### **Critical Design Review**

**Group 12** 

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# Bird Initiated Rubbish Disposal System



### **MEET THE TEAM**



Aiden Nipper
Photonic Science
and Engineering



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Computer Engineering

### Problem & Motivation



- Littering is still a problem
  - Current solutions to clean up public spaces involve extensive energy costs
    - Volunteer Cleanup
    - Convict Labor
    - Salaried Street Cleaners
- The best solutions are just out of reach
  - Access to and education on proper waste disposal
  - Discouraging single use items

"While humanity struggles to learn basic cleanliness, it falls upon birds to tidy up the mess"

—ChatGPT 3.5

- Birds are:
  - Omnipresent
  - Intelligent enough to form correlations
  - Cheap to bribe



### Goals & Objectives



- Promptly rewarding birds that bring trash
  - For the conditioning machine to have a chance at training birds, the most critical element is quick and decisive activation of the reinforcing stimuli
  - The BIRDS must be able to respond to a trash submission within a short period of time in order for the bird to associate the action of submitting trash with receiving food
- Consistent bird and trash detection verdicts
  - Other animals, such as squirrels, will not be able to interact with the BIRDS
  - Similarly, it would be disastrous for bird conditioning for a valid trash item to not be rewarded or the
    opposite of a non-valid object being rewarded. Such cases could threaten the conditioning or form
    unanticipated behaviors.
- The BIRDS should be robust
  - Should be able to survive outdoors for long periods of time
  - Most critical sections for weatherproofing are the food storage, so the food does not spoil and the internal circuitry of the processing boards







|           | Component          | Parameter              | Specification   |
|-----------|--------------------|------------------------|---|
|           | Power Supply       | Charge Density         | Power all functions for a week                          |
|           | PCB                | Central Control        | Low power draw  |
| Electrica | Motors             | Reliability            | Operate without fail<br>95% of the time                 |
| El        | Image<br>Processor | Computational<br>Power | Deliver a verdict within 2 seconds of taking the image. |
|           | Status Displays    | Human<br>Interface     | Simple, cheap, and weatherproof                         |

|         | Component                | Parameter           | Specification        |
|---------|--------------------------|---------------------|----------------------|
|         | Collimating Lens (Trash) | FOV                 | 25°                  |
|         | Collimating Lens (Bird)  | FOV                 | 90°                  |
|         | Focusing Lenses          | Focal Length        | 10-20 cm             |
| Optical | CMOS Sensors             | Resolution Size     | 640 x 480            |
| 0       | Red Laser Diode          | Spot Size           | 6mm @ 5m<br>distance |
|         | Photodiode               | Wavelength<br>Range | 400-1100nm           |
|         | Mirror                   | Wavelength<br>Range | 400-700nm            |

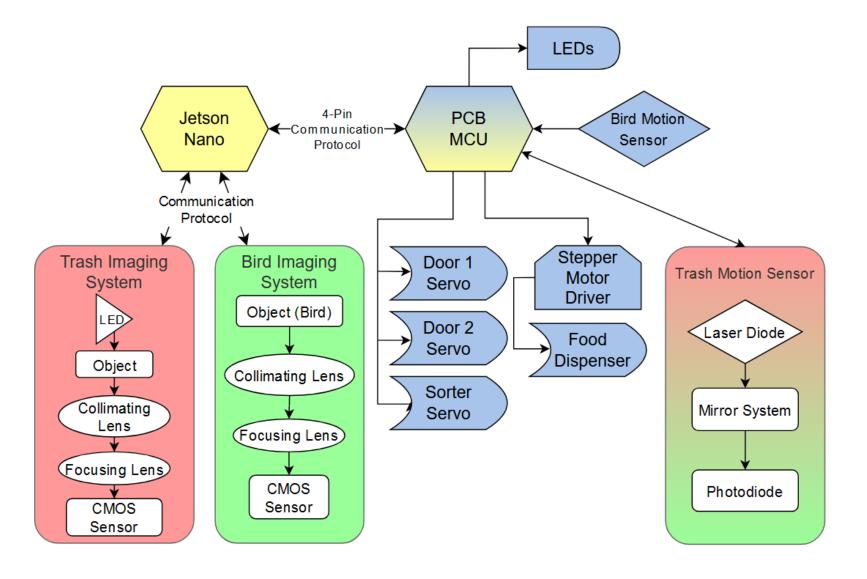


| System/Feature                                      | Requirement |
|---|-------------|
| Accuracy of Bird Image Recognition                  | ≥ 90%       |
| Accuracy of Trash Image Recognition                 | ≥ 85%       |
| Accuracy of Bird Motion Sensor                      | ≥ 50%       |
| Accuracy of Trash Motion Sensor                     | ≥ 90%       |
| Time duration of Input Door from Bird Detection     | ≤3 sec      |
| Time duration of Trash Sorter verdict               | ≤3 sec      |
| Time duration of Food Dispenser from Trash approval | ≤2 sec      |



### Hardware Block Diagram





Work Distribution

Aiden

Will

Rodrigo

Donovan



### Peripheral Devices

#### **Motion Detectors**

- Bird Detector
  - Uses a Capacitive Sensor to read changes in the dielectric properties surrounding a sensing conductor
  - We built our own by charging and reading the discharge time of a large conductor plate that will serve as the sensor
- Trash Detector
  - Uses a Laser and Photodiode to detect when an object breaks the laser path as it passes through

#### **Actuators**

- Food Dispenser
  - Uses a 5 Vdc Stepper Motor for its high torque and precise control
  - Will rotate a screw conveyer to dispense bird feed down to the food receptacle
- Door Actuators
  - Uses two 9-gram Servo Motor to operate two semicircle doors that grant access to the Analysis Chamber
- Trash Sorter
  - Uses one 9-gram Servo Motor to rotate the analysis plate towards Trash Storage or the Discard Chute

#### **Jetson Communication**

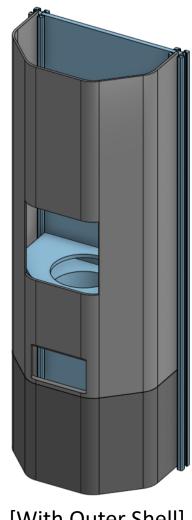
- Simple 4-pin connection to the Jetson Nano to activate and receive the verdict of the image processing cameras
  - 2-pins for "Activate Bird Camera" and "Is Bird?"
  - 2-pins for "Activate Trash Camera" and "Is valid item?"



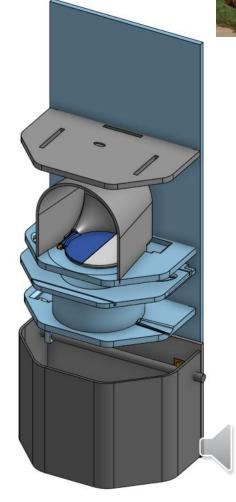
### A B.I.R.D.S. House

#### The body in which to house the internal organs

- Four general sections (from top to bottom):
  - Food Storage and Food Dispenser
  - Trash Intake, Bird Platform, and Food Delivery
  - Laser Break Plane
  - Analysis Chamber
  - Discard Chute & Trash Storage
- Removable Trash Storage
- Unwanted Items will be discarded out the Housing
- Will (probably) have a transparent Outer Shell
- Removable Outer Shell for ease of maintenance
- System Plates will go through Backplate and be permanently secured



[With Outer Shell]



[Without Outer Shell]

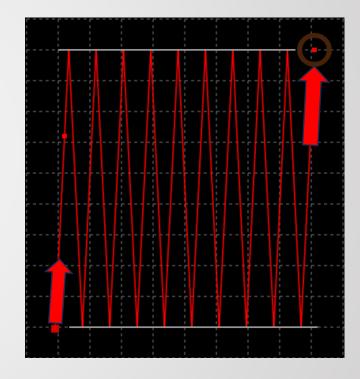
### Laser Motion Sensor Selection

### **Photodetector Specifications**

|                       | Solar Cell    | Photodiode     | Photoresistor |
|-----------------------|---------------|----------------|---------------|
| Data Rate             | 34.2 Mbps     | 2 Gbps         | N/A           |
| Rise Time             | N/A           | 1 µs           | 10 ms         |
| Spectral<br>Bandwidth | 400nm - 700nm | 500nm - 1000nm | 400nm - 900nm |
| Cost                  | \$15.50       | \$0.40         | \$0.20        |

Photodiode has fastest rise time and a low cost

#### Laser Grid

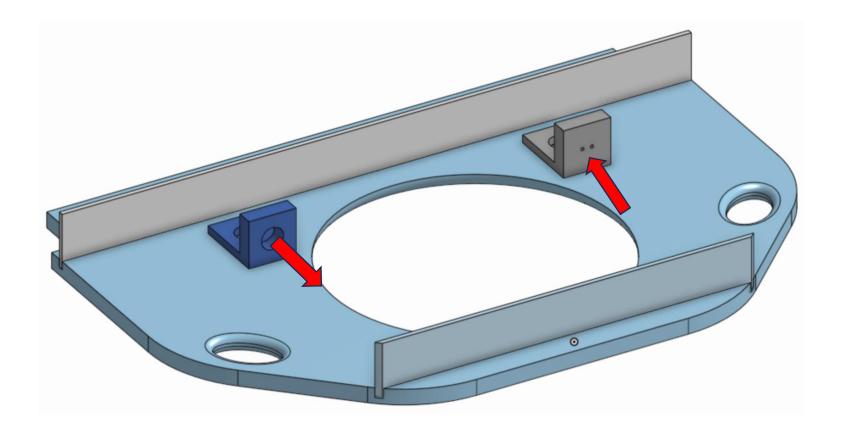


 Single laser diode will send laser through mirror system into photodiode



### Laser Motion Sensor Design

### Integration





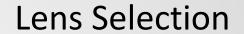


### Lens System Hardware Selection

### **Sensor Specification**

|                      | CMOS | CCD   |
|----------------------|------|-------|
| Readout<br>Noise     | 5 e  | 3 e-  |
| Power<br>Consumption | 2.5V | 12V   |
| Readout<br>Time      | 10ms | 500ms |
| Price                | \$50 | \$200 |

- There is more noise created by the CMOS sensor
- Readout time and power consumption are very low



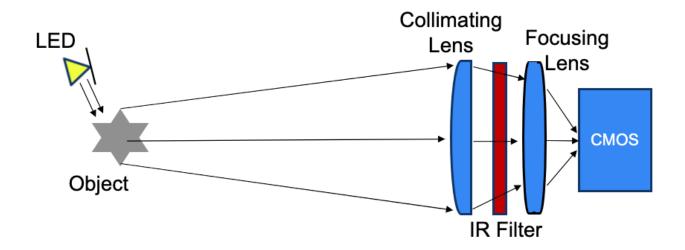
| Lens shape    | Ideal conjugate ratio                         |
|---------------|---|
| Biconvex      | < 5:1   |
| Plano-convex  | All   |
| Plano-concave | Infinite, larger finite (> 5:1)               |
| Biconcave     | < 5:1   |
| Meniscus      | Varies; dependent upon curvature and polarity |

- The objects to be imaged will be far away from the front of the lens
- The image formed by the lens will be close to the back of the lens
- Leads to a large or infinite ideal conjugate ratio
- If a second lens is used, it would have a much smaller ideal conjugate ratio



### Lens System Hardware Design

### General 2-Lens System



- An infrared (IR) filter is needed to block out all IR wavelengths that could interfere with the desired image
- 2-Lens system was designed to acquire greater minification of image onto CMOS sensor

### Bird vs. Trash

- FOV
  - Bird 90 deg
  - Trash 25 deg
- Focal Length
  - Bird 10-20cm
  - Trash 15cm

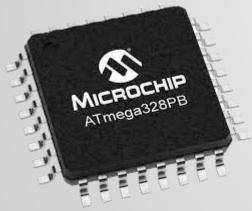


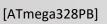




#### What we need:

- Low power mode to draw less power when reading sensors
- Enough GPIO pins to support our peripherals
- Interrupt capabilities
- Cheap
- Development board for easy testing and debugging







[MSP430]







- Because almost every microcontroller has the features that we are looking for, we decided to try to minimize cost.
- Another goal was to use Arduino IDE to make writing the embedded software very simple giving much more time and focus to the machine learning section.
- To minimize cost we wanted to pick a development board that someone in the group already owns
- Our final choice of development board was the Arduino Uno

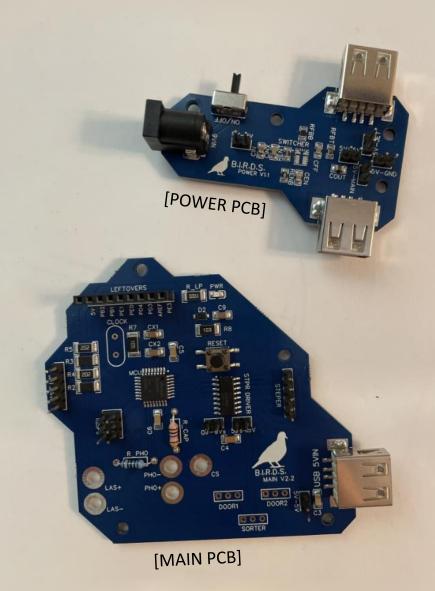
| Device               | Flash  | RAM   | GPIO<br>Pins | Programming<br>Language | Price          |
|----------------------|--------|-------|--------------|-------------------------|----------------|
| Arduino Mega<br>2560 | 256 KB | 8 KB  | 54           | Arduino IDE             | Already<br>Own |
| Arduino Uno          | 32 KB  | 2 KB  | 14           | Arduino IDE             | Already<br>Own |
| MSP430FR698<br>9     | 128 KB | 2 KB  | 83           | С                       | \$26.59        |
| MSP430G2553          | 16 KB  | .5 KB | 24           | С                       | Already<br>Own |

### Microcontroller Selection



- Since we are using Arduino, we are sticking with the ATmega series of MCU
- Flash memory is more than what we need
- EEPROM is only for storing specific data when powered off
- The I/O pins are important we have more than we need
- PWM channel the size doesn't matter as long as we have 3
- We may need a couple ADC channels but all of these have way more than necessary
- Price was the biggest selling point for the ATmega328PB

| Device      | Flash<br>KB | EEPROM<br>KB | RAM<br>KB | I/O<br>Pins | 16-bit<br>resolution<br>PWM<br>channels | Serial<br>UARTs | ADC<br>Channels | Price  |
|-------------|-------------|--------------|-----------|-------------|---|-----------------|-----------------|--------|
| ATmega328PB | 64          | 2            | 4         | 27          | 11 (10 bit resolution)                  | 1               | 11              | \$1.63 |
| ATmega640   | 64          | 4            | 8         | 86          | 12                                      | 4               | 16              | \$5.7  |
| ATmega1280  | 128         | 4            | 8         | 86          | 12                                      | 4               | 16              | \$16.2 |
| ATmega1281  | 128         | 4            | 8         | 54          | 6                                       | 2               | 8               | \$13.6 |
| ATmega2560  | 256         | 4            | 8         | 86          | 12                                      | 4               | 16              | N/A    |
| ATmega2561  | 256         | 4            | 8         | 54          | 6                                       | 2               | 8               | \$25.4 |



### PCB Design

#### **Power Board**

- Since we are powering two computing boards (Jetson Nano & Main), one board powers both
- Easier to debug and repair if something goes wrong
- Has multiple Jumper-Pin breakers for testing and control
- Takes in 9 Vdc from a 120 Vac to 9 Vdc wall adapter
   Outputs 5 Vdc to two USB-A female ports
- 96.608 % Efficient Buck Converter circuit from TI's WEBENCH application

#### Main Board

- Controls and coordinates all peripheral devices
- Programed through ICSP
- Has an isolatable stepper driver circuit that can be independently powered or powered by the Main
- Powered by 5 Vdc USB-A cable or ICSP pins
- Power LED (of course)





### Machine Learning Technology Comparison

#### What we need:

- Hardware capable of performing an object detection of a bird within our allotted time of 1 second
- GPIO pins to communicate with our processor on our custom PCB.
- Needs to be able to run 2 different machine learning models either in parallel or quickly swap.



#### **Nvidia Jetson Nano**

- Can run multiple machine learning models in parallel
- The most expensive option with original price
- The most versatile because of its gpu
- 4 USB ports for easy access of peripherals
- Has a prebuilt library for machine learning called jetpack

#### Final Consideration Options:



#### **Coral Dev Board**

- Works very well with TensorFlow architectures
- Has easy access to Wi-Fi and Bluetooth
- Has prebuilt machine learning models on TensorFlow hub that can be easily integrated



#### Raspberry Pi 4 + Coral Accelerator

- This was the first idea for machine learning
- Coral Accelerator helped match this to contenuers
- Is a very popular choice for projects
- Is very commonly sold out, or being resold for much more than its value
- Might run into issues with multiple pieces of hardware working together

### Machine Learning Technology Comparison

- We need a device that can work well on a number of architectures
- We need a device that can respond quick enough to meet our 1 second image processing goal
- We need a device that we can receive quickly to start testing
- We need a device that can handle 2 different models for a bird and for detecting a cigarette

#### Neural Net Framework Benchmark

#### Hardware Specifications Machine Learning

| Model        | Framework  | NVIDIA Jetson<br>Nano | Raspberry Pi 4 B<br>+ Coral | Coral Dev<br>Board |            | Coral Dev Board  | NVIDIA Jetson Nano Developer Kit                                       | Raspberry Pi 4 Model B<br>+ Coral USB Accelerator                      |
|--------------|------------|-----------------------|-----------------------------|--------------------|------------|--|--|--|
|              |            |                       | Accelerator                 |                    |            |  |  |  |
|              |            |                       |                             |                    | CPU        | Quad Cortex-A53, Cortex-M4F                                      | Quad-core ARM A57  | Quad core Cortex-A72   |
| MobileNet V1 | TensorFlow | 61.6                  | 14.9                        | 15.7               |            |  |  |  |
|              |            |                       |                             |                    | Clock Rate | 1.5 GHz  | 1.43 GHz   | 1.8 GHz  |
| MobileNet V2 | TensorFlow | 72.3                  | 18.2                        | 20.9               |            |  |  |  |
|              |            |                       |                             |                    | GPU        | Integrated GC7000 Lite Graphics<br>+ Google Edge TPU coprocessor | 128-core Maxwell @ 921 MHz   | Google Edge TPU coprocessor  |
| ResNet-50 V2 | PyTorch    | 27.7                  | 52                          | 56                 |            |  |  |  |
|              |            | -                     |                             |                    |            |  |  |  |
| ResNet-18    | PyTorch    | 200                   | DNR                         | DNR                | Cache      | 32KB L1 instruction and data cache, 1MB unified L2               | 48 KB L1 instruction cache and 32 KB L1 data cache, 2 MB of L2 Unified | 48 KB L1 instruction cache and 32KB of L1 data cache, 1MB of L2 Shared |
|              |            |                       |                             |                    |            |  | Cache  | Cacho  |
| YOLOv3-320   | Darknet    | 40                    | DNR                         | DNR                | RAM        | 1 or 4 GB LPDDR4 Memory  | 2 or 4 LPDDR4 Memory   | 1, 2, 4, or 8 GB LYDDR4 Memory   |
|              |            |                       |                             |                    | NAIVI      | 1 01 4 GB LF DDIX4 Wiemory                                       | 2 01 4 LF DDIA4 Memory   | 1, 2, 4, 01 6 GB Le DDI(4 Memory                                       |
| VGG – 19     | MXNet      | 100                   | 200                         | DNR                |            |  |  | 19   |
|              |            |                       |                             |                    | Flash      | 8 GB eMMC, with a MicroSD slot for expandable storage            | 16 GB eMMC, with a MicroSD slot for expandable storage                 | Expandable MicroSD slot for program and Operating system               |
| Unet         | Caffe      | 55.5                  | 200                         | DNR                |            |  |  |  |



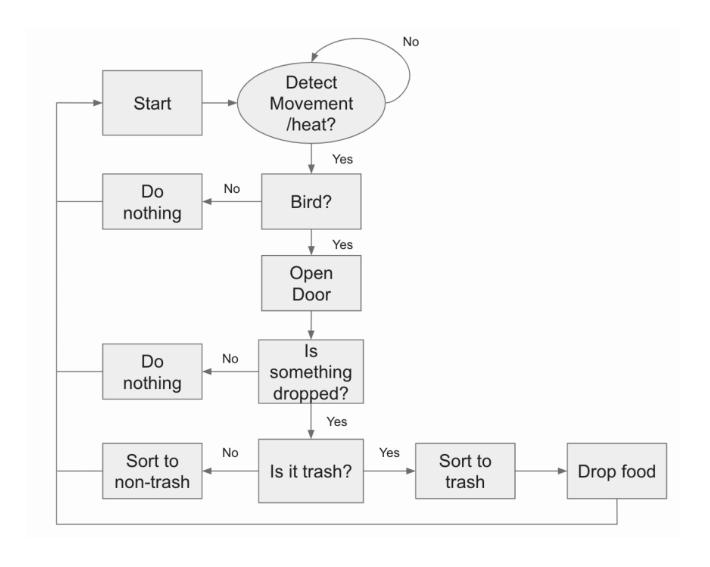


For our choice of machine learning technologies, we chose to go with the NVIDIA Jetson Nano 4GB.

- This board comes with an easy way of installing a wide array of libraries called Jetpack SDK
- This was the most versatile option guaranteeing performance
- Had a very easy setup with a microSD card using a tutorial from NVIDIA.
- A wide array of tutorials from NVIDIA on how to import prebuilt models and create your own object detection models.
- This was in stock and came within a few days of purchase.
- Very easy to add Wi-Fi and Bluetooth capabilities should we need them in the future



### Software Block Diagram



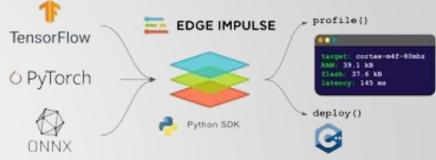




## Training Machine Learning Platform

- To be able to do object detection we need a place to train our custom model since the Jetson isn't powerful enough for training.
- We decided to pick a cloud-based platform that could train object detection models from prebuilt weights on a neural network framework.
- This pick was Edge Impulse and one of the biggest selling points was the ability to export the model to the specific jetson library







[Example shot of our Object Detection for cigarettes]

### Budget

| Item               | Price     | Quantity |
|--------------------|-----------|----------|
| Microcontroller    | \$13      | 1        |
| Nvidia Jetson Nano | \$150     | 1        |
| Servo Motor        | \$10      | 3        |
| Stepper Motor      | \$13      | 1        |
| Custom PCB         | \$20-\$30 | 2        |
| Housing Material   | \$200     | 1        |
| LED Light Source   | \$15      | 10       |

| Item             | Price   | Quantity |
|------------------|---------|----------|
| Collimating Lens | \$23.50 | 2        |
| Focusing Lens    | \$20    | 2        |
| IR Filter        | \$5     | 2        |
| CMOS Sensor      | \$25    | 2        |
| Laser Diode      | \$13.18 | 1        |
| Photodiode       | \$0.32  | 20       |
| Mirror           | \$0.28  | 50       |
| TOTAL            | \$790   | 6.58     |



### Work Distribution



Aiden Nipper



- Lens System Design (Trash Camera)
- Laser Motion Sensor Design
- Housing & Structural Design

Will DiSalvo



- Lens System Design (Bird Camera)
- Laser Motion Sensor Design
- Structural Design

#### Rodrigo Guerra



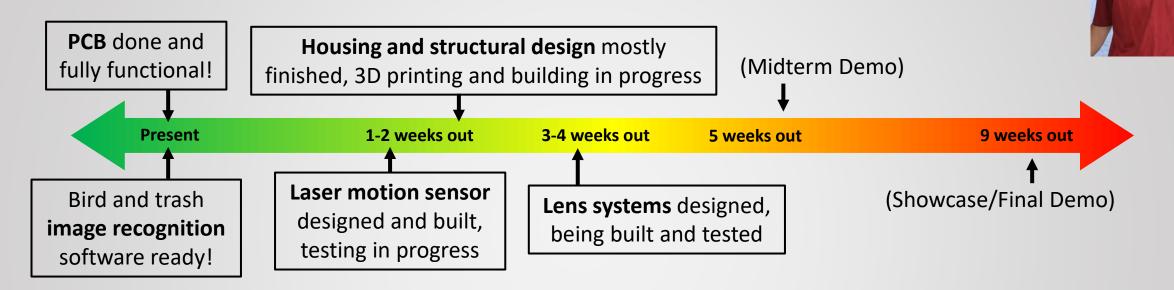
- PCB Design
- Wiring and Soldering
- Housing & Mechanical Design

#### **Donovan Reynolds**



- Machine Learning (Jetson)
- Embedded Software
- Website

### Progress and Future Milestones



- Anticipating system to be built at least one month before showcase, with hope of obtaining data through field-testing
- Hope for the project is that others improve design and use it to train birds in a specific area
- Having the BIRDS in major cities and/or universities could slowly begin to change the way our streets look

