

AI For Behavioral Discovery



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Motivation

Can machine learning learn the behavioral effects of EMF radiation on biological systems?



Machine Learning

- Models are trained to identify complex patterns that humans can not easily recognize
- Highly scalable
- Shortcut Learning
 - Models learn the easiest feature (covariates)
- Noise Reduction
 - Uncontrollable features (typically unrelated)



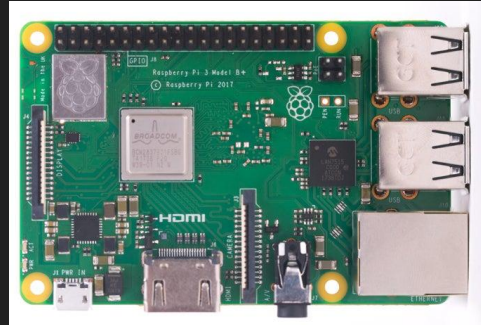
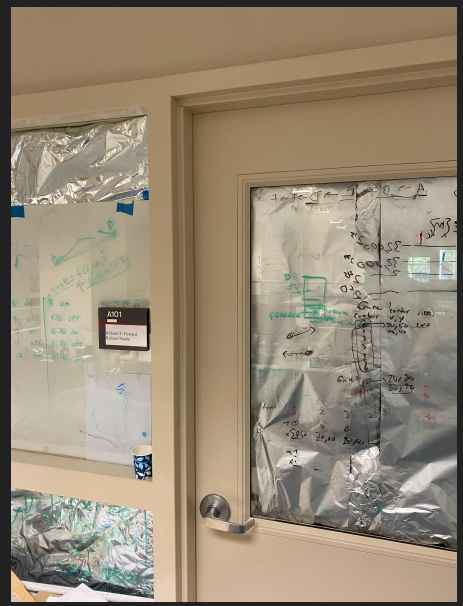
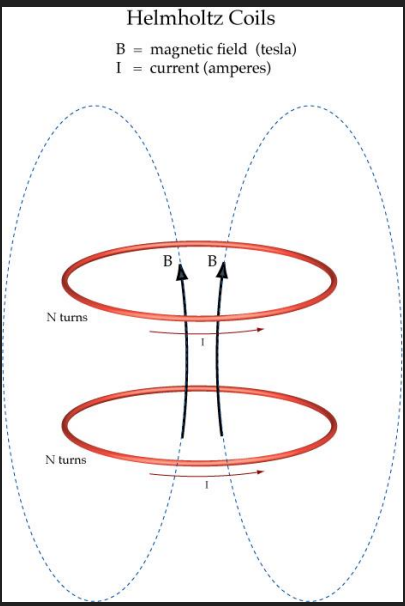
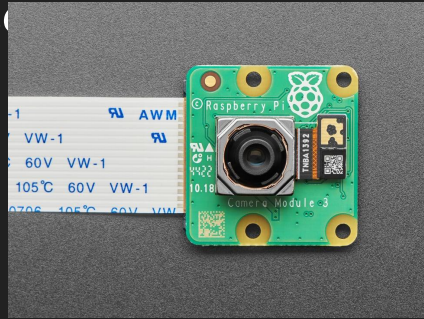
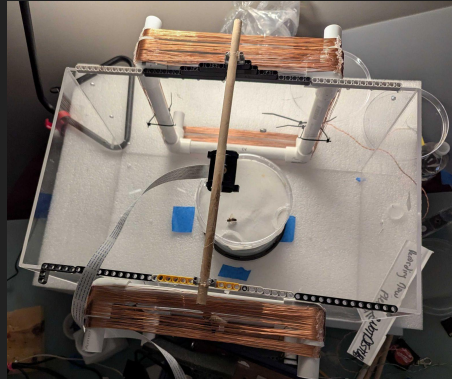
Our Approach

- Use simulation data to diagnose behaviors
- Camera recording data
- Using real video data starting with bees and ants
- Deep Neural Network to determine field presence
- Understand role of electromagnetic stimuli on multiple species



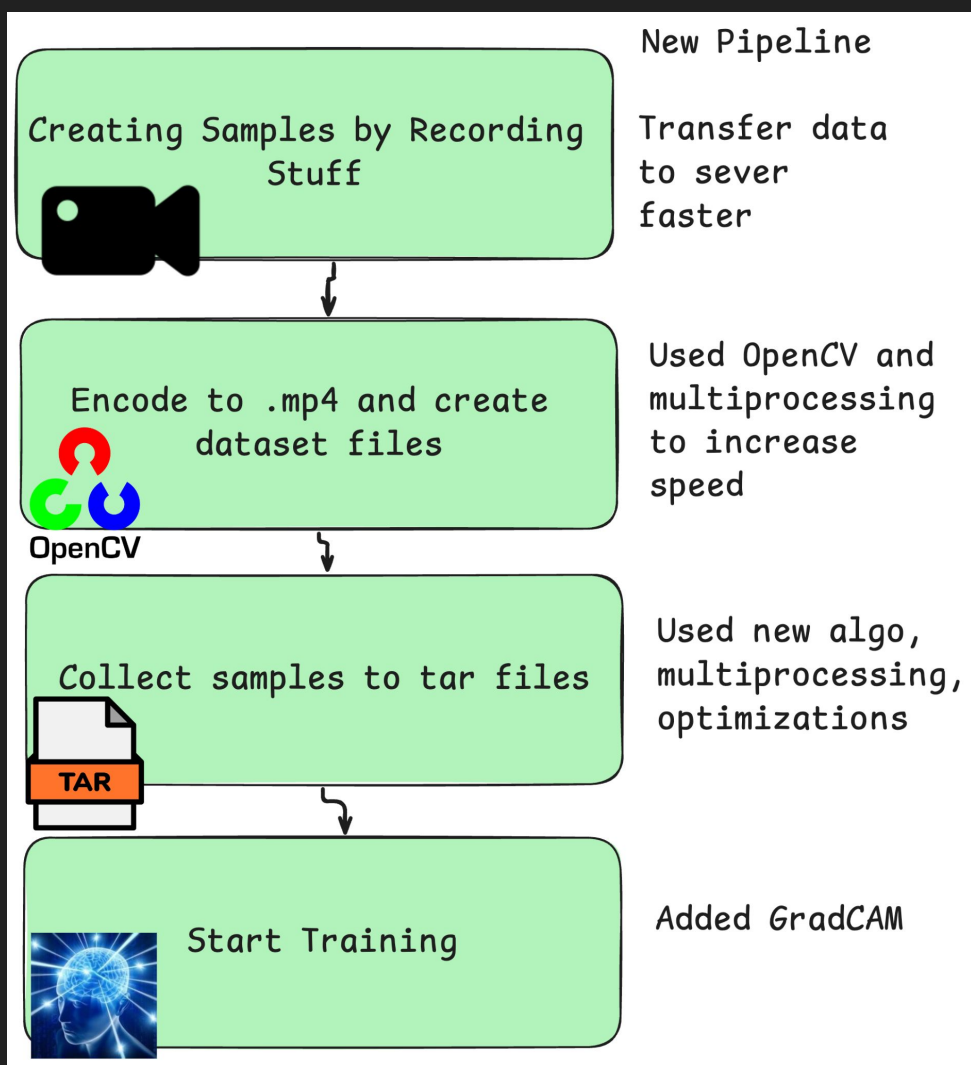
Experimental Setup

- Copper Helmholtz coils -> Uniform field
- Petri dish to hold insects
- Sealed off light cavities to reduce noise
- Raspberry Pi + Overhead camera



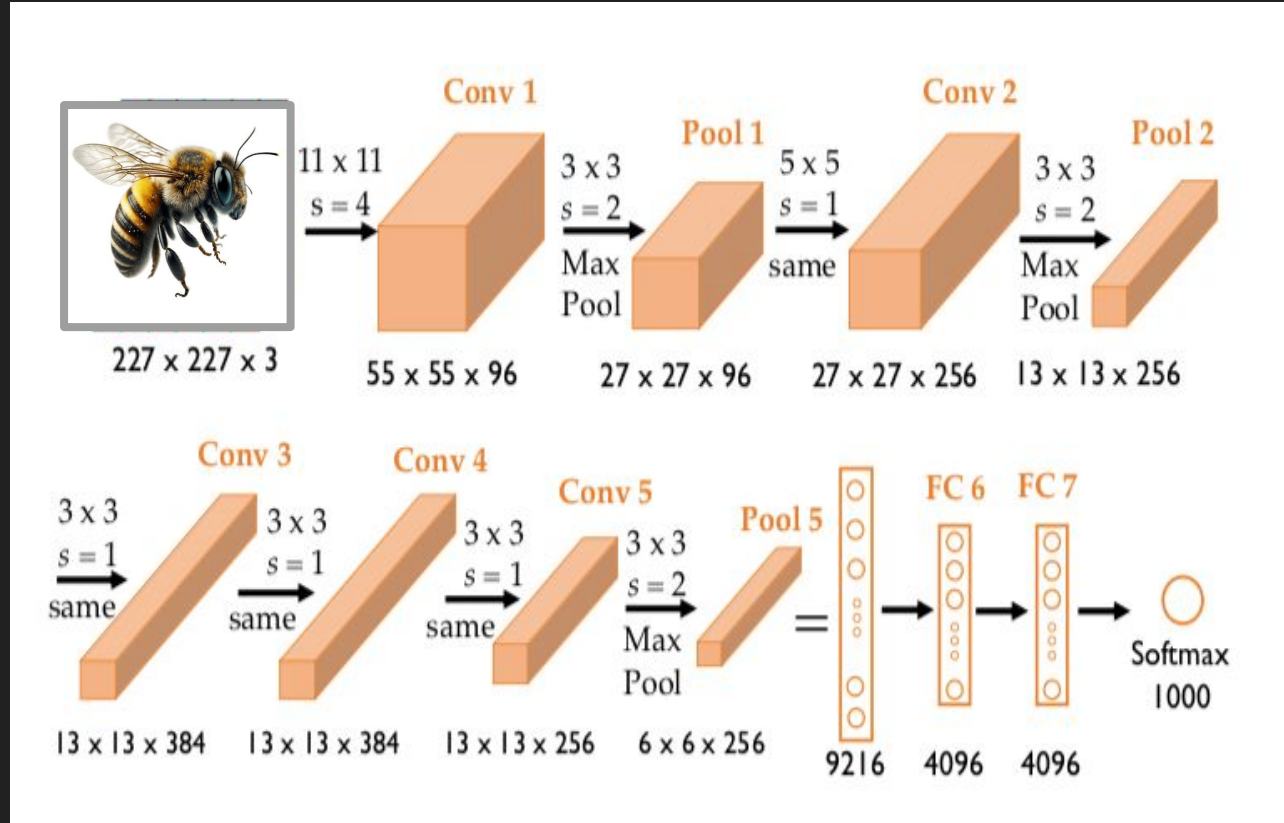
Software Pipeline

- Shell scripts
- Slurm to run scripts.
 - Access to GPUs
 - Unreliable
- K-Fold Cross Validation
 - Removes potential biases
- Finished new pipeline -> shorter execution times (multiprocessing)



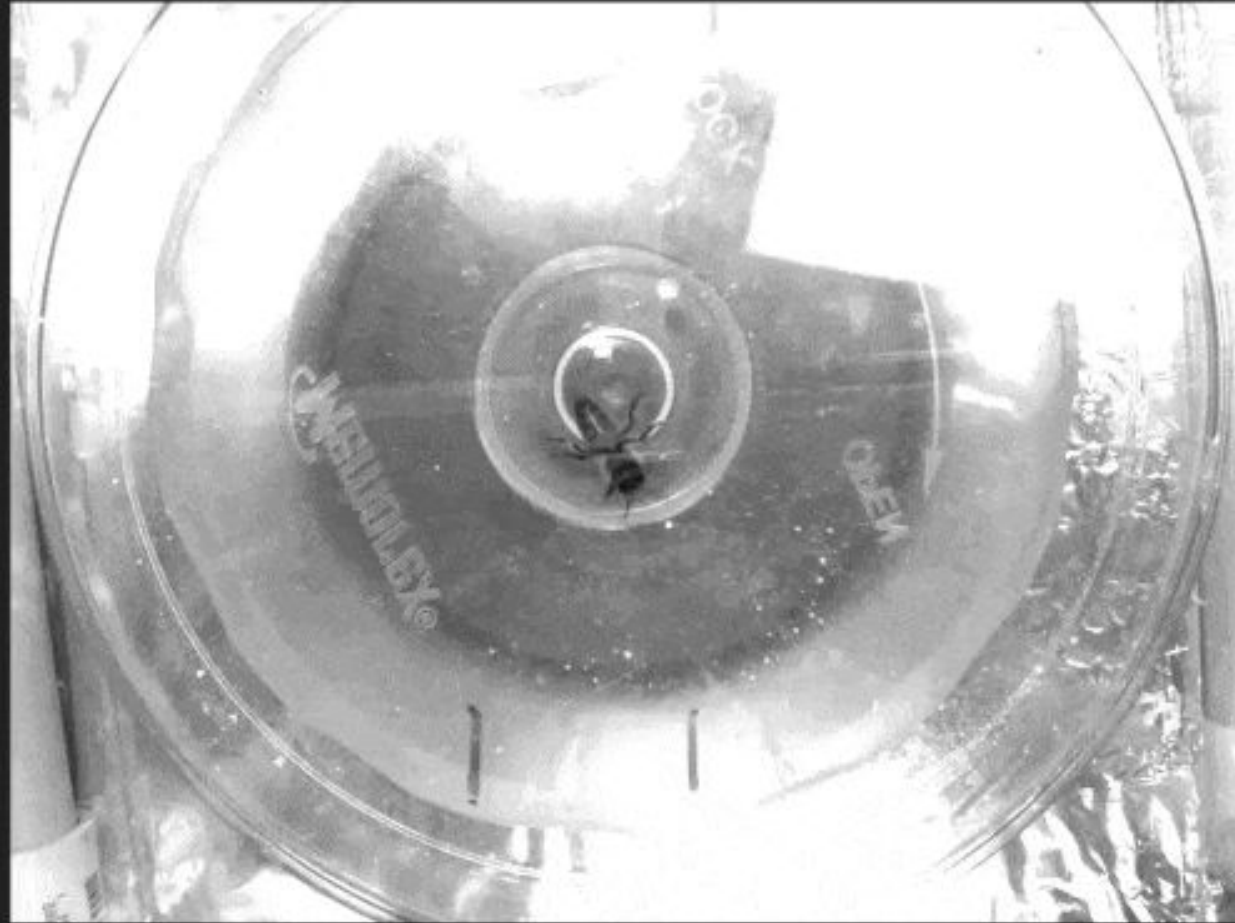
Model Architecture

- Used CNN (Alexnet); along with other architectures
- Multiple frames per sample -> added as channel dimension.



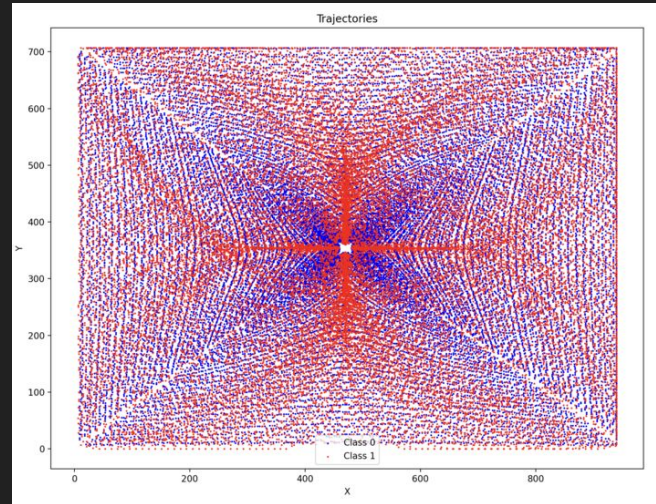
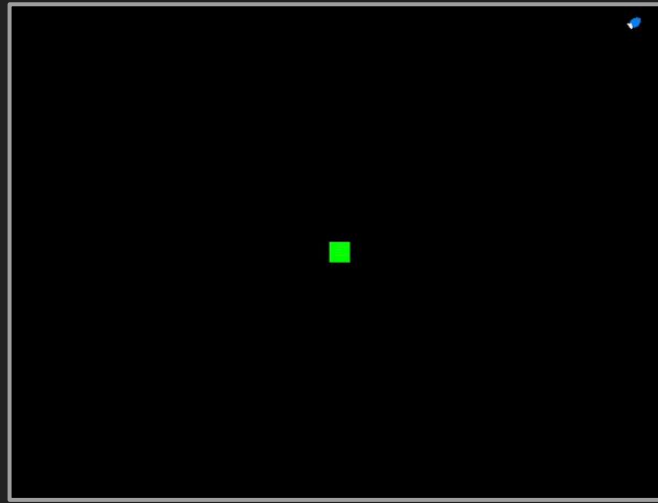
Grad-CAM plots

- Model Visualization tool
- Highlights regions that the model uses to learn
- Similar to heatmap
 - Blue - low significance
 - Red - high significance



Simulated Data

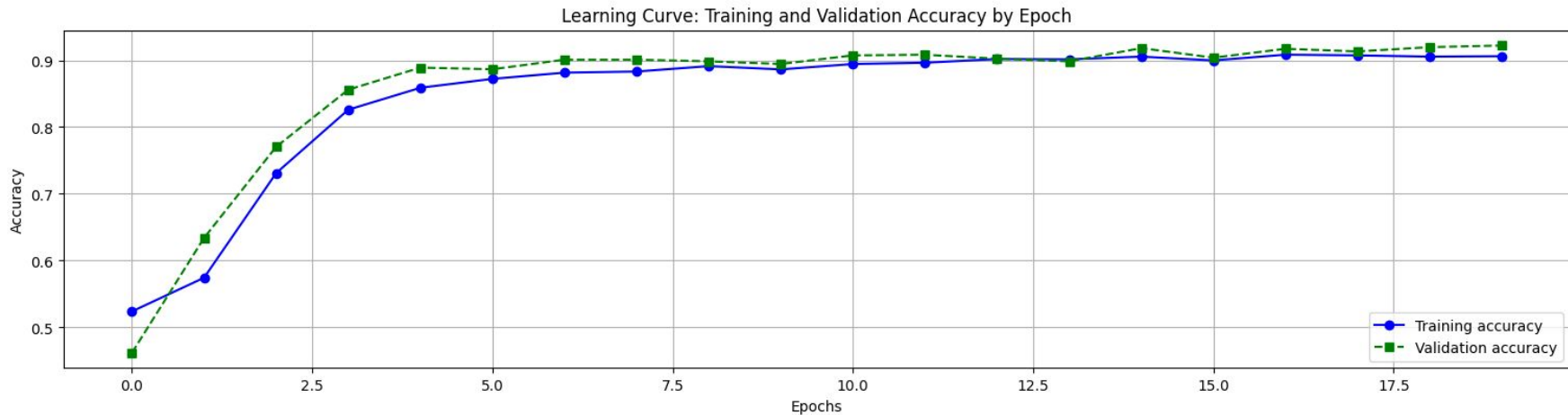
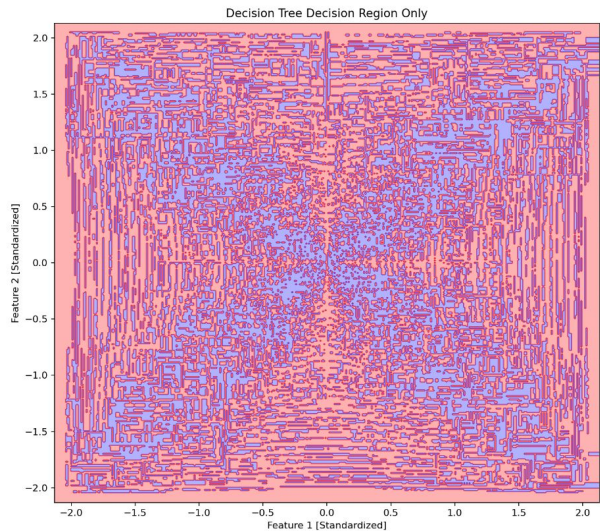
- 2 possible states: on/off
- Constant & normalized speed
- Varying features to increase data distribution
- 7 continuous frames per sample



Training Results

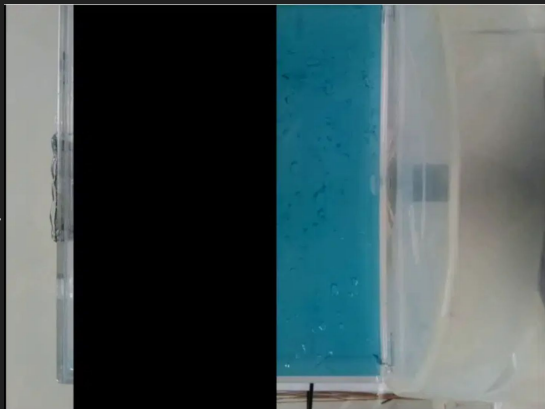
- Shortcut learning
- Attempted Naive Bayes Classification, collapsed

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Confusion Matrix:  
[[ 0 29587]  
 [ 0 36360]]  
Accuracy: 55.14%
```



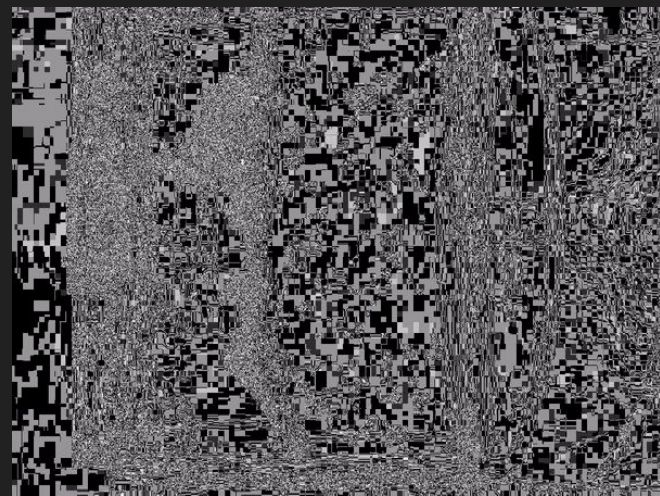
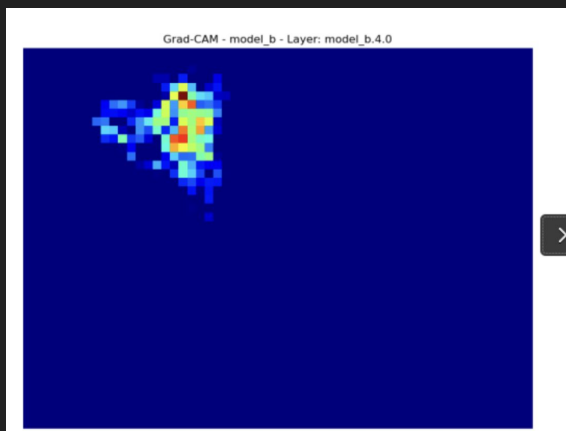
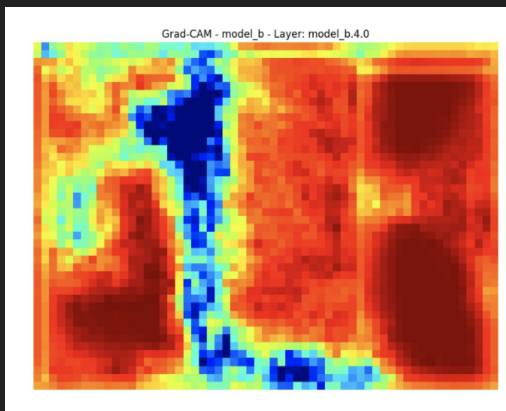
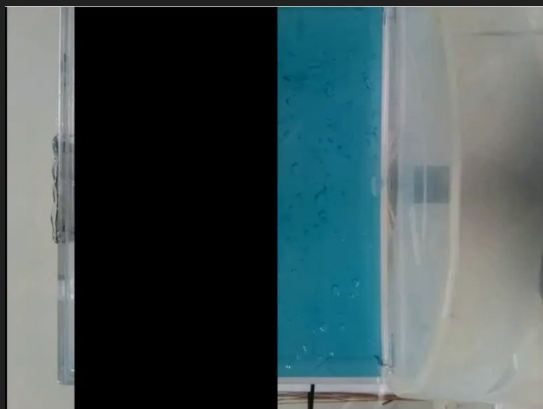
Real Data (Ants)

- Acrylic case filled with Ants and food-gel
- Unfamiliar environment
- Recorded camera data + trained on
 - Uncropped
 - No background
 - Control



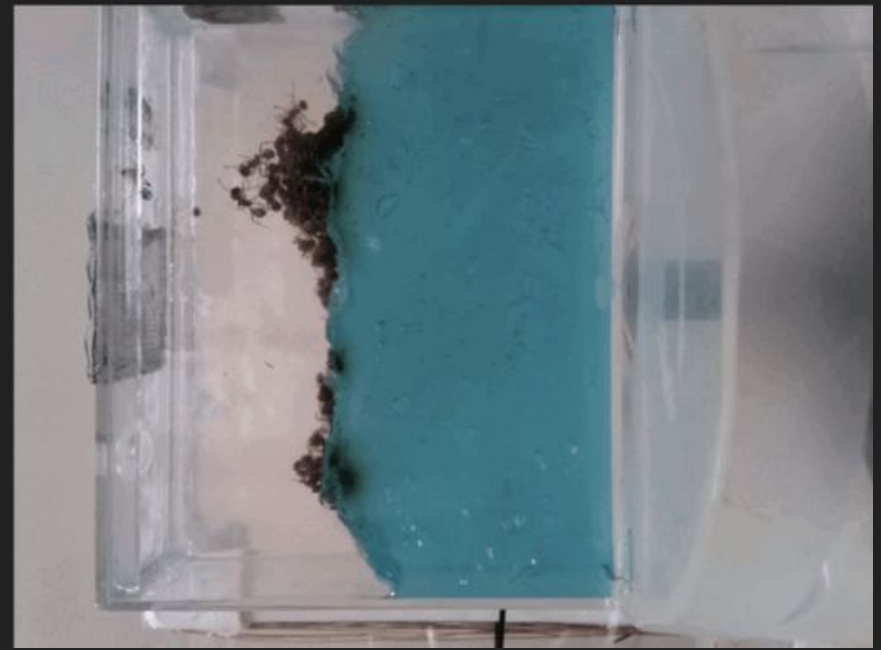
Training Results

- ~90% accuracy (Suspiciously high)
- Potential underlying issues
 - Lighting
 - Model not focusing on ants
- Addressing issues:
 - Background subtraction
 - Cropping
 - Grad-Cam analysis
 - Control Cropping

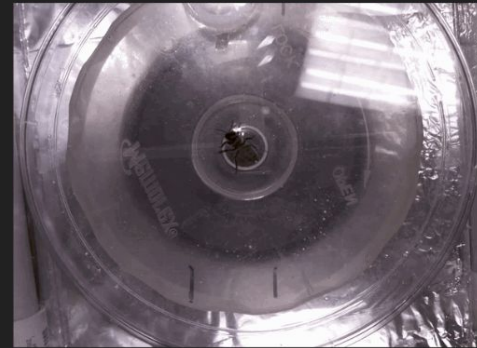


Conclusions + Realizations

- Promising results; further analysis is required
- Model is susceptible to shortcut learning
 - Easily affected by covariates
- Imperative that data is “clean”
 - Eliminate potential covariates
- Working with large data has limitations

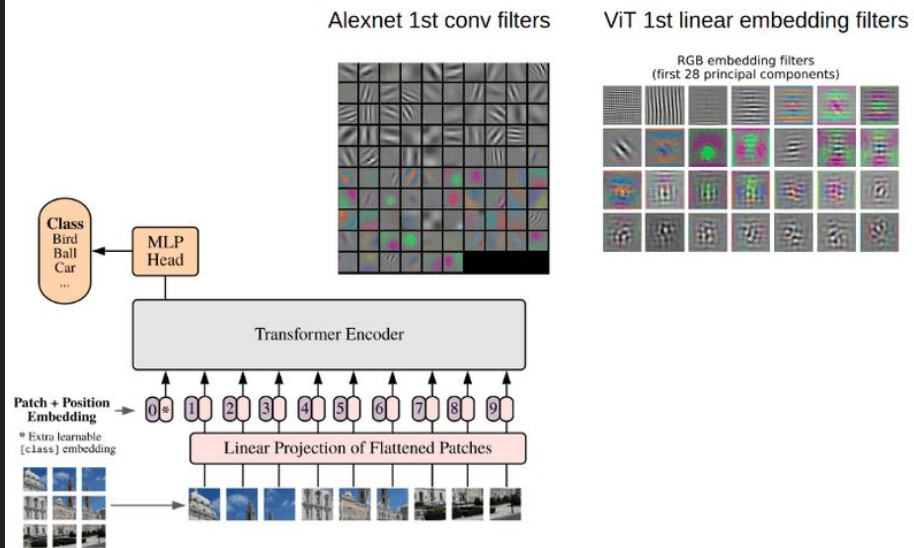


```
Accuracy: 0.8441747572815534
Class 0 precision=0.9067617662240157, recall=0.7347528517110267
Class 1 precision=0.8451539428212378, recall=0.9157872340425532
Class 2 precision=0.7537437603993344, recall=0.94375
Evaluating model.
Evaluation confusion matrix:
label 0:[19318, 3702, 3280]
label 1:[1551, 21521, 428]
label 2:[434, 238, 11328]
```



Future Work

- Reduce AI shortcut learning
 - Reduce noise
 - Make experimental setup closer to ‘ideal/perfect’ conditions
- Explore different models
 - Temporal capabilities



Thank you!

Any Questions?