

Object Tracking using Statistic-based Feature Fusion Technique

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Introduction

Goal

Automatically track objects in wide area motion imagery (WAMI).

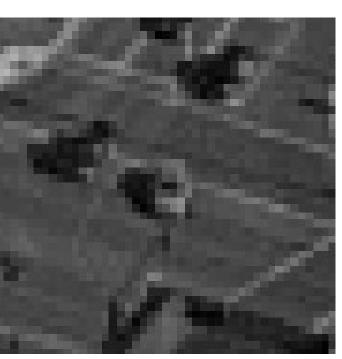
Constraints/Challenges

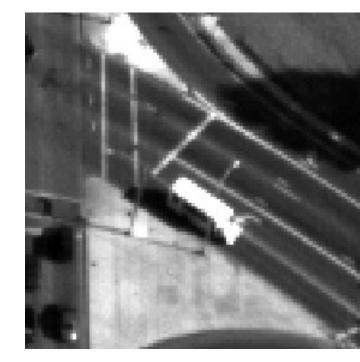
Very low resolution, presence of noise, illumination variation, occlusions, complex object motion, and complex object shapes.

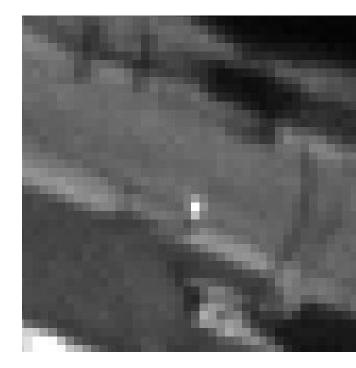
Proposed Innovation

To better combine features of the DRIFT algorithm through fusion based on past frame effectiveness





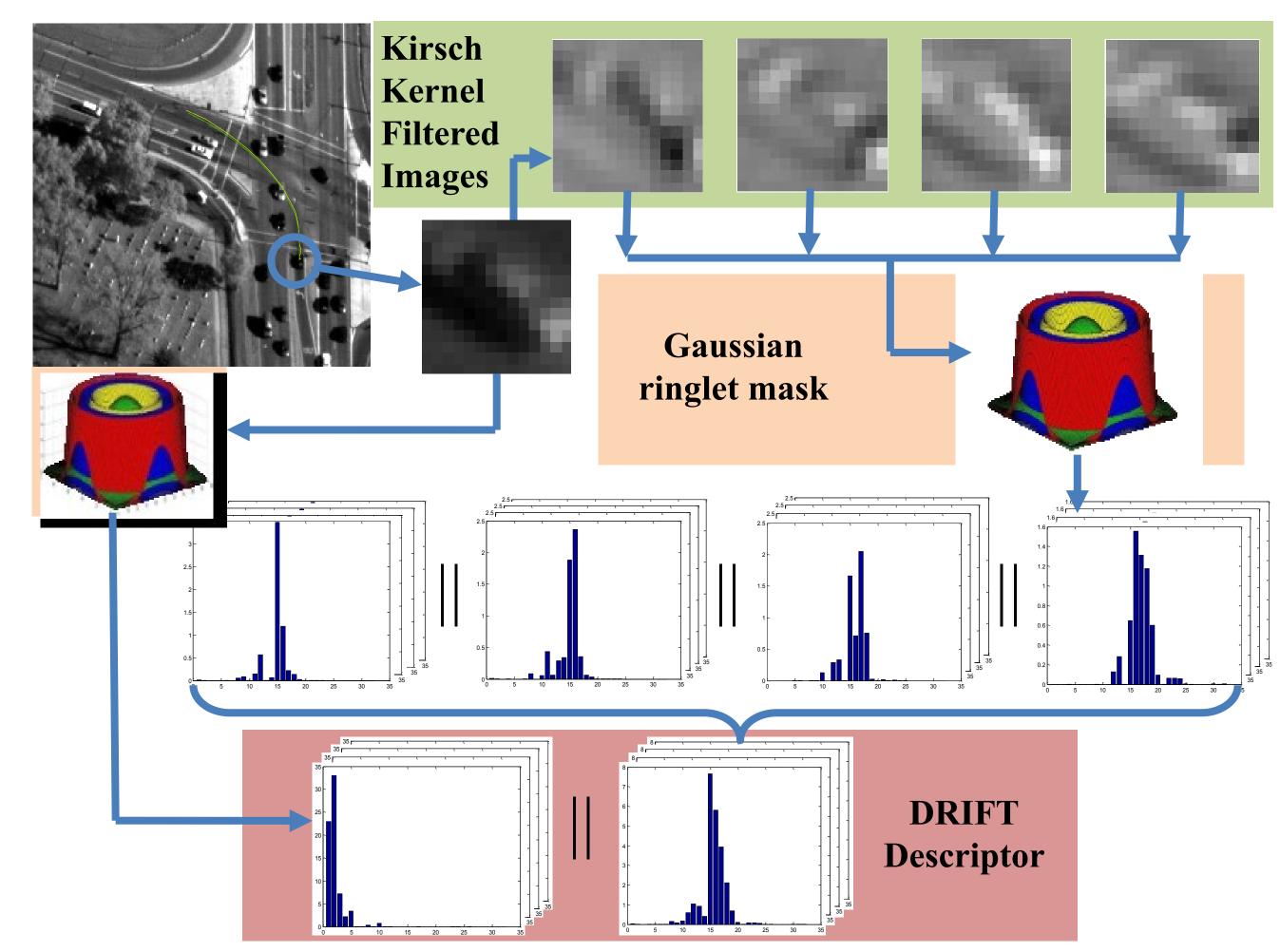




DRIFT Algorithm

Feature Extraction Method

- 1. Intensity image extracted from video frames
- 2. Four directional Kirsch kernels used to filter images
- 3. Gaussian ringlet masks used to create feature histograms for each Kirsch filtered image and the intensity image
- 4. Histograms from Kirsch filtered images concatenated to retain rotation invariance
- 5. All Histograms concatenated to create feature descriptor



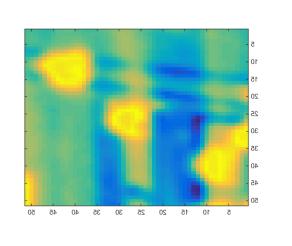
Fusion of Likelihood maps

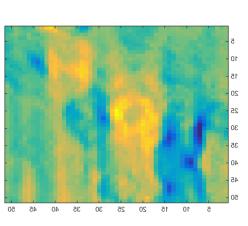
- Each feature histogram is classified using Earth Mover's Distance
- A likelihood map is created for each feature for the search area
- Weights selected from previous frame performance
- Fusion based on variance ratio between target and background:

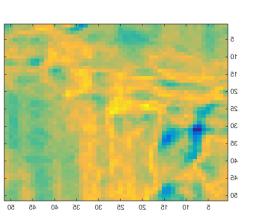
$$w_i = \frac{VR(L_i, p, q)}{\sum_{i=1}^{N} VR(L_i, p, q)}$$

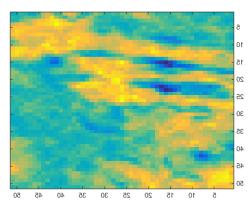
$$VR(L_{i}, p, q) = \frac{var(L_{i}, (p + q)/2)}{[var(L_{i}, p) + var(L_{i}, q)]}$$

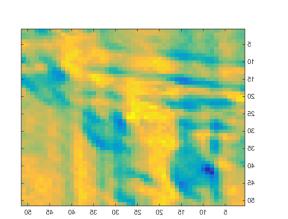
where object pixels, p, and background pixels, q



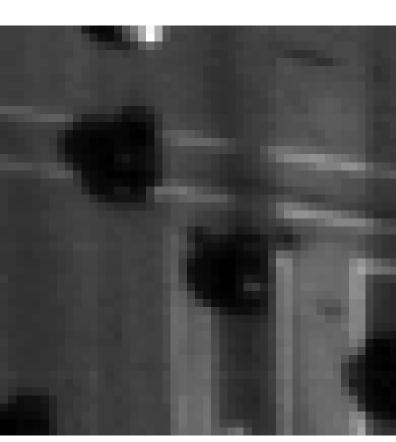


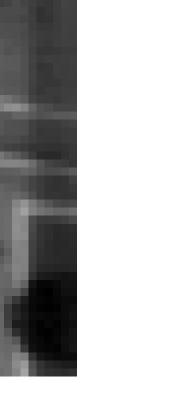


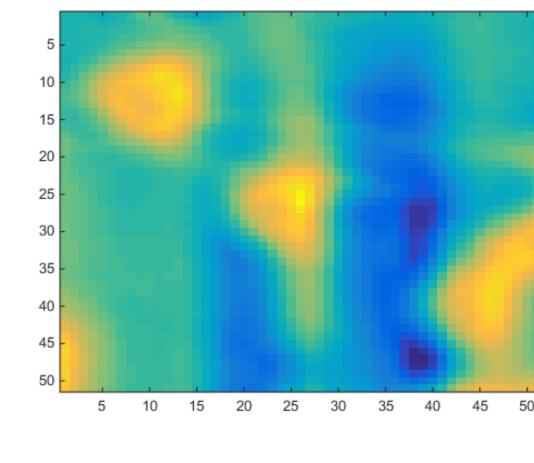




Feature Likelihood Maps



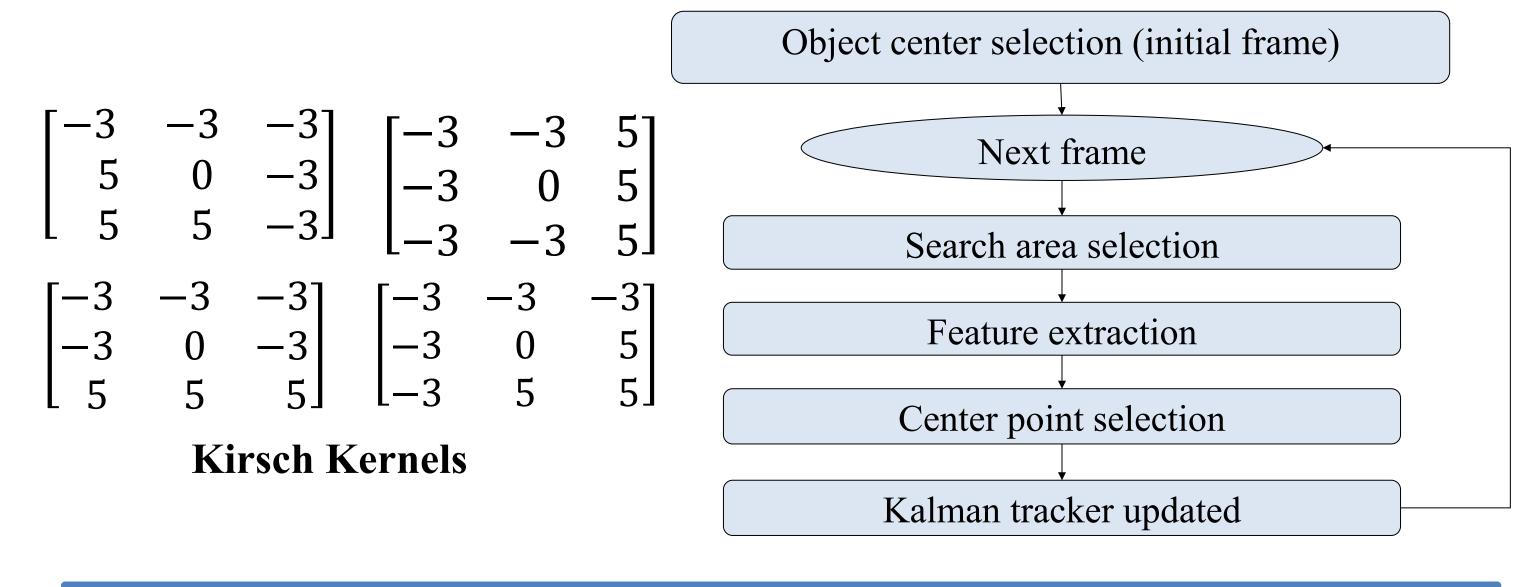




Search Area

Fused Likelihood Map

Tracking process



Kalman Tracker

Kalman tracker based on state equations of position and velocity to estimate position if an object is not detected

Each frame is processed by enhancement algorithm first. Search area is selected based on Kalman Tracker, camera properties, and target properties.

Feature is extracted using DRIFT descriptor.

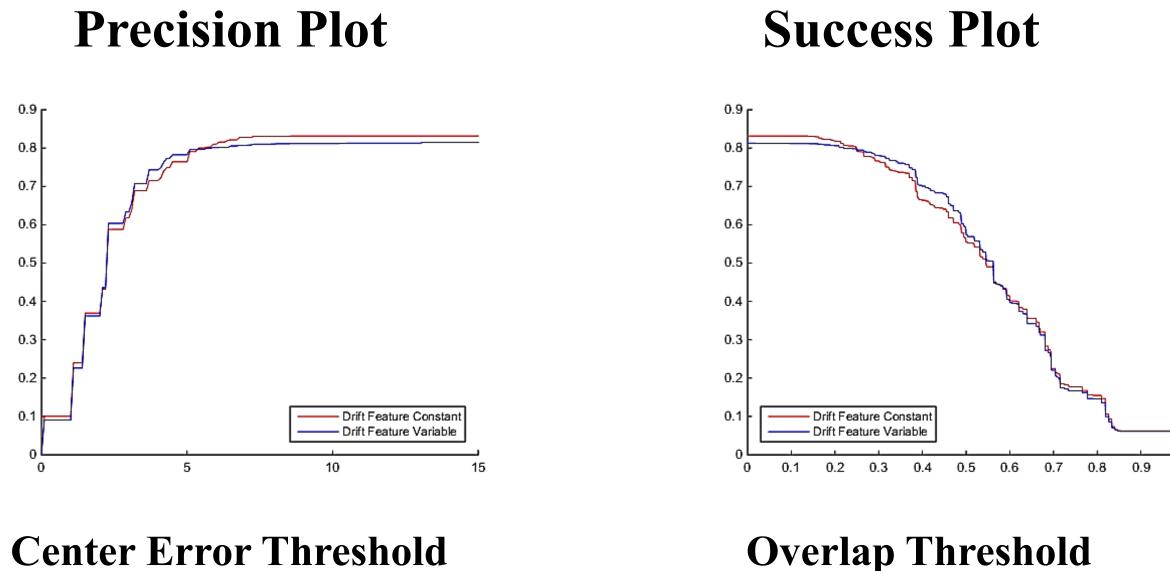
Classification is used to determine closest match.

The reference feature and Kalman Tracker are updated.

Results

- Results for constant weighting and variance ratio based weighting are compared
- 8 WAMI sequences were used for evaluations

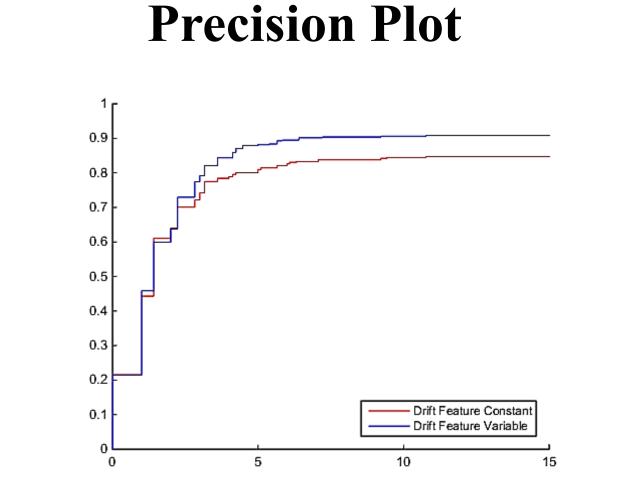
Spatial Robustness – Evaluated using different initial bounds

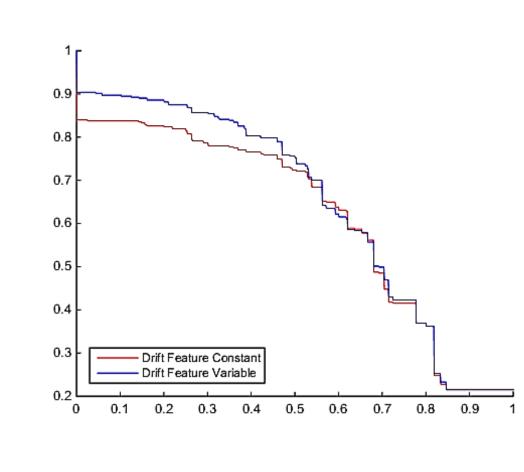


Center Error Threshold

Success Plot

Temporal Robustness — Evaluated using different starting frames





Center Error Threshold

Overlap Threshold

	Center Location Error			Frame Detection Accuracy	
	DRIFT	Drift		DRIFT	Drift
	Constant	variable		Constant	variable
CLIF1	11.22	1.31	CLIF1	0.66	0.74
CLIF2	1.55	1.86	CLIF2	0.72	0.68
CLIF3	0.43	0.49	CLIF3	0.88	0.86
CLIF4	1.76	1.16	CLIF4	0.71	0.78
BD1	40.79	42.66	BD1	0.36	0.36
BD2	0.70	0.63	BD2	0.83	0.85
BD3	1.68	1.96	BD3	0.72	0.69
BD4	25.96	12.71	BD4	0.58	0.70
Average	10.51	7.85	Average	0.68	0.71