3_encoder.pkl
- A4_Q4_encoder.pkl
- A4_Q5_encoder.pkl
- activity_classification_report.py
- activity_type.py
- ADHD_Subtype_encoder.pkl
- adhd_subtype_encoder.pkl
- app.py
- best_activity_model.pkl
- best_implicitly_model.pkl
- gender_encoder.pkl
- implicitly_classification_report.py
- implicitly_level.png
- model_results.csv
- model_selection.py
- model_training.py
- pscpd.pkl
- scales.pkl

The main area displays the content of the `model_selection.py` file:

```
y_implicitly = y_implicitly['encoder'].fit_transform(y_implicitly)

# Scale numerical features
scaler = StandardScaler()
X_scaled = scaler.fit_transform(X)

# Split data into train and test sets
X_train, X_test, y_activity_train, y_activity_test, y_implicitly_train, y_implicitly_test = train_test_split(
    X_scaled, y_activity, y_implicitly, test_size=0.2, random_state=42)

# Address Class Imbalance with SMOTE
smote = SMOTE(random_state=42)
X_train_activity, y_train_activity = smote.fit_resample(X_train, y_activity_train)
X_train_implicitly, y_train_implicitly = smote.fit_resample(X_train, y_implicitly_train)

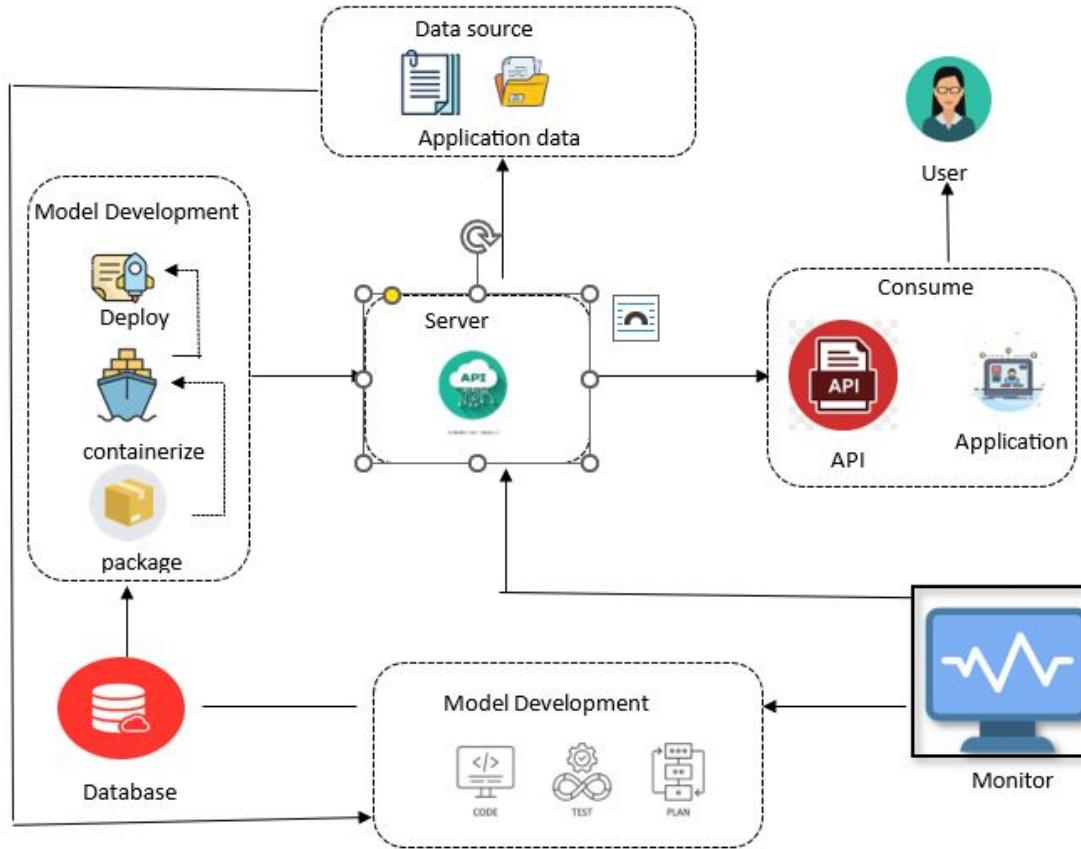
# Define models
models = [
    "Logistic Regression": LogisticRegression(max_iter=5000),
    "SVM": SVC(probability=True),
    "Random Forest": RandomForestClassifier(),
    "Neural Network": MLPClassifier(max_iter=5000)
]
```

Below the code editor is a terminal window showing the following output:

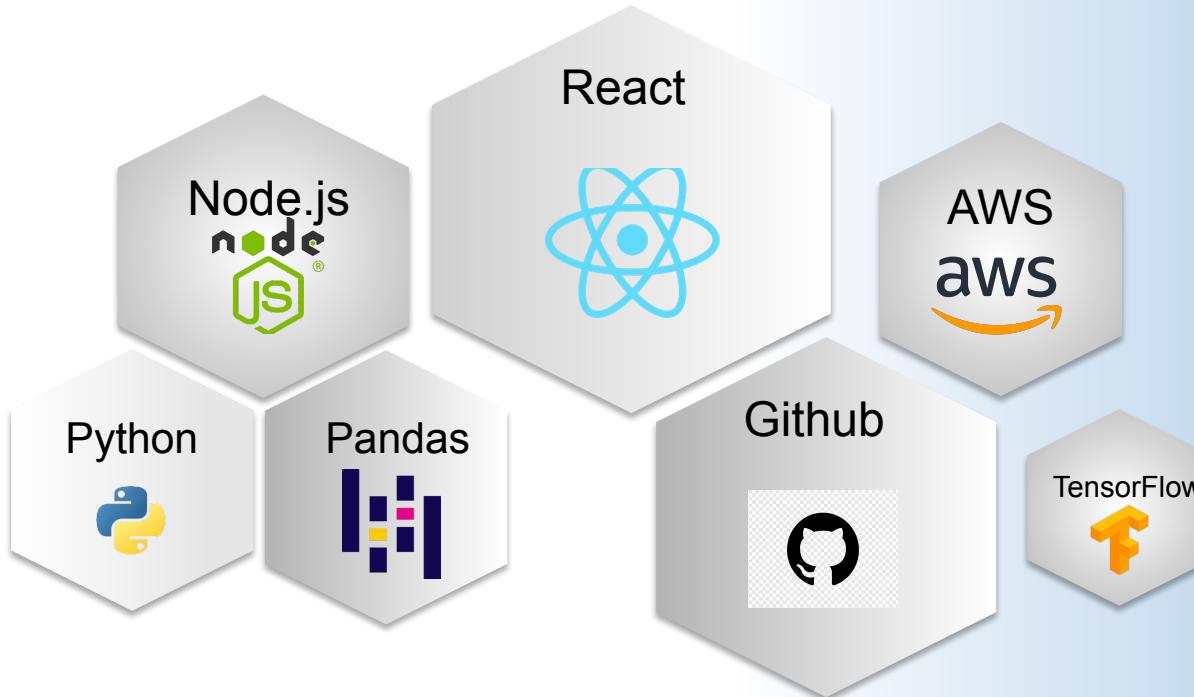
```
Requirement already satisfied: joblib<1.1.1 in c:\software\adh\questioner\venv\lib\site-packages (from imbalance-dlearn)
Requirement already satisfied: threadpoolctl<2.0.0 in c:\software\adh\questioner\venv\lib\site-packages (from imbalance-dlearn)
(3.5.0)
[notebook] A new release of pip is available: 24.0 -> 24.3.1
[notebook] To update, run: python -m pip install --upgrade pip
```

The status bar at the bottom right shows: In 11 Col 19 Spaces 4 UTR-8 15:57PM 15/09/24

SYSTEM DIAGRAM



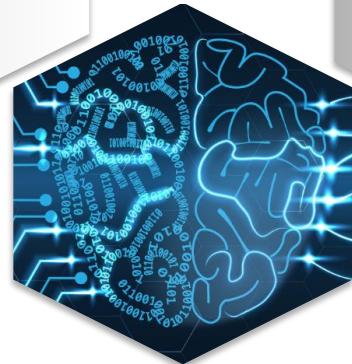
TECHNOLOGIES



KEY PILLARS

Machine
learning

Cloud
computing



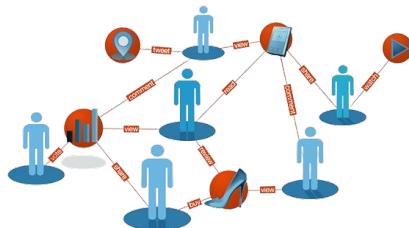
COMPONENT SPECIFIC REQUIREMENTS

Functional Requirement

Customizable Timetables



Activity Integration

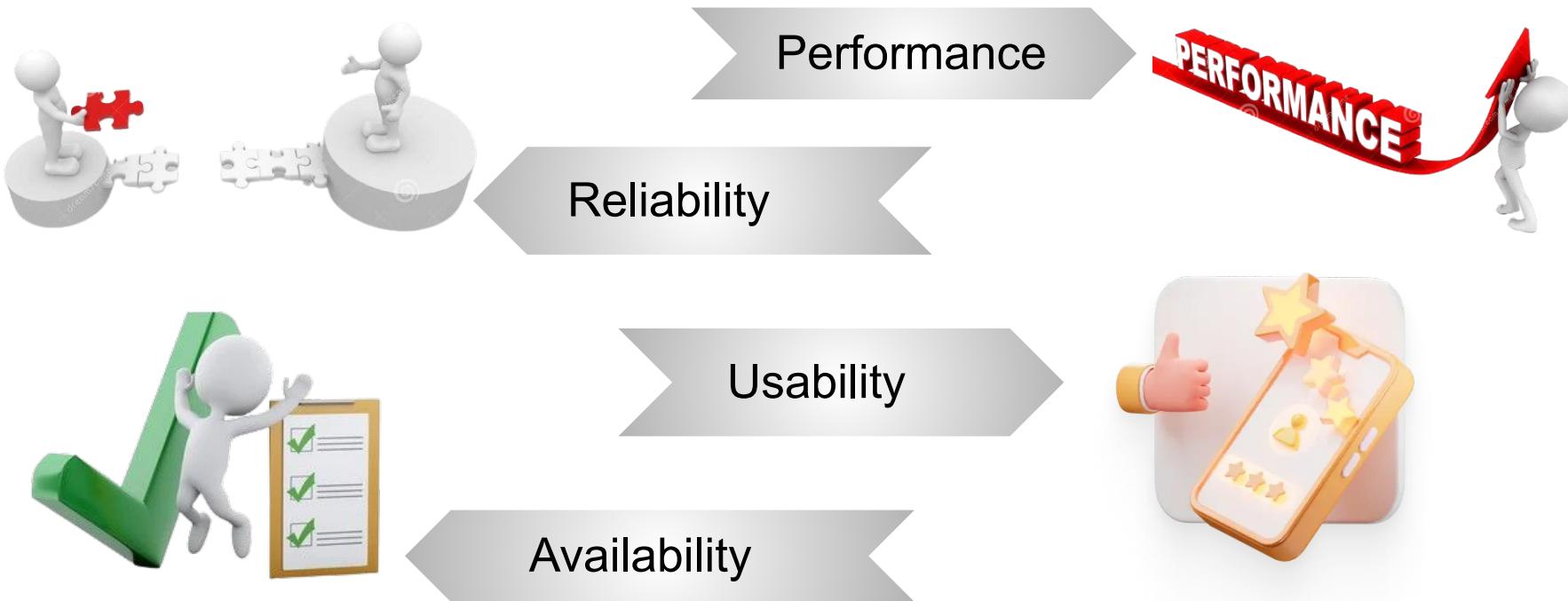


Tracking and Feedback

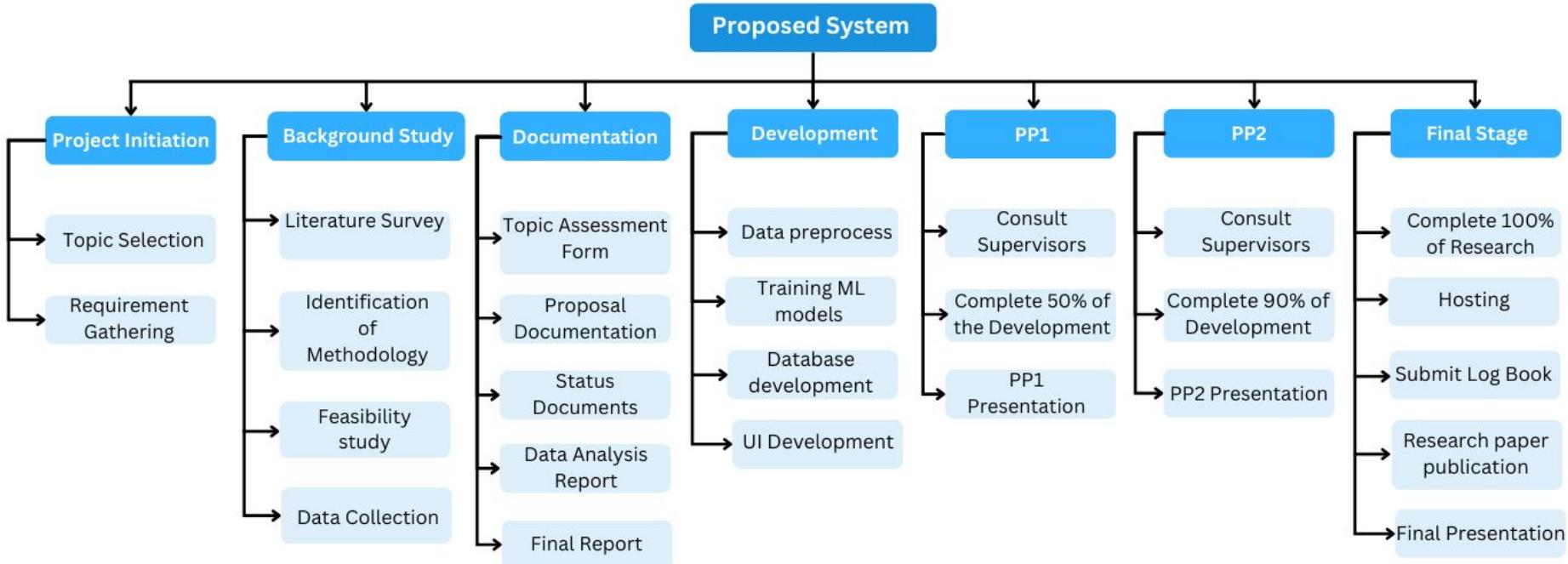
Notifications and Reminders

COMPONENT SPECIFIC REQUIREMENTS

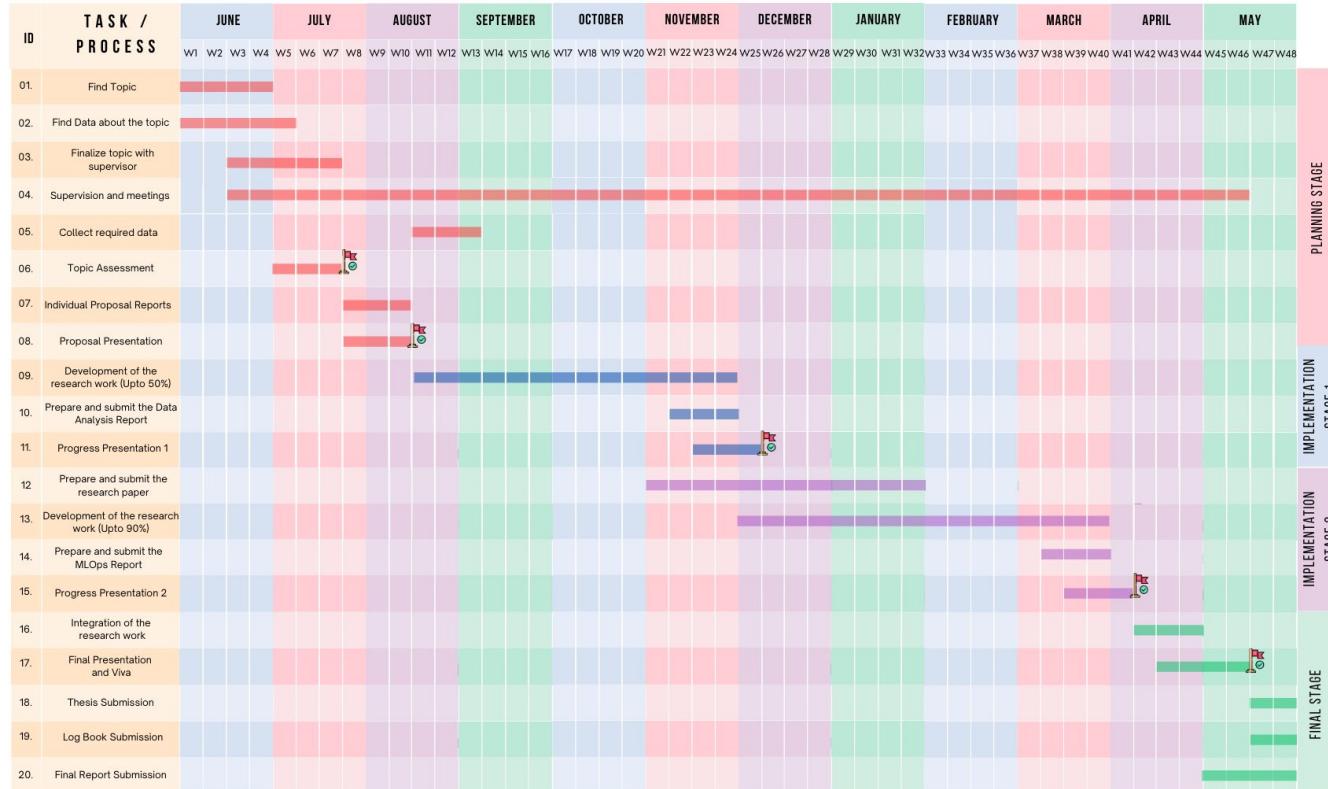
Non-functional Requirement



WORK BREAKDOWN CHART



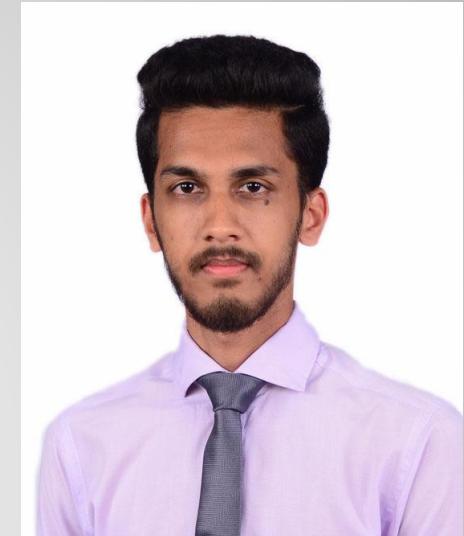
GANTT CHART



REFERENCES

- [11] S. McGrath, "Research Review - The Effectiveness of Parent Training Programs for Children with ADHD Aged 6-11 Years: A Systematic Review," *ResearchGate*, Oct. 2024. [Online]. Available: <https://www.researchgate.net/publication/382875375>. [Accessed: 16-Oct-2024].
- [12] <https://www.researchgate.net/publication/354311415>
Characterizing neuroanatomic heterogeneity in people with and without ADHD based on subcortical brain volumes
- [13] S. K. Beura, A. R. Panigrahi, and P. Yadav, "Development and Standardization of an Impulse Control Scale for Adolescents," *Published Online*, Jun. 12, 2019. Available: <https://www.researchgate.net/publication/360385680>

AI-Driven Adaptive Learning and Intervention System with future predictions for Personalized ADHD Management in Children



IT21380532 | Halliyadda H.U.M.S.
Specialization: Information technology

INTRODUCTION

The research aims to develop an intelligent system for data analysis, pattern recognition, **future challenge prediction** and **prevention mechanisms prediction** that dynamically adapts intervention strategies based on a child's interactions to provide personalized and effective support over time.



What we can predict?

- Progress in academic or learning tasks - **Academic Struggles**
- Attention span and focus duration - **Increased risk of anxiety**
- Impulsivity levels during tasks - **Increased or decreased impulsivity**
- Social interaction patterns and responses - **Social Challenges**



RESEARCH GAP

Component	[14]	[15]	[16]	Proposed System
Monitoring Child's interactions	✓	✓	✓	✓
Personalized Feedbacks	✗	✓	✓	✓
User Dashboard	✓	✗	✓	✓
Future Predictive Insights	✗	✗	✗	✓
AI-Driven Recommendations	✗	✗	✗	✓
Alert and Notifications	✓	✓	✗	✓

[14] January 2022 - An Overview of Predicting the Prevalence of ADHD

[15] June 2023 - Diagnosing attention-deficit hyperactivity disorder (ADHD) using artificial intelligence: a clinical study in the UK

[16] April 2020 - Machine-Learning prediction of comorbid substance use disorders in ADHD youth using Swedish registry data

RESEARCH PROBLEM

Current ADHD management methods are unable to adapt to individual needs, leading to inconsistent outcomes. An AI-driven system is needed to provide dynamic, personalized strategies and proactive solutions for better ADHD management.



OBJECTIVES

Monitor the child's interactions with the application and Collect data on key indicators such as response times, completion rates, attention span, and behavioral trends to assess the child's progress.



Develop a user-friendly interface for educators and caregivers to easily track progress, view key achievements, and identify areas that need further attention.



Leverage AI to predict potential challenges the child may face in the future based on current and historical performance data.

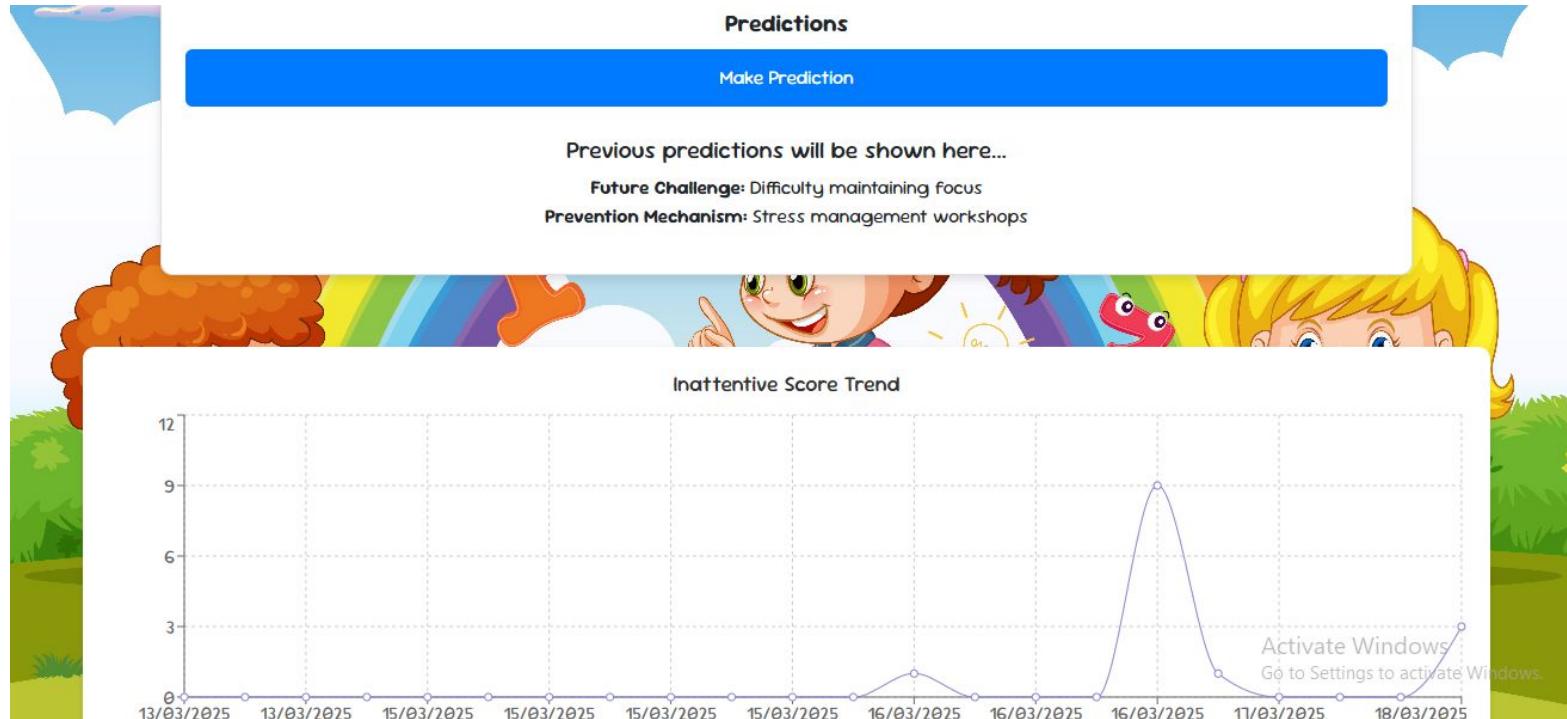


Deliver tailored feedback to both the child and caregivers, offering insights into strengths, areas for improvement, and specific intervention strategies.

- Implement frontend and backend
- Collect and generate dataset
- Train a future predictive machine learning model
- Implement user-friendly dashboard for monitoring
- Validate and improve model
- Implement user-friendly dashboard for monitoring
- Integrate dashboard with components to monitoring



PROOF OF COMPLETION





ADHD Predictions Form

Age
4

Gender
Female

ADHD Subtype
Select ADHD Subtype

Inattentive Score
3

Hyperactive-Impulsive Score
3

Combined Score

File Edit Selection View Go Run ... ↶ ↷ ADH Challenge

EXPLORER model_selection.py model_creation.py app.py

ADH CHALLENGE model > app.py

frontend

src

setupTests.js .gitignore package-lock.json package.json README.md

model

Academic_Grade_encoder.pkl ADHD_subtype_encoder.pkl adhd_children_dataset.csv

app.py

best_future_challenge.model.pkl best_logistic_regression.model.pkl best_prevention_mechanism_mode... challenge_encoder.pkl Current_Strategy_encoder.pkl Figure_1.png Gender_encoder.pkl model_creation.py model_selection.py prevention_encoder.pkl scaler.pkl shap_future_challenge.png Teacher_Feedback_encoder.pkl

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL PORTS

(Use 'node --trace-warnings ...' to show where the warning was created)

(node:15888) [MONGODB DRIVER] Warning: useUnifiedTopology is a deprecated option: useUnifiedTopology has no effect since Node.js (node:15888) [MONGODB DRIVER] Warning: useUnifiedTopology is a deprecated option: useUnifiedTopology has no effect since Node.js Driver version 4.0.0 and will be removed in the next major version

Server running on http://localhost:5981

MongoDB connected...

POST /api/predict 200 1788.478 ms - 170

POST /api/predict 200 38.227 ms - 185

MongoDB connected...

POST /api/predict 200 1788.478 ms - 170

POST /api/predict 200 38.227 ms - 185

POST /api/predict 200 1788.478 ms - 170

POST /api/predict 200 38.227 ms - 185

backend... model.m... frontend...

OUTLINE TIMELINE

master model_selection.py model_creation.py app.py

Col 22 Spaces: 4 LF Python Go Live Quokka Prettier

```
model_selection.py
import pandas as pd
from sklearn.model_selection import train_test_split, cross_val_score
from sklearn.ensemble import RandomForestClassifier
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.neural_network import MLPClassifier
from sklearn.pipeline import make_pipeline
from sklearn.svm import SVC
import numpy as np

# Load the dataset
file_path = "adhd_children_dataset.csv" # Update the path as needed
dataset = pd.read_csv(file_path)

# Drop unnecessary columns
dataset.drop(columns=["Child id", "Name"])

# Separate features (X) and targets (y)
X = dataset.drop(columns=["Future Challenge", "Prevention Mechanism"])
y_challenge = dataset["Future Challenge"]

# Encode categorical features and targets
categorical_columns = X.select_dtypes(include=[object]).columns
for column in categorical_columns:
    X[column] = LabelEncoder().fit_transform(X[column])

# Train the model
X_train, X_test, y_train, y_test = train_test_split(X, y_challenge, test_size=0.2, random_state=42)
grid_search = GridSearchCV(
    estimator=MLPClassifier(max_iter=1000),
    param_grid={
        "hidden_layer_sizes": [(10, 10), (20, 20), (50, 50)],
        "activation": ["relu", "tanh"],
        "solver": ["sgd", "adam"]
    },
    cv=5,
    scoring="accuracy"
)
grid_search.fit(X_train, y_train)
best_params = grid_search.best_params_
best_accuracy = grid_search.best_score_
print(f"Best accuracy: {best_accuracy * 100:.2f}%")
print(f"Best parameters: {best_params}")

# Make predictions
y_pred = grid_search.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print(f"Test accuracy: {accuracy * 100:.2f}%")

# Save the model
best_future_challenge_model.pkl
best_logistic_regression_model.pkl
best_prevention_mechanism_model.pkl
challenge_encoder.pkl
Current_Strategy_encoder.pkl
Figure_1.png
Gender_encoder.pkl
model_creation.py
model_selection.py
prevention_encoder.pkl
scaler.pkl
shape_future_challenge.png
Teacher_Feedback_encoder.pkl
```

```
model_selection.py
import warnings
from sklearn.exceptions import ConvergenceWarning
warnings.filterwarnings("ignore", category=UserWarning)
warnings.filterwarnings("ignore", category=ConvergenceWarning)

import pandas as pd
from sklearn.model_selection import train_test_split, GridSearchCV
from sklearn.preprocessing import LabelEncoder, StandardScaler
from sklearn.metrics import classification_report
from imblearn.over_sampling import SMOTE
from joblib import dump
from sklearn.linear_model import LogisticRegression
import numpy as np

# Load the dataset
data_path = "adhd_children_dataset.csv" # Replace with your dataset path
adhd_data = pd.read_csv(data_path)

# Drop irrelevant columns
columns_to_drop = ["Child id", "Name"]
X = adhd_data.drop(columns=columns_to_drop + ["Future Challenge", "Prevention Mechanism"])

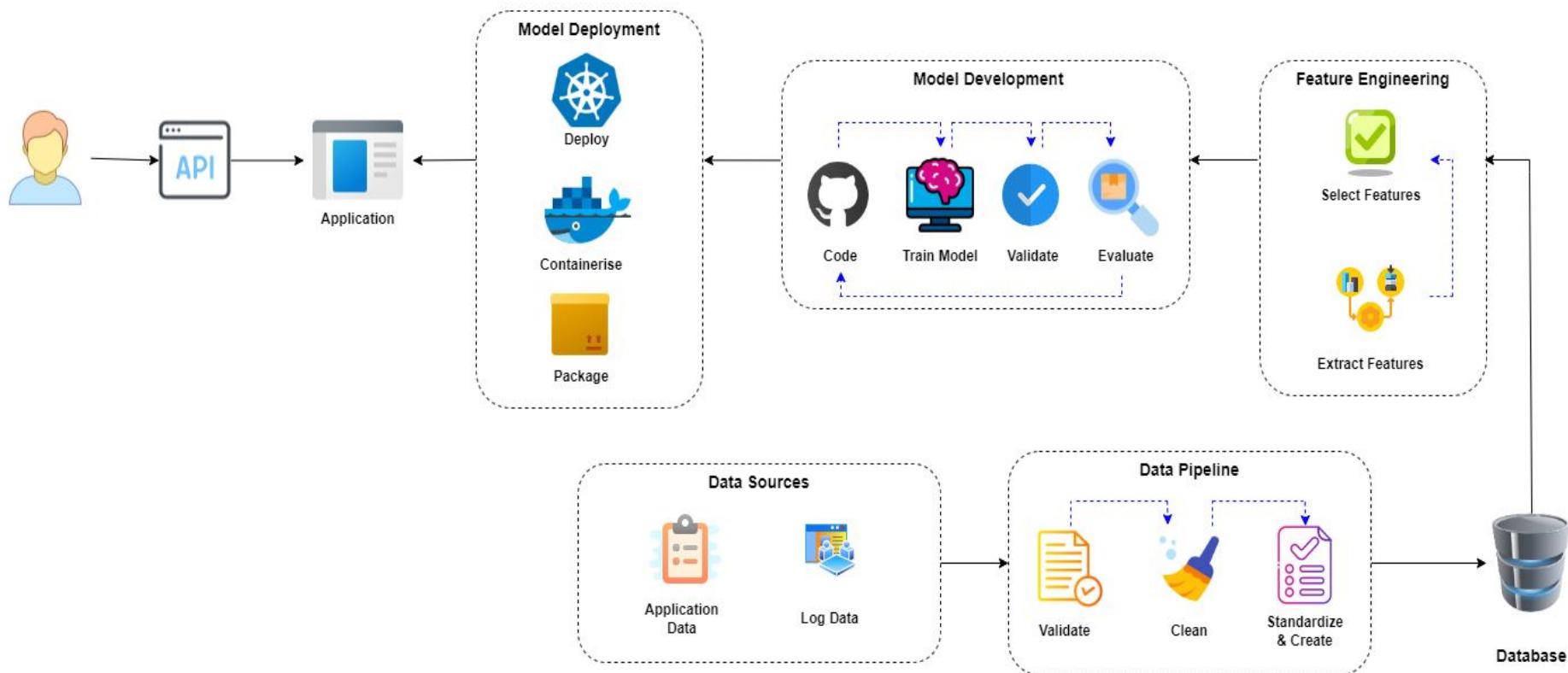
# Define target variables
y_challenge = X["Future Challenge"]

# Train the model
X_train, X_test, y_train, y_test = train_test_split(X, y_challenge, test_size=0.2, random_state=42)
grid_search = GridSearchCV(
    estimator=LogisticRegression(max_iter=1000),
    param_grid={
        "C": [0.001, 0.01, 0.1, 1, 10, 100]
    },
    cv=5,
    scoring="accuracy"
)
grid_search.fit(X_train, y_train)
best_params = grid_search.best_params_
best_accuracy = grid_search.best_score_
print(f"Best accuracy: {best_accuracy * 100:.2f}%")
print(f"Best parameters: {best_params}")

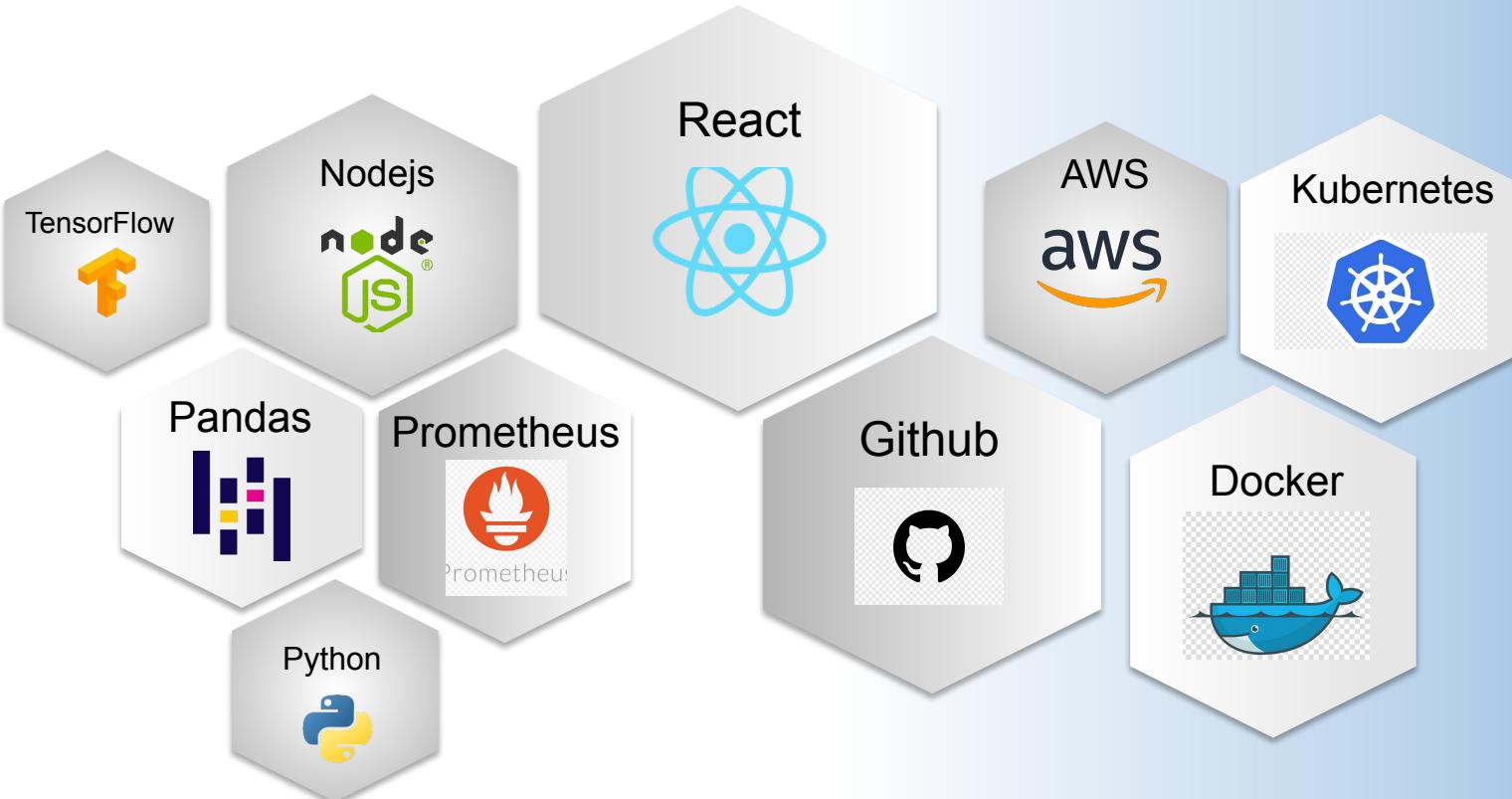
# Make predictions
y_pred = grid_search.predict(X_test)
accuracy = accuracy_score(y_test, y_pred)
print(f"Test accuracy: {accuracy * 100:.2f}%")

# Save the model
best_future_challenge_model.pkl
best_logistic_regression_model.pkl
best_prevention_mechanism_model.pkl
challenge_encoder.pkl
Current_Strategy_encoder.pkl
Figure_1.png
Gender_encoder.pkl
model_creation.py
model_selection.py
prevention_encoder.pkl
scaler.pkl
shape_future_challenge.png
Teacher_Feedback_encoder.pkl
```

SYSTEM DIAGRAM

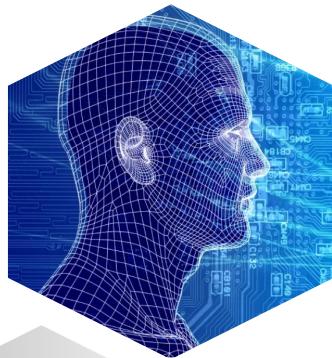


TECHNOLOGIES



KEY PILLARS

Cloud
Computing



Adaptive
Monitoring

Machine
Learning

COMPONENT SPECIFIC REQUIREMENTS

Functional Requirement



Data collection and feature engineering

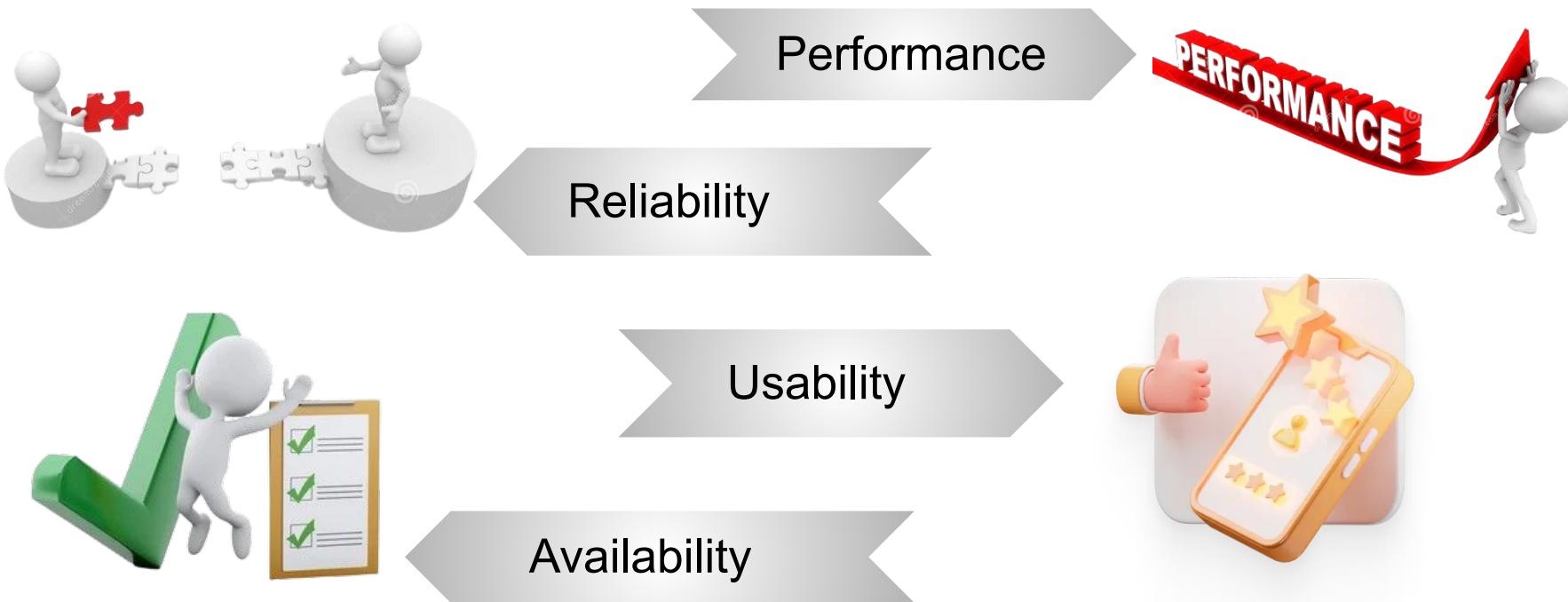
User-friendly Dashboard

Future Predictions and Risk Identification

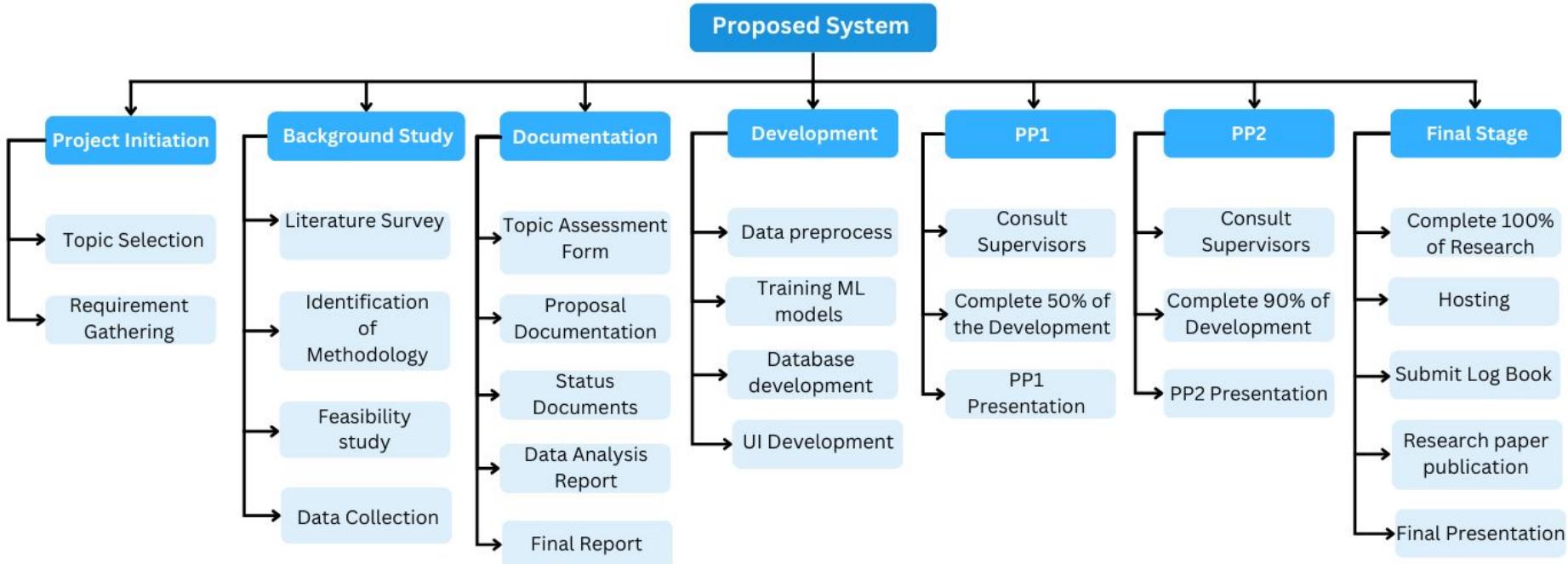
Personalized Feedback and Alerts

COMPONENT SPECIFIC REQUIREMENTS

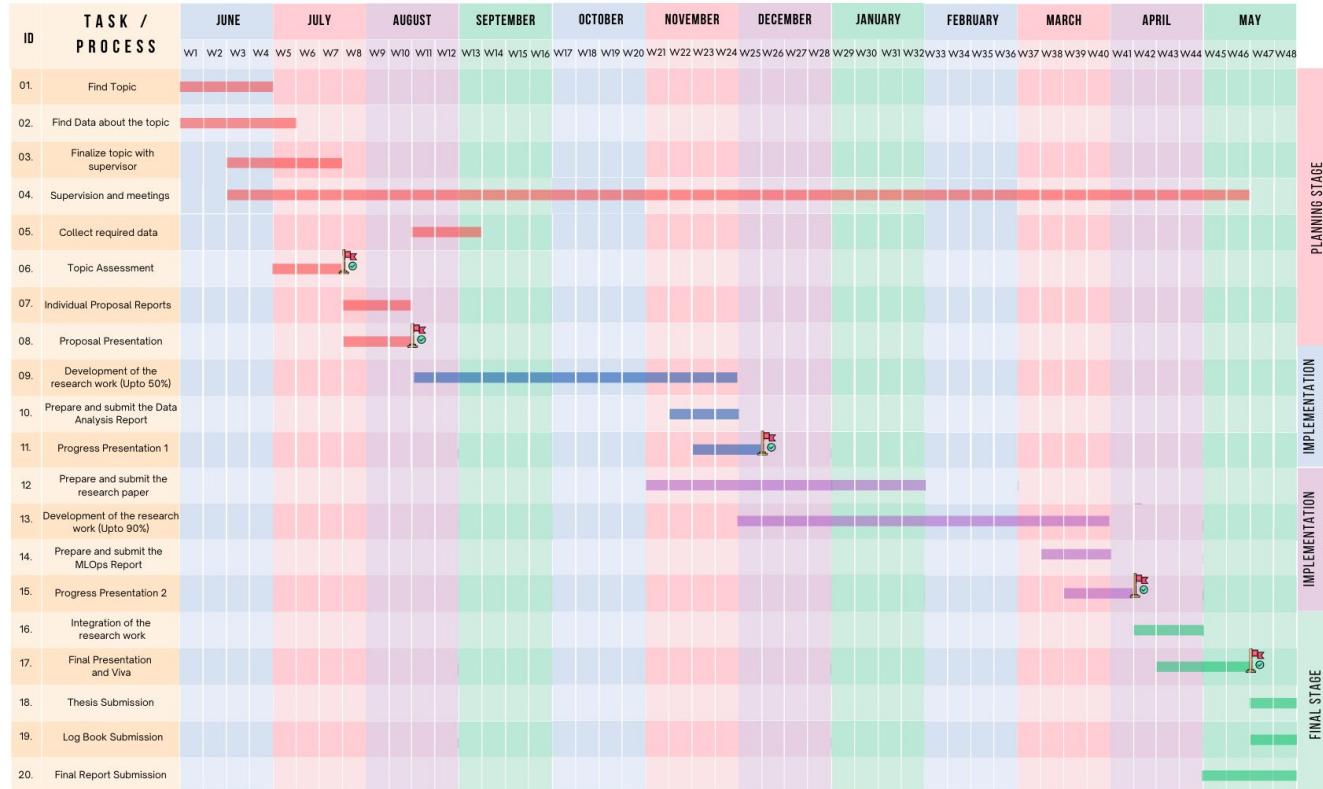
Non-functional Requirement



WORK BREAKDOWN CHART



GANTT CHART



REFERENCES

- [14] H. W. Loh, C. P. Ooi, P. D. Barua, E. E. Palmer, F. Molinari, and U. R. Acharya, "Automated detection of ADHD: Current trends and future perspective," *Computers in Biology and Medicine*, vol. 146, p. 105525, Jul. 2022, doi: <https://doi.org/10.1016/j.combiomed.2022.105525>.
- [15] T. Chen, I. Tachmazidis, S. Batsakis, M. Adamou, E. Papadakis, and G. Antoniou, "Diagnosing attention-deficit hyperactivity disorder (ADHD) using artificial intelligence: a clinical study in the UK," *Frontiers in Psychiatry*, vol. 14, Jun. 2023, doi: 10.3389/fpsyg.2023.1164433.
- [16] Y. Zhang-James, Q. Chen, R. Kuja-Halkola, P. Lichtenstein, H. Larsson, and S. V. Faraone, "Machine-Learning prediction of comorbid substance use disorders in ADHD youth using Swedish registry data," *Journal of Child Psychology and Psychiatry*, vol. 61, no. 12, pp. 1370–1379, Apr. 2020, doi: <https://doi.org/10.1111/jcpp.13226>.
- [17] J. Downs *et al.*, "Assessing machine learning for fair prediction of ADHD in school pupils using a retrospective cohort study of linked education and healthcare data," *BMJ Open*, vol. 12, no. 12, Dec. 2022, doi: <https://doi.org/10.1136/bmjopen-2021-058058>.
- [18] Y. Wei, "An Overview of Predicting the Prevalence of ADHD," Jan. 2022, doi: <https://doi.org/10.2991/assehr.k.220704.036>.

THANK YOU
