# CONTRIBUTIONS TO THE SEGMENTATION OF MOVING OBJECTS IN VIDEO SEQUENCES

#### Author:

## Maria Isabel de Castro Lopes Martins Pinto Ferreira

#### **Motivation of the work**

Unsupervised segmentation of moving objects in video sequences, by background subtraction (BS):

- fundamental step in many vision systems and critical factor for the success of the overall system,
- hard and challenging task.

Complex and challenging scenarios include poor lighting conditions, sudden illumination changes, nighttime videos, shadows, parasitic background motion, ...

#### **State-of-the-art methods** [1]:

- no method has been able to fully deal with all challenges,
- most widely used methods build a statistical model of background pixels [2],
- methods appear to be complementary in nature [3],
- better performance usually at the cost of significant increase in complexity and computational load.

#### **Comparison of methods:**

- most datasets do not contain a balanced set of videos presenting real challenges,
- hard or impossible to compare results computed on different datasets,
- metrics used to evaluate only average performance do not reveal performance frame by frame.

#### **Thesis Objectives**

In this PhD research we want to further explore some of the most efficient approaches to propose a more robust algorithm.

#### **Research Question:**

Is it possible to improve widely used approaches, such as GMM, by proposing more robust algorithms while keeping complexity low?

Can this method be further improved in order to cope with more difficult cases, such as nighttime videos, even if sacrificing the complexity?

Thesis Advisors: José Luís Alba Castro, Associate Professor, EET, U.Vigo, Spain Luís Corte-Real, Associate Professor, FEUP, U.Porto, Portugal

#### **Research Plan**

Yea

- ✓ State-of-the-art literature review.
- ✓ Development of a new scheme exploiting two different approaches: a bio-inspired motion detection method and a BS algorithm based on color information.
- ✓ Review of objective segmentation evaluation methods in order to identify different metrics that could be used in this context [7].
- ✓ Exploration of the discrimination capability of different color spaces in the context of a GMM-based algorithm.
- ✓ Development of a robust and computationally efficient method, based on GMM [6], suitable for real-time applications.
- ✓ Development of a new approach to model the local texture at the pixel neighborhood.
- ✓ Proposal of a new method, based on a combination of local texture and pixel color representations, to address the problem of moving objects segmentation in night videos.
- ✓ Preparation, submission and publication of a paper in an international journal
- ✓ Preparation and submission of a new paper to an international journal
- ✓ Writing of the PhD dissertation
- Thesis conclusion and submission (Oct/2018)
- PhD defense

**Next Year Planning** 

#### **Publications & International Conferences Presentations**

- Martins, I., Carvalho, P., Corte-Real, L., Alba-Castro, J.L. (2016) Bio-inspired Boosting for Moving Objects Segmentation. In: A. Campilho, F. Karray (eds) Image Analysis and Recognition. ICIAR 2016. Lecture Notes in Computer Science, vol. 9730, pp. 397-406, Springer, Cham.
  - > Oral presentation at ICIAR 2016, July 13-15, 2016, Póvoa de Varzim, Portugal.
- Martins, I., Carvalho, P., Corte-Real, L., Alba-Castro, J.L. (2017) BMOG: Boosted Gaussian Mixture Model with Controlled Complexity. In: L. A. Alexandre, J. S. Sánchez, J. M. F. Rodrigues (eds) Pattern Recognition and Image Analysis. IbPRIA 2017. Lecture Notes in Computer Science, vol. 10255, pp. 50-57, Springer Int. Publishing.
  - > Oral presentation at IbPRIA 2017, June 20-23, 2017, Faro, Portugal.

- Martins, I., Carvalho, P., Corte-Real, L., Alba-Castro, J.L. (2018) BMOG: boosted Gaussian Mixture Model with controlled complexity for background subtraction. Pattern Analysis and Applications (2018). DOI: 10.1007/s10044-018-0699-y.
- Martins, I., Carvalho, P., Corte-Real, L., Alba-Castro, J.L., Texture Collinearity Foreground Segmentation for Night Videos (provisional title), submitted to IEEE Transactions on Circuits and Systems for Video Technology on May 2018.

#### **Results & Discussions**

## **Bio-Inspired Boosting for Moving Objects Segmentation**

(presented and discussed in 2016 Annual Defense)

The main novelty introduced is the fusion of low-level information from the modeling of the human visual system [5] with state-of-the-art methods used in BS.

#### **BMOG:** boosted **Gaussian Mixture Model with** controlled complexity for background subtraction

Explores a **novel** classification mechanism that combines:

- Color space discrimination capabilities
- Pixel classification with hysteresis and
- **Dynamic learning** rate, independent for each pixel
- **Testing & Evaluation:** CDnet 2014 dataset;
- submitted to **CDnet** benchmark [4]; • compared with: MOG2 [6], RMoG,

SuBSENSE.

Mask 1120 Frame 1130 1100 1100 1110 1120 **Frame** 1130 1150 1100 1110 1120 **Frame** | 1130 \*\*\*\*\*\*\*\*\* 1120 **Frame** 1130 1150 Fig. 1. Visual example of the dynamic behaviour: (from top) sample frames from input video highway; corresponding segmentation

masks; pixel classification as BG/FG; decision threshold for channel L, th L; number of components in the mixture, M; learning rate  $\alpha$ .

Results show that BMOG consistently outperforms the widely used MOG2 method, and that it approaches top ranking, but much more complex, algorithms.

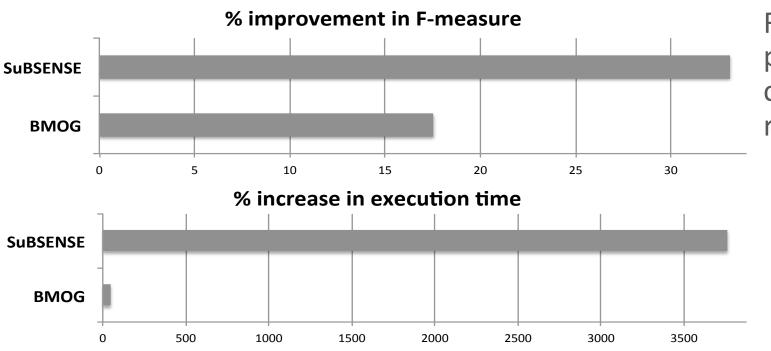


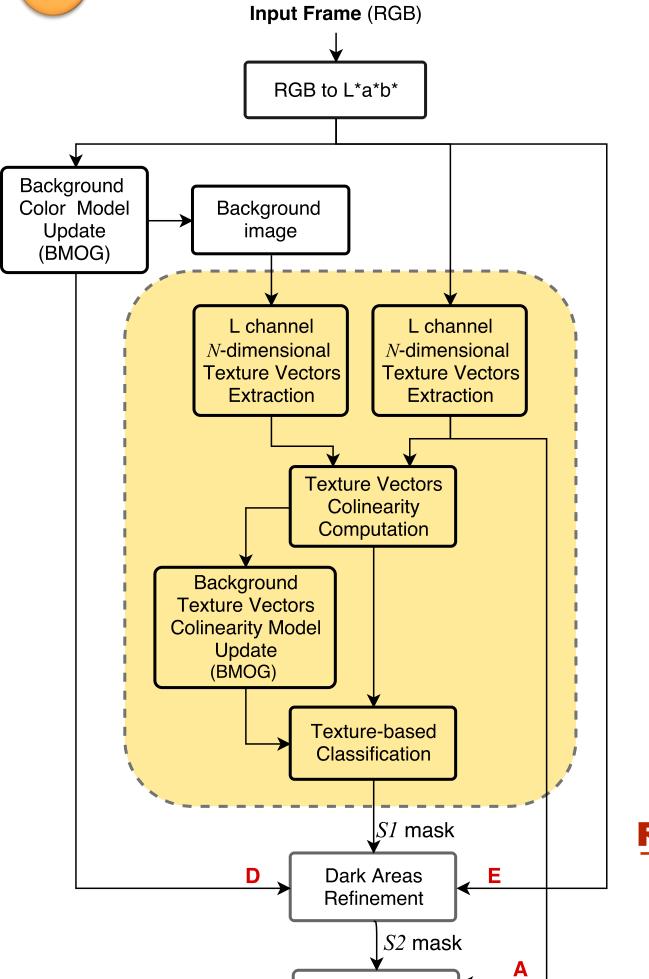
Fig. 2. Comparison of improvement in performance and increase in complexity for BMOG and a topranked method (reference: 0=MOG2).

- **Boosts the overall detection** accuracy while keeping complexity low.
- Good choice for real-time applications.

Presented and published at IbPRIA2017 proceedings. **NEW** Invited, as a revised and extended version, for Springer journal Pattern Analysis and Applications. Online: 02/4/2018.

#### **COLBMOG: Texture Collinearity Foreground Segmentation for Night Videos**

The proposed method is based on a local texture feature integrated with a parametric background model (BMOG).



Textureless FG

Post-Processing

Final COLBMOG Mask

Fig. 4. COLBMOG block diagram.

Input

S3 mask

CBM mask

**B** Previous

\_ (L\*a\*b\*)

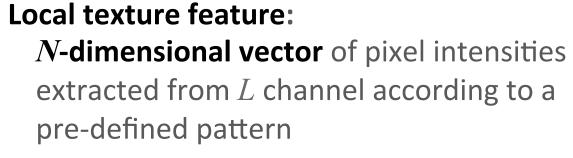


Fig. 3. Pattern of neighboring pixels representing local texture at pixel o.

### Measure of similarity between both textures at pixel *j*:

the collinearity between the corresponding texture vectors in the input frame,  $\vec{\chi}_i$ , and in the background image,  $\vec{b}_i$ .

Measure of collinearity: the angle between the texture vectors,  $\theta(\vec{x}_i, \vec{b}_i)$ .

A model of the collinearity between texture vectors of background pixels in successive frames, based on BMOG, is created and updated at every frame.

**Testing & Evaluation: CDnet 2014 NightVideos (NV)** category benchmark [4]

Table 1. F-measure across the overall set of videos for COLBMOG and EFIC. C-EFIC (the two top-ranked unsupervised methods in NV category) [4]

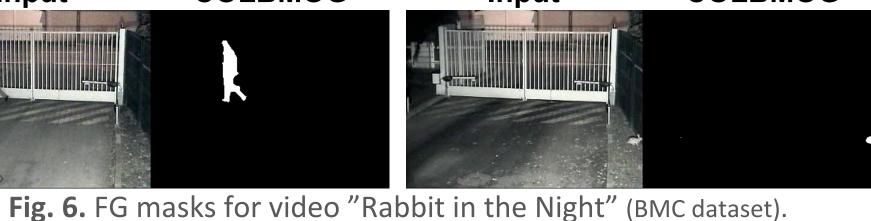
F-Measure	EFIC	C-EFIC	COLBMOG
Average	0,6548	0,6677	0,7564
St Dev	0,1245	0,1034	0,0435

#### **RANKS FIRST** in CDnet NV benchmark [4] for the unsupervised methods.

Not only the average F-measure is higher but also the standard deviation is significantly lower, meaning a more consistent performance across different challenges.

**NEW Tests** using the Background Model Challenge (BMC) dataset. **COLBMOG** Input

COLBMOG



"hard"

Fig. 5. Comparison of FG masks. Misclassified pixels are marked red.

**NEW** A paper with the provisional title *Foreground Segmentation for Night Videos* was submitted to IEEE Transactions on Circuits and Systems for Video Technology, on May 2018, and is now in the peer review stage.

#### References

- [1] Bouwmans, T.: Traditional and recent approaches in background modeling for foreground detection: An overview. Computer Science Review 11, 31-66, May (2014)
- [2] Stauffer, C., Grimson, E.: Adaptive background mixture models for real-time tracking. IEEE Int. Conf. Comput. Vision and Patt. Recogn. (CVPR) 2, 246–252, (1999) [3] Wang, Y., Jodoin, P.-M., Porikli, F., Konrad, J., Benezeth, Y., Ishwar, P.: CDnet 2014: An Expanded Change Detection Benchmark Dataset. In Proc. CDW-2014, at CVPRW-2014, pp. 387–394 (2014)
- [4] ChangeDetection.NET (CDNET), http://www.changedetection.net [5] Benoit, A., Caplier, A., Durette, B., Herault, J.: Using human visual system modeling for bio-inspired low level image processing. Comput. Vis. Image Underst. 114,(7), 758–773 (2010)
- [6] Zivkovic, Z., van der Heijden, F.: Efficient adaptive density estimation per image pixel for the task of background subtraction. Patt. Recogn. Lett 27(7), 773–780 (2006) [7] Cardoso, J. S., Carvalho, P., Teixeira, L. F., Corte-Real, L.: Partition-distance methods for assessing spatial segmentations of images and videos. Comput. Vis. Image Underst. 113 (7), 811–823 (2009)