



M I D E V A L U A T I O N P R E S E N T A T I O N

Object Detection, Tracking and Suspicious Activity Recognition for Maritime Surveillance using Thermal Vision

Group 03

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Motivation

CNA938CNA LifestyleCNA InsiderSingaporeAsiaWorldBusinessSportCommentaryNews ClipsVideo on DemandPodcAll Sections

05 Mar 2020 06:34PM

Asia

Sri Lanka seizes record US\$33 million drugs haul at sea

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POLICE NARCOTIC BUREAU

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Sri Lanka Coast Guard

Safe, Secure & Sereene Sea

HOMEABOUT USPROFILENEWSREGULATIONSPUBLIC

OVER 3KG CRYSTAL METHAMPHETAMINE (ICE) FOUND FLOATING AT SEA OFF TRINCOMALEE

Three sealed parcels of Crystal Methamphetamine, commonly known as 'Ice' found floating in seas off Trincomalee by a returning Sri Lankan multiday Fishing Vessel handed over to Sri Lanka Coast Guard on 14th February 2020. The parcels contained 3.172 kg of Ice with a street value over Rs. 30 million handed over to Police Station, Tangalle for further investigations.

Sri Lanka Coast Guard is playing a vital role under close supervision of Director General – SLOG, Rear Admiral Samantha Wimalathunge in making a Sri Lanka a drug-free society, has been initiating a number of steps to prevent drug trafficking into the country via sea routes and peddling of drugs in the country. The recovery of drugs is a result of a series of awareness programmes conducted by Sri Lanka Coast Guard for fishery community on illegal drug trafficking by sea and the monitoring fishing movements in all major fishery harbours in Sri Lanka.

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Friday, 20 March 2020 - 18:27

mnews.lkwww.sooriyanfmnews.lk

Trending News

ඉතා අවධානම් දක්වමින් 6ට අවර්ධව ඇදිරි නීතිය 01 April 2020

යාපනය, මරදාන සහ කුරුණෑගලින් තවත්

Event News >> Thirty (30) illegal Sri Lankan immigrants held by Navy in southern seas

Thirty (30) illegal Sri Lankan immigrants held by Navy in southern seas

Naval personnel attached to Fast Attack Craft (FACs) belonging to the Southern Naval Command, deployed on patrol waters, intercepted a suspicious trawler plying in southern seas this morning (07th March) and held 30 suspects onboard. Having spotted a suspicious boat movement at sea about 80 nautical miles off the Galle Lighthouse, two Fast Attack C directed to the location of the suspicious trawler. Accordingly, the naval personnel held 30 illegal Sri Lankan immigrants are due to be handed over to the Galle Harbour Police after a medical examination and preliminary naval investigation.

The Navy urges the general public not to involve in high risk sea-borne migration to overseas countries based on false promises that such attempts would finally end up behind bars.

Further the Navy reminds of its strong network of intelligence and regular patrols in place to nip such illegal migration strict measures to curb such attempts.

3

Modified Deliverables (After Feasibility Study)

Functioning Activity Detection algorithm

- Identifying at least 3-classes of pre-defined suspicious activities.
- With competitive mAP score.
- Using thermal image sequences.

Well documented software

- User-friendly

1

Functioning Object Detection algorithm

- For objects related to maritime environment such as boats, swimmers etc.
- Competitive mAP score on predicted bounding boxes
- Using thermal images.

2

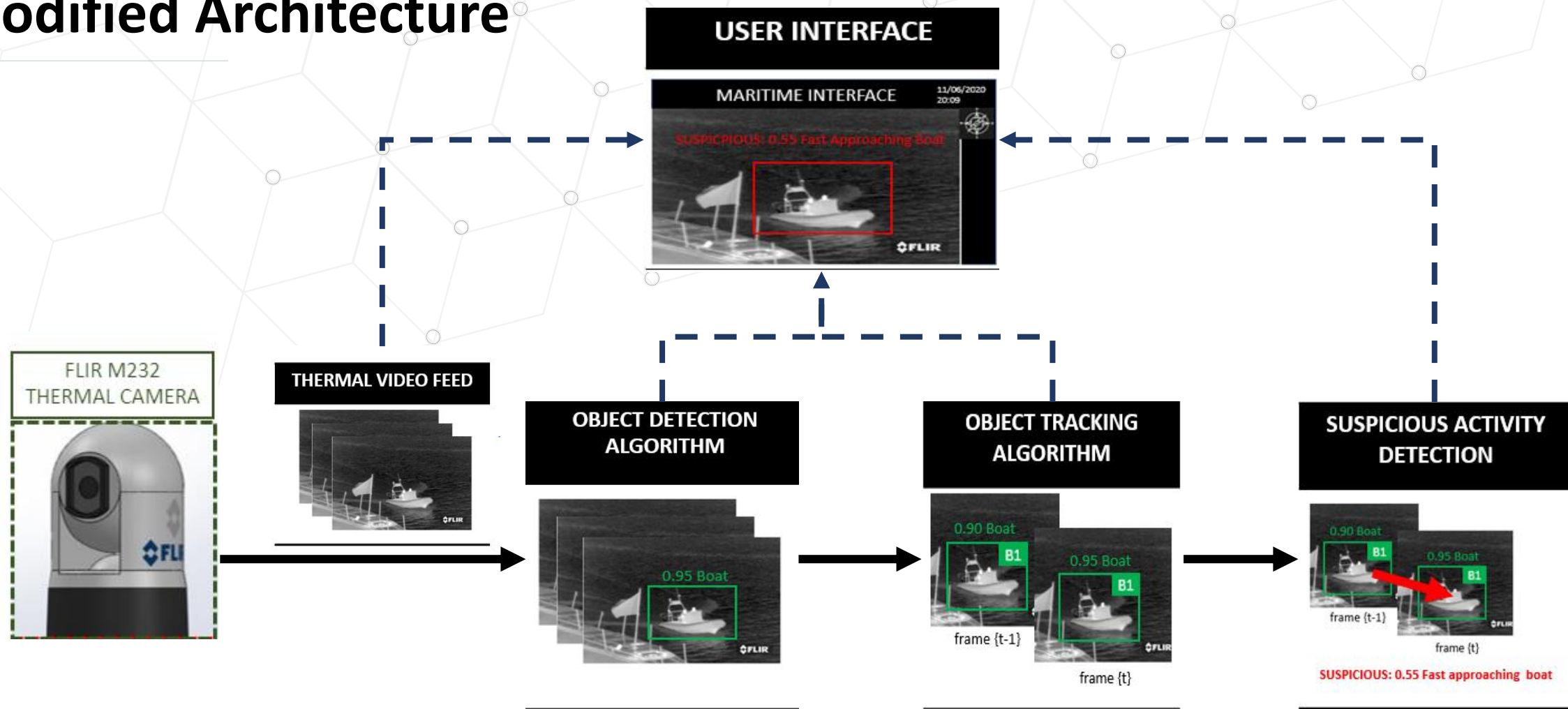
3

Annotated maritime dataset

- Capable of use for object detection and activity detection in maritime environment.
- Either a locally obtained thermal dataset or already published maritime thermal dataset.
- Publication in a well-known forum.

4

Modified Architecture



May include:

- Dumping objects
- Unauthorized fishing
- Loitering
- Evasive actions

Content



Maritime Object Detection

Alternative Datasets for Maritime environments

Datasets	Link	Images/ Video type	Description
Singapore Maritime Dataset	https://sites.google.com/site/dilipprasad/home/singapore-maritime-dataset	RGB and Near IR	<ul style="list-style-type: none">• RGB (Onshore and Onboard)• Near IR (Onshore)• NIR videos were captured using Canon 70D camera with hot mirror removed and Near-IR Bandpass filter (Different than actual thermal images)
SeaShips	http://www.lmars.whu.edu.cn/prof_web/shaozhenfeng/datasets/SeaShips(7000).zip	Only RGB	<ul style="list-style-type: none">• Contains 31455 images (Only 7000 images publicly available)• Annotations provided
IPATCH	http://ftp.pets.rdg.ac.uk/pub/PETS2016/MidHighLevelChallengeData/IPATCH/	Both RGB and Thermal	<ul style="list-style-type: none">• Contains a set of fourteen multi camera recordings (visible, thermal) collected off the coast of Brest, France• No annotations provided/ The categories of the objects

Maritime Object Detection

Alternative Frameworks for Maritime Object Detection

Dataset	Evaluation Criterion	Framework		
		SSD ^[1]	CornerNet-Lite (Squeeze) ^[2]	CenterNet ^[3]
SeaShips	mAP % @ IoU 0.5	28.4	59	81.8
	FPS	19	60	52
Singapore Maritime Dataset	mAP % @ IoU 0.5	27	55.3	60.7
	FPS	19	60	52

*Algorithms are trained under both SeaShips and Singapore Maritime dataset.



Video from an Onboard camera -SMD (Inferenced using CenterNet)



Video from an Onshore camera -Seaships (Inferenced using CenterNet)

/essel/ship0.8



Vessel/ship0.3



Vessel/ship0.8



2017-01-04 星期三 09:15:51

bulk cargo carrier0.5

A wide-angle view of a coastal city, likely Qinghai, with mountains in the background. The foreground is a body of water. In the middle ground, a city skyline is visible, including several tall buildings. A small orange box highlights a ship in the water, labeled "bulk cargo carrier0.5".

琴海北路北侧HJ005球机

Maritime Object Detection

Near IR version of Singapore Maritime Dataset



Video from an Onshore NIR camera (Without Haze) -SMD (Inferenced using CenterNet)

Video from an Onshore NIR camera (With Haze) -SMD (Inferenced using CenterNet)

Thermal Object Detection

Alternative Frameworks for Thermal Object Detection

Evaluation Criterion	Framework		
	SSD ^[1]	CornerNet-Lite (Squeeze) ^[2]	CenterNet ^[3]
mAP % @ IoU 0.5	28.8	81	88
FPS	19	60	52



Video from FLIR ADAS Thermal Dataset (Inferenced using CenterNet)

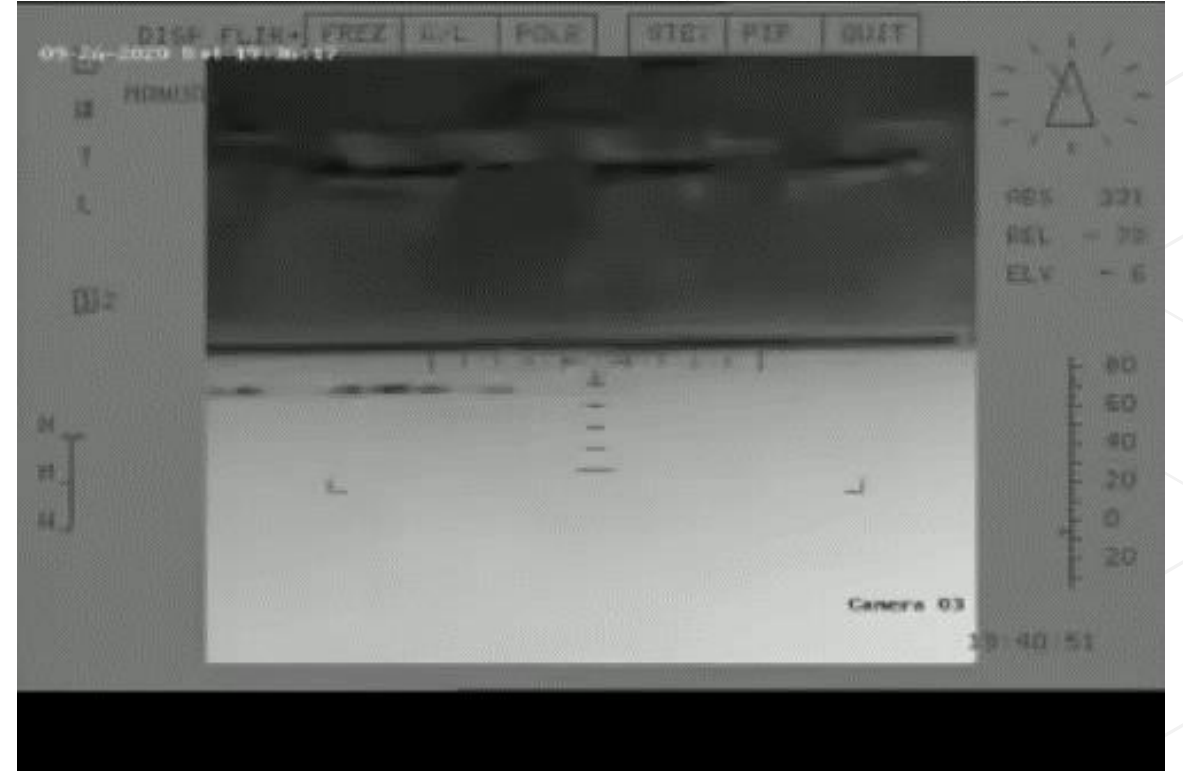
*Algorithms are trained under FLIR dataset.



Object Tracking

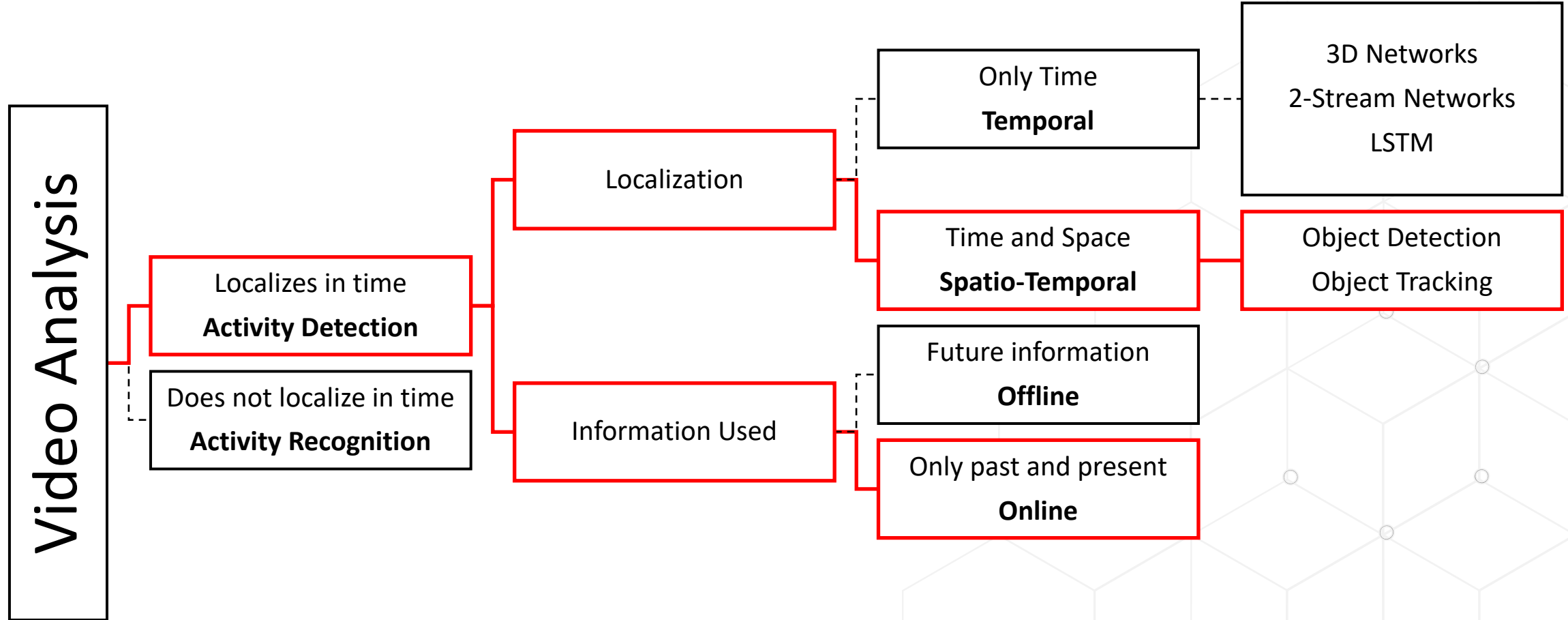
Alternative Algorithms

Tracker	Metric		
	Average IoU	Average CD	FPS
Boosting ^[4]	0.38	80	49
KCF ^[5]	0.22	15	102
Median Flow ^[6]	0.23	160	160
MIL ^[7]	0.35	60	20
MOSSE ^[8]	0.20	78	210
TLD ^[9]	0.24	110	27
CSRT ^[10]	0.50	50	25



Video from SL Navy (Tracker - CSRT)

Action Detection



Action Detection

Alternative Datasets

Datasets	Link	Images/ Video type	Description
UCF 101 - 24	https://www.crcv.ucf.edu/data/UCF101/UCF101.rar	<ul style="list-style-type: none">• RGB• Sports	<ul style="list-style-type: none">• Extracted from a large dataset (UCF 101) - 101 action categories• 13320 videos• Annotations provided• UCF 24 – Only sports with 24 categories
HMDB 21 - 51	https://serre-lab.clps.brown.edu/wp-content/uploads/2013/10/hmdb51_org.rar	<ul style="list-style-type: none">• RGB• Facial actions and Body movements	<ul style="list-style-type: none">• Extracted from a large dataset (HMDB 51) - 51 action categories• 928 videos• Annotations provided

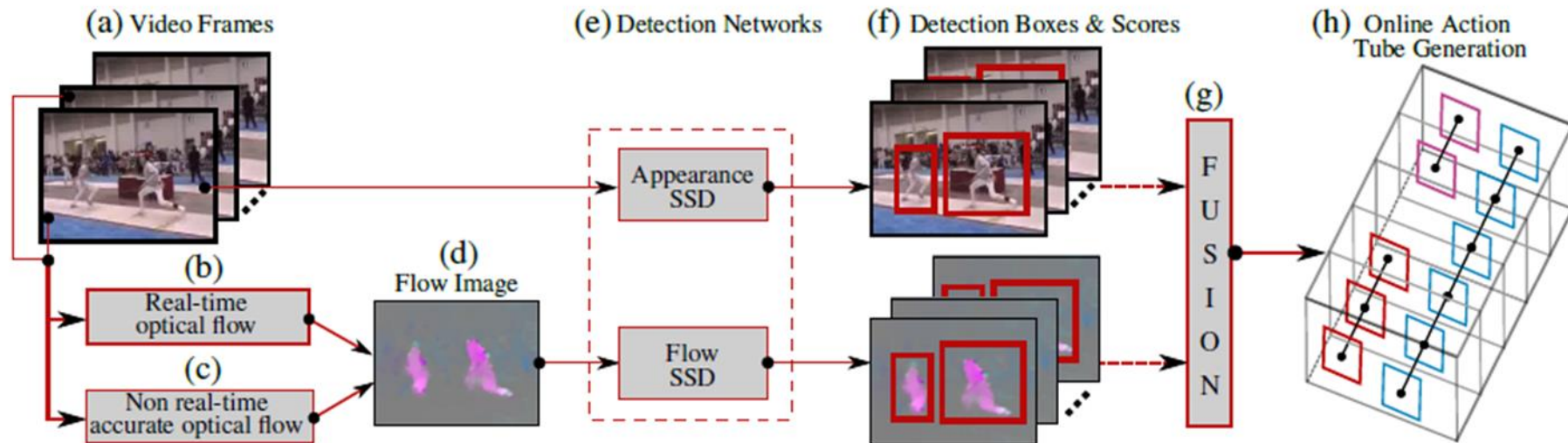
Action Detection

Evaluation of alternatives – Action Detection Algorithms

	Temporal Recurrent Networks (TRN)^[11]	Information Discrimination Unit (IDU)^[12]	Spatio-Temporal and Motion Encoding (STM)^[13]	A structured Model for Action Detection^[14]	Spatio-Temporal Progressive Learning(STEP)^[15]	Online Real-time Multiple Spatio-temporal Action Localization and Prediction (ROAD)^[16]
Temporal/Spatio-temporal	Temporal	Temporal	Temporal	Spatio-Temporal	Spatio-Temporal	Spatio-Temporal
Backbone	VGG-16 / ResNet-200	VGG-16 / ResNet-200	ResNet-50	ResNet-50/Mask-RCNN	VGG-16	VGG-16
Online/Offline	Online	Online	Offline	Offline	Offline	Online
FPS	24	24	-	12	21	28
Dataset	THUMOS'14	THUMOS'14	UCF101	UCF101	UCF101	UCF101
mAP score	47.2	60.3	96.0	77.9	75.0	43.0

Action Detection

ROAD Architecture



Issues	Solutions Worked-On
No implementation of linking algorithm in python	Implementing the linking algorithm in python (90% of the conversion is done)
Lack of end-to-end pipeline from action localizations to action linking	Building the end-to-end pipeline in python (100% implementation with SSD detector)
Duplication of results for Fast Optical Flow implementation is not provided	

Paper Submission on Action Detection Domain

IEEE International Conference of Image Processing (ICIP) - 2021

Paper Title: Online Real Time Spatio-temporal Action Detection Using Key Points

Our Contributions:

- A key-point based Spatio-Temporal (ST) Action Localization.
- A two-frame cascaded input to compute required temporal features.
- An improved tube-linking algorithm.

Paper Status

Paper Number: 1412

Paper Title: KEY-POINT DETECTION BASED ONLINE REAL-TIME SPATIO-TEMPORAL ACTION LOCALIZATION

Initial Submission Time Stamp: 1/11/2021 6:51:07 AM

Revised Submission Time Stamp: 1/25/2021 10:14:47 AM

Current Server Time: 3/17/2021 12:21:03 AM

Submission Topic

Code	Category	Topic
10.2	Image and Video Analysis, Synthesis, and Retrieval	Image & Video Interpretation and Understanding::
4.7	Applications of Machine Learning	Machine learning for image & video analysis, synthesis, and retrieval::

Status	Item	Result
COMPLETE	Copyright Form	Copyright form has been received.
COMPLETE	Paper Upload	Document has been uploaded.
COMPLETE	Paper Inspection	Document passed inspection.
IN PROGRESS	Technical Committee Review	Document is under review. No action is required by you at this time.
PENDING	Final Paper Upload	Document inspection review pending results.
PENDING	Final Paper Inspection	Document inspection pending final paper upload.

Paper Submission on Action Detection Domain

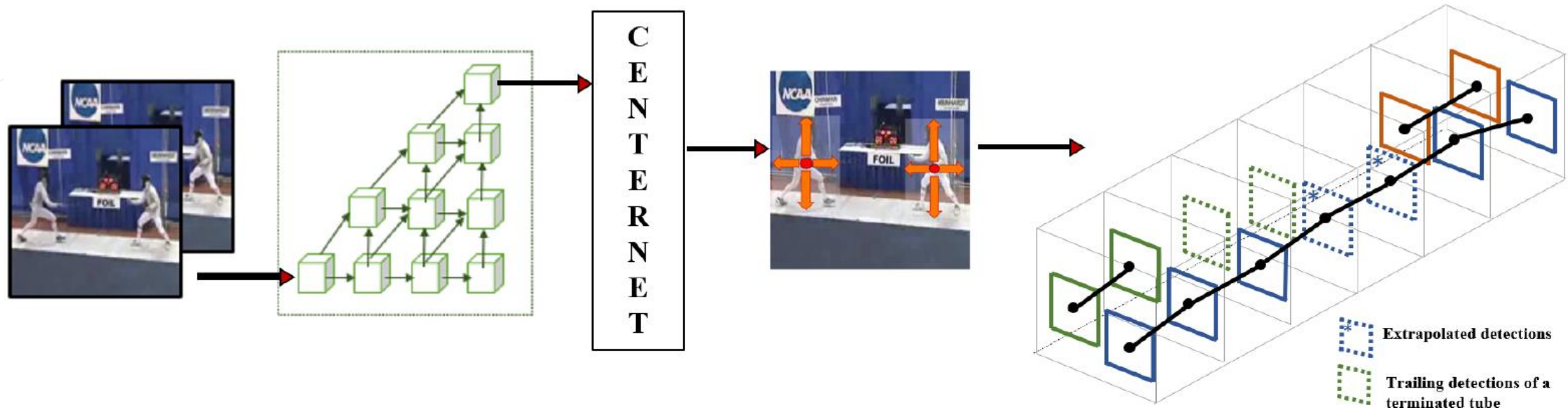
Proposed ST Action Localization Architecture

(a) Pair of consecutive video frames

(b) Object detector with DLA-34 feature extractor

(c) Center Points and Scores

(d) Online, real-time action tube generation

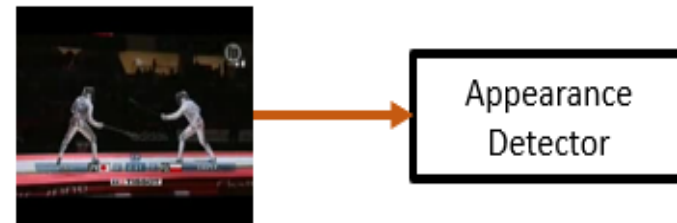


Paper Submission on Action Detection Domain

Novelty in our work

1. Cascaded two frame input for feature extractor to learn motion features.

Objective – To remove and replace the need of computationally expensive optical flow calculation



$$x \in \mathbb{R}^{h \times w \times 3}$$



$$x \in \mathbb{R}^{h \times w \times 3}$$

Flow Detector



Two Stream RGB and Optical Flow Input



$$x \in \mathbb{R}^{h \times w \times 6}$$

Appearance
Detector



Proposed Cascaded Input

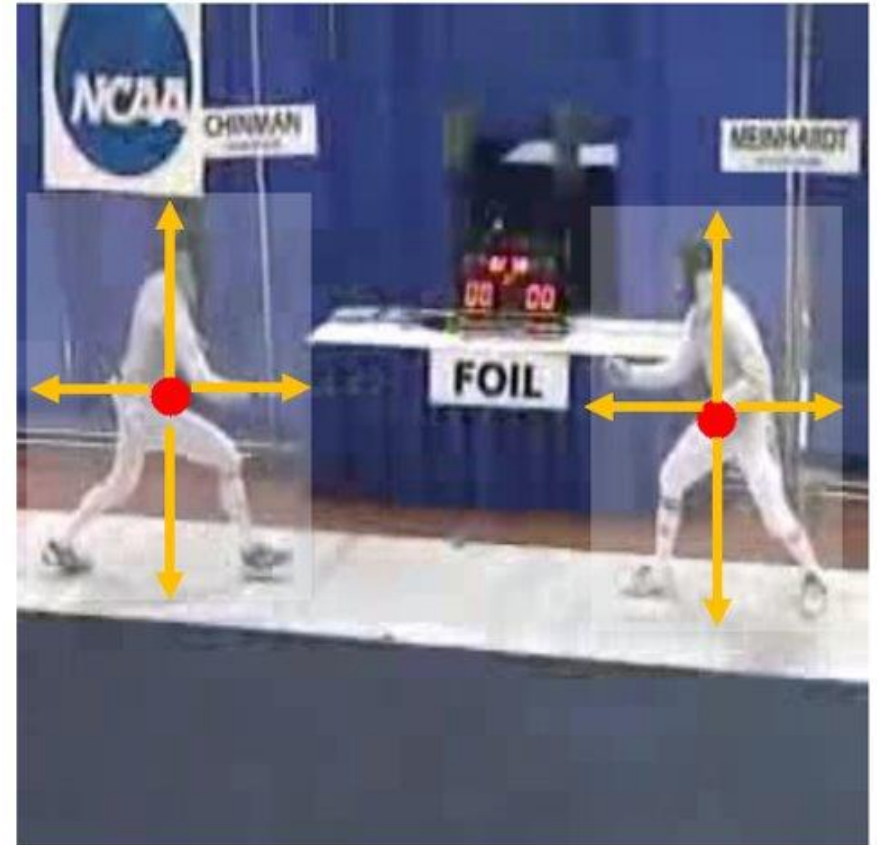
Paper Submission on Action Detection Domain

Novelty in our work

2. Key points for the Action Detection

Key-point based Action Localization has not being exploited in any of the past work.

Objective – To reduce the complexity and to improve inference time with key point detection using CenterNet with DLA-34.



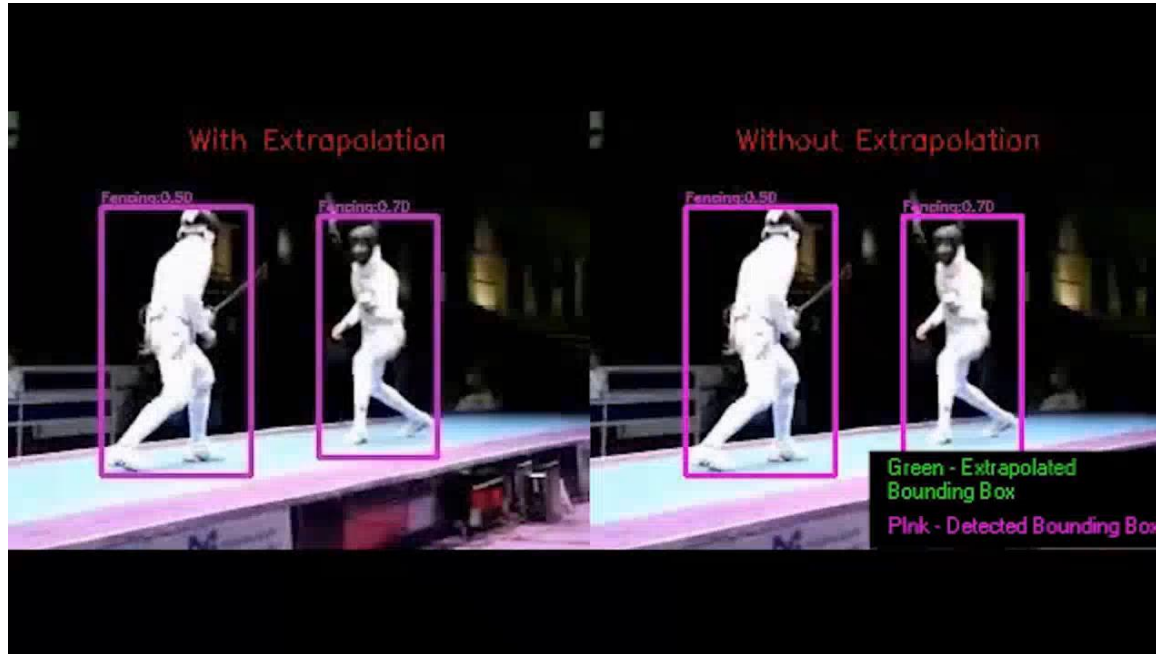
Key Point Detection of Action Instances

Paper Submission on Action Detection Domain

Novelty in our work

3. Improved Tube Linking Algorithm

Objective – To link the actions localized in the video frames and reduce the miss rate through extrapolation based on movement of the bounding boxes of past detections and their class scores



Demonstration of the Proposed Linking Algorithm

Algorithm 1: Online tube generation

```
Input:  $\mathcal{T}^{t-1}, \mathcal{D}^t, c, \lambda, k$   
Output:  $\mathcal{T}^t$   
for  $T_j^{t-1} \in \mathcal{T}^{t-1}$  do  
     $s \leftarrow 0$ ;  
     $m \leftarrow 0$ ;  
    for  $D_i^t \in \mathcal{D}^t$  do  
        if  $\text{IoU}(b_{D_i}^t, b_{T_j}^{t-1}) \geq \lambda$  and  $s < s_{D_i}^t(c)$  then  
             $b_{T_j}^t \leftarrow b_{D_i}^t$ ;  
             $s \leftarrow s_{D_i}^t$ ;  
             $m \leftarrow i$ ;  
             $\tau \leftarrow 0$ ;  
        end  
    end  
    if  $m = 0$  and  $\tau < k$  then  
         $b_{T_j}^t \leftarrow \text{predict\_bbox}(b_{T_j}^{t-1}, b_{T_j}^{t-2})$ ;  
         $\tau \leftarrow \tau + 1$ ;  
    end  
     $s_{T_j}^t, c_{T_j} \leftarrow \text{update\_label}(s_{T_j}^{t-1}, s_{D_m}^t)$   
end
```

Proposed Linking Algorithm

Paper Submission on Action Detection Domain

Results Obtained

Experiments were done on two datasets:

1. UCF-101-24 dataset : Challenging dataset with multiple action instances per video
2. J-HMDB-21 : Challenging dataset with single action instance per video

Table 1. ST action localization results (v-mAP) on UCF-101-24 dataset. Last two columns compare the f-mAP and FPS.

Method	v-mAP				f-mAP	FPS
	0.2	0.5	0.75	0.5:0.95	@0.5	
Saha <i>et al.</i> [1] [◇]	66.6	36.4	7.9	14.4	-	4
Peng(w/ MR) <i>et al.</i> [2] [◇]	72.9	-	-	-	65.7	-
Zhang <i>et al.</i> [8] ^{◇*}	74.8	46.6	16.7	21.9	67.7	37.8
ROAD (w/ AF) [9] [‡]	73.5	46.3	15.0	20.4	-	7
ROAD (w/ RTF) [9] ^{‡*}	70.2	43.0	14.5	19.2	-	28
Our (A+AF)[‡]	72.9	46.7	16.2	20.9	70.8	7.7
Our (A+RTF)^{‡*}	69.6	42.1	15.5	19.3	69.6	37.9
ROAD (A) [9] ^{†*}	69.8	40.9	15.5	18.7	-	40
Ours (A)^{†*}	70.2	44.3	16.6	20.6	71.8	52.9
Ours^{†*}	71.6	44.1	17.0	20.7	74.4	52.9

◇ Offline * Real-time † Online with no OF ‡ Online with OF

Table 2. ST action localization results (v-mAP) on J-HMDB-21 dataset. Last two columns compare the f-mAP and FPS.

Method	v-mAP				f-mAP	FPS
	0.2	0.5	0.75	0.5:0.95	@0.5	
Saha <i>et al.</i> [1] [◇]	72.6	71.5	43.3	40.0	-	4
Peng(w/ MR) <i>et al.</i> [2] [◇]	74.3	73.1	-	-	58.5	-
Zhang <i>et al.</i> [8] ^{◇*}	-	-	-	-	37.4	37.8
ROAD (w/ AF) [9] [‡]	70.8	70.1	43.7	39.7	-	7
ROAD (w/ RTF) [9] ^{‡*}	66.0	63.9	35.1	34.4	-	28
Our (A+AF)[‡]	68.8	67.6	49.9	43.7	46.9	7.7
ROAD (A) [9] ^{†*}	60.8	59.7	37.5	33.9	-	40
Ours (A)^{†*}	59.3	59.2	48.2	41.2	51.2	52.9
Ours^{†*}	57.2	55.9	48.1	39.9	47.9	52.9

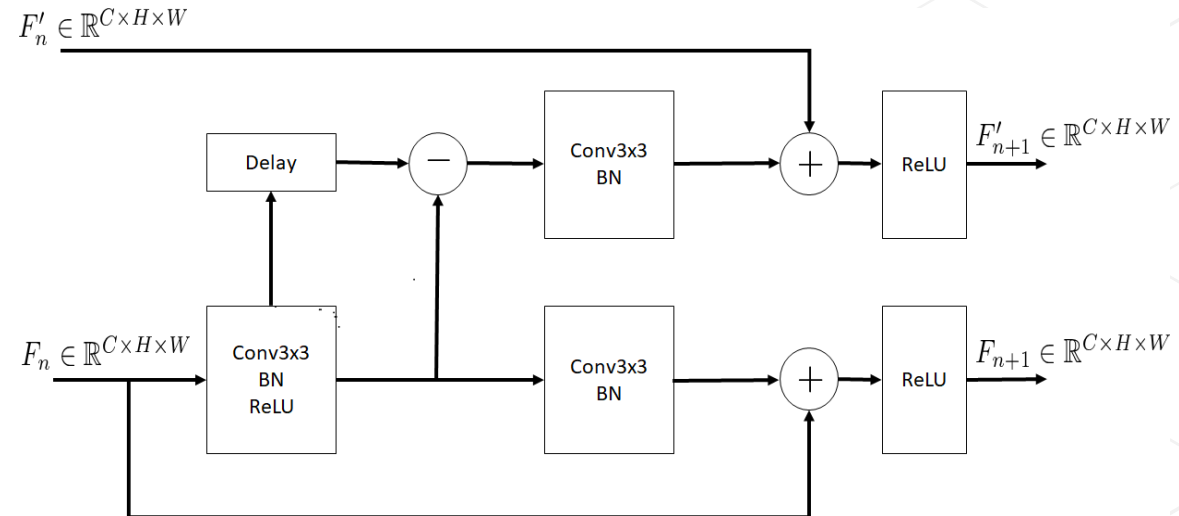
◇ Offline * Real-time † Online with no OF ‡ Online with OF

Action Detection

Short-Term Motion Module

- Feature Extractor Structure: ResNet-50
- Detection Head: CenterNet
- Pretrained Weights on: ImageNet

STMTrack



STM Module Architecture



Image $I^{(t)}$



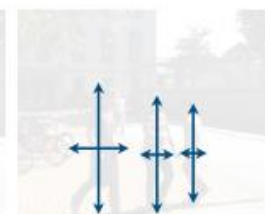
Image $I^{(t-1)}$



Tracks $T^{(t-1)}$



Detections $\hat{Y}^{(t)}$



Size $\hat{S}^{(t)}$



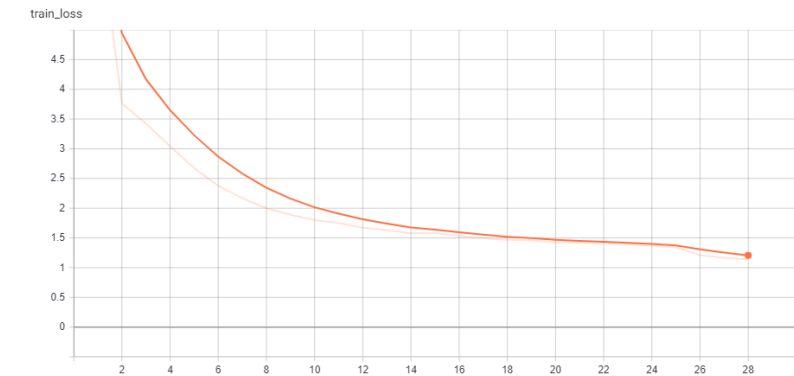
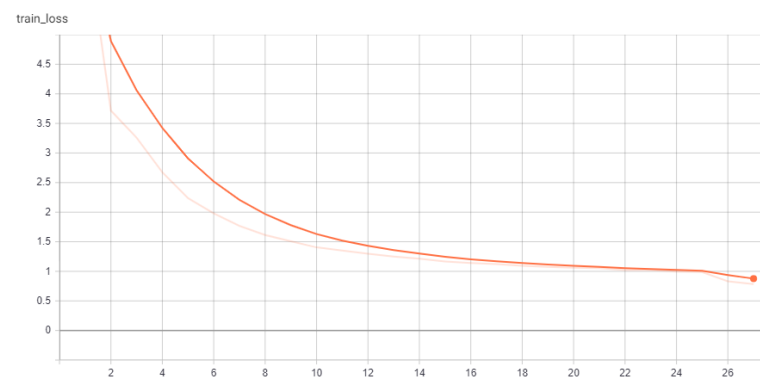
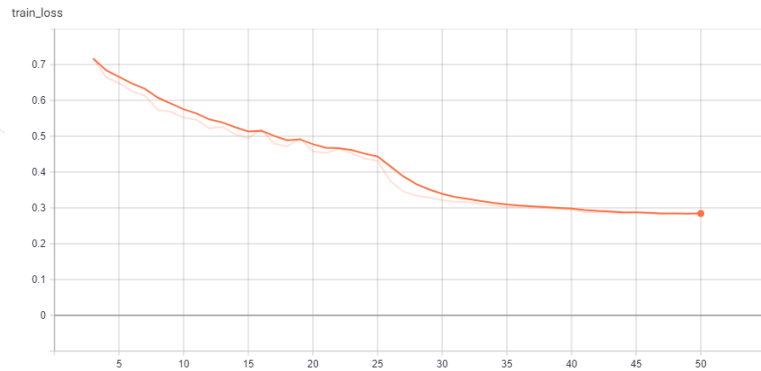
Offset $\hat{O}^{(t)}$

Action Detection

26

STMNet vs. ResNet vs STMTrack on Action Classification on HMDB-21

ResNet	STMNet	STMTrack
39.0%	33.4%	27.5%



Action Detection

Thermal dataset creation for Action Detection

- Experimental Technique.
- Used Generative Adversarial Network (GAN) based frameworks to generate synthetic images.
- Used Pix2Pix^[17] framework.
- Challenges :
 1. High Noise.
 2. Different thermal signatures.

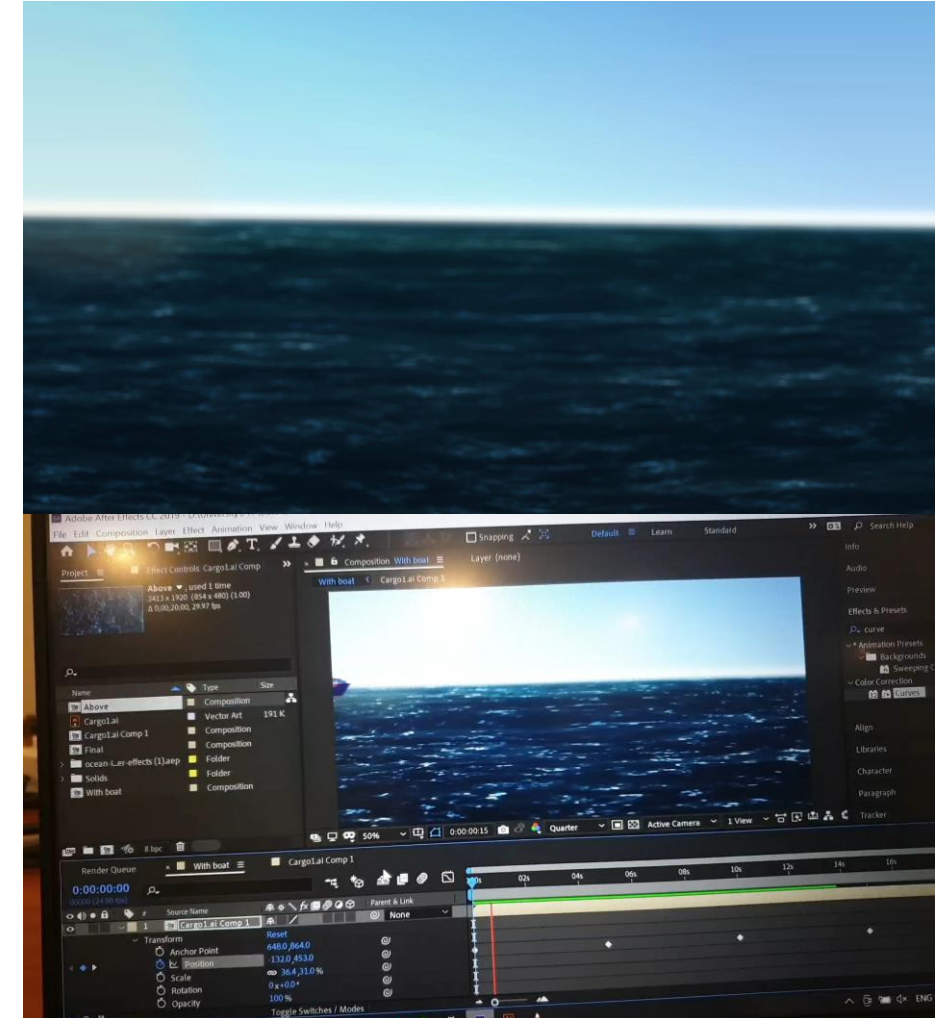


Converted RGB video using Pix2Pix

Action Detection

Maritime dataset creation for Action Detection

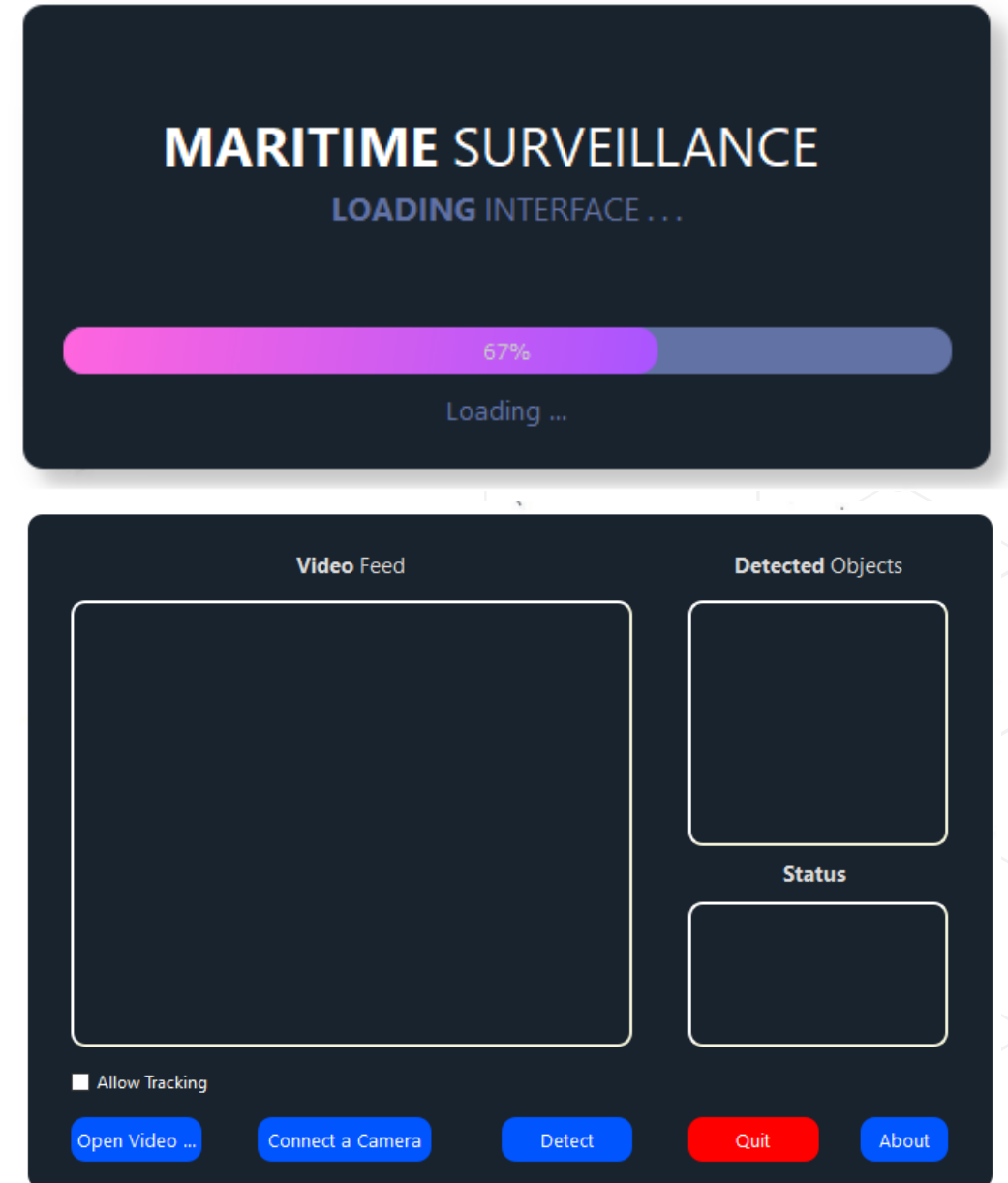
- Experimental Technique with the intention of creating a maritime dataset with action instances.
- Used Adobe After Effects to develop videos from scratch.
- Challenges:
 1. High rendering/creating time for each video.
 2. Repetition of patterns in the sea waves, which might affect the deep learning algorithms.



Custom made maritime action classification data using After Effects.

User Interface

- Developing using PyQt5.
- Python based interface with the ability to display detected objects, activities and track those in real-time.
- Displays detected objects on the side.



Interface created using PyQt5

Challenges

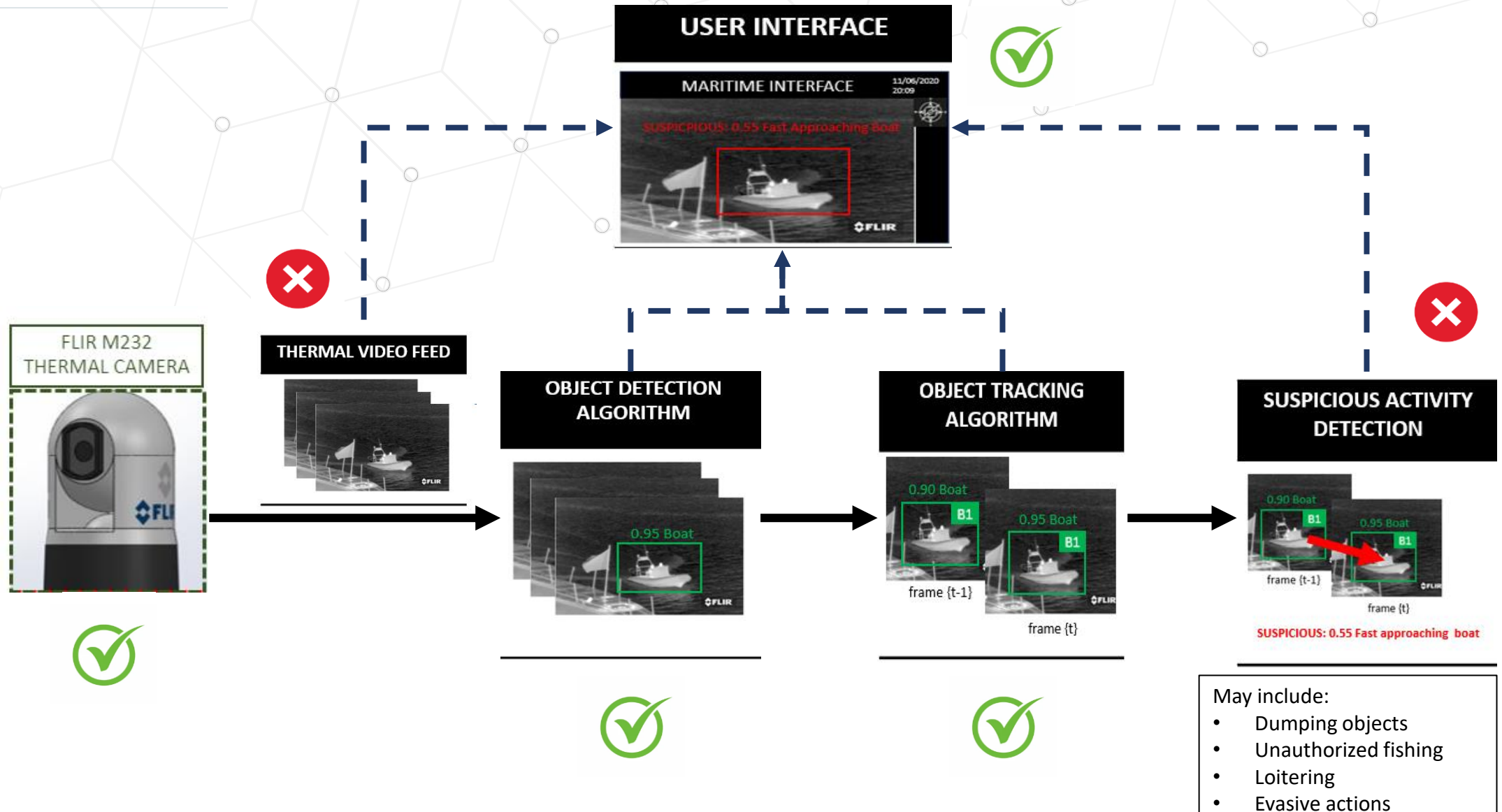
- Obtaining the Thermal Camera.
- Collaboration with the Sri Lanka Navy.
- Suitable, publicly available action detection datasets.



Meeting conducted with Sri Lanka Navy on obtaining the maritime dataset and planning the next steps.

Modified Architecture

31



Object Detection

100%

Thermal Object
Detection



Maritime Object
Detection



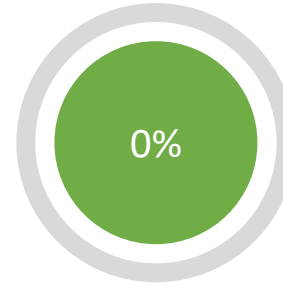
Action Detection

70%

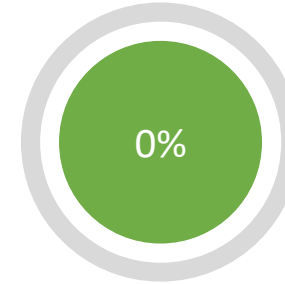
Online Real-Time
ST Action
Localization



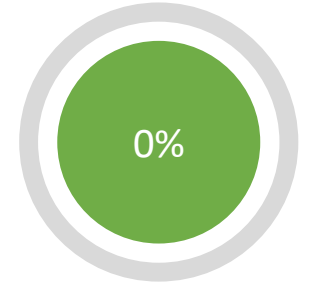
Thermal Action
Detection



Maritime Action
Detection



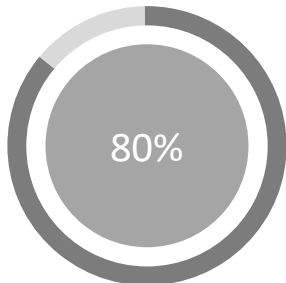
Suspicious Action
Detection in
maritime
environments



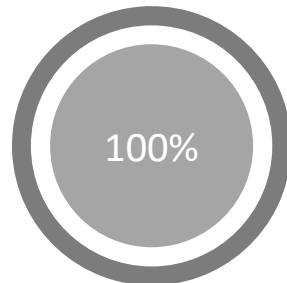
Object Tracking

90%

Thermal Object
Tracking



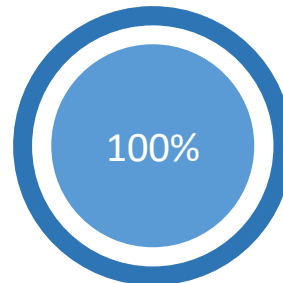
Maritime Object
Tracking



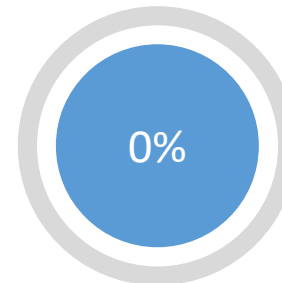
Paper Submission

100%

ICIP 2021



BMVC 2021



User Interface

60%

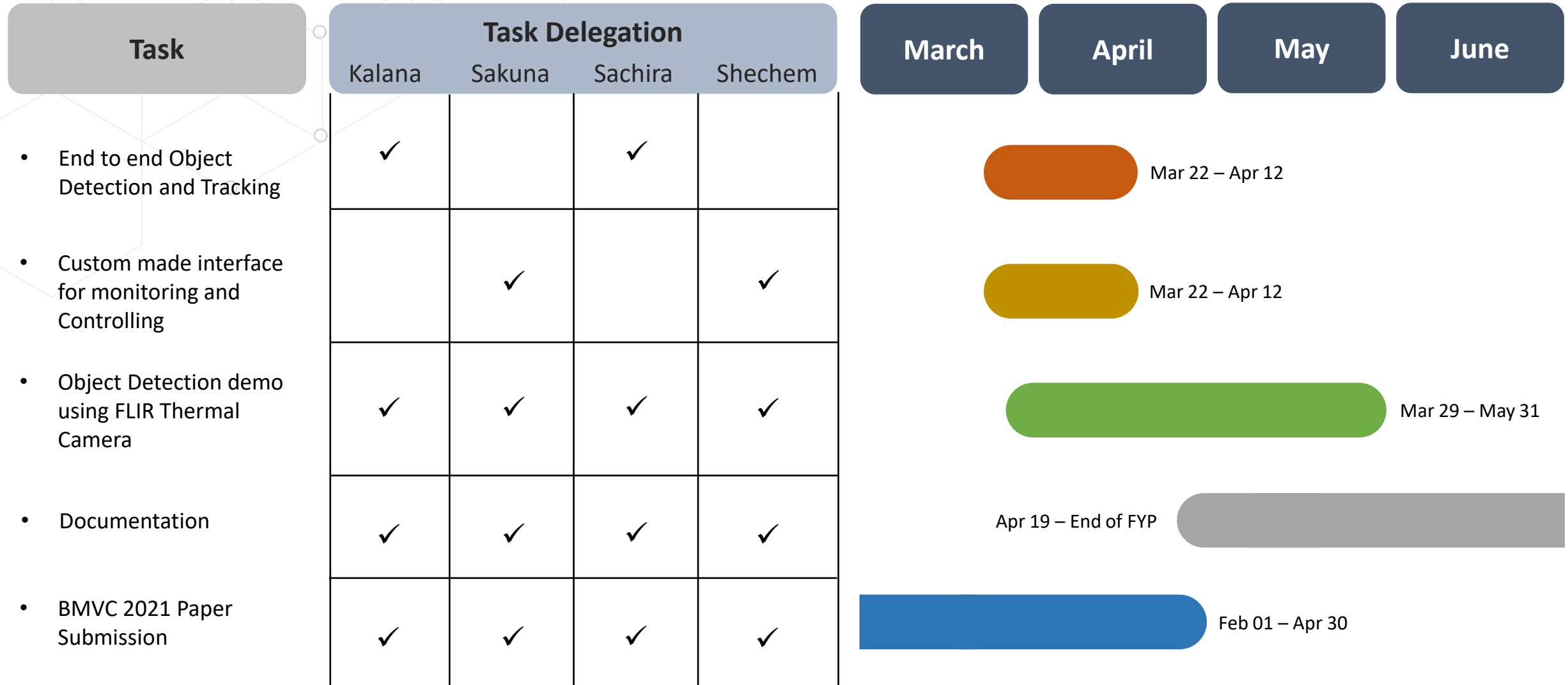
Thermal Camera Demonstration Plan

- Training the object detector algorithm with FLIR ADAS dataset
- Use our FLIR M232 thermal camera to collect our own thermal urban driving videos both in day and nighttime.
- Use trained object detector for inference on the collected videos.



FLIR M232 Thermal Camera

Final Completion Plan





Thank you.

References

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- [2] CornerNet-Lite: Efficient Keypoint Based Object Detection - Law et al.
- [3] Objects as Points – Zhou et al.
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- [9] Tracking-learning-detection – Kalal et al.
- [10] Discriminative correlation filter with channel and spatial reliability – Lukezic et al.
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