

Animal Aware - Smart Trail Camera

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MENTOR: Dr. David Coe

CPE 495: COMPUTER ENGINEERING DESIGN I
CPE 488: CYBERSECURITY ENGINEERING CAPSTONE I

ELECTRICAL AND COMPUTER ENGINEERING
UNIVERSITY OF ALABAMA IN HUNTSVILLE

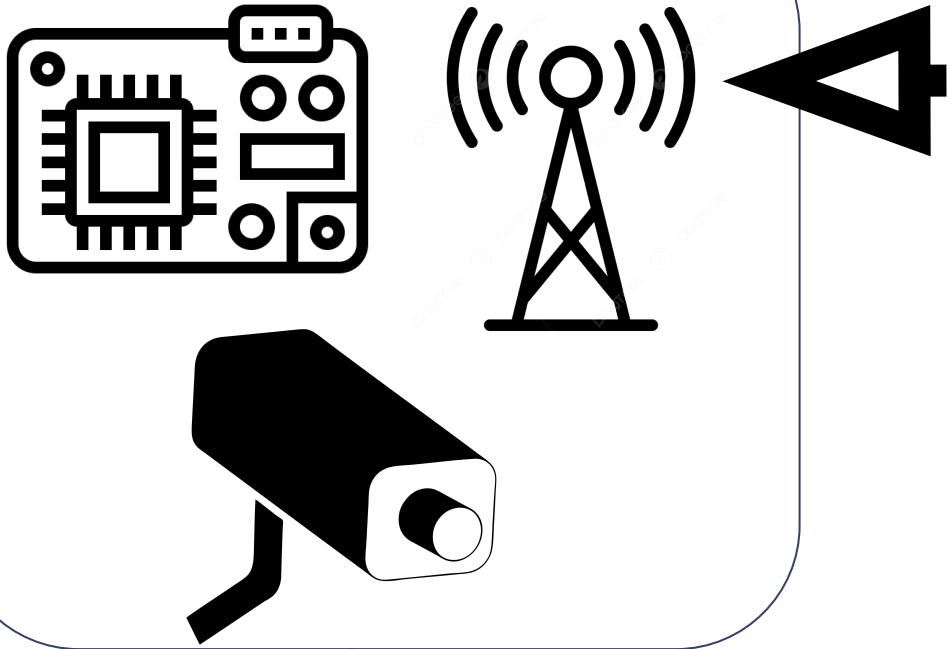
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Smart Monitoring and Detection

- Common users of Trail Cams are: Campers, Wildlife Monitors, Hunters, Park Rangers
- The Smart Trail Camera (STC) provides benefits to monitoring, protection, and analysis
- Employs new novel techniques for high classification accuracy and response time
- Utilizes inexpensive technologies lowering cost
- Allows for more personalized use of trail camera software

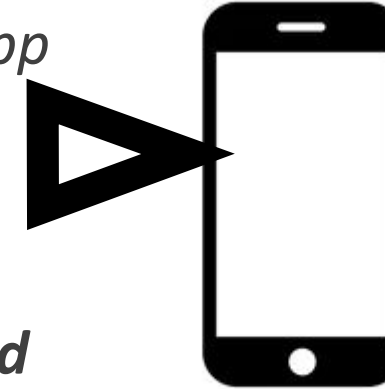
Smart Trail Camera: Animal Image Classification

STC Components



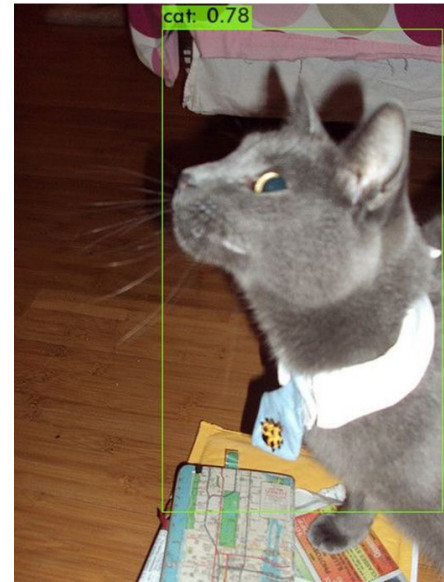
Alert Transmitted to App

■ ■ ■ ■ ■
Settings for STC



*App
Interface*

*Animal Detected
by STC*



Marketing Requirements

| Marketing Requirements | |
|------------------------|--|
| M1 | Identification of a variety of common animals (Where users can select phone notifications according to detected animal.) |
| M2 | App customization that supports at least four smart trail cameras that can generate alerts simultaneously |
| M3 | Tamper-proof app interface |
| M4 | Motion activated so that the cameras can maintain a good battery life. |
| M5 | Works even without direct network access |
| M6 | Supports multiple connections from cameras with the app. |
| M7 | Alert source must be viewable in the application |

Engineering Requirements

Engineering Requirements

| | |
|------------|--|
| ER1 | Animal classification with 50% accuracy against separate specifically chosen testing data. |
| ER2 | Animal classification in wild with 25% accuracy in environments with proper setup, and lighting. |
| ER3 | Secure and encrypted multi-camera connection and configuration within app interface. |
| ER4 | 30 Day battery life under average daily load (4% active SoC computation in a day). |
| ER5 | Activation of SoC within 0.8 seconds of Motion Detection. |
| ER6 | Connectivity with up to a 25-meter (~90 feet) range. |
| ER7 | Simultaneous multi-camera connections without data loss. |
| ER8 | Real-time alerts to the Android device indicating the node information within 6 seconds. |

Survey: Market & Competition

Three key aspects the STC aims to achieve is: performance, price, and functionality.

WiseEye MiniCam [<https://wiseeyetech.com/product/wiseeye-mini-cam/>]

- Automatic Species Recognition
- 32GB SD Card for Storage
- Long Range Cellular Antenna
- 65 ft. IR Flash
- \$199.99 / \$139.99(on-sale)

Analysis - Very Expensive, Unnecessary Features add to cost



Tactacam REVEAL Pro 3.0 [<https://www.amazon.com/TACTACAM-Cellular-Required-RV-TC-XPRO3-Surveillance/dp/B0D7HZKFCQ>]

- 4K UHD Photo
- LTE Cellular Network
- \$154.99

Analysis - Very Expensive, No Image Classification, Requires Mobile Cell Carrier



Survey: Existing Projects

- WiseEye [1], a trail camera company, was looking to develop a user-customizable open source platform that would allow more flexible camera traps for wildlife researchers.
- A user named Trail Camera Training [2] was developing an animal detection system using trail cameras.
- A study [3] by the Department of Forest Resources Management in Canada looked at object detection in camera trap object identification. Can you put authors &/or paper title
- A study titled “Innovations in Camera Trapping Technology and Approaches: The Integration of Citizen Science and Artificial Intelligence” [4] looked into the integration of artificial intelligence into trail cameras to create neural networks to identify different animal species.

1. <https://journals.plos.org/plosone/article?id=10.1371/journal.pone.0169758>
2. <https://universe.roboflow.com/trail-camera-training/trailcam-detection>
3. <https://www.sciencedirect.com/science/article/pii/S2351989422001068>
4. <https://www.ncbi.nlm.nih.gov/pmc/articles/PMC7023201/>

Survey: Existing Patents and Projects

- There was a paper in 2017 where the authors were looking for a way to automate the processing the large amount of data the “camera traps” produced. They aimed to create a neural network where their experimental results produced 96.6% and real world results of 90.4%. [Animal Recognition](#) [1]
- Patent: [#9,626,580](#) [2] Defining region for motion detection
- Patent: [US11537891B2](#) [3] Intelligent recognition and alert methods and systems

1. <https://ieeexplore.ieee.org/abstract/document/8259762>
2. <https://patents.justia.com/patent/9626580>
3. <https://patents.google.com/patent/US11537891B2/>

Possible Approaches

1. Perform inference on edge device, then send results mothernode which is used for interop to phone.
2. Gather video/images from edge device and stream to mothernode to perform compute.
3. Gather video from edge device and send to mothernode for interop to phone to perform compute.

Favorable - Perform inference on edge device, then send results mothernode which is used for interop to phone.

Possible Approaches (AI/ML)

- Pugh Chart over where to perform model inference/analysis.
 - Favorable is Inference on Mothernode

| | | Inference on Edge Node | Inference on Mothernode | Inference on Phone |
|------------------------------|----------|------------------------|-------------------------|--------------------|
| Cost | 3 | 0 | -1 | +1 |
| Reliability | 4 | 0 | +1 | 0 |
| Experience | 2 | +1 | +1 | -1 |
| Easy to build/develop | 1 | -1 | +1 | 0 |
| Score | | 1 | 4 | 1 |

Possible Approaches (Network Connectivity)

- Pugh Chart over what form the network connectivity should take.
 - Favorable is Radio

| | | Radio | LoRa | ZigBee |
|-------------|---|-------|------|--------|
| Cost | 4 | 0 | +1 | -1 |
| Reliability | 3 | +1 | 0 | 0 |
| Distance | 3 | +1 | +1 | 0 |
| Bandwidth | 6 | +1 | -1 | 0 |
| Score | | 12 | 1 | 4 |

Risk Analysis

- Project components and associated risks:

| Component | Potential Risks |
|--------------------|--|
| Camera module | poor image/video quality, power drain, lens malfunction |
| LED | burning out, harder to make covert |
| Mobile application | application bugs or crashes, security vulnerabilities, compatibility issues |
| Motion Sensor | false positives, false negatives, failure to detect motion, environmental impact |

Risk Analysis

- Project components and associated risks:

| Component | Potential Risks |
|-------------------------|--|
| Android Phone | incompatibility with application, battery drain |
| Microcontroller | firmware bugs, overheating, limited processing power |
| Network Connectivity | loss of signal, security breaches, high cost |
| Artificial Intelligence | false positives/false negatives, accuracy, dataset poisoning |

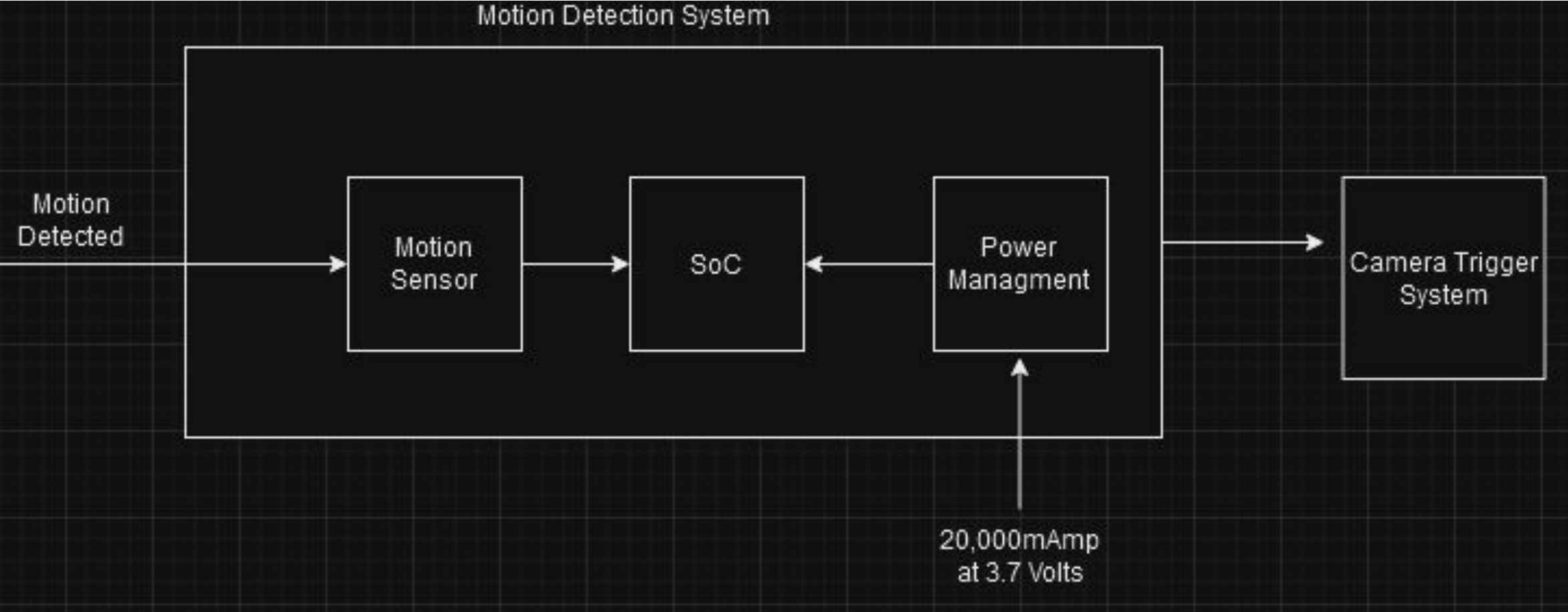
Project Summary

- Targeted Impact:
 - Aims to provide the market with new and innovative animal aware smart trail camera
 - Implementation of novel AI classification techniques on low compute edge devices
 - Reduction in price due to low compute environment
- Significance:
 - Automating Animal Identification
 - Customizable Alerts
 - Real-time Access
- Degree of innovation:
 - AI/Machine Learning integration
 - Expansion on existing trail cameras in terms of price and function
 - Multi-use functionality, including hunting, wildlife conservation and research, and property security.

Functional Decomposition:

| | |
|---------------|---|
| Module | Motion Detector |
| Inputs | Sensor detects motion inside the environment |
| Outputs | Signal interrupts SoC waking camera from sleep Trigger camera to start recording |
| Functionality | Detect motion to activate the camera for capturing images or video. |

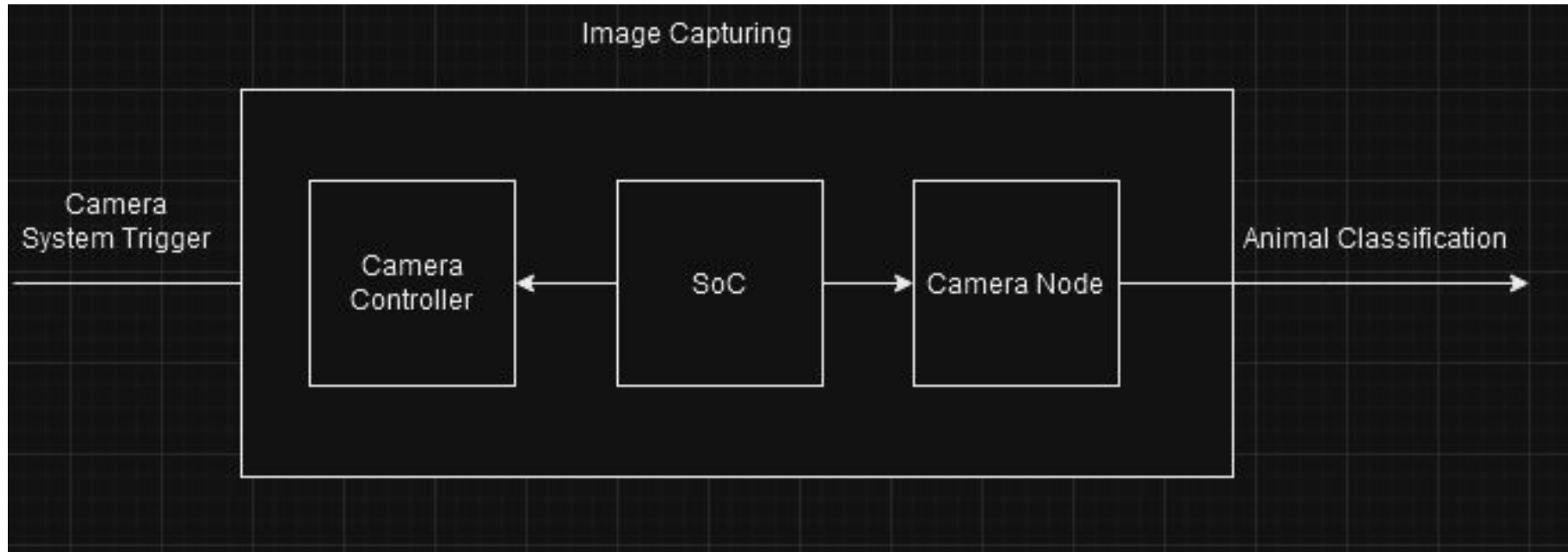
Functional Decomposition



Functional Decomposition:

| | |
|---------------|---|
| Module | Image/Video Capture Device |
| Inputs | Trigger SoC to initiate recording |
| Outputs | Unprocessed image or video data |
| Functionality | Capture images or video upon motion detection |

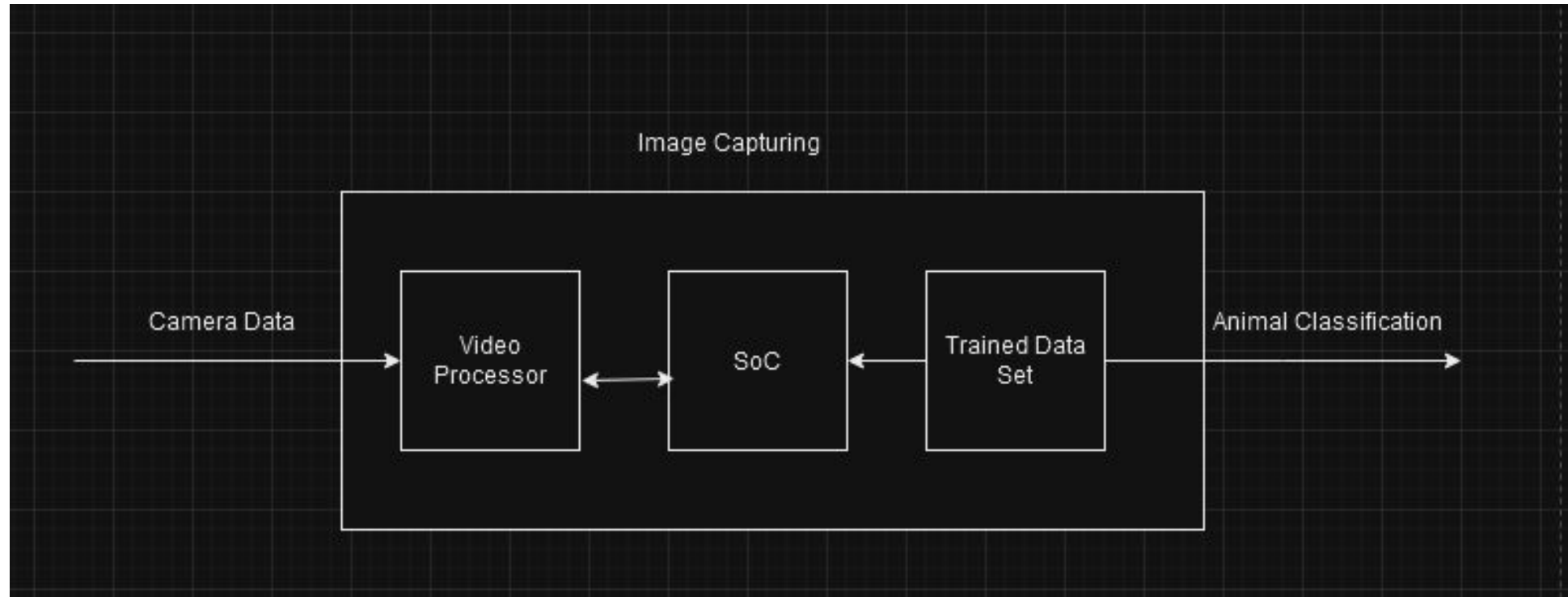
Functional Decomposition:



Function Decomposition

| | |
|---------------|--|
| Module | Image Classifier |
| Inputs | Raw image or video data from the camera sensor. |
| Outputs | Bounding box confidence score for each assessed classification. (i.e. <Bird, 90%>) |
| Functionality | Capture images or video upon motion detection |

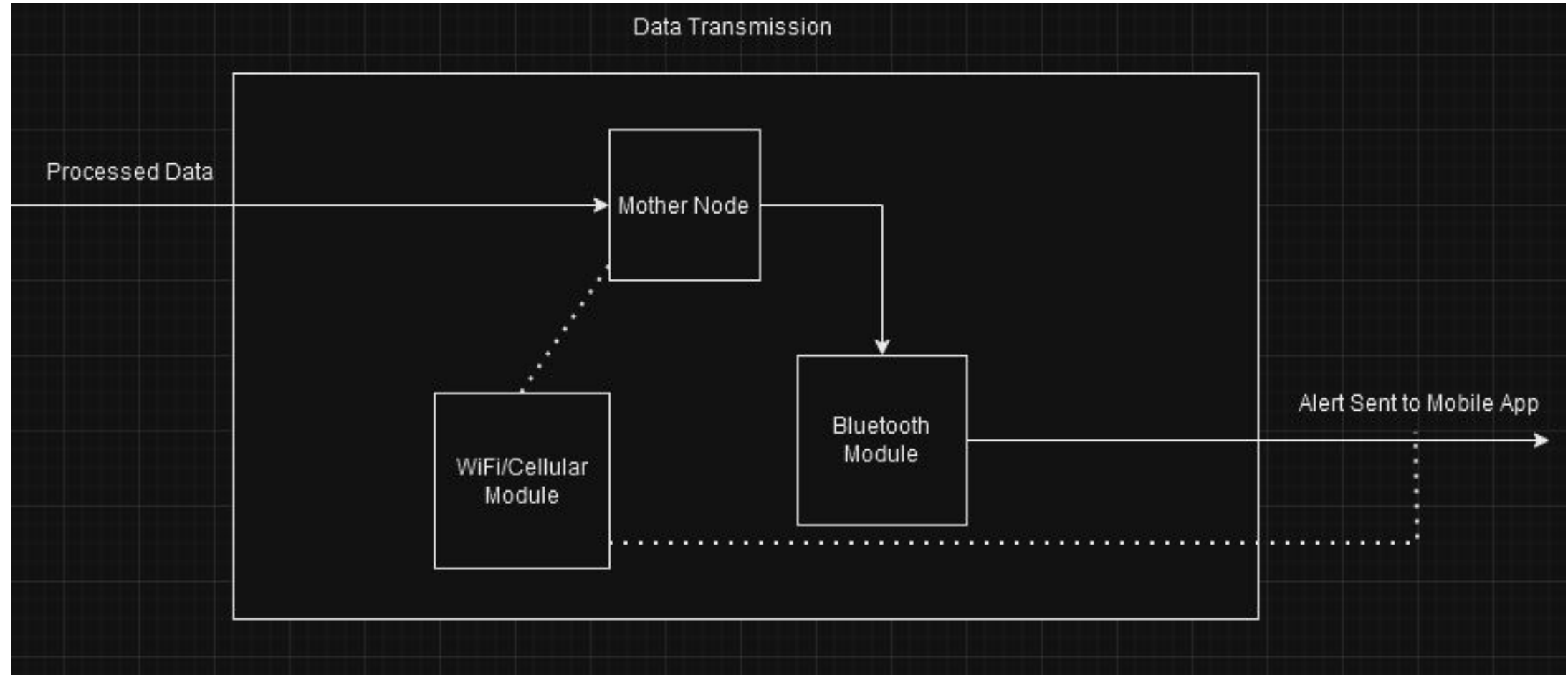
Functional Decomposition:



Functional Decomposition:

| | |
|---------------|---|
| Module | Mothernode - Data Transmission and Communication |
| Inputs | Radio Transmission of classification results and captured data from the camera. |
| Outputs | Alerts and data sent to the mobile app. |
| Functionality | Transmit alerts, video, and classification results to the mobile app via Bluetooth or WiFi Radio. |

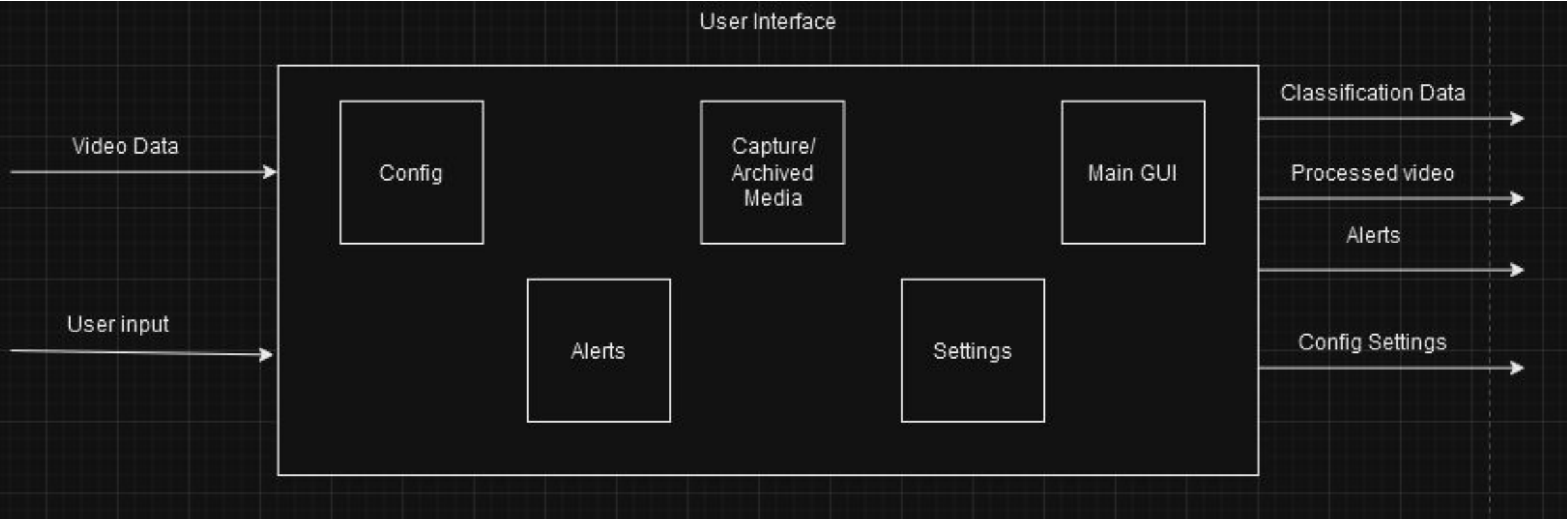
Functional Decomposition:



Functional Decomposition:

| | |
|---------------|---|
| Module | User Interface |
| Inputs | <p>Network - Data transmitted from the camera (via Bluetooth or WiFi).</p> <p>GUI - User configuration and camera settings.</p> |
| Outputs | <p>Display captured data, classification results, and alerts.</p> <p>Confirmation of updated configuration settings from received node to the camera.</p> |
| Functionality | Allow users to configure camera settings, view alerts, and review captured data. |

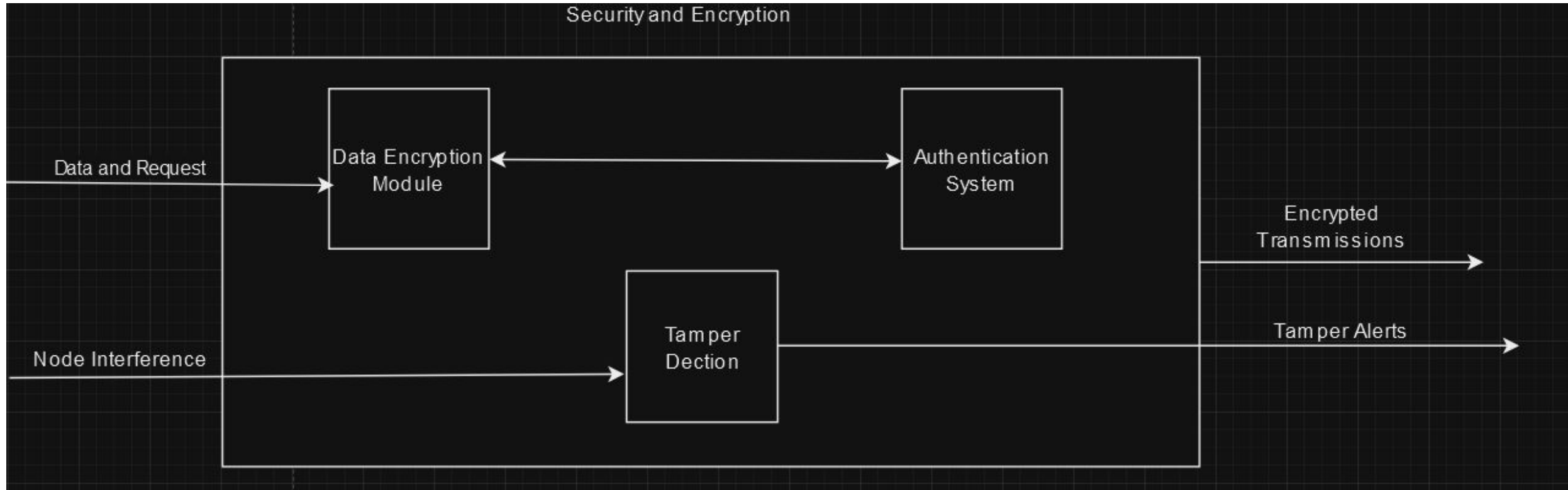
Functional Decomposition:



Functional Decomposition:

| | |
|---------------|--|
| Module | Security and Data Encryption |
| Inputs | Video Data User Request |
| Outputs | Encrypted data transmissions to the mobile app. Tamper alerts if unauthorized access is detected. |
| Functionality | Ensure secure communication between the camera and the mobile app to prevent unauthorized access |

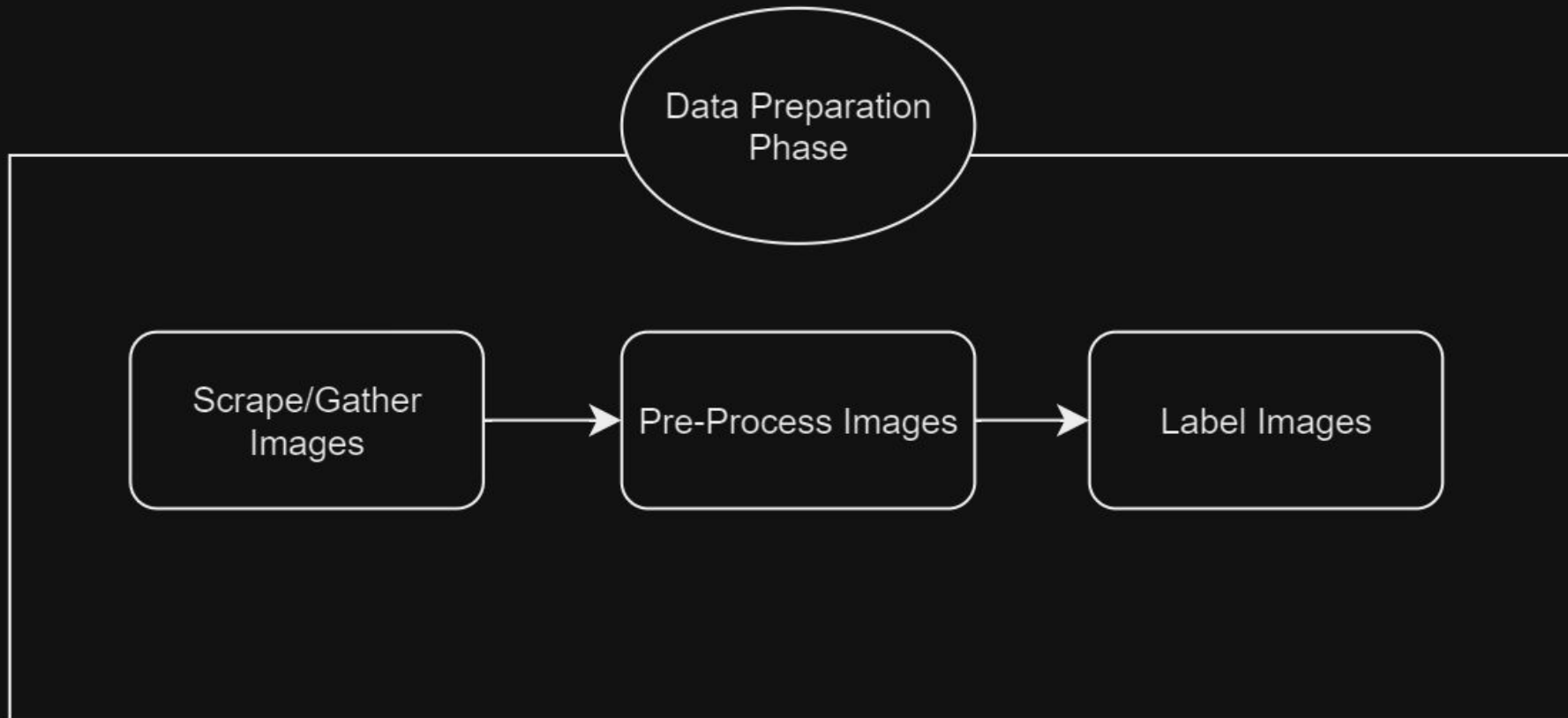
Functional Decomposition:



Functional Decomposition:

| | |
|---------------|--|
| Module | Data Preparation |
| Inputs | Video or Images |
| Outputs | Labeled and Preprocessed Images to fit into Database |
| Functionality | Prepares Images for utilization in database/model training |

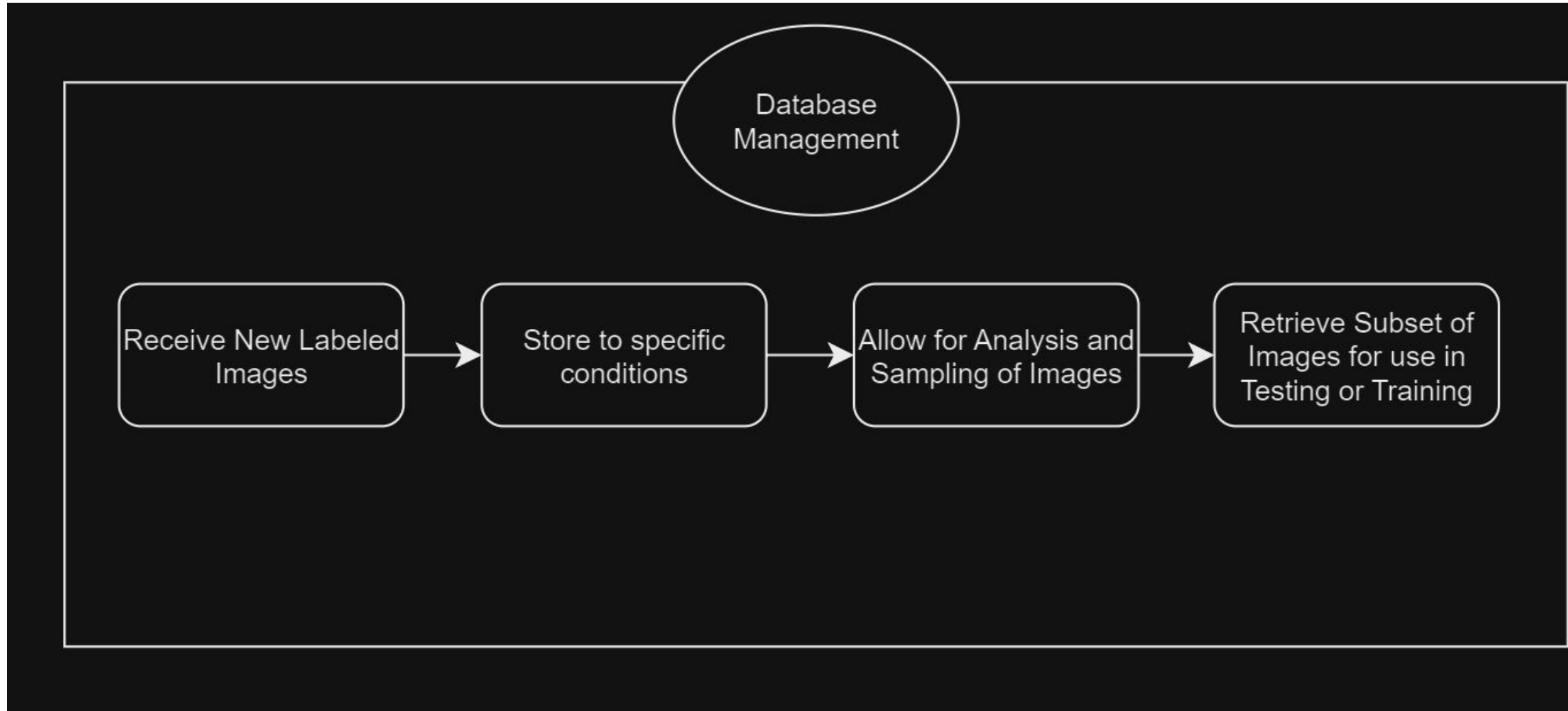
Functional Decomposition:



Functional Decomposition:

| | |
|---------------|---|
| Module | Database Management |
| Inputs | Labeled Images |
| Outputs | Subsets of Labeled Images and Configuration files based on their form. |
| Functionality | Allow easy access and management of gathered labeled images for utilization in model training and analysis. |

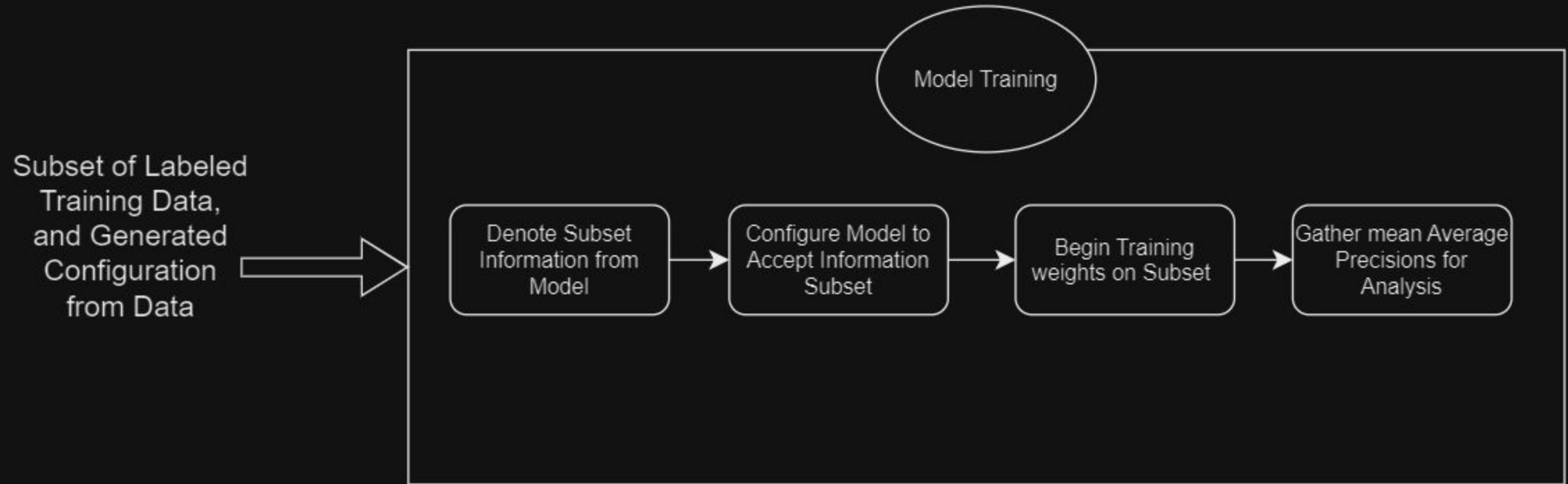
Functional Decomposition:



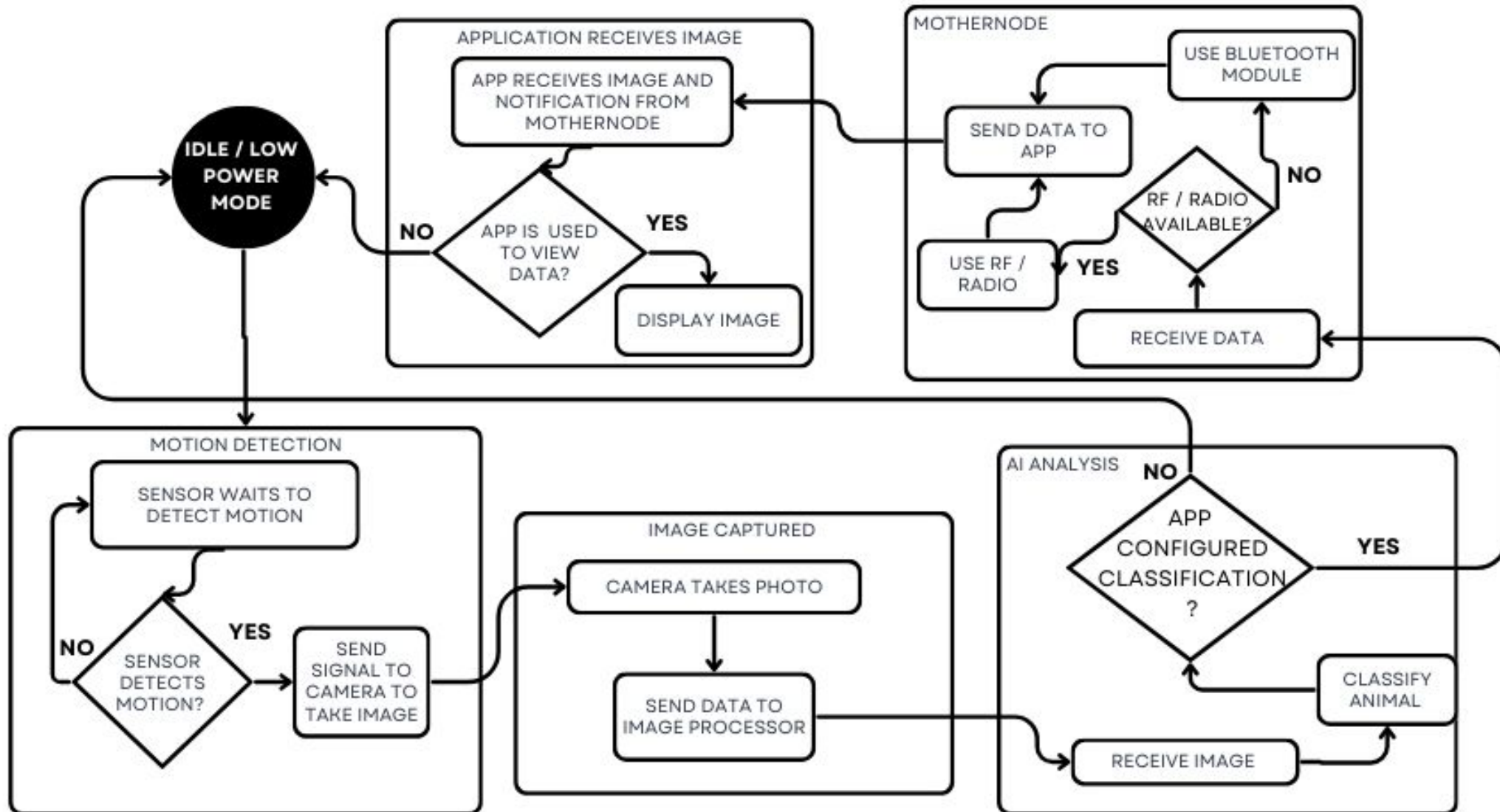
Functional Decomposition:

| | |
|---------------|--|
| Module | Model Training |
| Inputs | Labeled Image Set and Configuration File Prior weights if applicable. |
| Outputs | Generated Weights, and Mean Average Precision (mAP) |
| Functionality | Weights are able to be loaded to edge devices for inference, and mAP provides information about performance of model. |

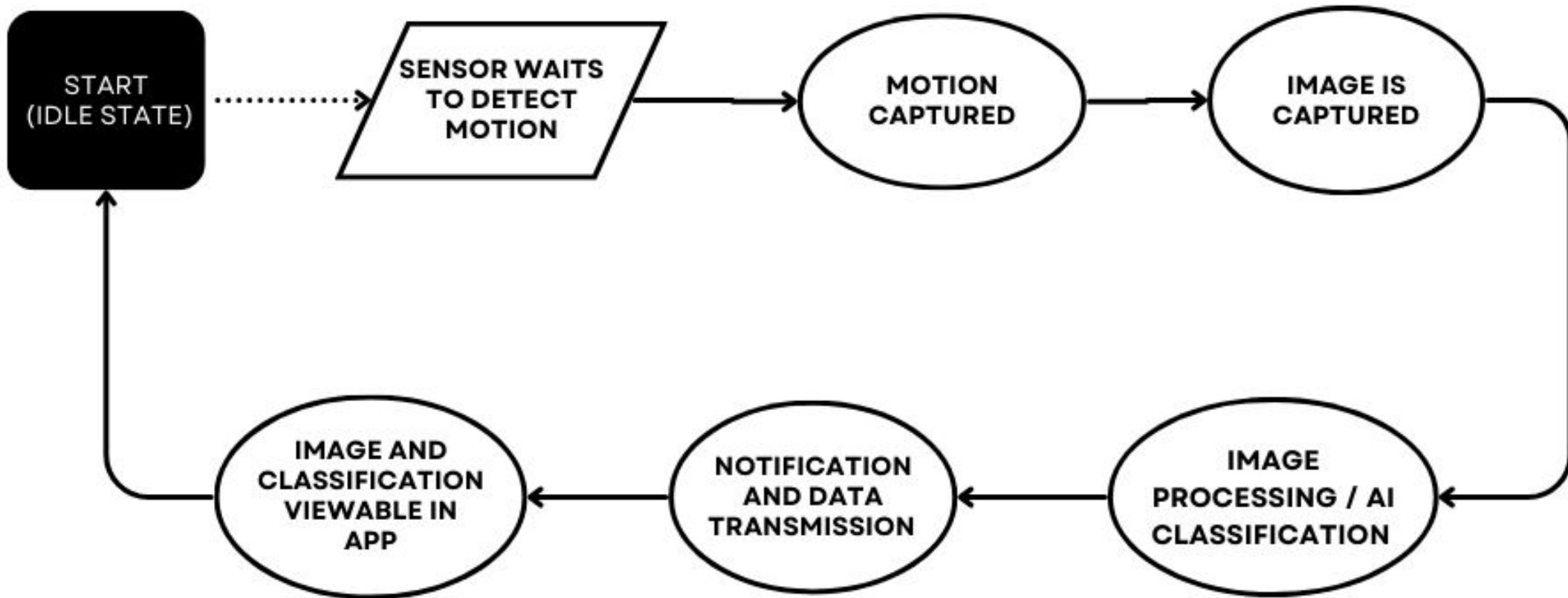
Functional Decomposition:



Behavioral Decomposition - State Machine



Behavioral Decomposition - Flowchart



Testing Plan - Unit Testing

Motion Detection:

- **Objective:** Ensure that the camera activates within 0.8 seconds when motion is detected.
- **Test:** Simulate motion in front of the camera and measure activation time.
- **Expected Result:** Camera triggers within 0.8 seconds.

Animal Detection & Classification (ER1):

- **Objective:** Verify that the classifier accurately detects animals in images.
- **Test:** Feed test images of various animals (including false positives like inanimate objects) to the model and check classification, and classification accuracy.
- **Expected Result:** At least 50% accuracy across all animal types.

Battery Consumption (ER3):

- **Objective:** Ensure the camera can last for 30 days with average motion detection/triggers.
- **Test:** Simulate an average day of activity and monitor battery life then extrapolate.
- **Expected Result:** Battery lasts at least 30 days in low-power mode when idle and motion-activated when needed.

Testing Plan - Unit Testing cont.

AI/ML Classification Analysis:

- **Objective:** Ensure that during the training process the AI/ML approach is able to classify various selected testing images.
- **Test:** Provide AI/ML with a set of images that the model has not seen in training before.
- **Expected Result:** AI/ML to classify the animals with the correct label.

AI/ML Realtime Computation:

- **Objective:** Ensure that during a real scenario, the AI/ML is able to produce outputs quick enough to be running at near real time speed.
- **Test:** Run AI/ML on expected hardware performing inference on various images and videos.
- **Expected Result:** AI/ML is able to produce classification output in near-realtime.

Testing Plan - Integration

Motion Detection with Animal Classification:

- **Objective:** Test that when motion is detected, the camera captures the footage, processes the video/image, and identifies animals correctly.
- **Test:** Trigger motion and verify that the camera captures video/images and processes the classification in real-time.
- **Expected Result:** Motion activated camera, video is recorded, and classification is performed successfully.

Multiple Cameras/Nodes (ER5):

- **Objective:** Ensure that the app can handle simultaneous connections from at least four cameras.
- **Test:** Connect two-four nodes and trigger events from each. Verify that the app processes all incoming data.
- **Expected Result:** App correctly receives alerts from all cameras with minimal data loss or latency.

Battery Life with Motion Activation (ER3):

- **Objective:** Test that the motion detection and real-time processing don't drain the battery prematurely.
- **Test:** Simulate field-like conditions over several days and measure battery consumption under normal use.
- **Expected Result:** The camera maintains sufficient battery for at least 30 days.

Testing Plan - Acceptance

Full System Test in Field Conditions (ER1, ER3, ER4):

- **Objective:** Verify the entire system in field conditions, simulating real-world scenarios
- **Test:** Deploy cameras in the field for a 2-week period, simulating typical field conditions. Test app configurations, real-time alerts, battery life, and offline functionality.
- **Expected Result:** Cameras function correctly, providing accurate alerts, battery lasts through the test, and the system works without internet access.

Tamper-Proof App Interface (ER2):

- **Objective:** Ensure that the mobile app configuration settings are secure, and multiple cameras can be configured simultaneously without unauthorized access.
- **Test:** Attempt to tamper with app settings while verifying encryption and secure communication. Test camera configuration for all four cameras simultaneously.
- **Expected Result:** App settings are protected from tampering, and all cameras can be controlled without conflict or data loss.

Multi-Camera Management (ER5):

- **Objective:** Ensure that the app can manage and display alerts from multiple cameras at once.
- **Test:** Set up four cameras and verify that the app properly displays real-time alerts and direction of the source camera.
- **Expected Result:** The app correctly manages alerts from four cameras and displays the source of the alert.

Testing Plan - Failure Plan

Unit Test Plan: If a unit test fails, immediately create a backlog of the bug. Assign priorities based on importance (e.g., battery life would be high priority). Slack time will be used for debugging and fixing issues. Re-run failed tests after fixes.

Integration Plan: If integration tests fail, assign bugs to the appropriate teams (e.g., hardware or software) and prioritize fixes. Use slack time to retest the integrated components.

Acceptance Plan: If any part of the acceptance test fails, re-assess the system requirements. Minor bugs will be logged and fixed in an additional sprint, but critical issues will trigger a full system review. Slack time is allotted for adjustments and retesting.

Gantt Chart: Fall Semester

Fall Semester

Smart Trail Cam

Legend:

On track

Low risk

Med risk

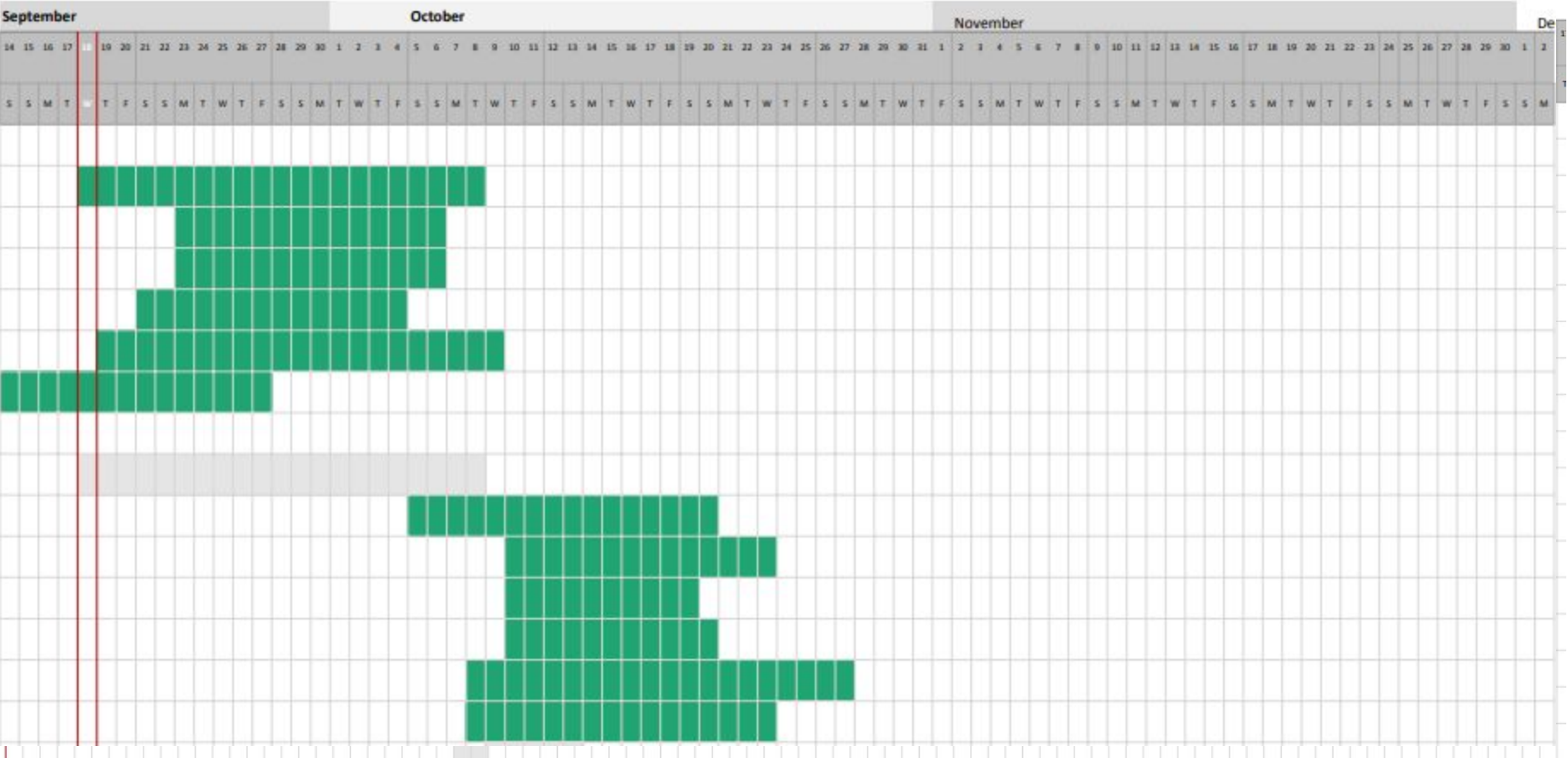
High risk

Unassigned

Group 10

Project start date: 9/14/2024

| Milestone description | Category | Assigned to | Progress | Start | Days |
|---|----------|---------------|----------|------------|------|
| Initial Research | | | | | |
| Research Machine Learning/AI Approaches | On Track | Rio, Alyson | 25% | 9/18/2024 | 21 |
| Research Necessary Security Components | On Track | All | 0% | 9/23/2024 | 14 |
| Research Alert Systems for prototyping | On Track | Bobby, Tate | 0% | 9/23/2024 | 14 |
| Research Mobile Application Development (Android) | On Track | Bobby, Alyson | 0% | 9/21/2024 | 14 |
| Research Trail Cameras & Hardware | On Track | Rio, Tate | 10% | 9/19/2024 | 21 |
| Research Network Connectivity | On Track | Rio, Tate | 15% | 9/14/2024 | 14 |
| Design Phase | | | | | |
| Milestone Briefing (Oct. 9) | | All | 0% | 9/18/2024 | 21 |
| Mobile Application Design | On Track | Bobby, Alyson | 0% | 10/5/2024 | 16 |
| Training Dataset Collection | On Track | Rio, Alyson | 0% | 10/10/2024 | 14 |
| Dataset labeling | On Track | All | 0% | 10/10/2024 | 10 |
| Database Design | On Track | Bobby, Alyson | 0% | 10/10/2024 | 11 |
| Hardware & Prototype Design | On Track | Rio, Tate | 0% | 10/8/2024 | 20 |
| Design Prototyping Network Connectivity | On Track | Rio, Tate | 0% | 10/8/2024 | 16 |



Legend:

On track

Low risk

Med risk

High risk

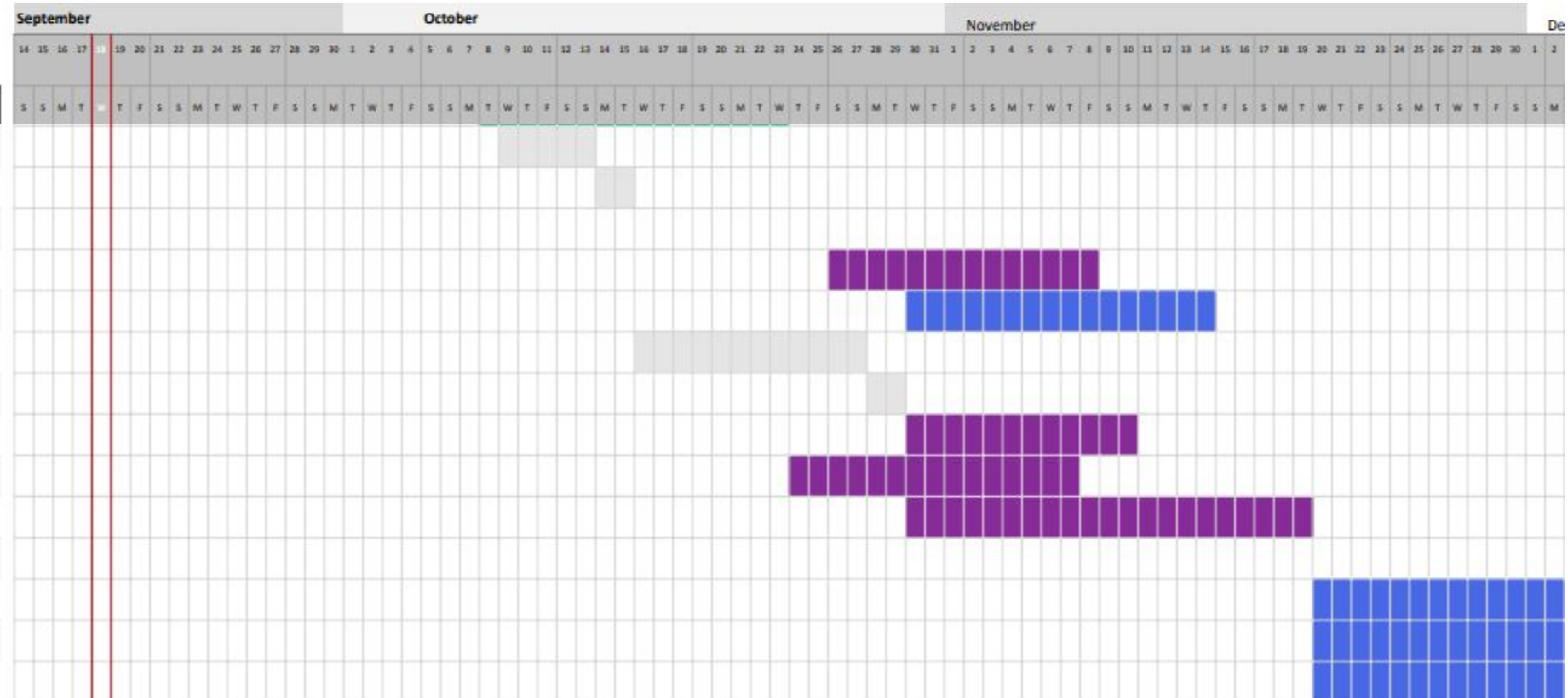
Unassigned

Group 10

Project start date:

9/14/2024

| Milestone Description | Category | Assigned to | Progress | Start | Days |
|---|-----------|---------------|----------|------------|------|
| Milestone Briefing (Oct. 14) | | | | 10/9/2024 | 5 |
| Milestone Briefing (Oct. 14) | | All | | 10/14/2024 | 2 |
| Development and Integration Phase of Prototypes | | | | | |
| Mobile Application Development | High Risk | Bobby, Alyson | 0% | 10/26/2024 | 14 |
| Alert System Prototype Development | Med Risk | Bobby, Tate | 0% | 10/30/2024 | 16 |
| Milestone Briefing (Oct. 28) | | All | 0% | 10/16/2024 | 12 |
| Milestone Briefing (Oct. 30) | | All | 0% | 10/28/2024 | 2 |
| Training Model on Labeled Dataset | High Risk | Rio, Alyson | 0% | 10/30/2024 | 12 |
| Network Connectivity Development | High Risk | Rio | 0% | 10/24/2024 | 15 |
| Hardware Development | High Risk | Rio, Tate | 0% | 10/30/2024 | 21 |
| Testing and Validation Phase of Prototype | | | | | |
| Motion Sensor Unit Testing | Med Risk | Bobby | 0% | 11/20/2024 | 15 |
| Animal Detection & Classification | Med Risk | Rio, Alyson | 0% | 11/20/2024 | 14 |
| Closed Internal System Test | Med Risk | All | 0% | 11/20/2024 | 15 |



Gantt Chart: Spring Semester

- Milestones:
 - Complete deliverables
 - Two or more fully functional nodes
 - Android phone/tablet with function application
 - Gitlab repository with all artifacts
- Focus:
 - Spend initial portion of the semester doing any additional research required
 - Continue design and development of hardware
 - Focus on full cybersecurity model and anti-tampering portions
 - Complete full-field testing with all components
 - Complete documentation and creation of artifacts for GitHub Repository.

Individual Responsibility

- Responsibilities:
 - Mobile Application, Database, Cybersecurity - Bobby Calhoun
 - Hardware, Network Connectivity, Cybersecurity - Tate Forcella
 - Machine Learning/Dataset Training, Hardware, Cybersecurity - Rio Laney
 - Mobile Application, Machine Learning, Cybersecurity - Alyson Williams
- Contingency Plan:
 - Loss of a team member will result in a redistribution of workload where the remaining team members will split up the missing member's tasks. We will have at least two people on each task to ensure that at least one other member is familiar with the content.
 - If an event occurs where a lockdown is put into place, we will move into a fully remote work environment. For physical components, a meeting place will need to be decided or simulations will be conducted in their place.
 - In the event of weather conditions, our group will meet remotely and discuss how to proceed once conditions have lessened.

Parts - Cost Estimation cont.

Compute Module

- 29\$ - Grove Vision AI v2 Kit [1]

Radio/Network Connectivity

- 11\$ 3x Wireless Transceiver Module, RF Transceiver Module, 2.4G 1100m [2]

Motion Sensor

- \$24.95 - JemRF Wireless Motion sensor kit [3]

Estimated price for a development of a single node is ~150\$

Additional component is an Android Phone/Tablet. A recent android phone is:

Samsung Galaxy A03s Cell Phone - 70\$

1. <https://www.seeedstudio.com/Grove-Vision-AI-V2-Kit-p-5852.html>
2. <https://www.amazon.com/Wireless-Transceiver-NRF24L01-Antenna-Raspberry/dp/B07QRZHZGY>
3. <https://www.jemrf.com/collections/rf-sensors/products/wireless-motion-sensor-kit>
4. <https://www.amazon.com/Samsung-Unlocked-Smartphone-Expandable-Infinite/dp/B09SM24S8C>

Parts - Cost Estimation

Camera

- 15-25\$ - Raspberry Pi NoIR Camera Module V2 [1]
- 15\$ - Bright Pi [2]

LED

- 3.5\$ (7\$) - 2x Infrared LED Light Board Module [3]

Battery

- 20\$ - Portable Charger Power Bank 30000mAh [4]

SoC

- 55\$ - Raspberry Pi 4 Model B 4GB / Board Only [5]

1. <https://www.pishop.us/product/raspberry-pi-noir-camera-module-v2/?src=raspberrypi>
2. <https://www.pishop.us/product/bright-pi-bright-white-and-ir-camera-light-for-raspberry-pi/>
3. <https://www.amazon.com/Infrared-Illuminator-Adjustable-Resistor-Raspberry/dp/B07FM6LL3V>
4. <https://www.amazon.com/Portable-Charger-Power-Bank-30000mAh/dp/B0CDLMKN35>
5. <https://vilros.com/products/raspberry-pi-4-model-b-1?src=raspberrypi>

Cost Estimation

- We expect the University of Alabama in Huntsville to pay for the necessary materials needed to complete this project.
- Fixed:
 - Camera, Motion sensor, battery, microcontroller, android device, transceiver module
- Variable:
 - No costs will vary throughout the project duration
- Labor:
 - 40-60 hours a week, 10-15 per group member

Resources and Deliverables

- Resources
 - Camera
 - Motion sensor
 - Network Transceiver
 - Battery
 - SoC/Microcontroller
 - Android Device
- Report
 - Field Test Results
 - Gitlab Artifact
 - Database of Labeled Training Set
 - Classification Accuracy Report for Each Wildlife Category
- Deliverables
 - Hardware
 - Smart Trail Camera
 - Alert System Prototype
 - Software
 - Mobile App Interface
 - Gitlab Repository with artifacts

CPE495/496 Milestones

- CPE 495 Milestones:
 - Background research over:
 - Machine Learning, Hardware, Database creation, Mobile application development
 - Project Proposal
 - Begin Design of hardware, application, network
 - Initial Design Review
 - Development of Application and Prototype
 - Final Report
- CPE 496 Milestones:
 - Continue development of hardware, connections, cybersecurity component
 - Project Status Meeting
 - Conduct testing and documentation for all artifacts
 - Final Design Reviews

The Smart Trail Cam (STC) is an animal aware trail camera that performs classification, analysis, and alerting of detected animals. It incorporates multiple edge nodes to perform monitoring, and a 'mother' node to provide an interop between these edge nodes and an Android application which receives these analysis's and alerts to showcase to the user. Furthermore, the STC Android application allows configuration of edge nodes through selection of critical animal classification alerts, location, and battery life.

Appendix

Engineering Requirements - Full List

| | |
|------------|--|
| ER1 | Animal classification with 50% accuracy against separate specifically chosen testing data. |
| ER2 | Animal classification in wild with 25% accuracy in environments with proper setup, and lighting. |
| ER3 | Secure and encrypted multi-camera connection and configuration within app interface. |
| ER4 | 30 Day battery life under average daily load (4% active computation in a day). |
| ER5 | Activation of SoC within 0.8 seconds of Motion Detection. |
| ER6 | Connectivity with up to a 25-meter (~90 feet) range. |
| ER7 | Simultaneous multi-camera connections without data loss. |
| ER8 | Real-time alerts indicating the camera's location within 6 seconds. |

| Engineering Requirements - Continued | |
|--------------------------------------|---|
| ER9 | Connectivity for up to four trail cameras |
| ER10 | Classification data covers at least 4 animals including but not limited to squirrels, cats, dogs, and birds |
| ER11 | Application will support at least one priority notification which allows the user to get notifications of one specific creature |
| ER12 | Trail camera will function independently of network access and will store up to 20 images while connection is unavailable |
| ER13 | Application will maintain a log of alerts, timestamps, and identified animal |

Questions

- We are now open to any questions