

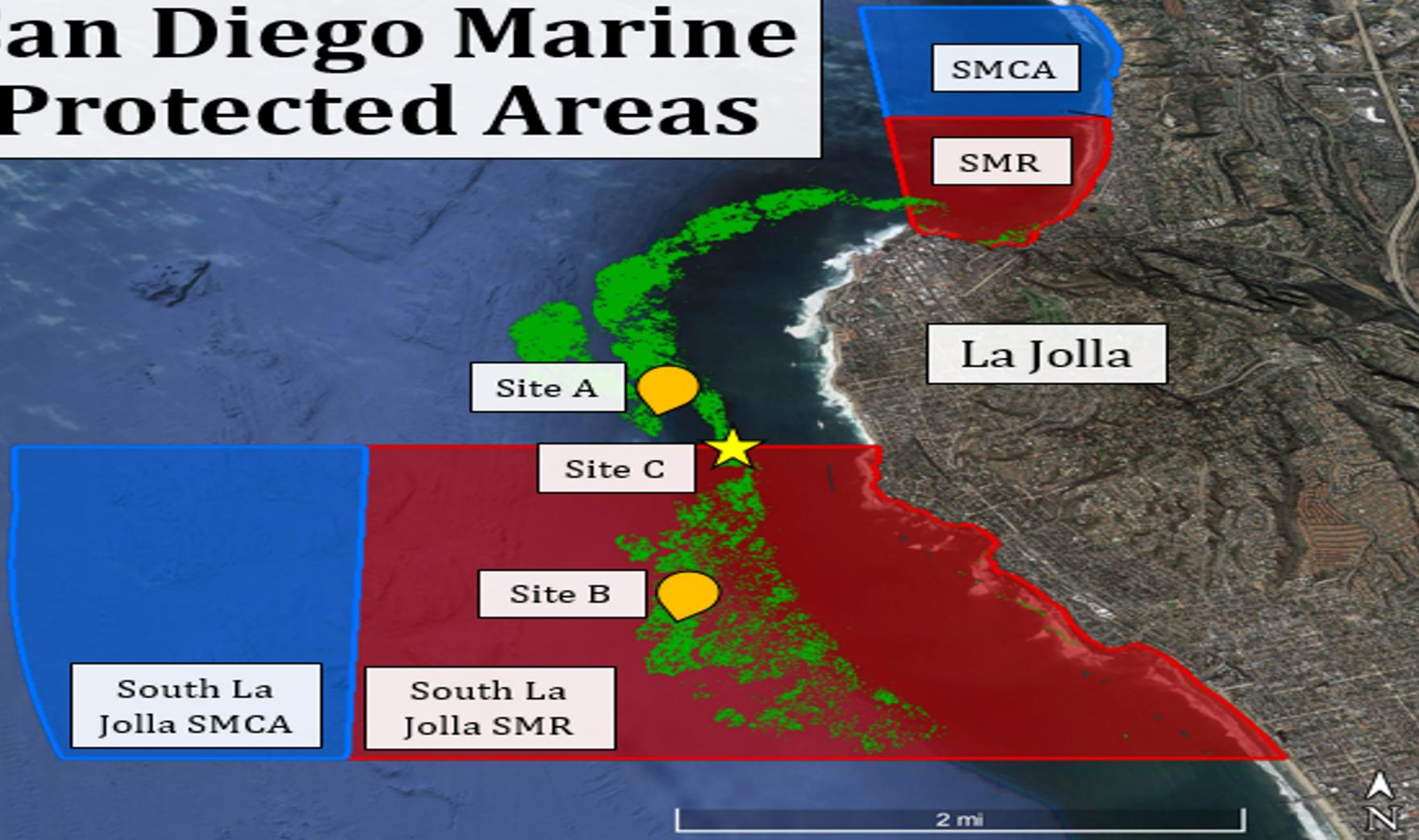


What's That Fish?

Monitoring & identification of fish in the La Jolla Kelp Forest

Kian Bagheri & Helen Cai

San Diego Marine Protected Areas





Motivations

- How do we know if MPAs are working?
- How can we better understand resident species?

Active monitoring: trawling, diver surveys, costly efforts

Passive monitoring: use noninvasive image and acoustic recording.



Objectives

Kian

- Deploy/ Recover passive monitoring system
- Process and Analyze images

FishOASIS

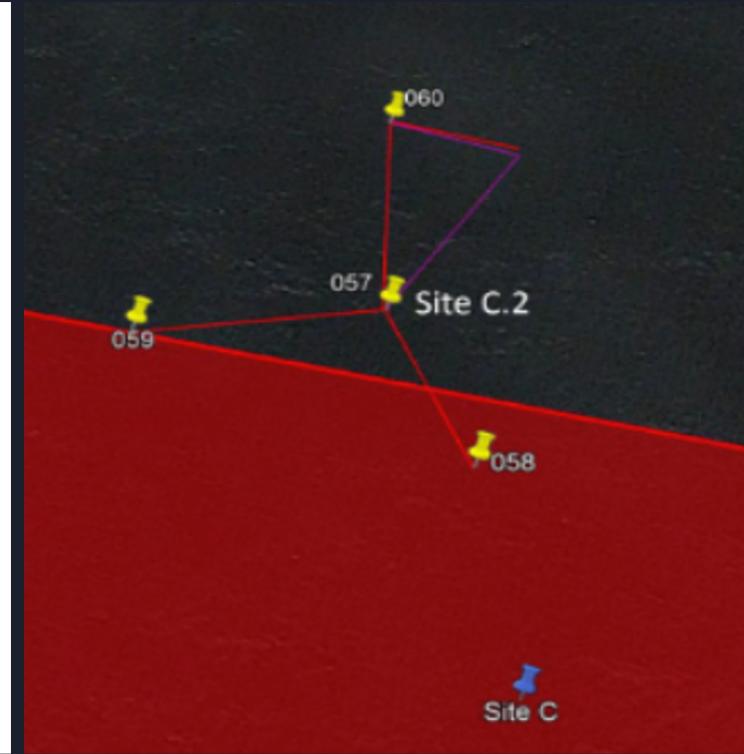
- Fish Optical and Acoustic Sensor Identification System
- Verify identity of fish
- Correlate biologic acoustics with individual species



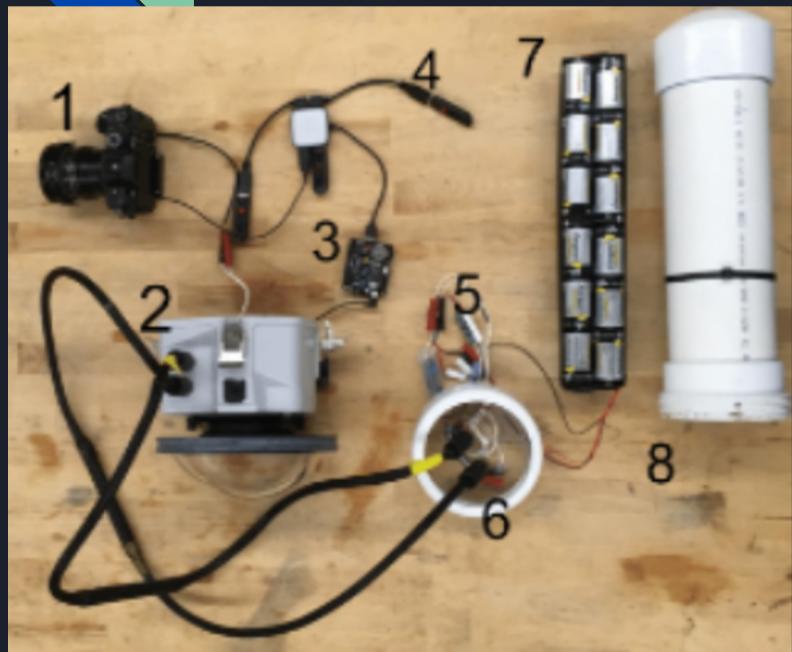


Array Setup

Site C



Camera: Hardware

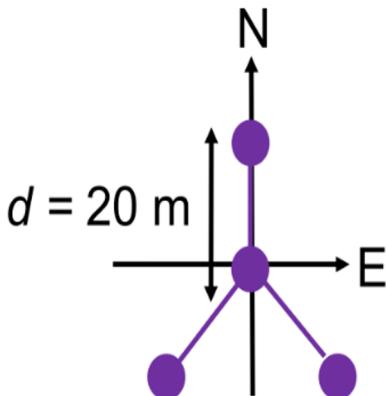


Deployment



Hydrophone Array

Passive Acoustic System



Components:

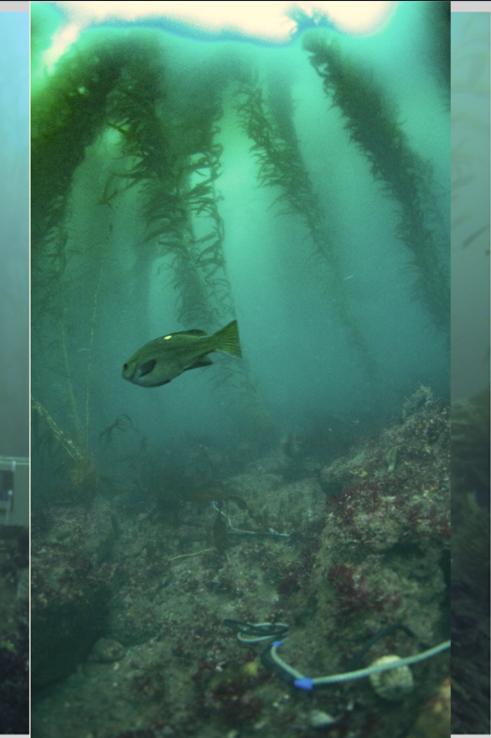
- SoundTrap ST4300 256 GB acoustic recorder
- four HTI-96-MIN hydrophones

Capabilities:

- sampling at 48-288 kHz
- up to 14 days continuous recording at 48 kHz



Image Processing: Raw Image



"SOUTHERN" KELP BED ROCKY REEFS

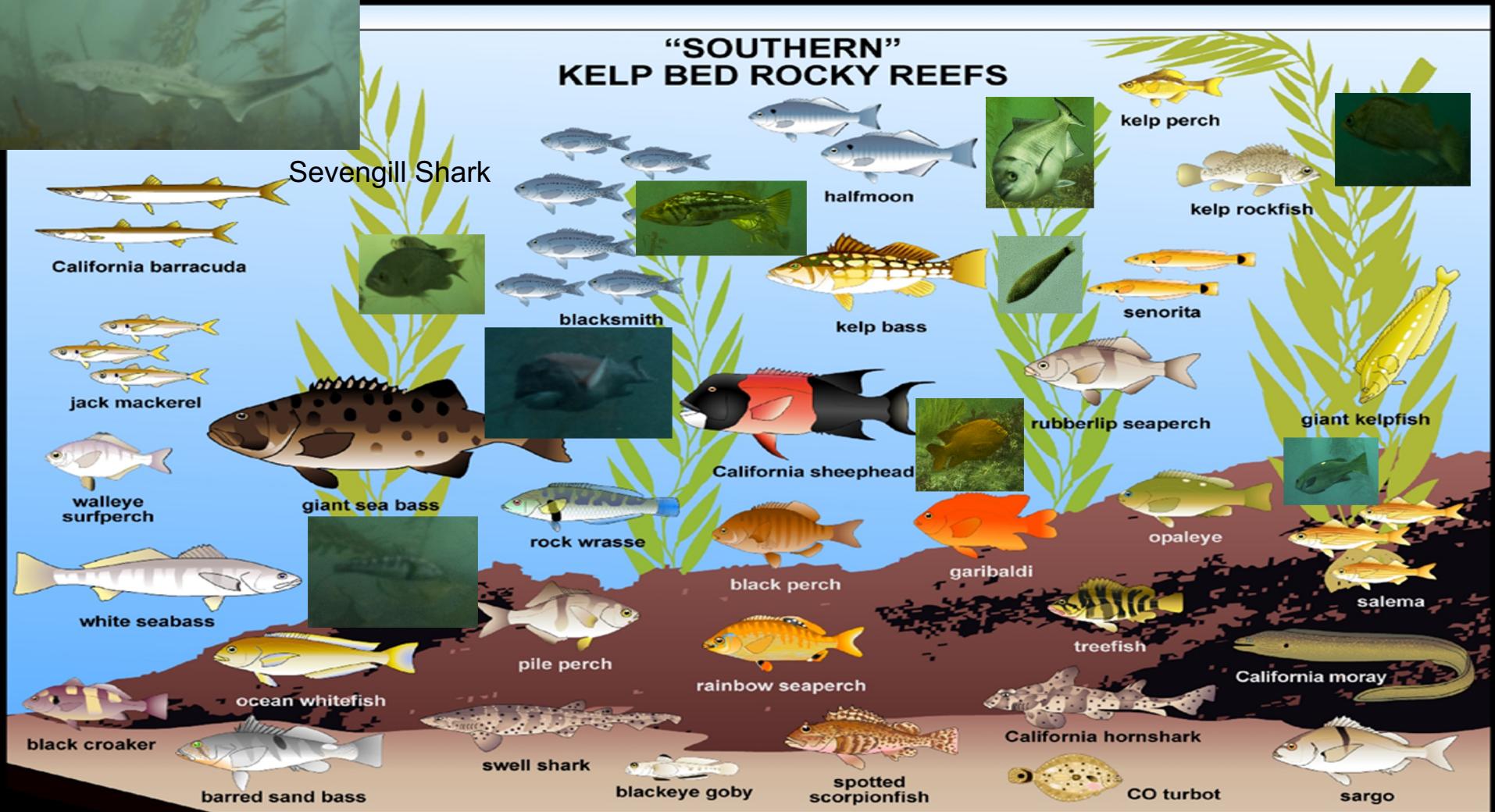


Image Analysis

Image Processing v.4

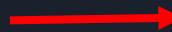
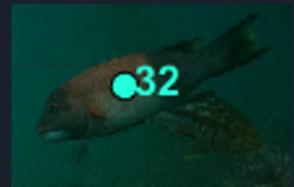
File

species

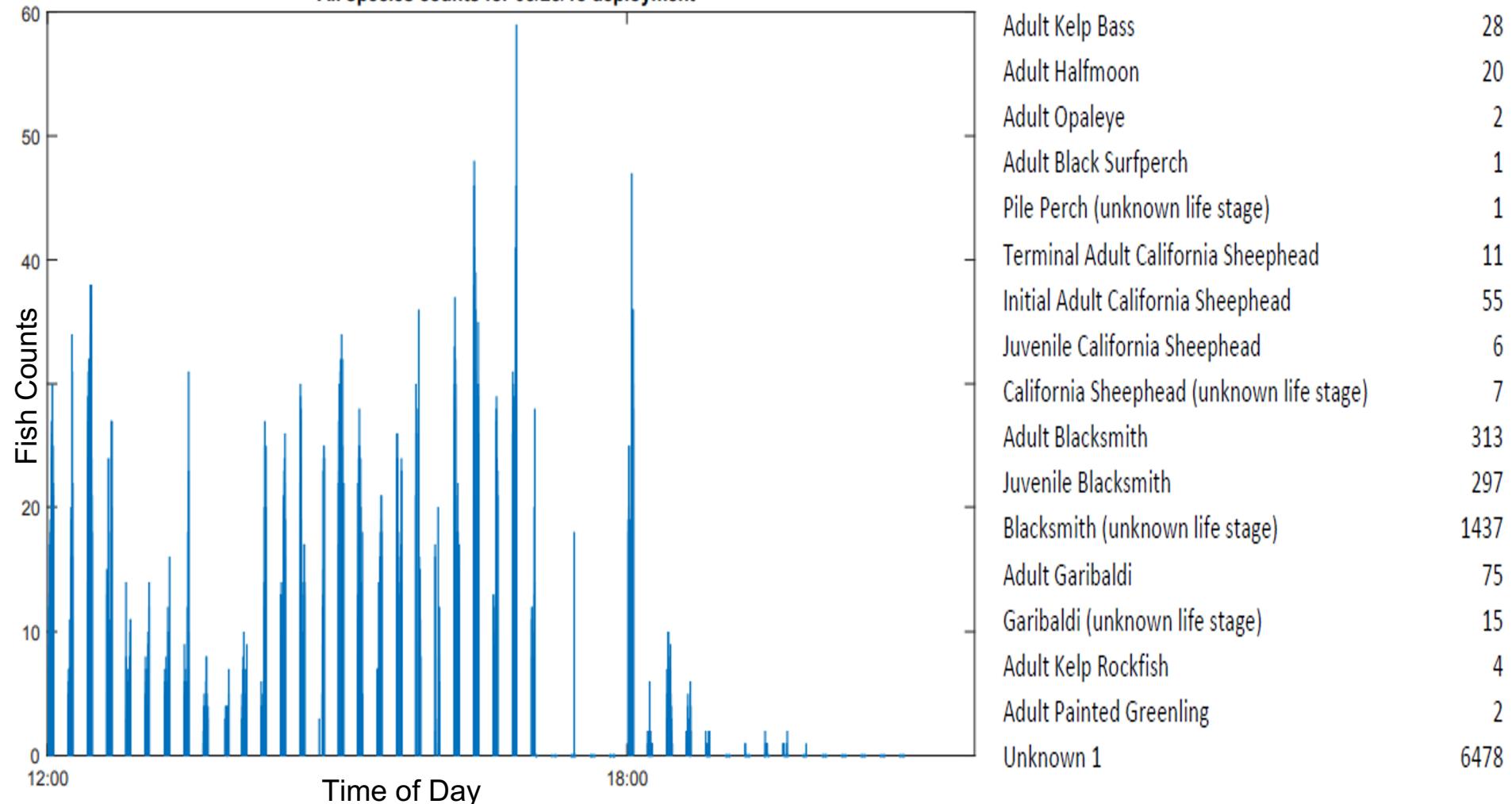
1 Spp0F	0
2 Spp0M	0
3 Spp0J	0
4 Spp0U	0
5 ProlA	0
6 ProlI	0
7 ProlU	0
8 ProBA	0
9 ProBI	0
10 ProBU	0
11 McPA	0
12 McAU	0
13 McAL	0
14 CnPA	0
15 CnBI	0
16 CnBU	0
17 CnPB	0
18 PloyA	0
19 FertI	0
20 RycA	0
21 RycI	0
22 RycU	0
23 HerA	0
24 HerI	0
25 HerU	0
26 HerBI	0
27 HsemI	0
28 HsemU	0
29 OcaI	0
30 OcaU	0
31 OcaU	0
32 SquIA	0
33 SquAI	0
34 SquU	0
35 SquU	0
36 CognA	0
37 CognI	0
38 CognU	0
39 HrubA	0
40 HrubI	0
41 HrubU	0
42 SatI	0
43 SatJ	0
44 SatU	0
45 ScatA	0
46 ScatJ	0
47 ScatU	0
48 SdalA	0
49 SdalI	0

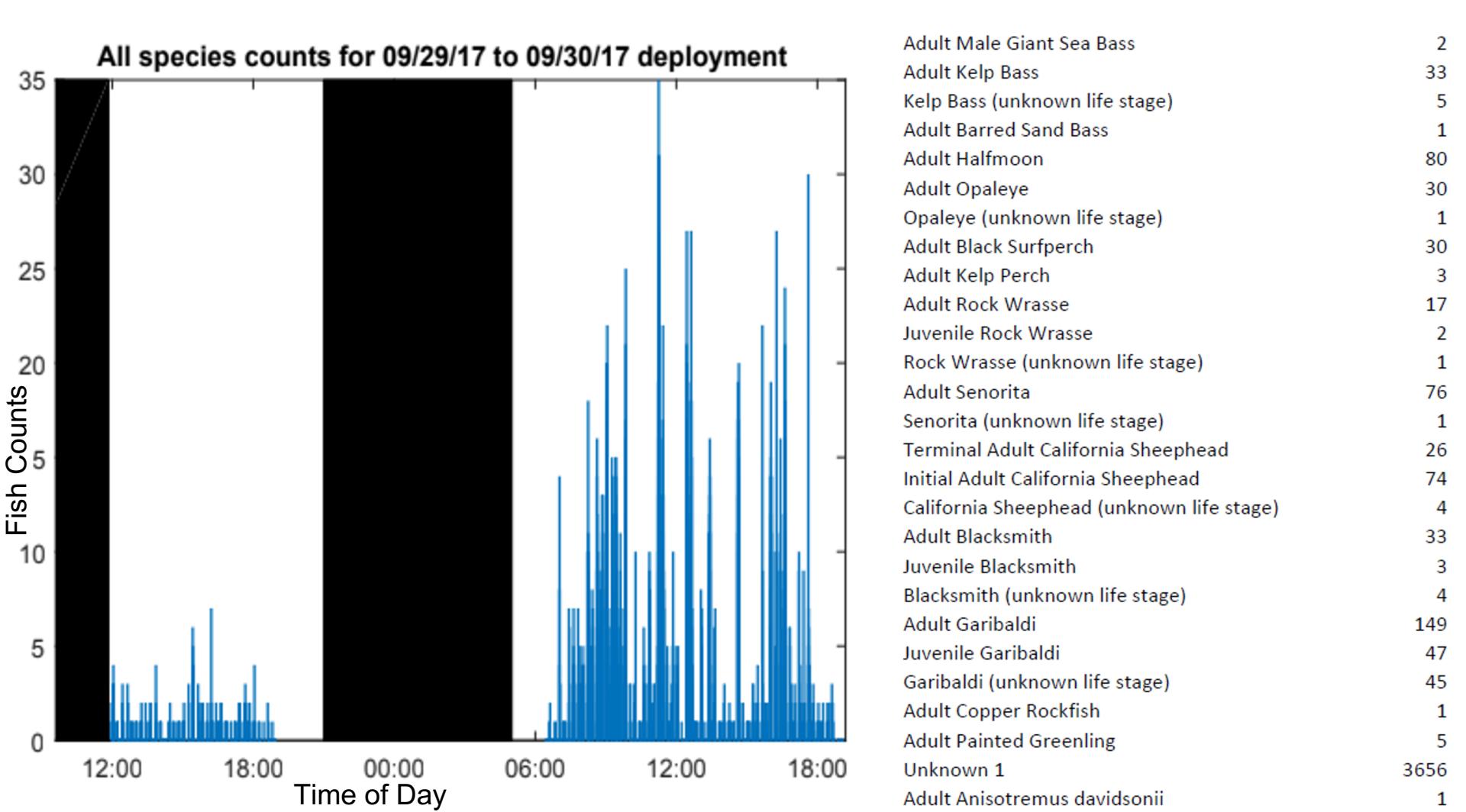
Comments

enter comments here



All species counts for 05/25/18 deployment





Outcomes

- Total # of Dives: 21
- Advancement of depth: Now eligible to dive between 60-100 feet
- Lead diver on three dives
- Greater knowledge of the species that reside within the kelp forest
- Giant sea bass spotting: 1!
- Total # of Pictures analyzed: ~3600



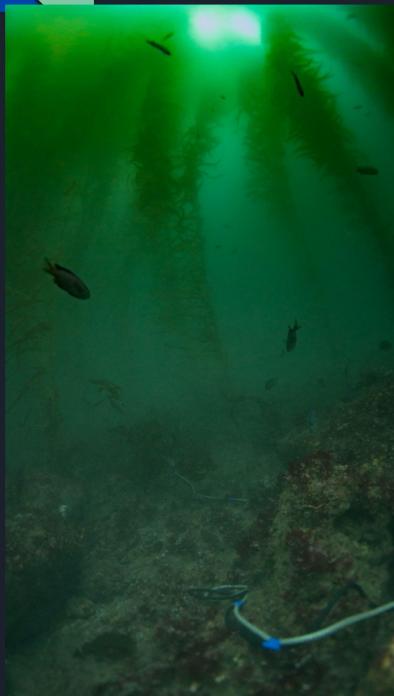


Objectives

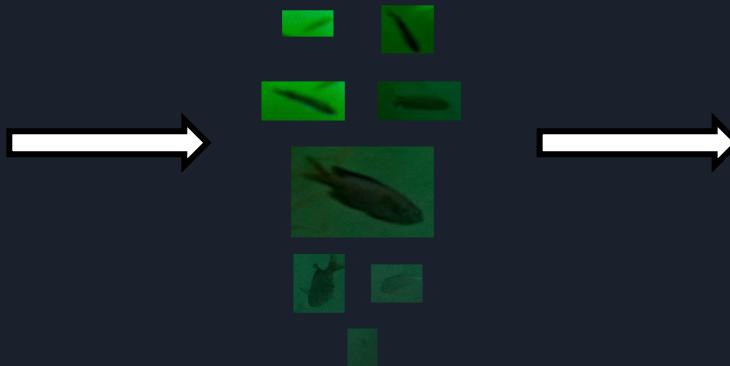
Helen

- Analyze camera data
- Automate the process of identifying fish

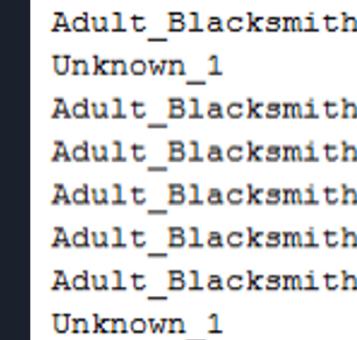
Automated processing



Raw image files



Object detection

A list of species classifications. The list consists of eight entries, each preceded by a small green square icon. The entries are:

- Adult_Blacksmit
- Unknown_1
- Adult_Blacksmit
- Adult_Blacksmit
- Adult_Blacksmit
- Adult_Blacksmit
- Adult_Blacksmit
- Unknown_1

Species classification

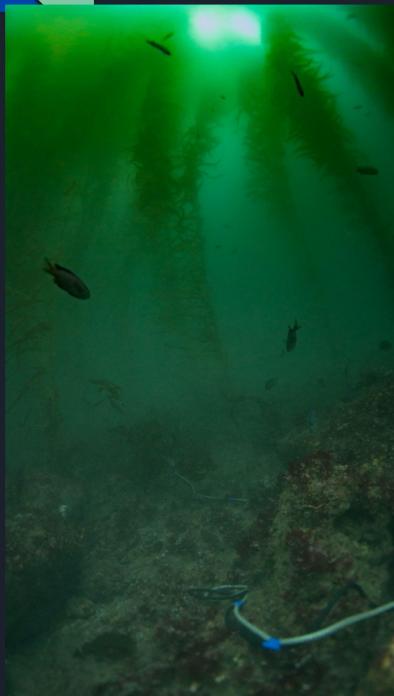
Automated processing



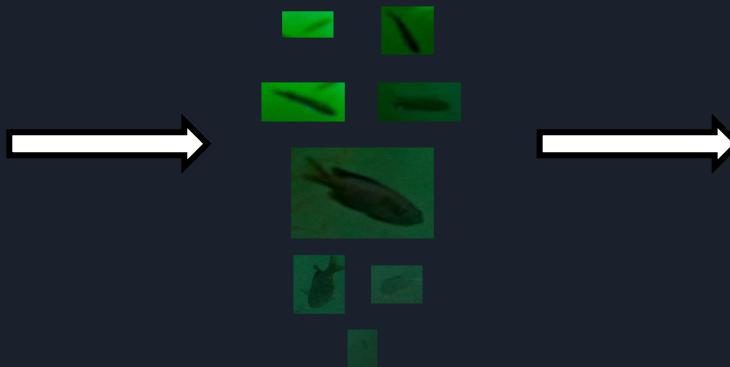
When there's 1500 pictures of fish to look thru



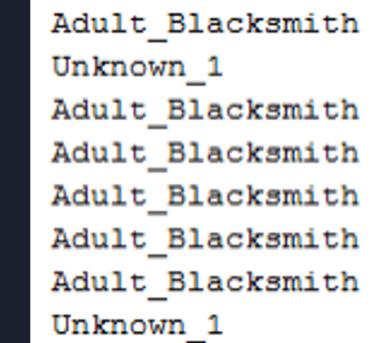
Automated processing



Raw image files



Object detection

A vertical list of text entries representing the species classification results. The entries alternate between "Adult_Blacksmit" (in blue) and "Unknown_1" (in red). There are eight entries in total, indicating multiple detections of Blacksmith fish and one unknown entry.

- Adult_Blacksmith
- Unknown_1
- Adult_Blacksmith
- Adult_Blacksmith
- Adult_Blacksmith
- Adult_Blacksmith
- Adult_Blacksmith
- Unknown_1

Species classification



An Open Source Framework for Underwater Image Processing

VIAME

Video and Image Analytics for a Marine Environment In cooperation with NOAA's Automated Image Analysis Strategic Initiative ([AIASI](#)), Kitware has developed VIAME, an open-source system for analysis of underwater video and imagery for fisheries stock assessment. VIAME will enable rapid, low-cost integration of new algorithmic modules, datasets and workflows.



[View Project on GitHub](#)

Why VIAME?

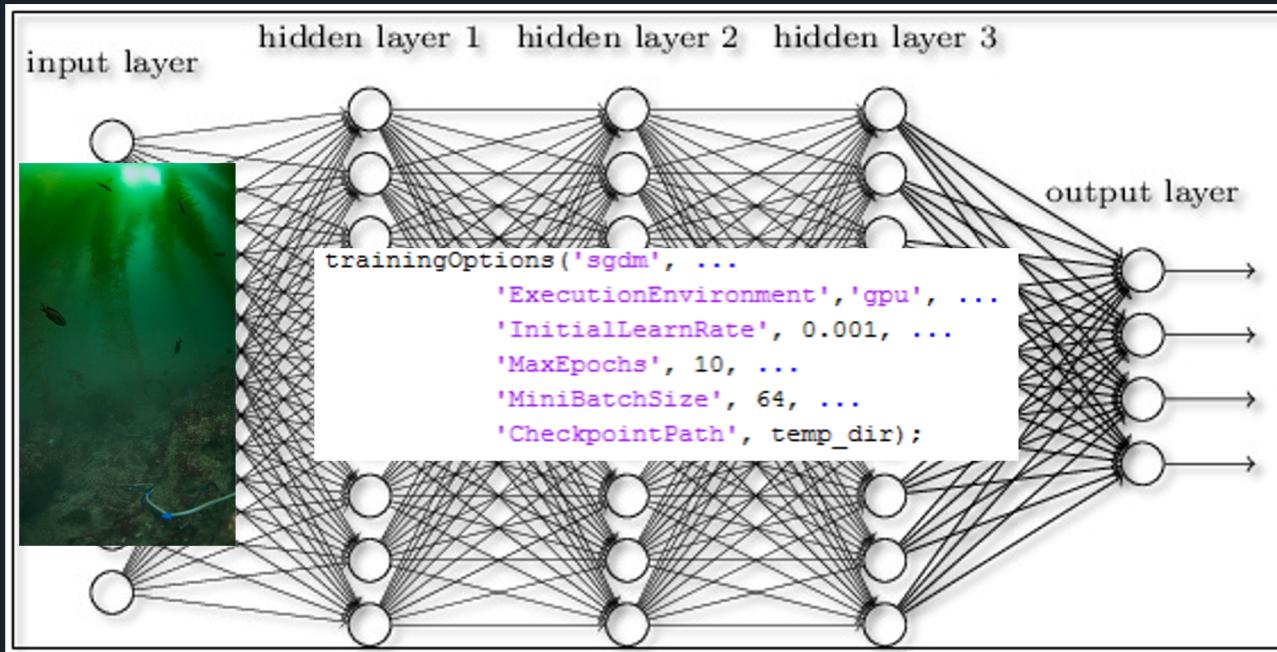
- **Modular Dataflow Architecture** enables straightforward integration of a wide variety of third party analytics, libraries and tools.
- **Open Source** basis fosters easy collaboration and transparent development for the widest possible community involvement.
- **Based on KWIVER**, the [Kitware Video and Image Exploitation and Retrieval toolkit](#), which includes a pipelined image processing framework, baseline algorithms, and wrappers around algorithms from the broader computer vision community.



VIAME

- Weren't able to build it on Windows or Linux
- Not guaranteed to work : not tested on our “type”
- Maybe in the future

Building from scratch



Convolutional neural networks (CNNs)



Building from scratch

1. Write the neural network
2. Train the neural networks on a known data set
(Kian and Addison)
3. Optimize performance & accuracy
4. Use on real data



Writing the neural network

- MATLAB: Neural Network Toolbox makes things very convenient
- AlexNet is pretrained on 1000+ categories
- Transfer learning: modifying existing neural networks to output new categories



Writing the neural network

Need to write two neural networks:

boxnet, for object detection

fishnet, for species classification

Training object classification

```
preds =  
  
297×1 categorical array  
  
velvet  
grey whale  
grey whale  
electric ray  
dugong  
dugong  
dugong  
dugong  
grey whale  
grey whale  
velvet  
grey whale  
grey whale  
grey whale  
great white shark  
dugong  
dugong  
grey whale  
sea snake  
electric ray  
sea snake  
grey whale  
grey whale  
grey whale  
tench  
rock beauty
```



```
preds =  
  
1147×1 categorical array  
  
Adult_Blacksmith  
Unknown_1  
Adult_Blacksmith  
Adult_Blacksmith  
Adult_Blacksmith  
Adult_Blacksmith  
Adult_Blacksmith  
Adult_1  
Adult_Blacksmith  
Adult_Blacksmith  
Adult_Blacksmith  
Adult Blacksmith
```

AlexNet

fishnet



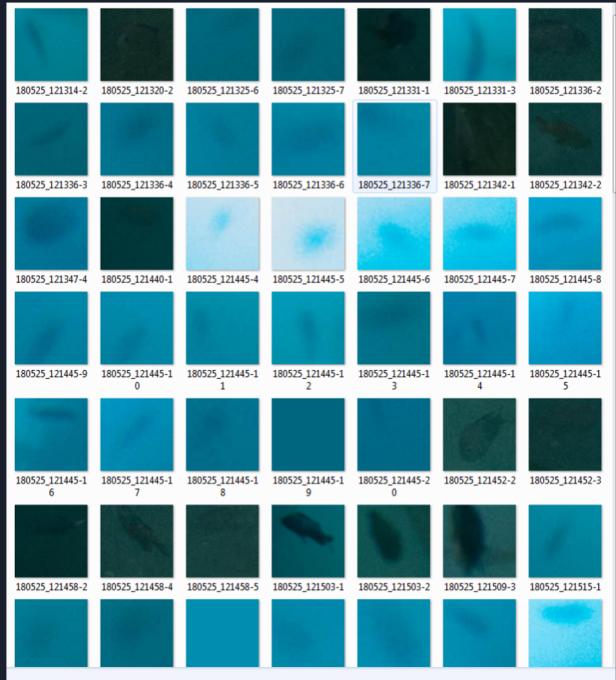
Training object classification (fishnet)

Split labeled images into different categories:

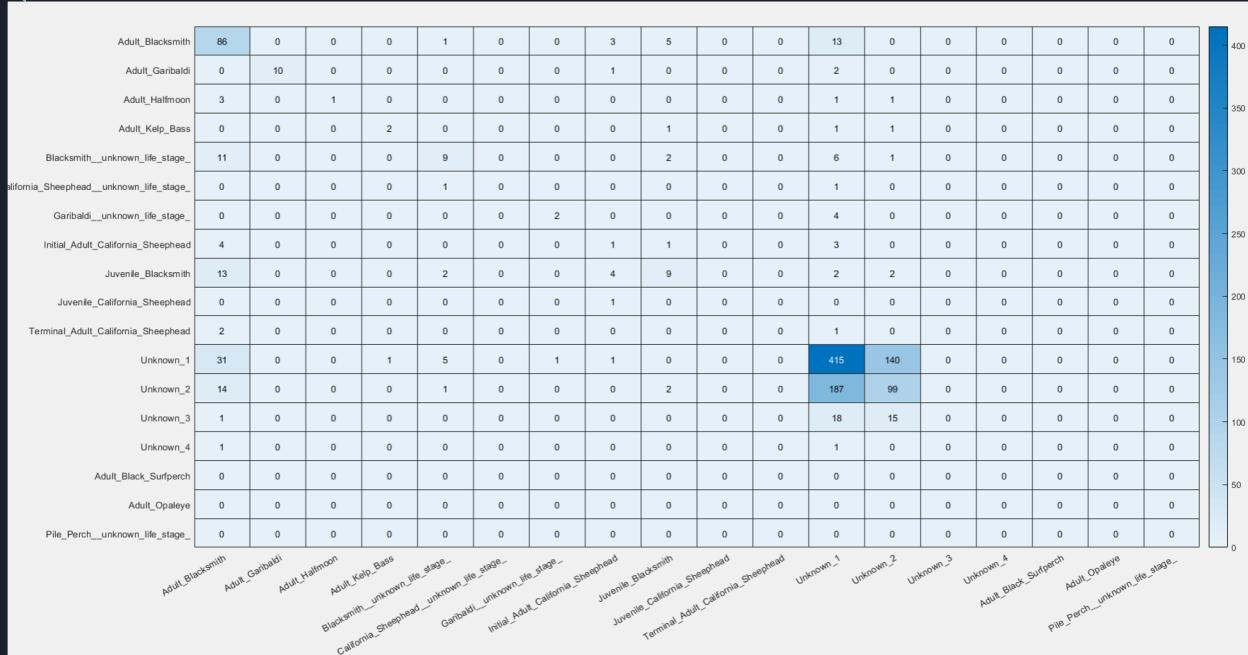
- training set
- validation (test) set

Training object classification (fishnet)

- Zoom in/out to standard size
- Make different labels based
 - on different groups
- More is more



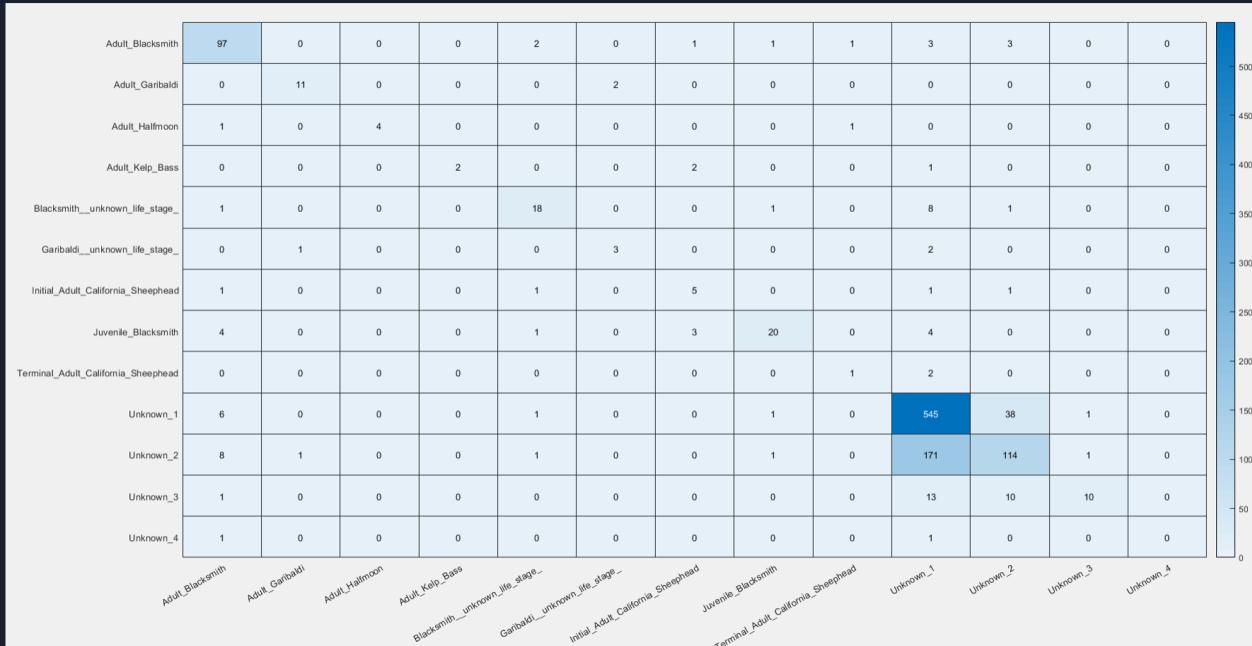
Optimizing performance



F1 = 0.3292

Confusion matrix

Optimizing performance

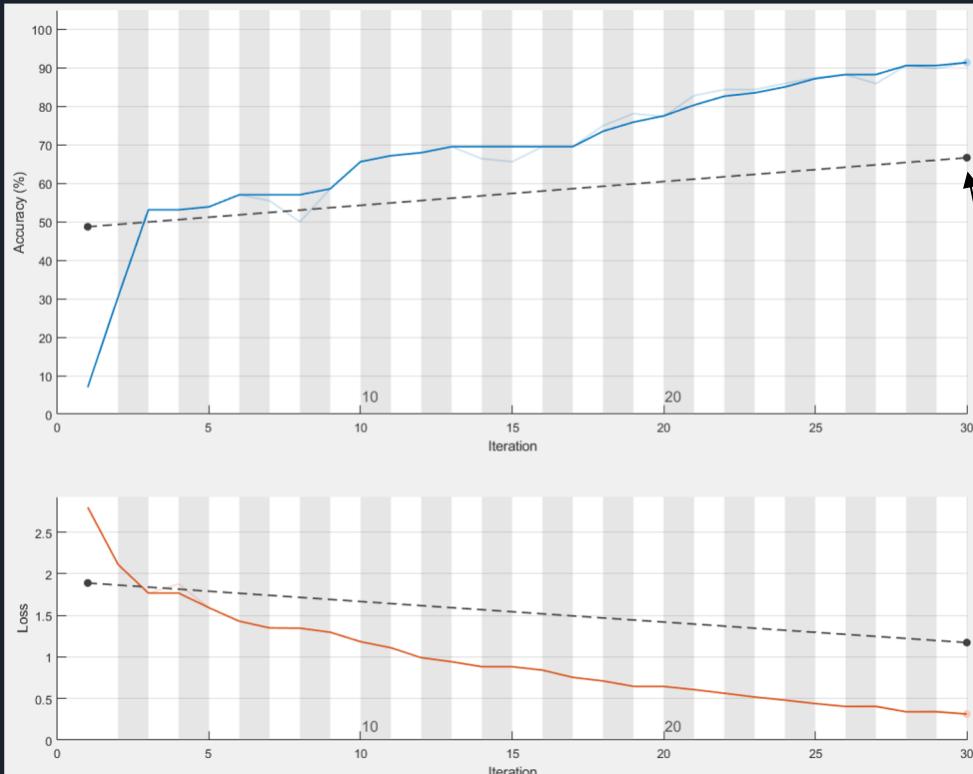


F1 = 0.6039

Optimizing performance



Optimizing performance



Training accuracy = 92%

Validation accuracy =
67%



Optimizing performance

- Increasing size of training set
- Data augmentation
- Adjusting layers of the neural network (*in progress*)

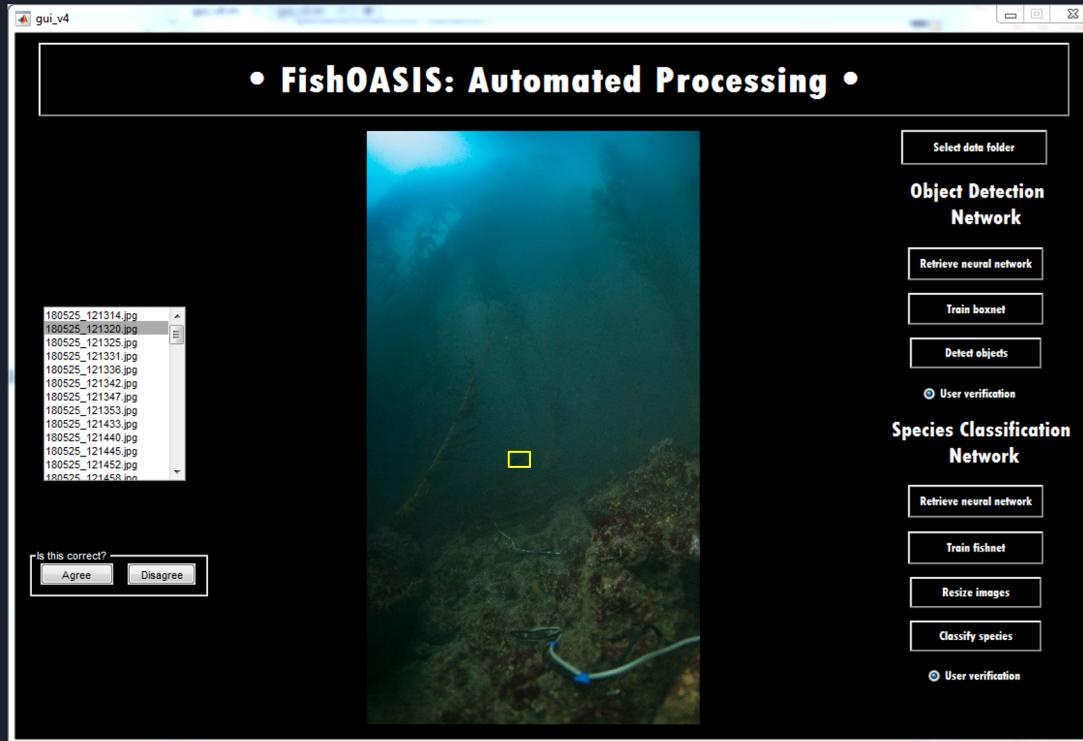


Training object detection (boxnet)

- Faster R-CNNs require a lot of memory and processing ability
- Our images are 2384 x 4240 pixels
- Strategically downsize & resample
- Use new computing resources:

Google Cloud machine with GPU

Use on real-world data



Use on real-world data





Use on real-world data

- Automated detection is at an impasse
- Object detection is still done by hand
- Species classification performs relatively well on novel data



Conclusion

- Lots of image analysis
- Lots of deployment and data collection



Big picture

- Identify and count fish in pictures
- Correlate image IDs with acoustic data
- Figure out what is in the MPA



Acknowledgements

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