

Introduction

- ❖ Moving object detection in a PTZ Camera
 - Video segmentation, motion clustering, semantic segmentation...
 - Background-centric approach
 - Motion compensation + Background subtraction
 - VERY fast, Unsupervised, Suitable for pre-processing



Problem

- Naive approach: Many false positive by compensation error



- Existing problem of conventional approach
 - Foreground Loss problem
 - Background Contamination Problem



Proposed Method

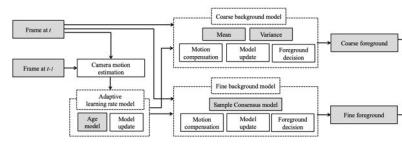
- ❖ Fine background model
 - Extend the Vibe for moving camera
 - Spatio-temporal update is applied
 - Model initialization and update rule is changed by camera movement



$$C_i = \sum_{j \in S_i} \sum_{n=1}^N \mathbf{1}(D(I_i, \tilde{v}_j^{(n)}) < R),$$

spatio-temporal
update rule

❖ Combine Two backgrounds characteristics



Algorithm 1: Updating the fine background model

for Each pixel i on current frame $I^{(t)}$ do

if $t = 0$ then

M_i is initialized to $\{I_i^{(0)}, I_i^{(1)}, \dots\}$

else

Motion compensation is applied to M_i .

if $t < N$ then

$\tilde{v}_i^{(t)}$ is removed from M_i

$I_i^{(t)}$ is inserted to M_i

Compute C_i in equation (8)

if $C_i \geq \#_{min}$ then

$P \leftarrow \min(\alpha_i, \phi)$

$p \sim \text{Uniform}(0, P - 1)$

if $p = 0$ then

$n \sim \text{Uniform}(0, N - 1)$

$\tilde{v}_i^{(n)}$ is removed from M_i

$I_i^{(t)}$ is inserted to M_i

$p_2 \sim \text{Uniform}(0, P - 1)$

if $p_2 = 0$ then

$k \sim \text{Uniform}(0, K)$

$j \leftarrow S_i(k)$

$n \sim \text{Uniform}(0, N - 1)$

$\tilde{v}_j^{(n)}$ is removed from M_j

$I_i^{(t)}$ is inserted to M_j

Experiments

❖ F-measure of Moving camera dataset

Method	walking	skating	woman	woman2	fence	ground1	ground2	ground3	ground4	ground5	average
ViBe* [6]	0.0375	0.2229	0.0375	0.0929	0.1042	0.5656	0.4733	0.4118	0.0299	0.1309	0.2107
FIC* [8]	0.0613	0.2373	0.0361	0.1345	0.0954	0.4543	0.4108	0.1538	0.0453	0.1319	0.1761
BMRI-ViBe* [9]	0.0438	0.2402	0.0400	0.0921	0.1104	0.4249	0.3868	0.2161	0.0383	0.1377	0.1730
MCD NP [25]	0.4351	0.4164	0.4935	0.5791	0.2691	0.2773	0.3750	0.1222	0.1969	0.3540	0.3519
MCD 5.8ms [26]	0.7349	0.2447	0.3395	0.3448	0.7357	0.6573	0.7177	0.1531	0.5274	0.0678	0.4523
Stochastic approx [28]	0.8335	0.6543	0.3986	0.8783	0.8788	0.2221	0.2792	0.0181	0.0111	0.2181	0.4392
FP Sampling [27]	0.7058	0.8539	0.7268	0.5828	0.7654	0.7977	0.8306	0.1396	0.4226	0.8212	0.6646
SC MCD [29]	0.7496	0.8560	0.6650	0.6311	0.7637	0.8965	0.9118	0.8843	0.8824	0.9326	0.8173
uNLC [32]	0.0158	0.1419	0.0178	0.0487	0.0346	0.0570	0.0342	0.0216	0.0031	0.0143	0.0389
OSVOS [1]	0.3397	0.5344	0.0121	0.1260	0.7033	0.7697	0.5447	0.9696	0.0050	0.1224	0.4127
CIS [33]	0.0538	0.3036	0.1522	0.4681	0.1180	0.1545	0.0862	0.0581	0.0046	0.0184	0.1418
BASNet [34]	0.3433	0.9379	0.0205	0.2289	0.2119	0.6039	0.9564	0.9586	0.9439	0.9829	0.6188
Proposed method	0.7809	0.9600	0.7269	0.7065	0.8081	0.9037	0.9032	0.8700	0.9080	0.9793	0.8546

❖ Video object segmentation measure

Measure	Mean \mathcal{J}	Recall \mathcal{J}	Mean \mathcal{F}	Recall \mathcal{F}
ViBe* [6]	0.2095	0.1364	0.1717	0.0773
FIC* [8]	0.1701	0.0607	0.2256	0.1337
BMRI-ViBe* [9]	0.1640	0.0553	0.1703	0.0817
MCD NP [25]	0.2634	0.0580	0.5569	0.7090
MCD 5.8ms [26]	0.3736	0.3756	0.5427	0.6100
Stochastic approx [28]	0.3398	0.7889	0.4003	0.4245
FP Sampling [27]	0.4294	0.5009	0.6031	0.7156
SC MCD [29]	0.5213	0.5952	0.7021	0.8200
uNLC [32]	0.1073	0.1002	0.1416	0.1181
OSVOS [1]	0.2547	0.2259	0.4129	0.3068
CIS [33]	0.1583	0.0591	0.2356	0.1253
BASNet [34]	0.5540	0.6204	0.6696	0.6880
Proposed method	0.5603	0.6541	0.7214	0.8378

\mathcal{J} : region-based segmentation similarity \mathcal{F} : contour-based accuracy

❖ Ablation study of two backgrounds

Method	precision	recall	F-measure
coarse BG model	0.9084	0.7655	0.8248
fine BG model	0.5669	0.7833	0.6095
combined model	0.9286	0.8041	0.8546

❖ Computational loads

CPU only, 320 x 240, 45.5fps

Module	Time (millisecond)
Motion estimation	2.207
Motion compensation	5.117
Age map update	0.595
Background model update	13.902
Foreground combining	0.671
Total	22.492

❖ Robustness test to noise image

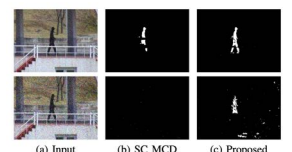
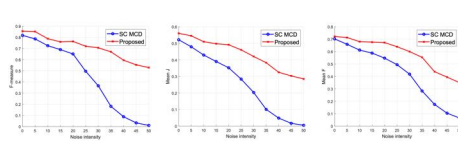
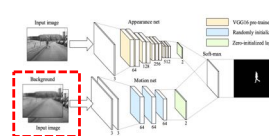


Fig. 7. Example results for the noisy images. Each row shows the experimental results when the noise mean of image is 25 and 50, respectively.

❖ Combined with supervised method (AMNet)



Method	F-measure
AMNet [37] using MCD 5.8ms [26]	0.8789
AMNet [37] using SC MCD [29]	0.9175
AMNet [37] using Proposed BG	0.9529

References

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