PROJECT

Project Title:

HematoVision: Advanced Blood Cell Classification Using

Transfer Learning

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Team Members:

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Phase-1: Brainstorming & Ideation

Objective:

To develop an Al-powered blood cell classification system using transfer learning that enhances diagnostic accuracy, reduces manual analysis time, and supports healthcare applications such as automated diagnostics, remote consultations, and medical education.

Key Points:

1. Problem Statement:

- Microscopic analysis of blood cells plays a critical role in diagnosing conditions such as leukemia, anemia, and infections. However, manual analysis is time-consuming and subject to human error.
- HematoVision utilizes transfer learning to classify blood cell types from microscopic images accurately. By integrating Al into hematovision system enhances diagnostic speed, precision, and reliability, supporting early disease detection and improves patient outcomes.

2. Proposed Solution:

- AnAl-powered application that uses transfer learning to accurately classify blood cells from microscope images in real-time.
- Ithelps doctors by providing quick, reliable results, reducing manual effort and improving the speed and accuracy of diagnosis.

3. TargetUsers:

- Pathologists and lab technicians who need quick and accurate blood cell analysis.
- Hospitals and diagnostic centers aiming to automate blood smear classification.
- Doctors offering remote consultations who require fast diagnostic support.
- Medical students and trainees learning about blood cell identification.

4. Expected Outcome:

- A functional Al-powered application that accurately classifies blood cells from images using transfer learning.
- Provides fast, reliable diagnostic support to improve efficiency in medical analysis and education.

Phase-2: Requirement Analysis

Objective:

Define the technical and functional requirements for the HematoVision application.

Key Points:

1. Technical Requirements:

- ProgrammingLanguage:Python
- Python Packages:
 NumPy, Pandas, Scikit-learn, Matplotlib, SciPy,

Seaborn, TensorFlow, Flask

- Frameworks: Flask for web integration, TensorFlow for deep learning
- Pre-trained Model: VGG16 (used for transfer learning)
- DevelopmentTools: Command Line (pip install)

2. Functional Requirements:

- Ability to upload microscopic blood cell images through the web interface.
- Classify blood cells into types like eosinophils, lymphocytes, monocytes, and neutrophils using a trained model.
- Display classification results along with prediction confidence.
- Provide an easy-to-use interface for doctors, students, and lab technicians.

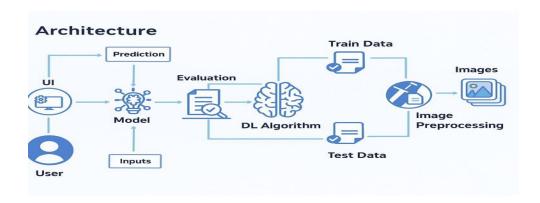
3. Constraints & Challenges:

- Handling imbalanced data across different blood cell classes.
- Managing low-quality or blurry microscope images that affect prediction accuracy.
- Optimizing model size and performance for faster processing and easy deployment.
- Ensuring the web interface is responsive and user-friendly on all devices.

Phase-3: Project Design

Objective:

Develop the architecture and user flow of the application.



Key Points:

1. System Architecture:

- The user uploads a blood cell image through the web interface (UI).
- The image is sent as input to the trained deep learning model.
- The model uses a pre-trained CNN (like VGG16) to process the image and predict the blood cell type.
- The prediction is evaluated and displayed back on the UI with confidence levels.
- Initially, the model is built using preprocessed images from the dataset, split into training and testing sets.
- The training data is used to fine-tune the DL algorithm using transfer learning, while test data is used for evaluation.

2. User Flow:

Step 1: User opens the web application.
 Step 2: Uploads a microscope image of a blood smear.
 Step 3: The model processes the image and classifies the cell type.
 Step 4: The predicted cell type and confidence score are displayed.
 Step 5:User can repeat with other images or use the results for diagnostic

insight.

Phase-4: Project Planning

Objective:

Breakdown development tasks for efficient completion.

Sprint	Task	Priority	Duration	Deadline	Assigned to	Dependencies	Expected outcome
Sprint 1	Environment Setup & Package Installation	□ Ну	3hours	Day 1	Member 1	Anaconda , Python	Project environment ready
Sprint 1	Dataset Collection & Preprocessing	□₩у	4hours	Day 1	Member 2	Dataset access	Clean,prepar ed image dataset
Sprint 2	Model Building using Transfer Learning Flask Web	□ High	5hours	Day 2	Member 3	Preprocessed data, TensorFlow	Trained classification model
Sprint 2	App Integration	□ Medium	3hours	Day 2	Member 1 & 4	Trained Model,Flask installed	Working web interface
Sprint 3	Testing & Debugging	□ Medium	2 hours	Day 2	Member 2 & 3	Complete System	Bug-free and responsive system
Sprint 3	Final Presentation & Deployment	□ Low	1hour	End of Day 2	Entire Team	Working application	Project deployed and demo-ready

Sprint Planning with Priorities Sprint 1 - Setup & Preparation (Day 1) □

High Priority

- Set up the environment using Anaconda Navigator.
- Install all required Python packages (TensorFlow, Flask, etc.).

Collect and preprocess the blood cell image dataset.

Sprint 2 – Model Development & Integration (Day 2)

High Priority

- Build and train the blood cell classification model using transfer learning (e.g., VGG16).
- Integrate the trained model with the Flask web application.

Sprint 3 – Testing, Deployment & Submission (Day 2)

Medium Priority

• Test the app functionality, fix bugs, and improve UI responsiveness.

☐ Low Priority

• Finalize deployment and prepare presentation/demo materials.

Phase-5: Project Development

Objective:

Implement the core features of the HematoVision application using transfer learning for blood cell classification.

Key Points:

1. Technology Stack Used:

• Frontend: HTML (via Flask templates)

Backend: Flask Framework

• **Deep Learning:** TensorFlow with pre-trained VGG16 model

Programming Language: Python

2. Development Process:

- Built and trained a blood cell classification model using transfer learning (VGG16).
- Preprocessed a dataset of 12,000 labeled blood cell images.
- Integrated the trained model into a **Flask web application**.
- Developed a user interface to upload images and display classification results with prediction confidence.

3. Challenges & Fixes:

- Challenge: Model overfitting on certain blood cell types.
 Fix: Used data augmentation and dropout regularization.
- Challenge: Large model size caused slow predictions.
 Fix: Applied model optimization and saved in.h5 format for faster loading.
- Challenge: Image quality variation.
 Fix: Added preprocessing steps like resizing and normalization before prediction.

Phase-6: Functional & Performance

Objective:

Ensure that the HematoVision application performs accurately, reliably, and consistently across various test cases and environments.

Test Case ID	Category	Test Scenario Upload	Expected Outcome Correct cell	Status	Tester
10-001	Functional Testing	image of eosinophils	type identified with confidence score		Tester 1
TC-002	Functional Testing	Upload mixed N WBC/RBC image	fulti-class classification displayed correctly	∀ Passed	Tester 2
TC-003	Performance Testing	Check model R response time	displayed under 2 secods	⚠ Needs Optimization	Tester 3
TC-004	Bug Fix Validation	Image with poor lighting	System still makes reasonable prediction Layout	∀ Fixed	Developer
TC-005	UI Responsiveness	Test on mobile browser	adjusts properly on mobile App loads	X Failed	Tester 2
TC-006	Deployment Testing	Hosted on local server and accessed remotely	and predicts successfully online	□ Deployed	DevOps