

# Master in Computer Vision Barcelona

Project Module 4 Coordination

**Week 2: Tasks Description** 

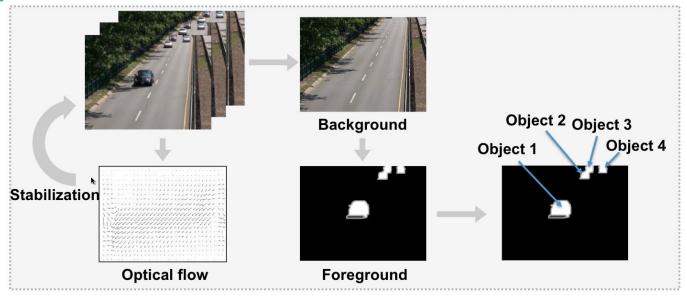
Video Surveillance for Road Traffic Monitoring J. Ruiz-Hidalgo / X. Giró

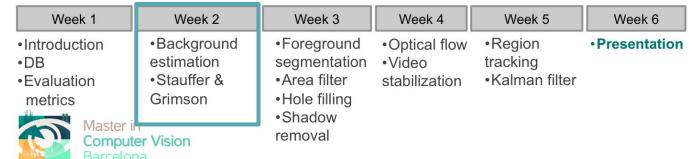
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**Project Schedule** 



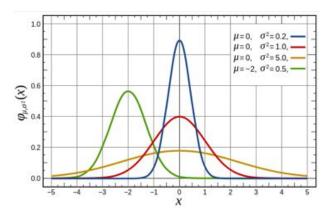


## Goals Week 2

#### Background estimation

- Model the background pixels of a video sequence using a simple statistical model to classify the background / foreground
  - Single Gaussian per pixel
  - Adaptive / Non-adaptive
- The statistical model will be used to preliminary classify foreground

 Comparison with more complex models (Stauffer and Grimson)



## **Tasks**

## Mandatory

- Task 1: Gaussian distribution
- Task 2 & 3: Evaluate results
- Task 4: Recursive Gaussian modeling
- Task 5: Evaluate and compare to non-recursive

## Optional

- Task 6: Compare with S&G
- Task 7: Color sequences

# **Sequences**

ID	FRAME RANGE	TYPE
Highway	1050 - 1350	Baseline
Fall	1460 - 1560	Dynamic background
Traffic	950 - 1050	Camera jitter







## **Reminder: metrics**

### The groundtruth images contain 5 labels namely

o 0 : Static

50 : Hard shadow

85 : Outside region of interest

• 170 : Unknown motion (usually around moving objects, due to semi-transparency and motion blur)

o 255 : Motion

#### We will use:

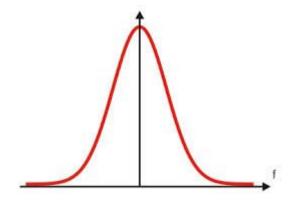
o Background: 0, 50

o Foreground: 255

Unknown (not evaluated): 85, 170

# Task 1: Gaussian modelling

- 1 Gaussian function to model each background pixel
  - First 50% of the test sequence to model background
  - Mean and variance of pixels



### Second 50% to segment the foreground

```
for all pixels i do

if |I_i - \mu_i| \ge \alpha \cdot (\sigma_i + 2) then

pixel \to Foreground

else

pixel \to Background

end if

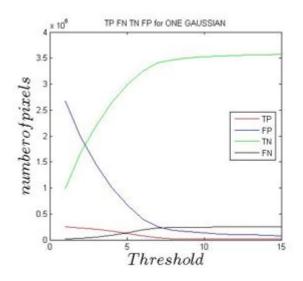
end for
```

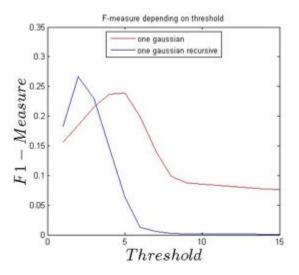
 $\triangleright +2$  to prevent low values of  $\sigma_i$ 

## **Task 2: Evaluation**

#### Evaluate Task 1

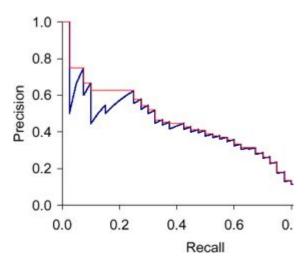
True Positive, True Negative, False Positive, False Negative, Precision, Recall, F1-score vs alpha





## **Task 3: Evaluation**

- Evaluate Task 1
  - Precision vs Recall curve
  - Area Under the Curve (AUC)



# Task 4: Adaptive modelling

### Adaptive modelling

- First 50% frames for training
- Second 50% left background adapts

$$\begin{aligned} & \textbf{if pixel } i \in \text{Background then} \\ & \mu_i = \rho \cdot I_i + (1 - \rho) \cdot \mu_i \\ & \sigma_i^2 = \rho \cdot (I_i - \mu_i)^2 + (1 - \rho) \cdot \sigma_i^2 \\ & \textbf{end if} \end{aligned}$$

### • Best value of $\alpha$ , p to maximize F1-score

- o Two methods:
  - Obtain best  $\alpha$  non-recursive and then obtain p for the recursive cases
  - Optimize them together

# **Task 5: Comparison**

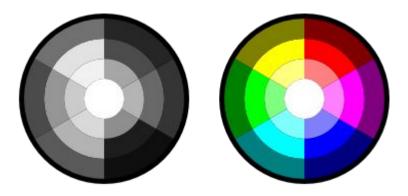
- Compare both the adaptive and non-adaptive version and evaluate them for all 3 sequences proposed
  - F1-score / AUC

# **Task 6: Optional**

- Compare with Stauffer and Grimson
  - Implementation
    - Matlab → Computer Vision Toolbox (preferred) or provided (StGm.
       zip)
    - Python → OpenCV
    - Select the best number of Gaussians (3 to 6)
  - Evaluate precision vs recall to comment which method (single Gaussian programmed by you or S&G) performs better
  - Evaluate the sequences than benefit more of the multiple Gaussians and try to explain why

# Task 7: Optional

- Update your implementation to support color sequences
  - Decide color space? RGB vs YUV?
  - Number of Gaussians needed?



## **Deliverables**

- Google drive with slides per tasks
- Code used for the week assignment

- 6th January
  - Upload link to GitHub
  - Fill the intra-group evaluation

# **Scoring Rubric**

Grade is assigned based on the satisfactory accomplishment of...

Grade	Common meaning	Succesfully completed tasks
9-10	Excellent	All mandatory and two optional tasks
7-9	Very good	All mandatory and one optional tasks
5-7	Average	All mandatory tasks
3-5	Difficulties	All mandatory tasks but one
0-3	Fail	All mandatory tasks but two or more