

Video and Image Analytics for Marine Environments (VIAME) An Open-Source, Do-It-Yourself AI Toolkit

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NOAA: Benjamin Richards, Dvora Hart, George (Randy) Cutter, Elizabeth Clarke, Charles Thompson, Kresimir Williams, Bill Michaels, Erin Moreland, Katie Sweeney, Abigail Powell

CFF: Liese Siemann



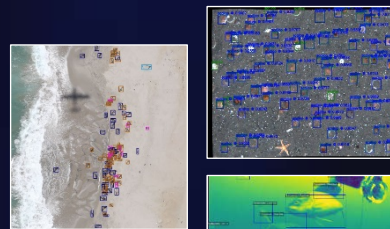
Kitware

What is VIAME?



- A do-it-yourself (DIY) AI toolkit which can be applied to multiple types of imagery or video
- Can be run by people with no programming or machine learning background in both web and desktop interfaces
- Released as fully open-source with a permissive license, see viametoolkit.org
- Specializations to maritime processing such as motion fusion, stereo measurement, image enhancement, and object tracking which other software (e.g. Amazon SageMaker) lack

Object Detection



Object Tracking

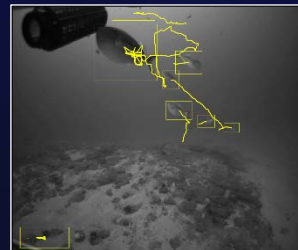
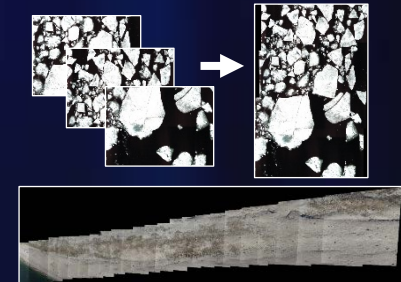


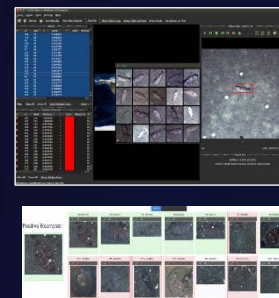
Image Enhancement



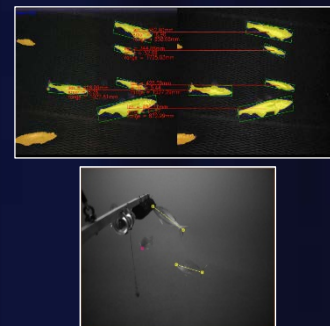
Image Registration and Mosaicing



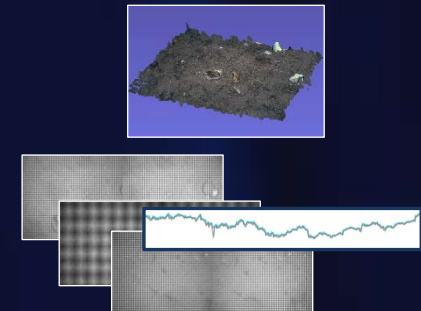
Video Search and Rapid Model Generation



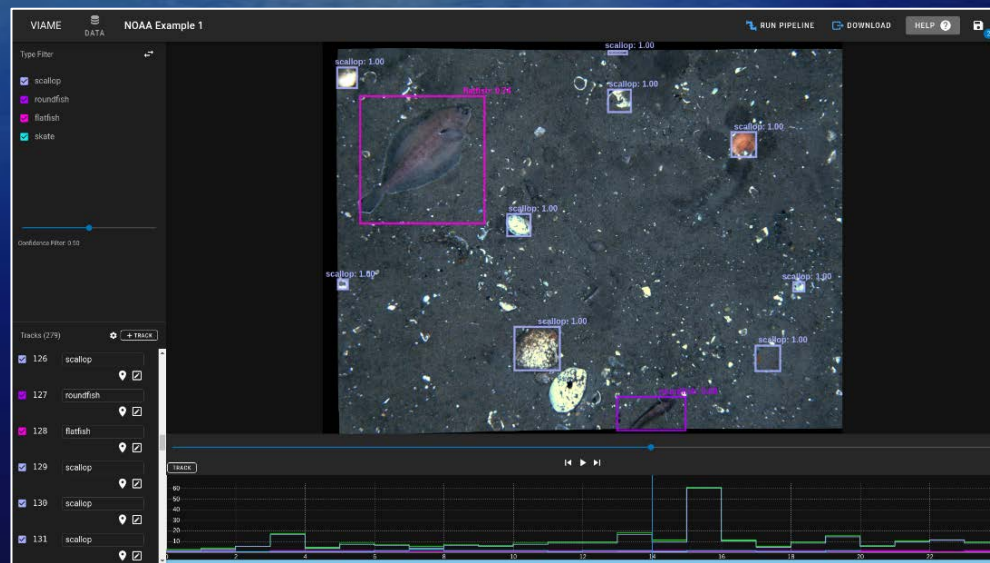
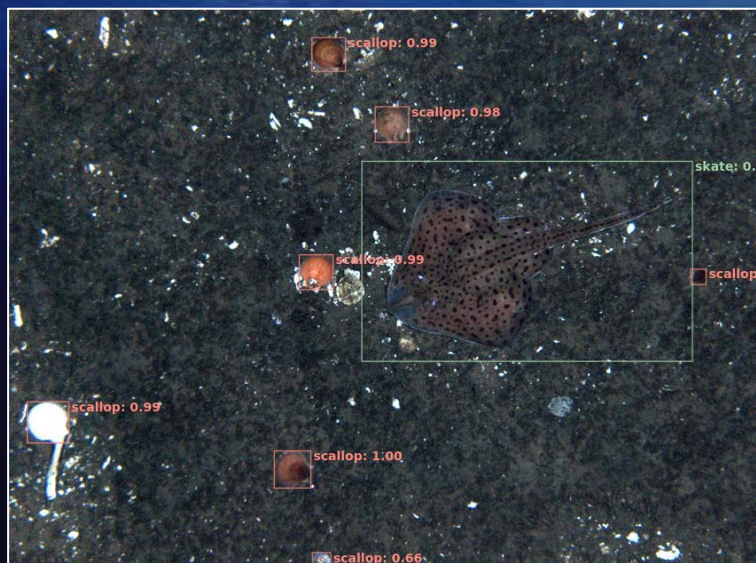
Stereo Measurement



Calibration, 3D and Altitude Estimation



Downward-Facing Underwater Cameras (AUVs, Trawls)



Source: HabCam (NEFSC, CFF, WHOI)



Source: NWFSC AUV Data



Source: NWFSC AUV Data



Source: UK P. Maximus Images

Aerial Surveys (Manned Fixed-Wings, UAVs)



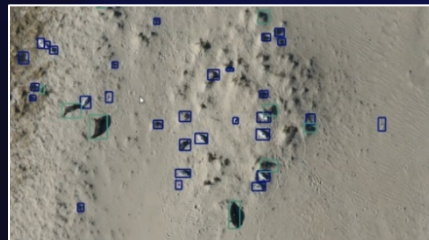
Source: AKFSC MML Stellar Sea Lion



Source: AKFSC MML Arctic Seal



Source: AKFSC MML Stellar Sea Lion



Source: NEFSC Gray Seal



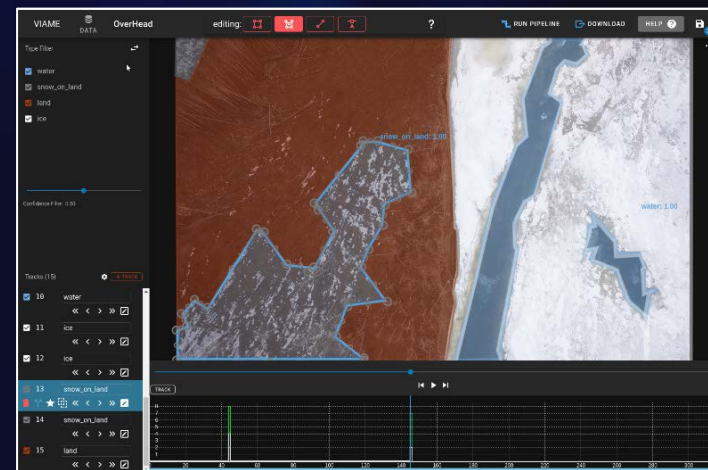
Source: SWFSC



Source: NEFSC Harbor Seal

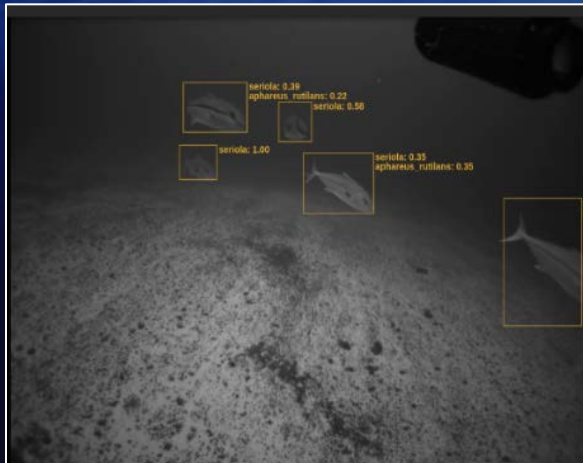


Source: SWFSC Penguin Aerial Data



Sources: University of Alaska Data
Scene Segmentation

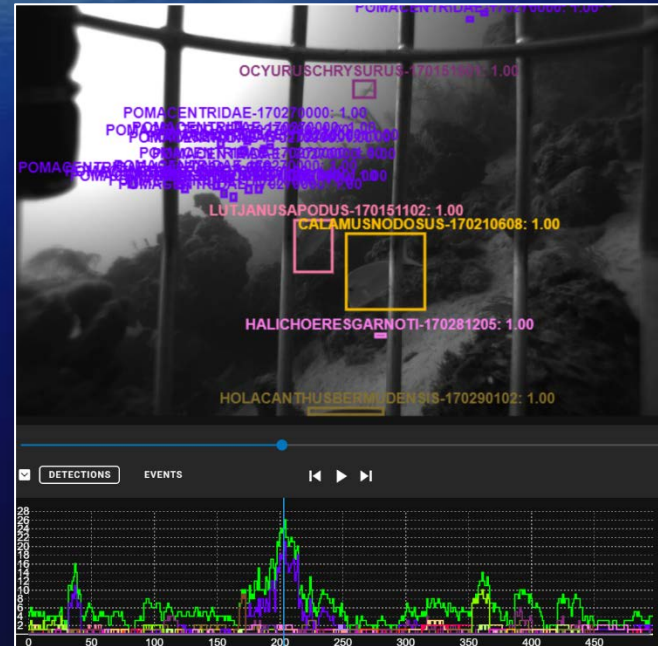
Outward Facing Cameras (Underwater, Ship-Based)



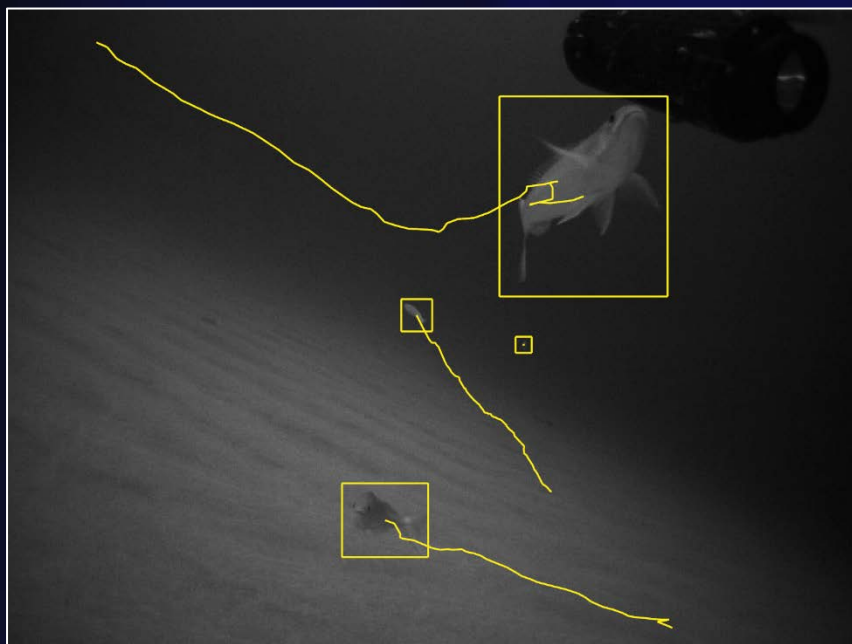
Source: PIFSC MOUSS



Source: SWFSC Penguin Cam



Source: SEFSC Quadcam

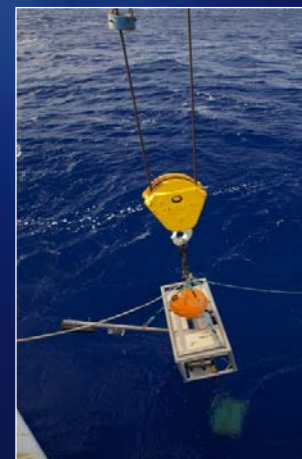
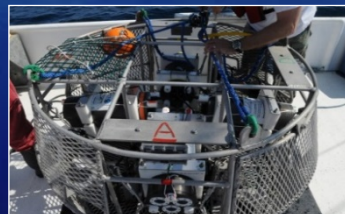
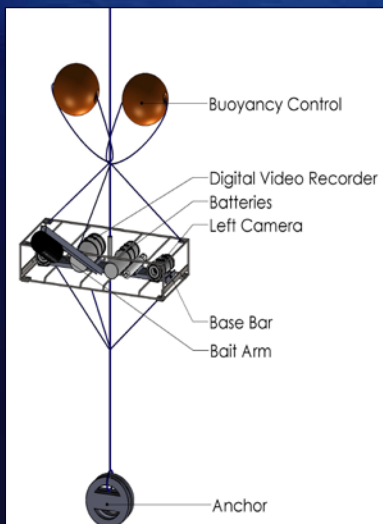


Source: PIFSC MOUSS



Source: PIFSC EM Data

Example Platforms

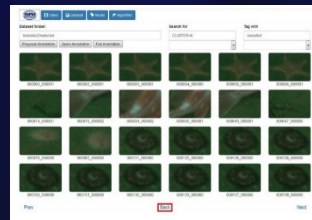
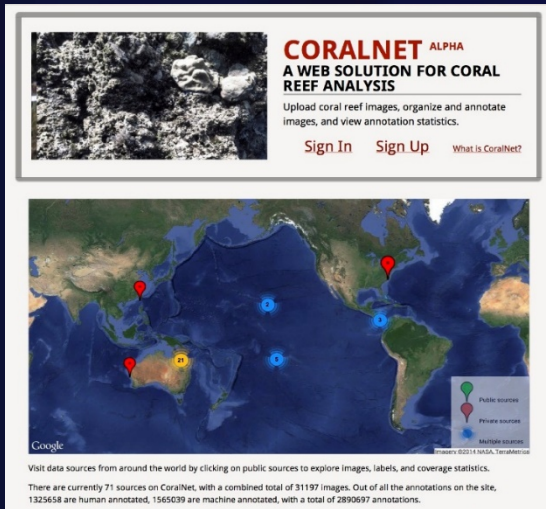


NMFS Strategic Initiative on Automated Image Analysis

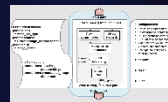
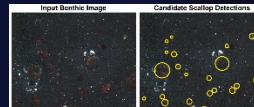
Mission: Develop guidelines, set priorities, and fund projects to develop broad-scale, standardized, and efficient automated analysis of still and video imagery for use in underwater stock assessment

Funded VIAME and CoralNet from 2015 to present

- Benjamin Richards (*chair*)
NOAA Pacific Islands Fisheries Science Center
- Alexandra Branzan Albu
University of Victoria
- Elizabeth Clarke
NOAA Northwest Fisheries Science Center
- George “Randy” Cutter
NOAA Southwest Fisheries Science Center
- Duane Edgington
Monterey Bay Aquarium Research Institute
- Dvora Hart
NOAA Northeast Fisheries Science Center
- Anthony Hoogs
Kitware, Inc.
- David Kriegman
University of California, San Diego
- Clay Kunz
Google
- Michael Piacentino
SRI International
- Lakshman Prasad
Los Alamos National Laboratory
- Charles Thompson
NOAA Southeast Fisheries Science Center
- Kresimir Williams
NOAA Alaska Fisheries Science Center



Flask, Michael Piacentino, SRI



Misc. Analytics (e.g. LANL)



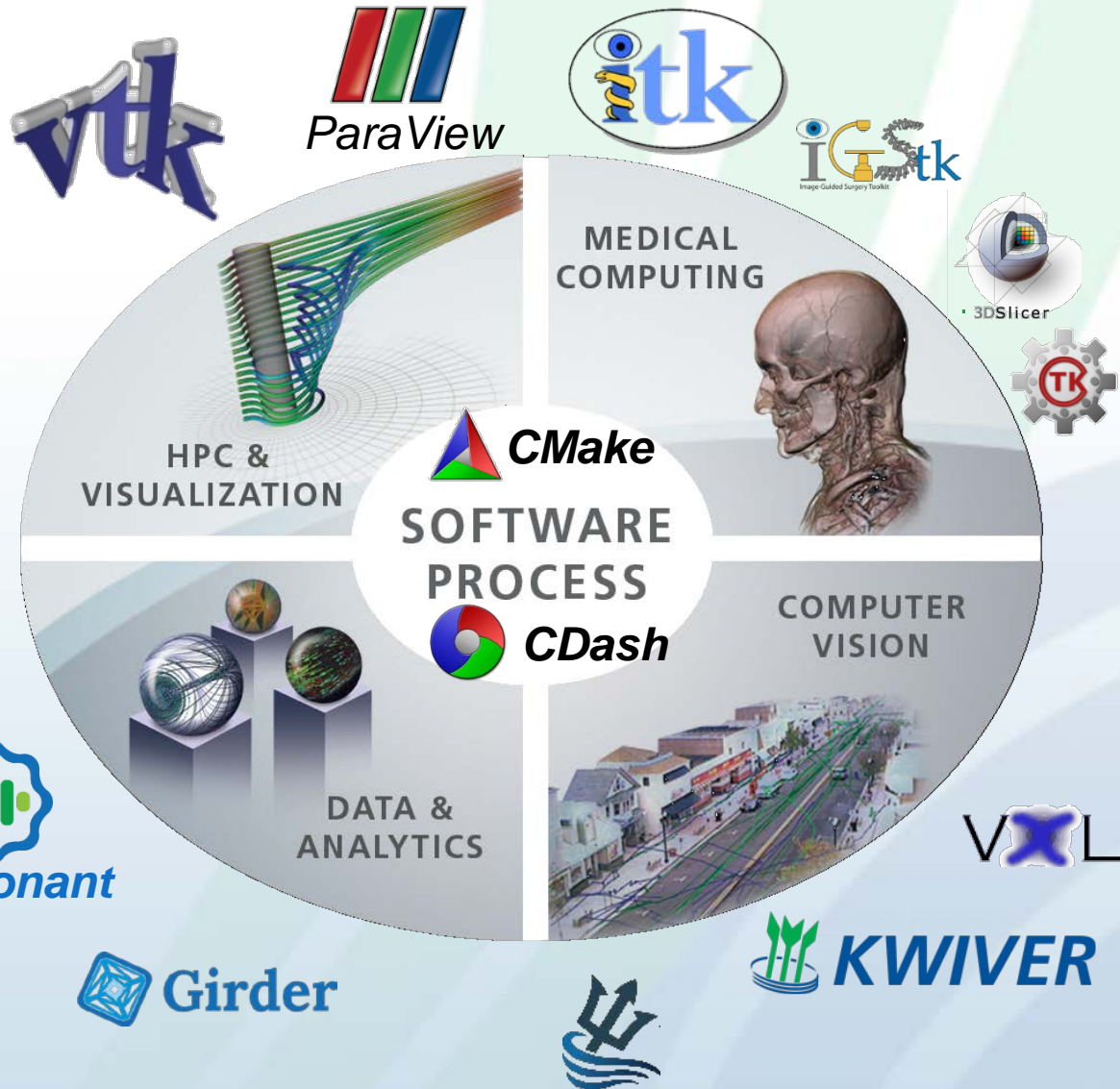
2019 Department of Commerce Gold Medal Awarded to NOAA Members of AIASI for VIAME and CoralNet

<http://coralnet.ucsd.edu>, D. Kriegman, UCSD

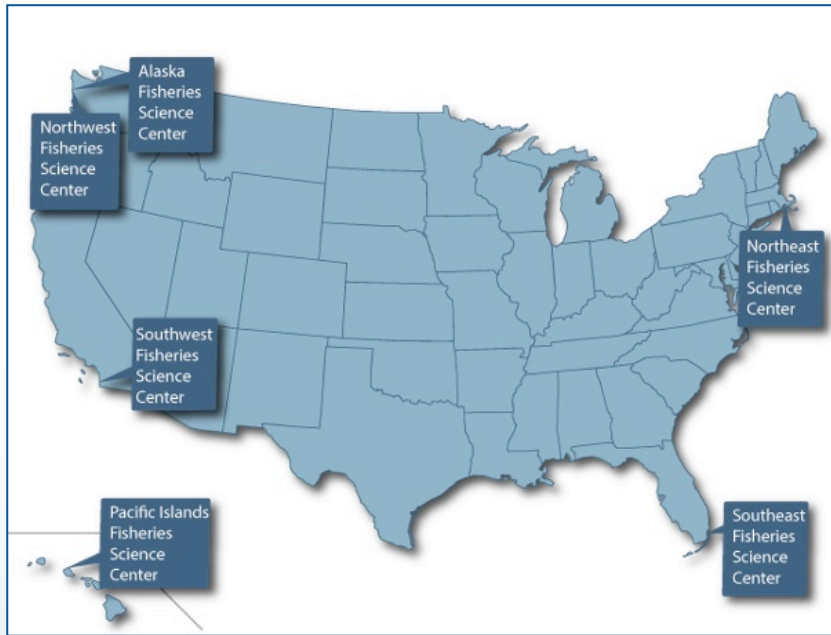
Other Funded Initiatives

Kitware

- Collaborative software R&D: algorithms & applications, image & data analysis, support & training
- Best known for open source toolkits and applications
- 150+ employees:
 - 1/3 masters
 - 1/3 PhD
- Founded in 1998
- Offices in Albany, NY; Chapel Hill, NC; Santa Fe, NM; Minneapolis, MN; Arlington, VA; Lyon, France

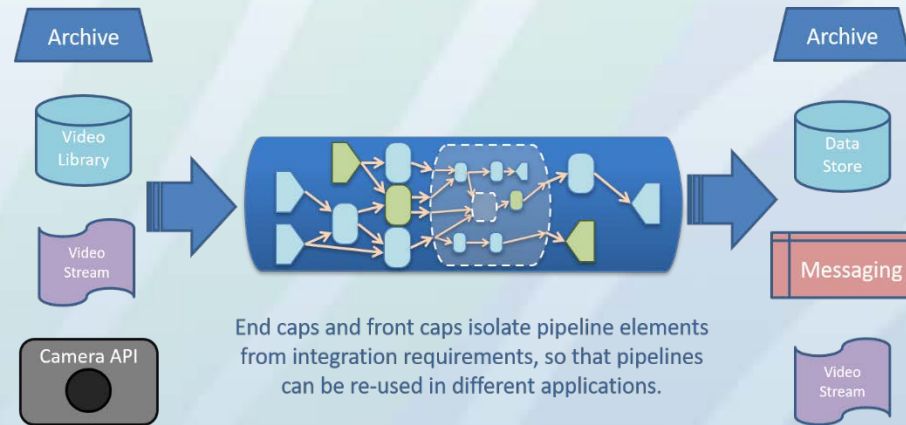
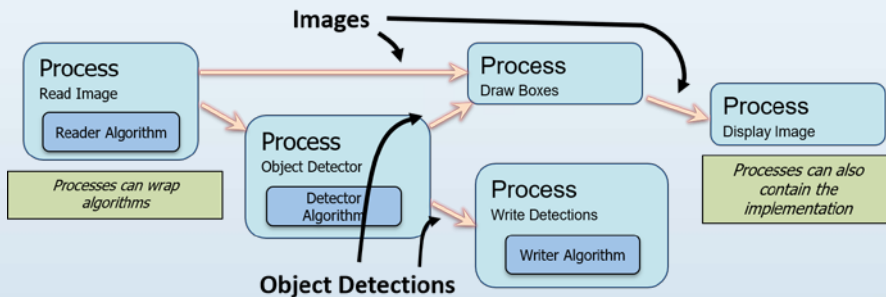


History – Algorithm Integration Platform



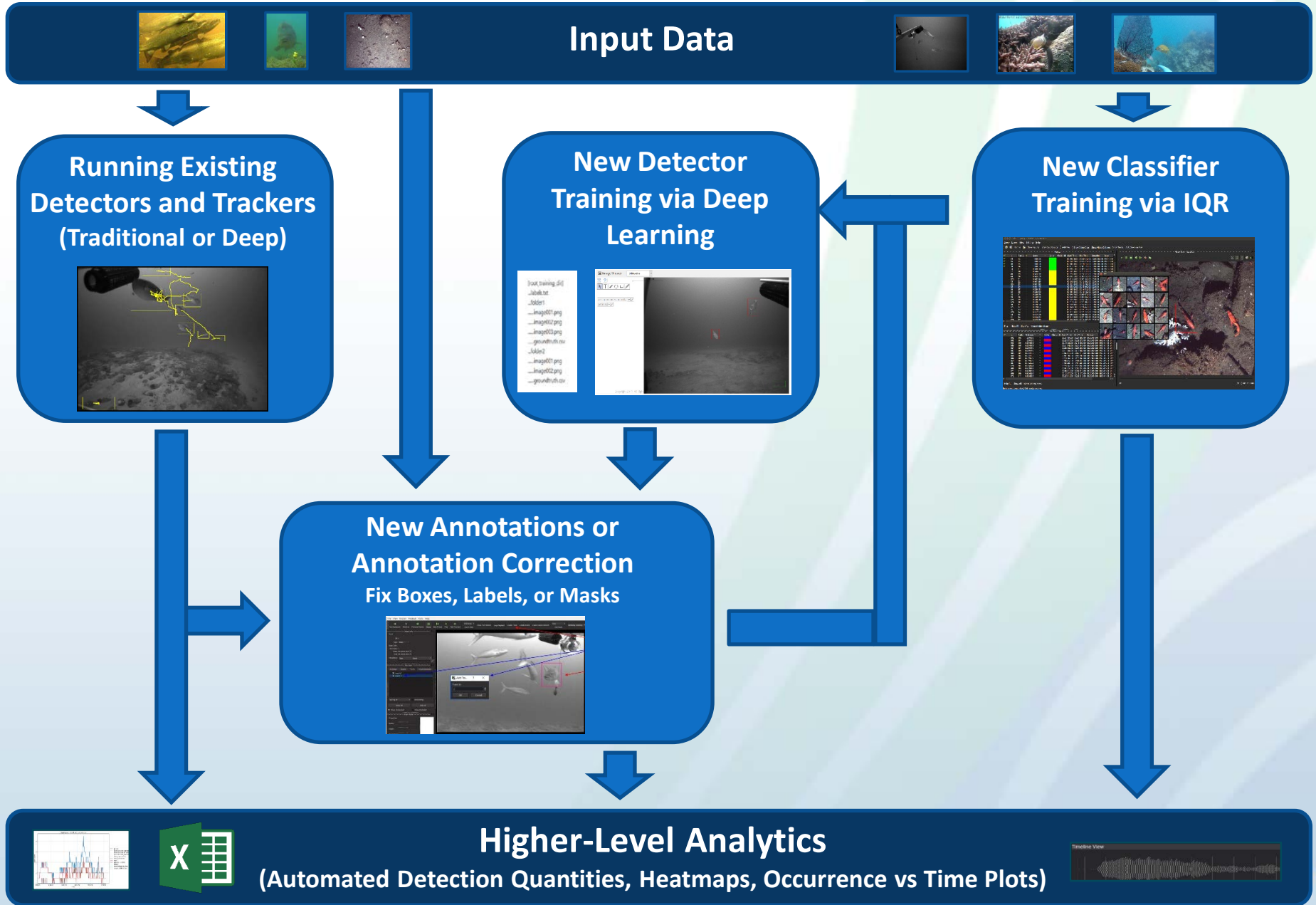
Capability	Primary data source	POC	Stereo calibration	Stereo processing	Video	Color, contrast correction	scallop detection	fish detection	fish length, sizing	fish tracking	fish classification	anomaly det.	habitat classification	image segmentation
NW SC CamTrawl	Cam Trawl	Williams	yes	yes	4 Hz	no, grayscale	yes	automatic	yes (UW student)	yes	yes			
ROV video fish detection and tracking	SWFSC ROV video	Cutter	no	no	30 Hz	yes	yes, DPM (UW)	no	manual	no	desired	desired		
ROV stereo fish measurement	SWFSC ROV GigE stereo	Cutter	yes	yes	2-4 Hz	yes	no	manual	no	no				
WHOI/NEFSC scallop detector	HABCAM towed rig	Dvora	yes	yes	no	yes	yes							
RPI/Kitware scallop detector	HABCAM towed rig	Hoogs	no	no	no	yes	yes							
SRI fish detection, classification, size	PI FSC MOUSS/BotCam	Ben/Mike	yes, acceptal files	yes	30 Hz	no, grayscale	yes			yes	yes			
SEFSC stereo proc	Drop cams from SEFSC	Thompson	yes	yes	yes		yes, basic background	manual	no	no				
Toyon SBIR I	Drop cams from SEFSC	Thompson	yes	yes	yes		yes, basic HOG	manual	yes	yes				
LANL segmentation and shape analysis	HABCAM towed rig	Lakshman	no	yes	no	no	yes	yes	no	no	yes (image)	yes	yes (polygonal)	
Toyon SBIR II	Still Images AUV, drop, towed	Clarke	yes	yes	no	yes (Hanu)	yes	yes	no	yes				
WHOI/NEFSC habitat classifier	HABCAM towed rig	Dvora	yes	yes	no	yes						yes	partially	
NWFSC clustering	AUV and MOUSS	Clarke	no	no	no							yes	partially	

Green well-implemented; quantified, comparative performance assessment; ready for integration
Yellow Existing implementation as mature research code; some performance quantification
Red preliminary research code with ongoing work against major problems
Gray idea or concept; no implementation

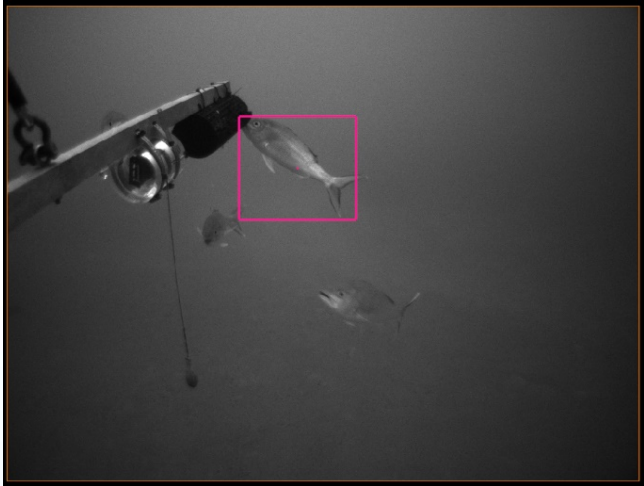


- Base classes for common operations (image filters, object detectors, trackers, ...)
- Derivation of base classes in C/C++, CUDA, Python, or Matlab
- Backend coded in C++ for efficiency, automatically multi-threaded

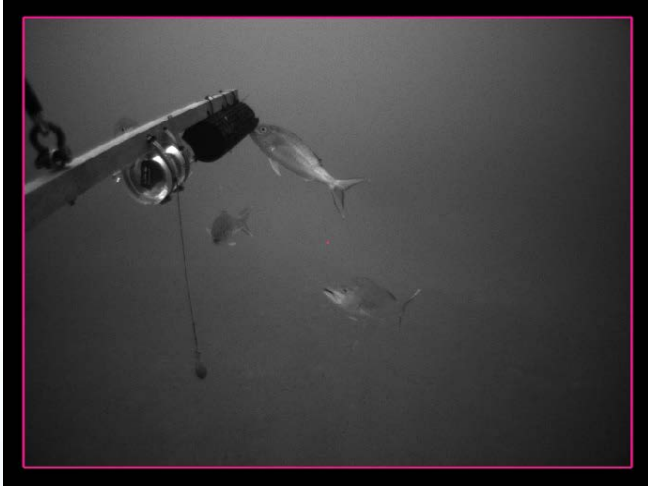
Three Detection Workflows



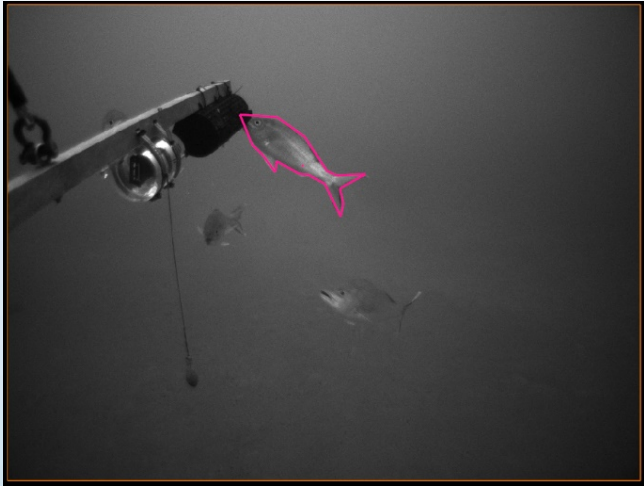
Types of Annotation and Detection Models



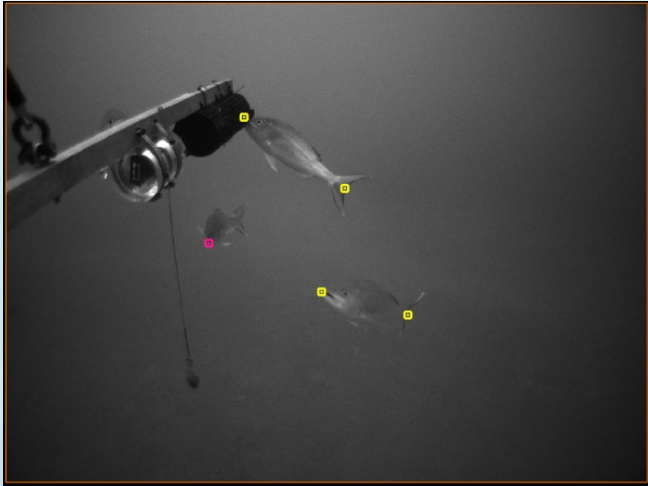
Box-Level



Frame-Level



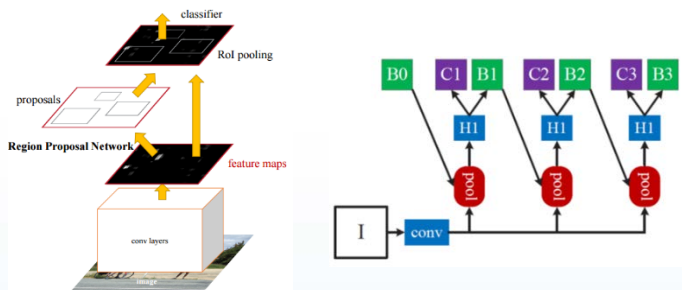
Pixel-Level



Keypoints

Baseline Object Detectors

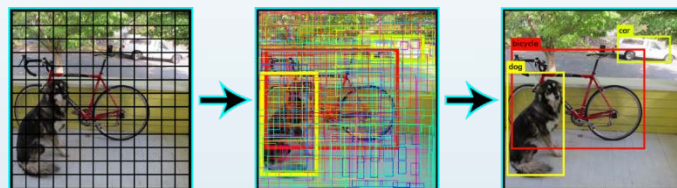
Cascade Faster R-CNN [1]



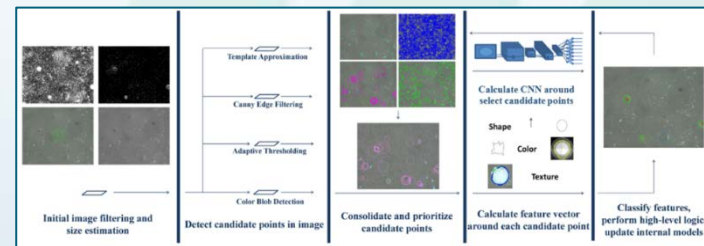
Cascade Mask Faster R-CNN [2]



YOLOv3 and v4 [3,4]



Scallop-TK



VIAME contains multiple baseline general purpose detectors from the larger computer vision community for wide applicability, but then specializations and other functionality added specific to domains of interest

[1] Cai, Zhaowei, et al. "Cascade R-CNN: Delving into High Quality Object Detection." CVPR 2018.

[2] Chen, Kai et al. "MMDetection: Open MMLab Detection Toolbox and Benchmark." arXiv preprint 2020.

[3] Redmon, Joseph, and Ali Farhadi. "YOLOv3: An Incremental Improvement." arXiv preprint 2018.

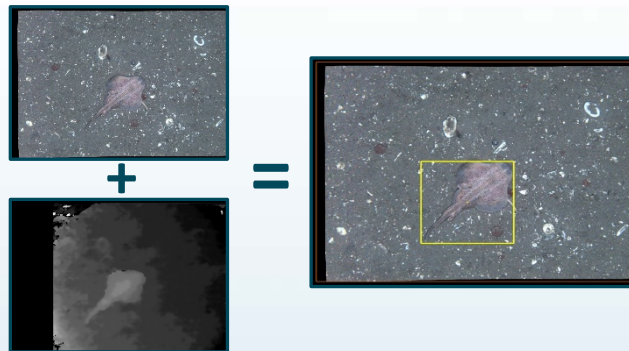
[4] Bochkovskiy, Alexey et al. "YOLOv4: Optimal Speed and Accuracy of Object Detection." arXiv preprint 2020.

Automatic Parameter Optimizations

- Automatically handle LR stepping based on validation loss, running multiple hyperparameter sets, and early stopping criteria
 - NetHarn: <https://gitlab.kitware.com/computer-vision/netharn>
- Automatically choose whether to grid detectors over image
- Utilities to turn dot annotation into boxes

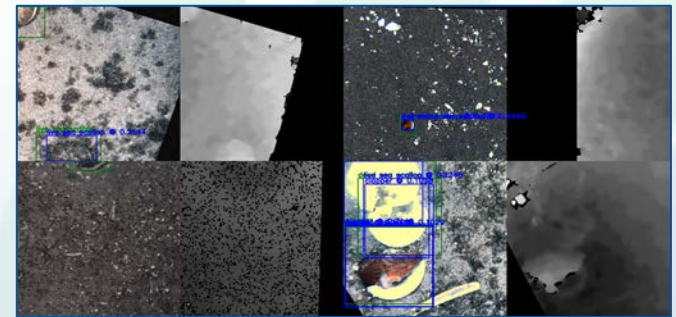
Auxiliary Data Fusion

Fuse depth and motion maps into object detectors



Extra Augmentation

Selectively augment channels differently



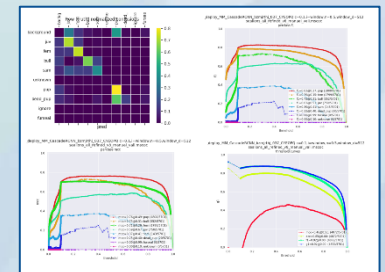
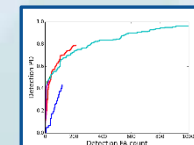
Ensemble Classifiers

Fuse output of detectors from different frameworks



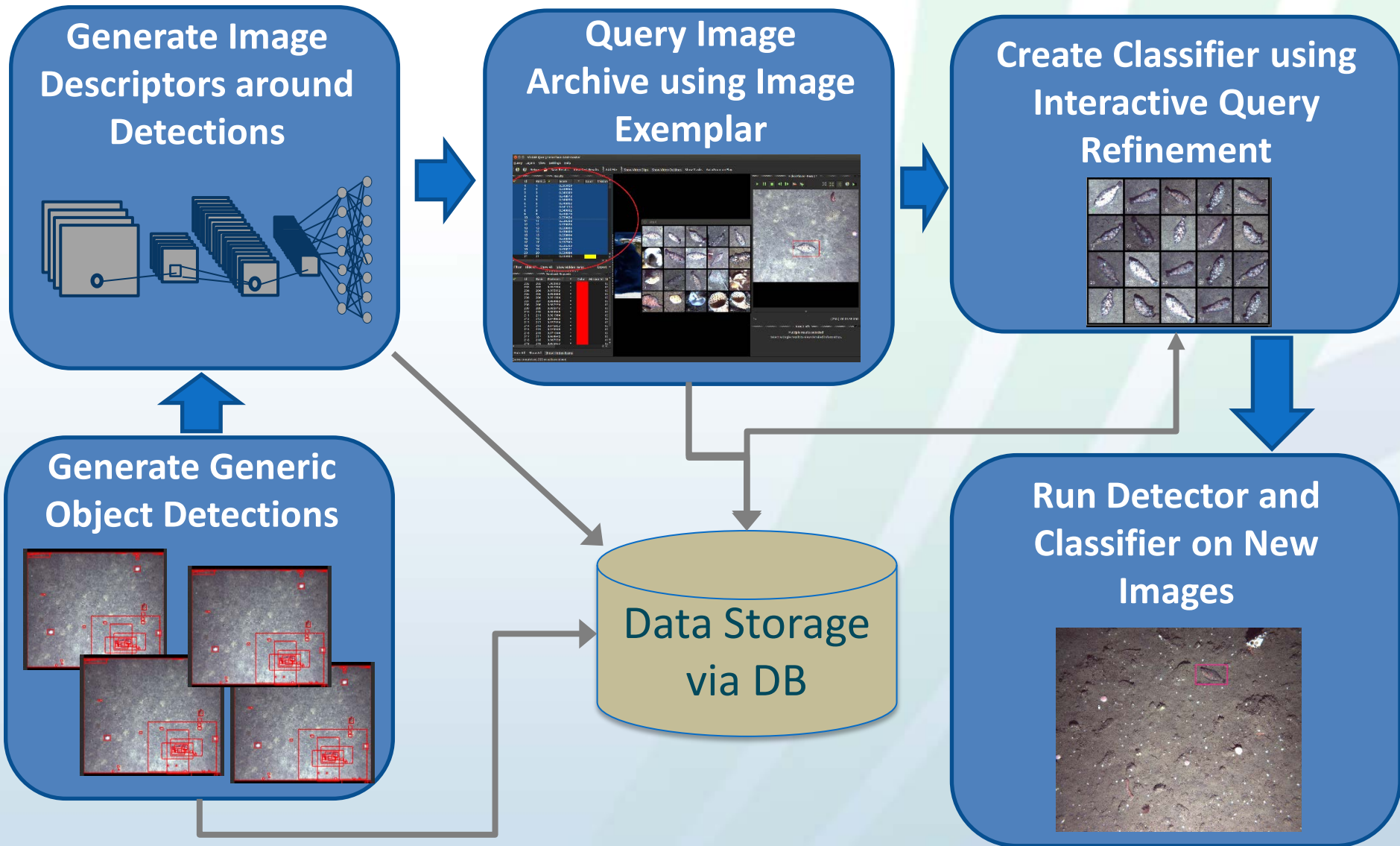
Scoring Utilities

Evaluation via ROCs, PRCs, track metrics, ...



```
Detection-Pd: 0.791209
Detection-FA: 213
Detection-PFA: 0.515738
Frame-NFAR: not computed
Track-Pd: 0.791209
Track-FA: 213
Computed-track-PFA: 0.515738
Track-NFAR: not computed
Avg track (cont., purity): 1.34, 1
Avg target (cont., purity): 1.47, 0.79
Track-frame-precision: 0.5
DEPTH-Hash: "a2123ede"
```

Interactive Search and Rapid Model Generation



Interactive Search and Rapid Model Generation

The screenshot displays the VisGUI Query Interface 2.0.0-master. The interface includes a menu bar (Query, Layers, View, Settings, Help), a toolbar with buttons for Refine, Save Results, View Best Results, Add File, Show Video Clips, Show Video Outlines, Show Tracks, and AutoZoom on Play. A video player window shows a video of a dark, cratered surface with a red bounding box around a specific feature. Below the video player is a grid of 20 thumbnails, with the 19th thumbnail highlighted in red. On the left, there are two tables of search results. The top table has columns for Id, Rank, Score, Color, Mission Id, and Start. The bottom table has columns for Id, Rank, Preference, Color, Mission Id, and Start. Both tables show a list of results with a yellow and a blue color column respectively.

Id	Rank	Score	Color	Mission Id	Start
2	2	0.601095	Yellow		00:16
3	3	0.599851	Yellow		00:02
4	4	0.592139	Yellow		00:06
5	5	0.586647	Yellow		00:03
6	6	0.583385	Yellow		00:15
7	7	0.580398	Yellow		00:06
8	8	0.578843	Yellow		00:03
9	9	0.576041	Yellow		00:05
10	10	0.574787	Yellow		00:07
11	11	0.570809	Yellow		00:06
12	12	0.568628	Yellow		00:24
13	13	0.566879	Yellow		00:07
14	14	0.566052	Yellow		00:00
15	15	0.565552	Yellow		00:06
16	16	0.565048	Yellow		00:07
17	17	0.564532	Yellow		00:07
18	18	0.563615	Yellow		00:06
19	19	0.561714	Yellow		00:04
20	20	0.561407	Yellow		00:06

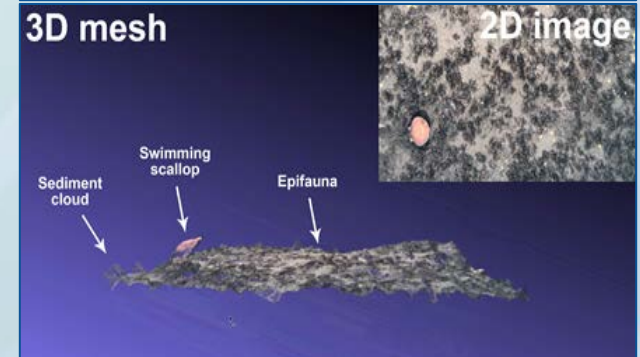
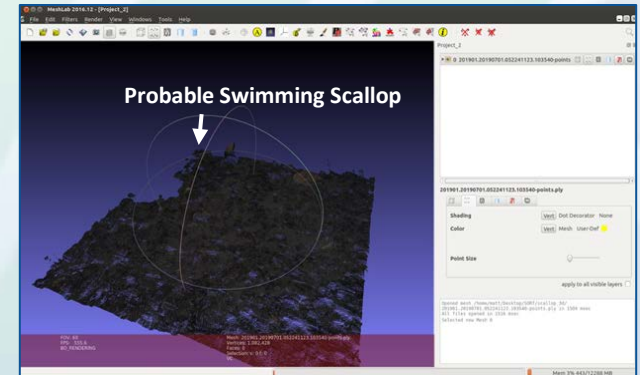
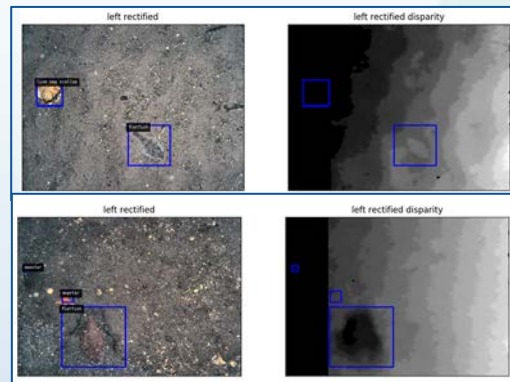
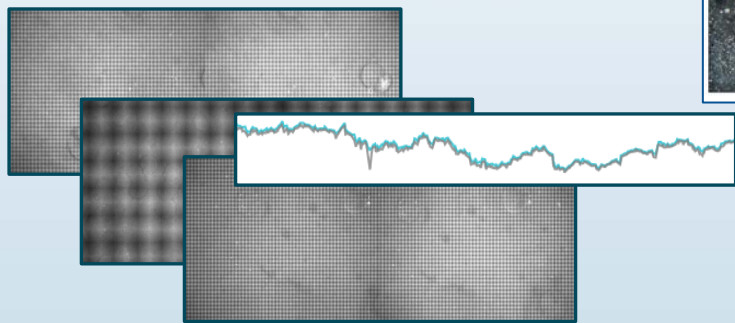
Id	Rank	Preference	Color	Mission Id	Start
202	202	1.000000	Blue		00:03
203	203	0.997996	Blue		00:29
204	204	0.995992	Blue		00:13
205	205	0.993988	Blue		00:14
206	206	0.991984	Blue		00:25
207	207	0.989980	Blue		00:12
208	208	0.987976	Blue		00:23
209	209	0.985972	Blue		00:14
210	210	0.983968	Blue		00:06
211	211	0.981964	Blue		00:23
212	212	0.979960	Blue		00:05
213	213	0.977956	Blue		00:23
214	214	0.975952	Blue		00:16
215	215	0.973948	Blue		00:13
216	216	0.971944	Blue		00:20
217	217	0.969940	Blue		00:13
218	218	0.967936	Blue		00:23

User provides initial image query:



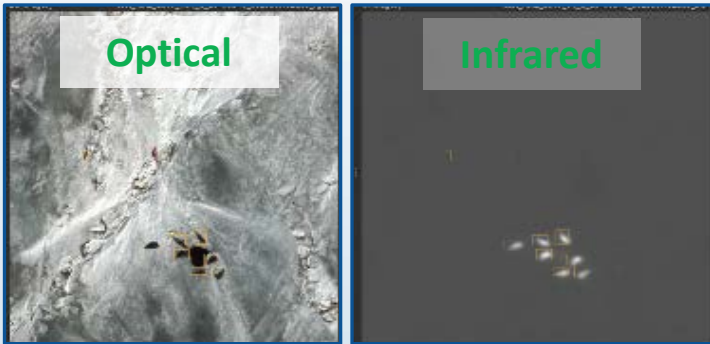
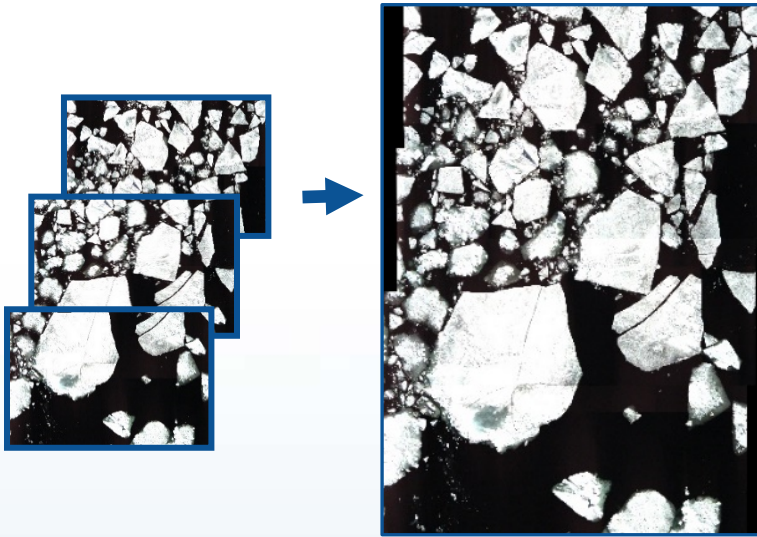
User corrects system returns on subsequent iterations through iterative query refinement (IQR)

Enhancement, Calibration and Depth Estimation

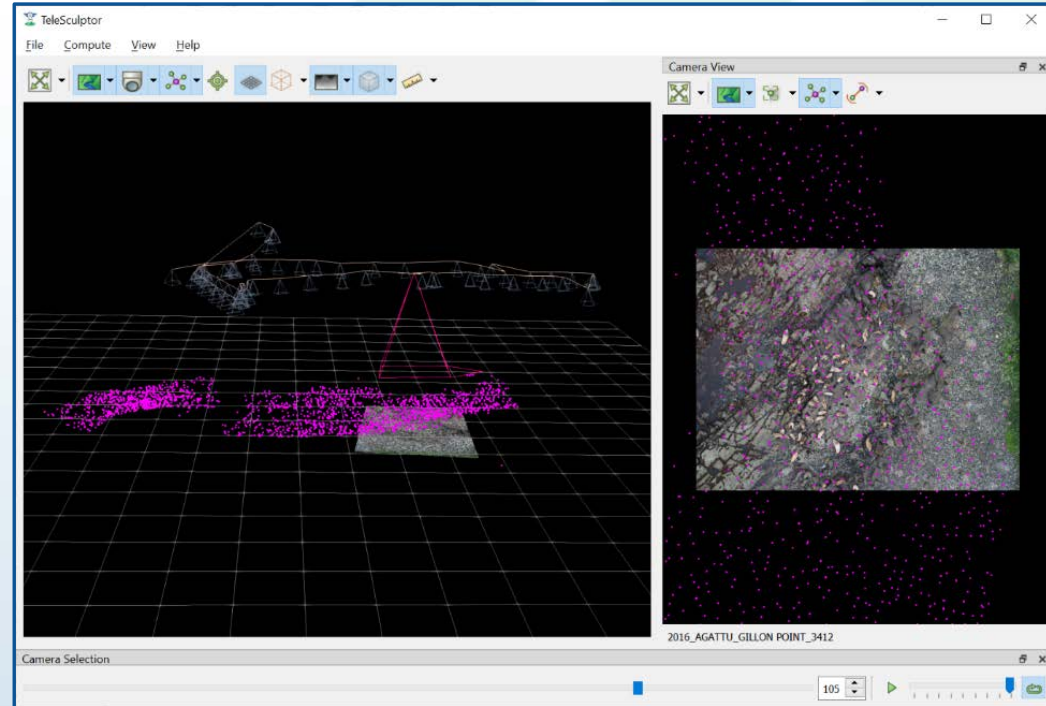


Work performed in conjunction with
Coonamesett Farm Foundation (CFF) and NEFSC

Registration and Mosaicing



Multi-Camera and Multi-Modality Registration



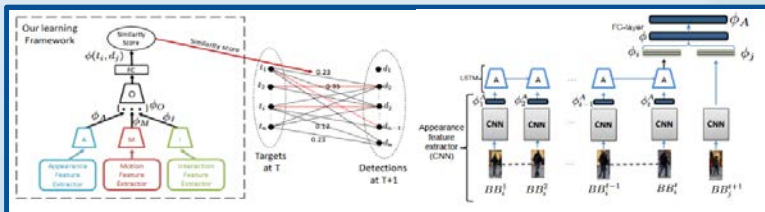
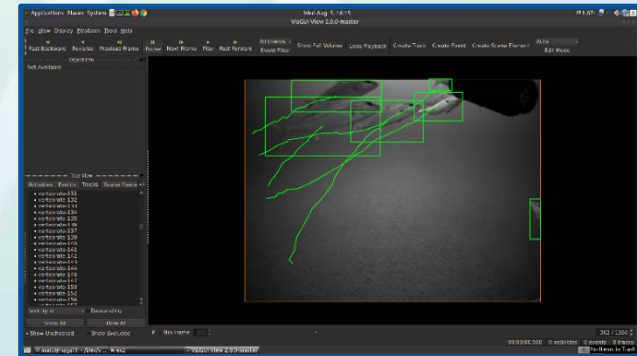
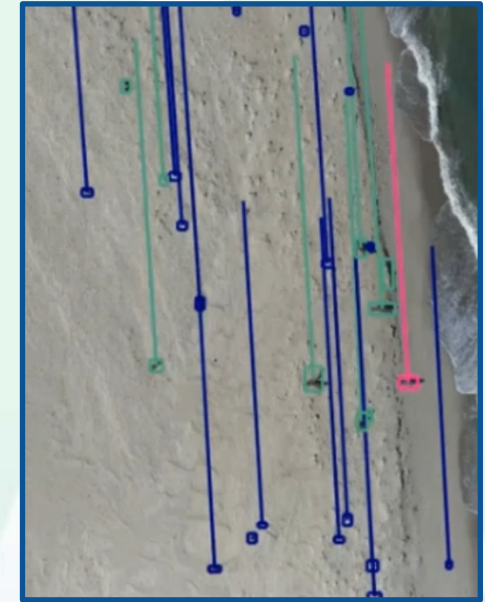
Above Source: TeleSculptor Open-Source 3D Geometry,
<https://github.com/Kitware/TeleSculptor>
Contact: matt.leotta@kitware.com

Work performed in conjunction with
NOAA AKFSC Marine Mammals Lab

Object Tracking

Have integrated multiple trackers into VIAME:

- Registration-Only Based (Aerial Pinniped)
- Deep Learning Based LSTM Tracker Detector Linkers [5]
- Non-Deep Learning Kalman Filter Detector Linkers
- Single Target Trackers (e.g. SiamRPN++ [6]) for annotation assist (below)

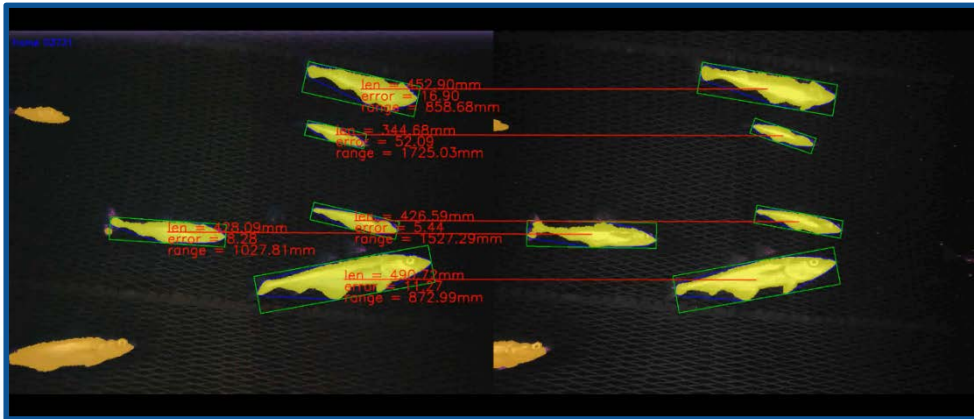


Code developed in conjunction with Air Force Research Laboratory (PRR# 88ABW-2019-4904, Distribution Statement A - Approved for public release: distribution unlimited).

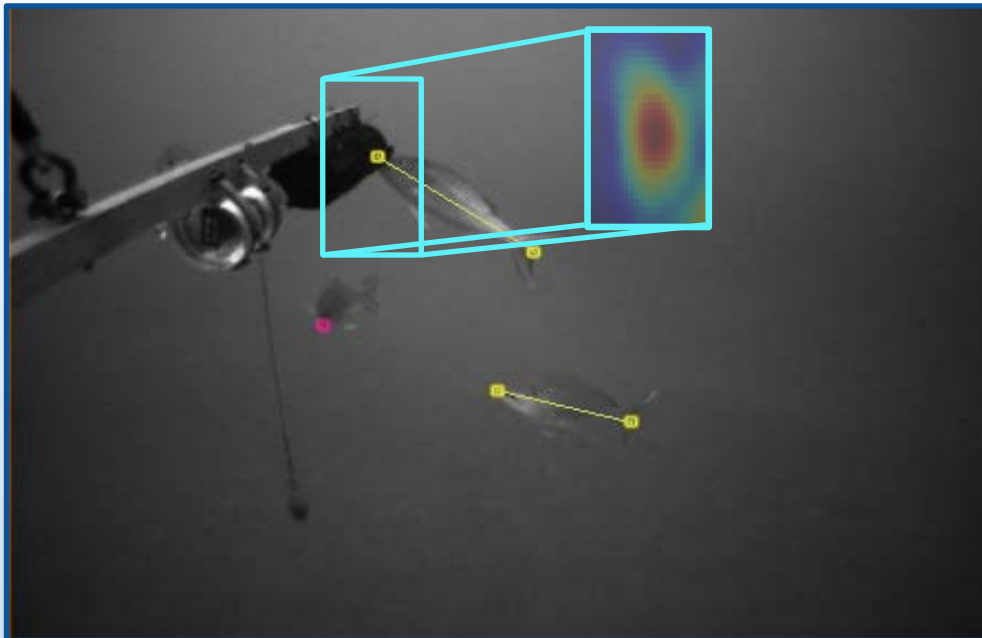
[5] Sadehian, Amir, et al. "Tracking the untrackable: Learning to track multiple cues with long-term dependencies." CVPR 2017.

[6] Li, Bo, et al. "SiamRPN++: Evolution of siamese visual tracking with very deep networks." CVPR 2019.

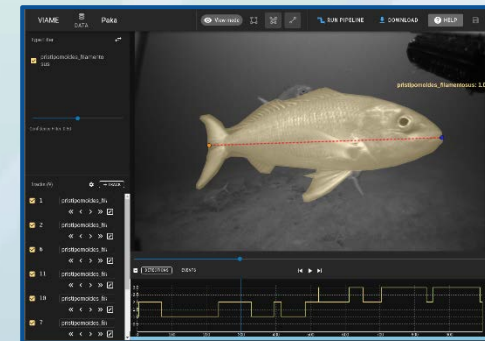
Stereo Measurement



Method #1 - Modeled off camtrawl process (Williams et al 2010), python port of matlab code. Use centroids of smaller sides of oriented bounding boxes as head/tail positions.

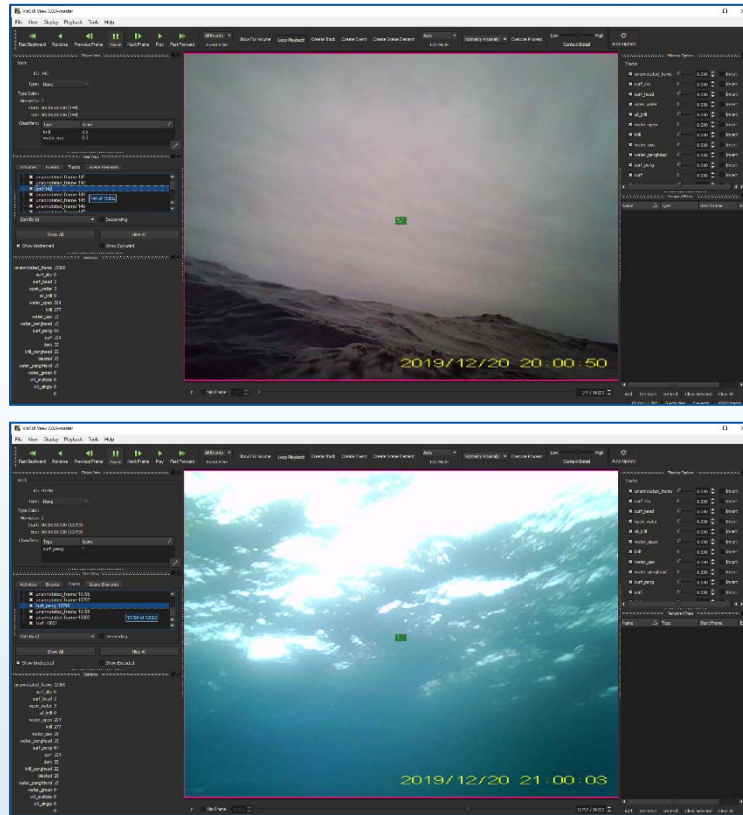


Method #2 - Feature point detection using dedicated CNN keypoint detectors, either in the same network or separate dedicated network (e.g. heavily modified version of [7]). Ongoing: collecting additional annotations.



Full Frame Classification

```
unannotated_frame 10000
surf_sky 0
surf_head 2
open_water 3
all_krill 0
water_open 204
krill 277
water_uso 16
water_penghead 15
surf_peng 64
surf 104
dark 55
krill_penghead 22
blasted 25
water_pengfriend 15
```



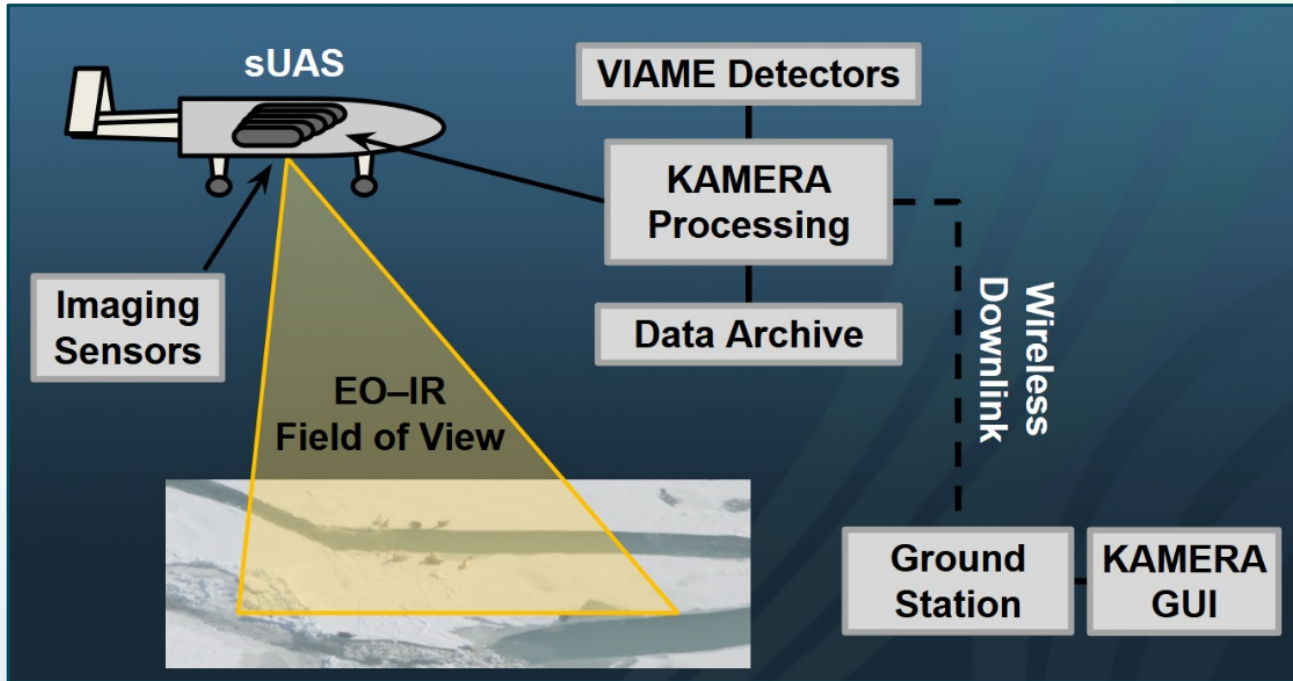
Substrate = X

Substrate = Y

Method #1 – Typical deep training pipeline (ResNet50 [8] – better for cases that have a lot of manual groundtruth)

Method #2 – SVM on fixed feature vector, similar to image search and rapid model generation pipeline (better for less training samples)

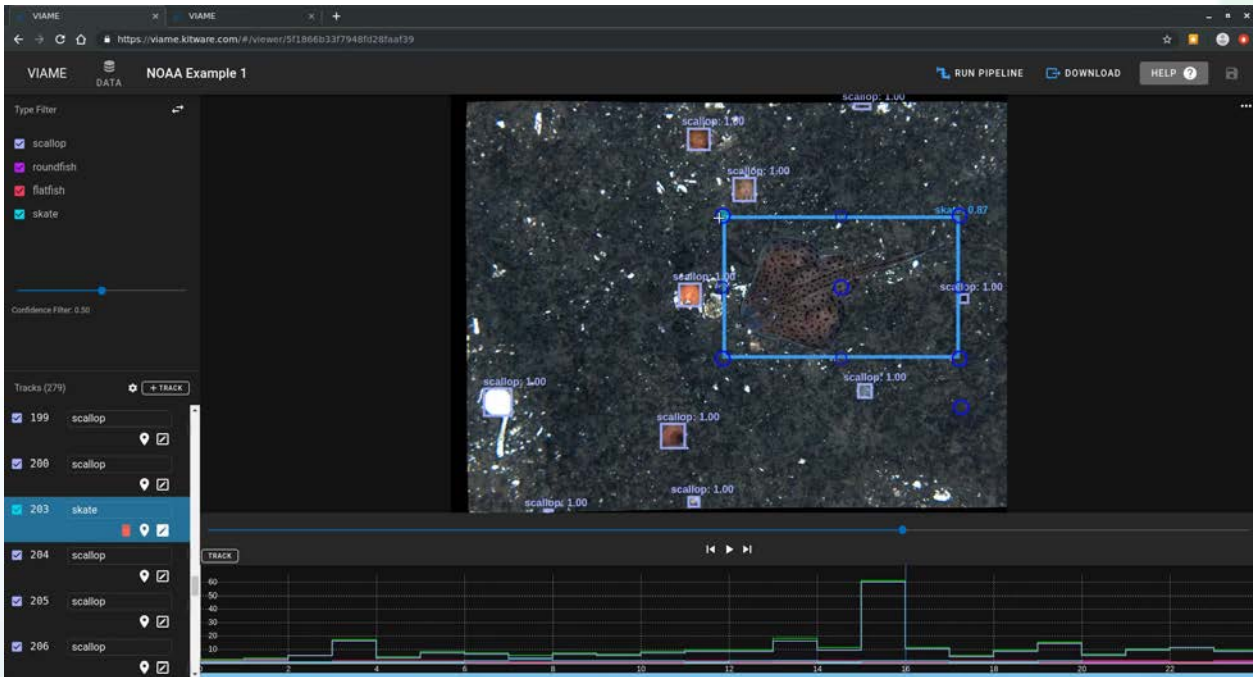
Embedded Processing



PI: matt.brown@kitware.com

Funded by NOAA ADAPT SBIR Phase I and NOAA AKFSC MML

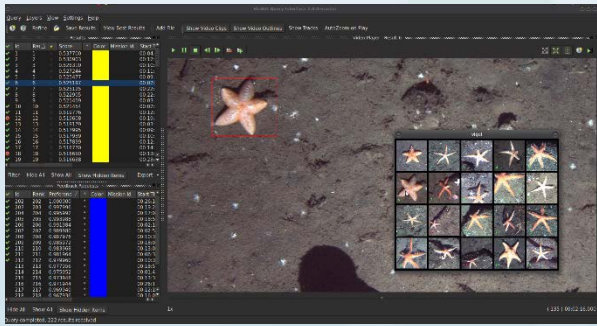
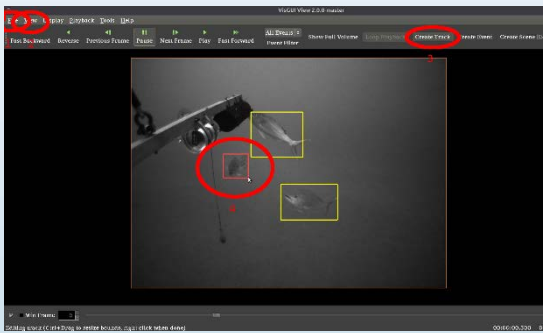
Graphical User Interfaces



Web vs Desktop

- Both wrap arbitrary processing pipelines and have tracking support (split, merge)
- Web currently has better pixel classification annotation, though some in desktop
- Desktop can currently drive user initialized tracks, web can't, though adding shortly
- Desktop currently has search and rapid model generation and multi-view display, not in web
- Normalizing features across all versions in progress

VIAME-Web Annotator

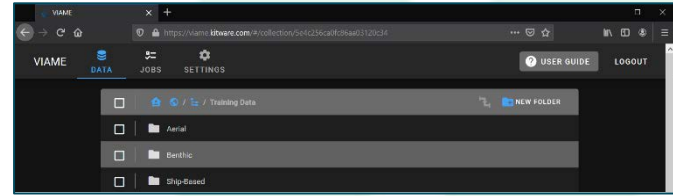
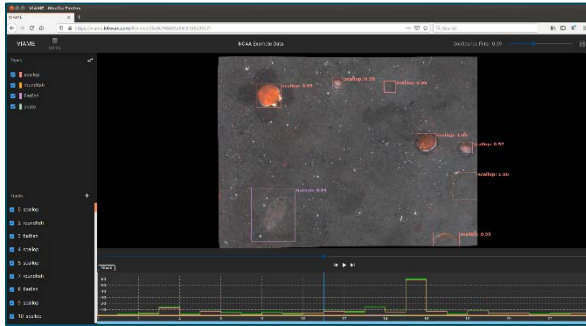


Desktop Default Annotator

Desktop Search Engine

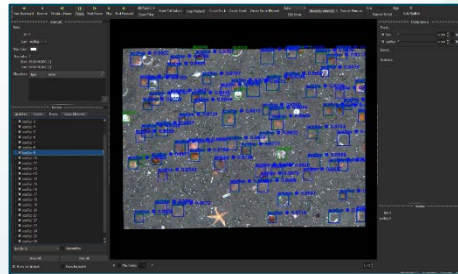
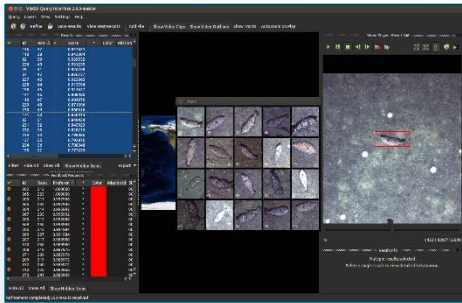
Stereo/Multi-View Annotator

Web Application and Annotation Archive



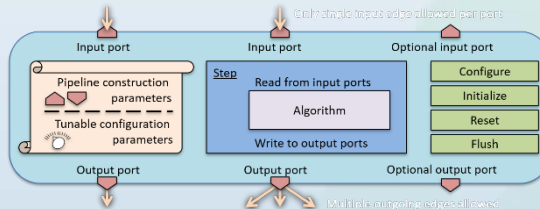
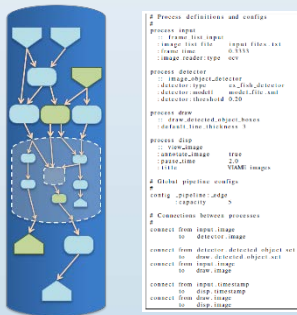
Online Example: viame.kitware.com
Server Manages Data and Annotations

Desktop Applications



User Manages Data and Annotations

Command-Line Tools and APIs:



Full Feature Support
More Customization Ability
Useful for Embedded Platforms

Complexity

Covered in Tutorial on Sept 22nd, 12 pm to 2 pm EST

Public VIAME Server

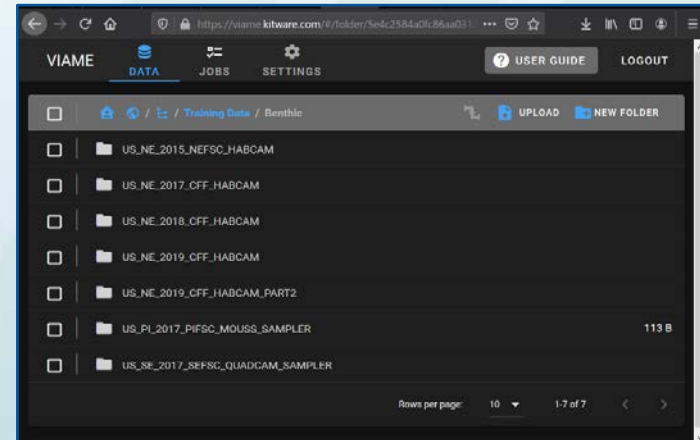
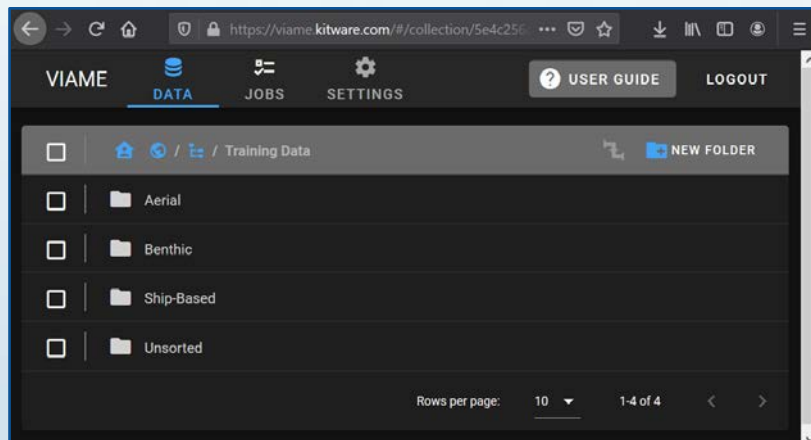
Example VIAME-Web instance: <https://viame.kitware.com>

Public data and annotation store provided by Kitware for:

- Storing and Sharing of Annotations and Imagery related to VIAME
- Performing Annotation
- Running Detectors on Data
- Limited Detector Training

15 Tb open training data store (raided for backup), 2 GPUs

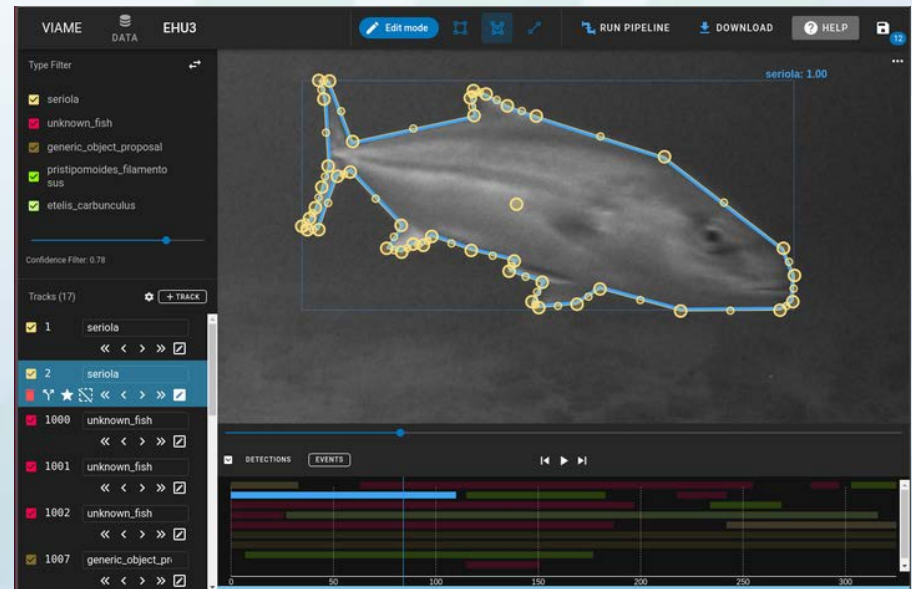
- 3 Tb used so far, though still sorting annotations to post online
- Users can annotate data, run multiple pre-trained detectors, and (shortly) use limited training capabilities (FIFO queue on 1 GPU) for new categories



Code and docker containers: <https://github.com/VIAME/VIAME-Web>

Current and Future Work

- Full feature support in web GUIs, improved desktop GUIs
 - Web GUI is very focused towards object detection and tracking, adding in support for other auxiliary features, such as mosaic generation, image enhancement, etc...
 - Make cross-use of other projects at Kitware performing similar work (IARPA, DARPA, DoD)
 - Pixel-classification utility pipelines (e.g. boxes to masks)
- Additional algorithm specializations
- More documentation
- Behavior and event detection
- Acoustic data processing
- Electronic monitoring

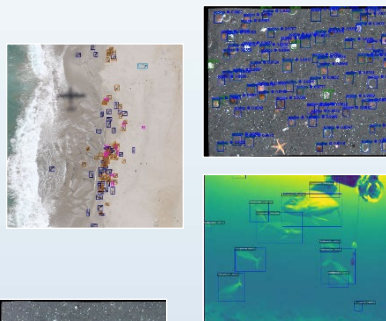




Conclusions

- VIAME is a do-it-yourself (DIY) AI toolkit which can be applied to multiple types of imagery or video
- Can be run by people with no programming or machine learning background in both web and desktop interfaces
- Released as fully open-source with a permissive license
- Specializations to maritime processing such as motion fusion, stereo measurement, image enhancement, and object tracking which other software (e.g. Amazon SageMaker) lack

Object Detection



Object Tracking

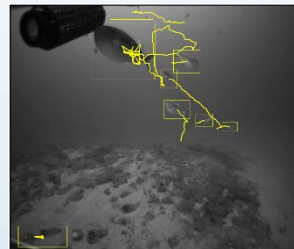
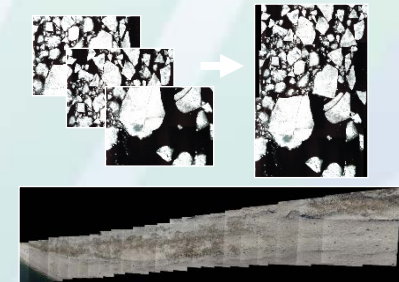


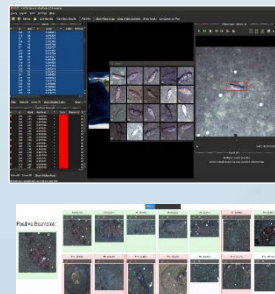
Image Enhancement



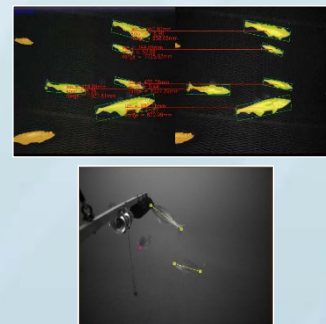
Image Registration and Mosaicing



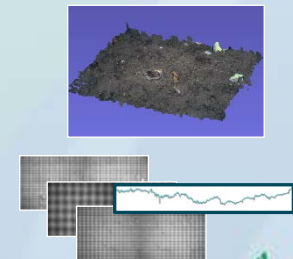
Video Search and Rapid Model Generation



Stereo Measurement



Calibration, 3D and Altitude Estimation



Special Thanks to:

- NOAA AIASI, NOAA AKFSC MML and CFF for funding VIAME
- Image annotators and testers across various organizations

See tutorial (Sept 22nd) and
<https://viametoolkit.org/> for more information

Tutorial Agenda

- | | | |
|-------------------|---|---|
| 12:00 to 12:15 pm | – | VIAME Overview
Desktop Installation
Web vs Desktop vs APIs |
| 12:15 to 12:45 pm | – | VIAME Web Edition
Types of Annotation
Model Training |
| 12:45 to 1:45 pm | – | VIAME Desktop Edition
Core Functionalities
Auxiliary Features |
| 1:45 to 2:00 pm | – | APIs and Configs [Advanced] |
| 2:00 to 2:30 pm | – | Open Discussion |