

On-Line Object Behaviour Analysis for Surveillance Systems

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Surveillance system

- net of sensors (cameras, microphones)
- public and private places monitoring
- actually low level of signal processing in practice
- no possibility for later querying the surveillance data
- many human-resources “exploring” the data

Specificity

- surveillance video system for underground scenarios
- data characteristics
 - high noise, low resolution
 - bad but stable lighting conditions
- algorithm demands
 - work in real-time
 - robustness

Event types

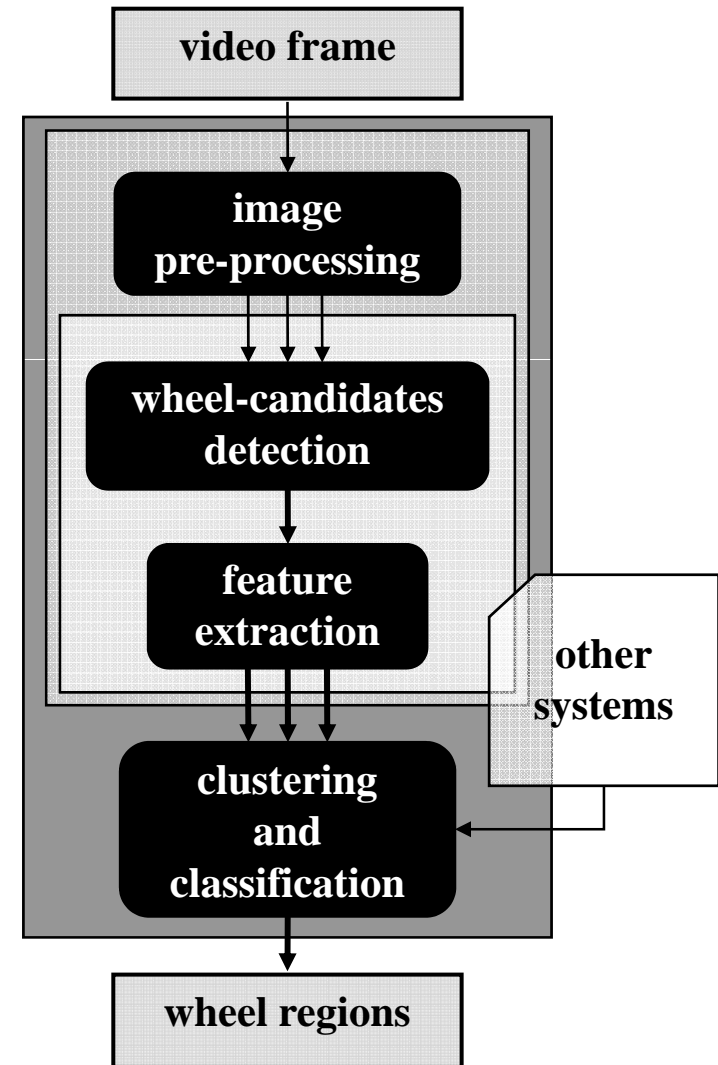
- statistical approaches for ***simple event*** detection
 - ad-hoc *bicycle detector*
 - *dog detector* based on AdaBoost classifier
 - *trajectory classification* of general moving objects based on HMM
- spatio-temporal rules for ***compound event*** detectors
 - *dangerous occurrence of the person on the platform edge*

Bicycle Detection



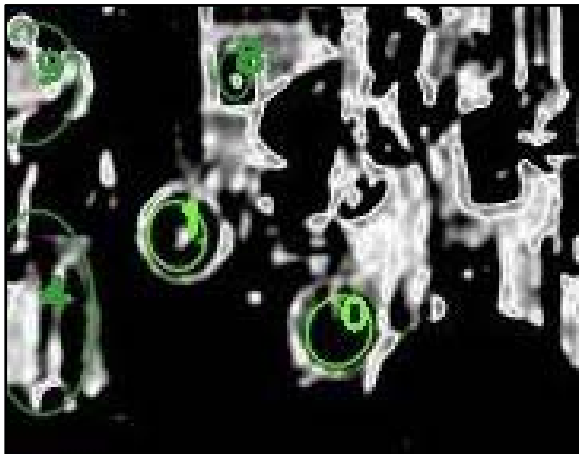
Bicycle Detection

- image pre-processing
 - several probability levels based on wheel colour model
- wheel candidates on each level
- classification & validation
- candidate's clustering over levels



Candidates Classification

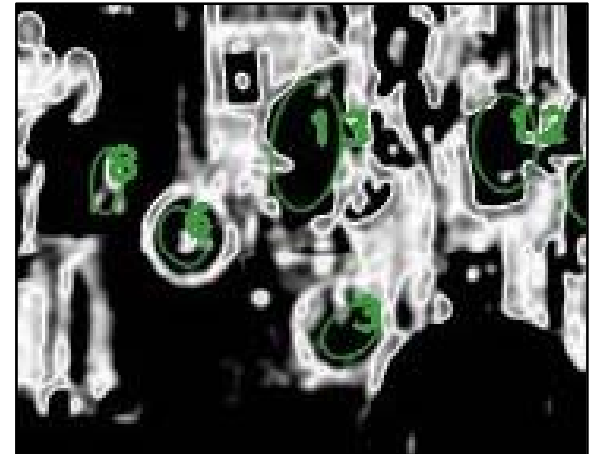
- feature extraction
 - geometrical constraints
 - multi-scale template matching
 - ellipse profiling



Level 0



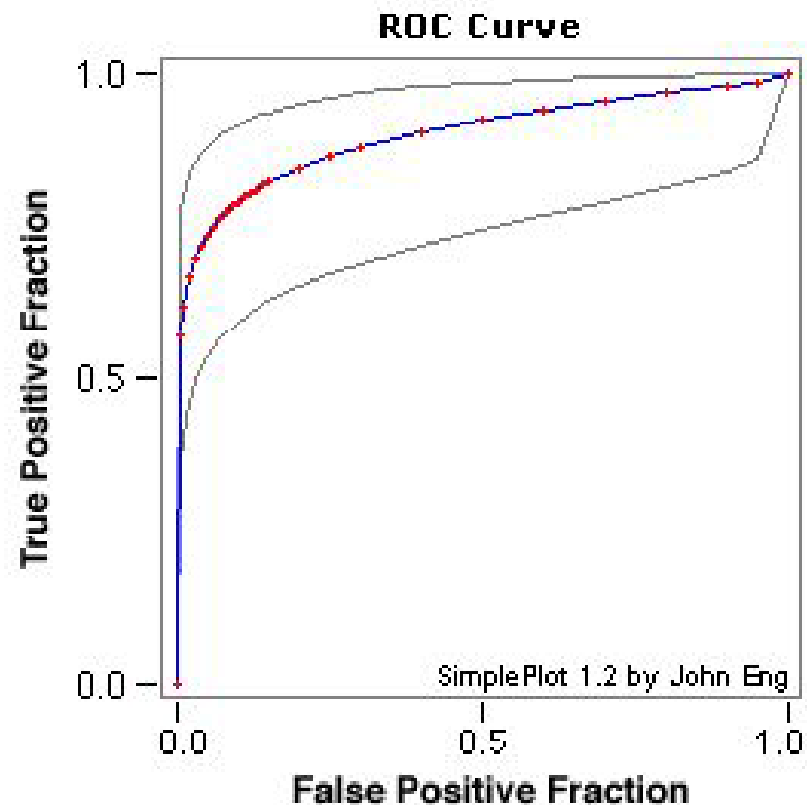
Level 1



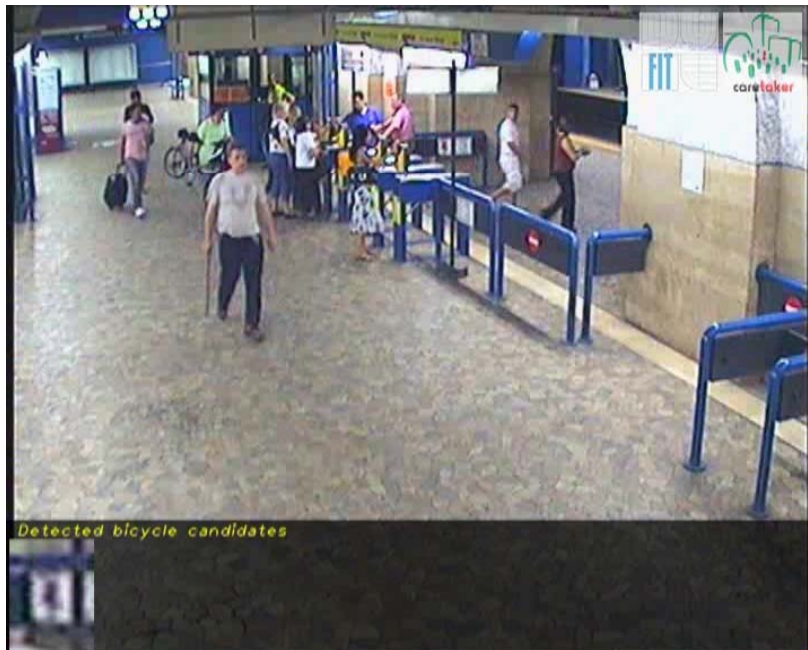
Level 2

Evaluation

- images from video of Roma underground
- resolution 720x576 using format 24bits/pixel
- RGB colour model
- evaluation dataset
 - 60 still images containing bicycles
 - manually annotated
- about 120 positive samples (wheels)



Examples



Dog Detection



AdaBoost – Adaptive Boosting

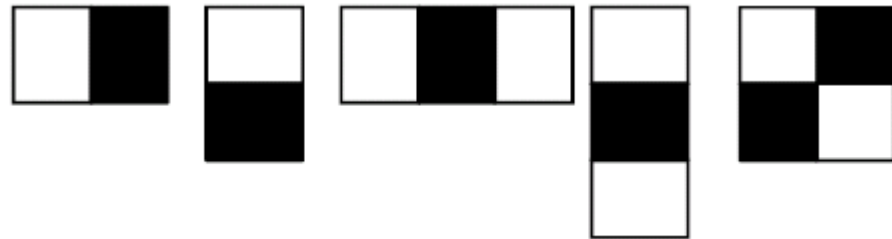
- linear combination of *weak classifiers*

- error < 0.5

$$H(\mathbf{x}) = \text{sign}[\alpha_1 h_1(\mathbf{x}) + \alpha_2 h_2(\mathbf{x}) + \alpha_3 h_3(\mathbf{x}) \dots \alpha_T h_T(\mathbf{x})]$$

- iterative training – selection of weak classifiers

- *Haar wavelets*

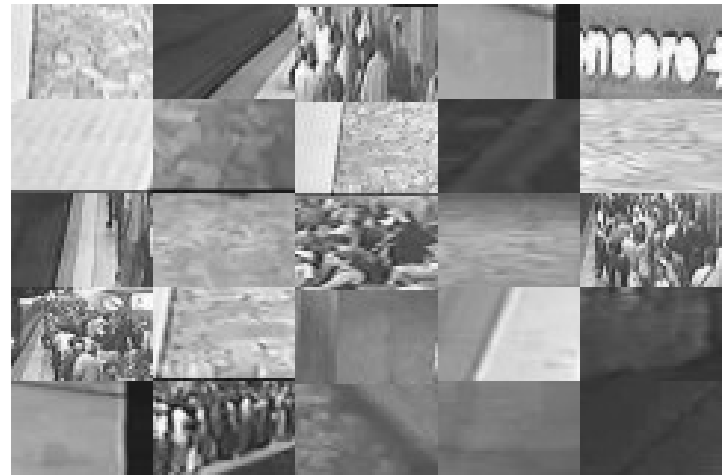


- *cascade* of classifiers – early results
- based on *integral image*

Training and Evaluation

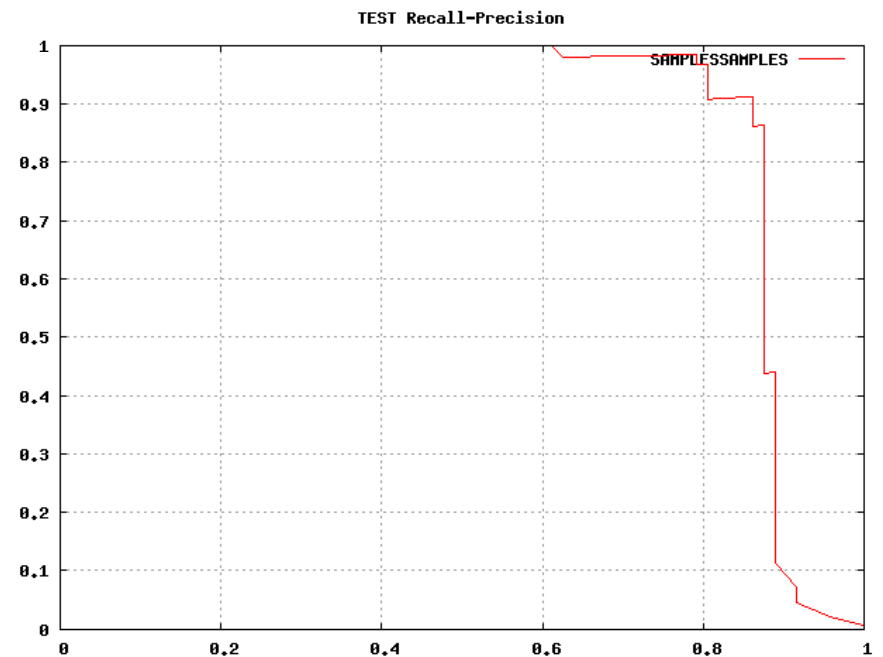
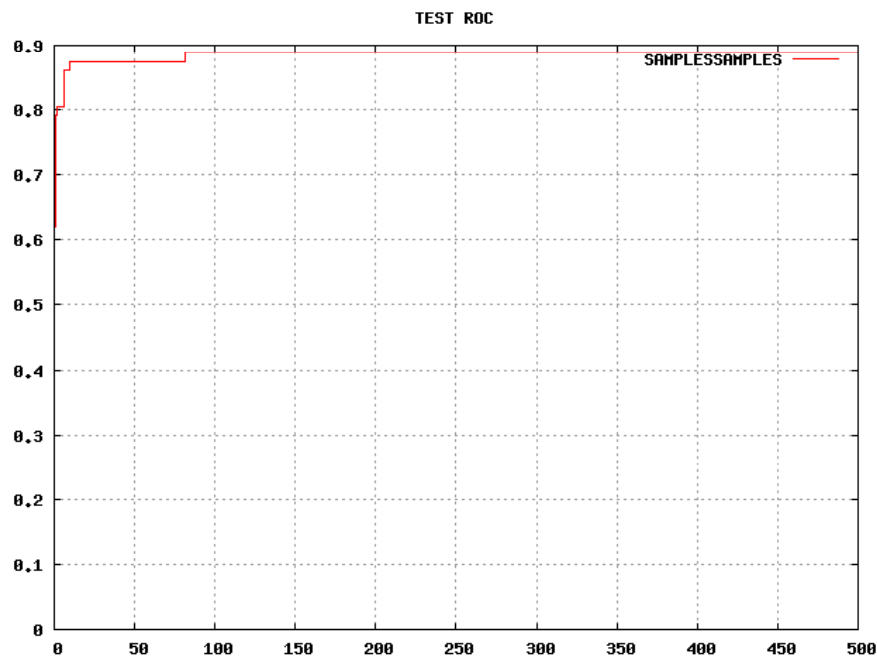
- not ideal data for underground detection
 - mostly taken from outdoor scenes
- positive samples
 - several hundreds of dog images
- negative samples (background)
 - random sub-windows from underground videos (practically unlimited amount)

	Training	Validation	Testing
dog	236	190	162
background	21292	21291	10646



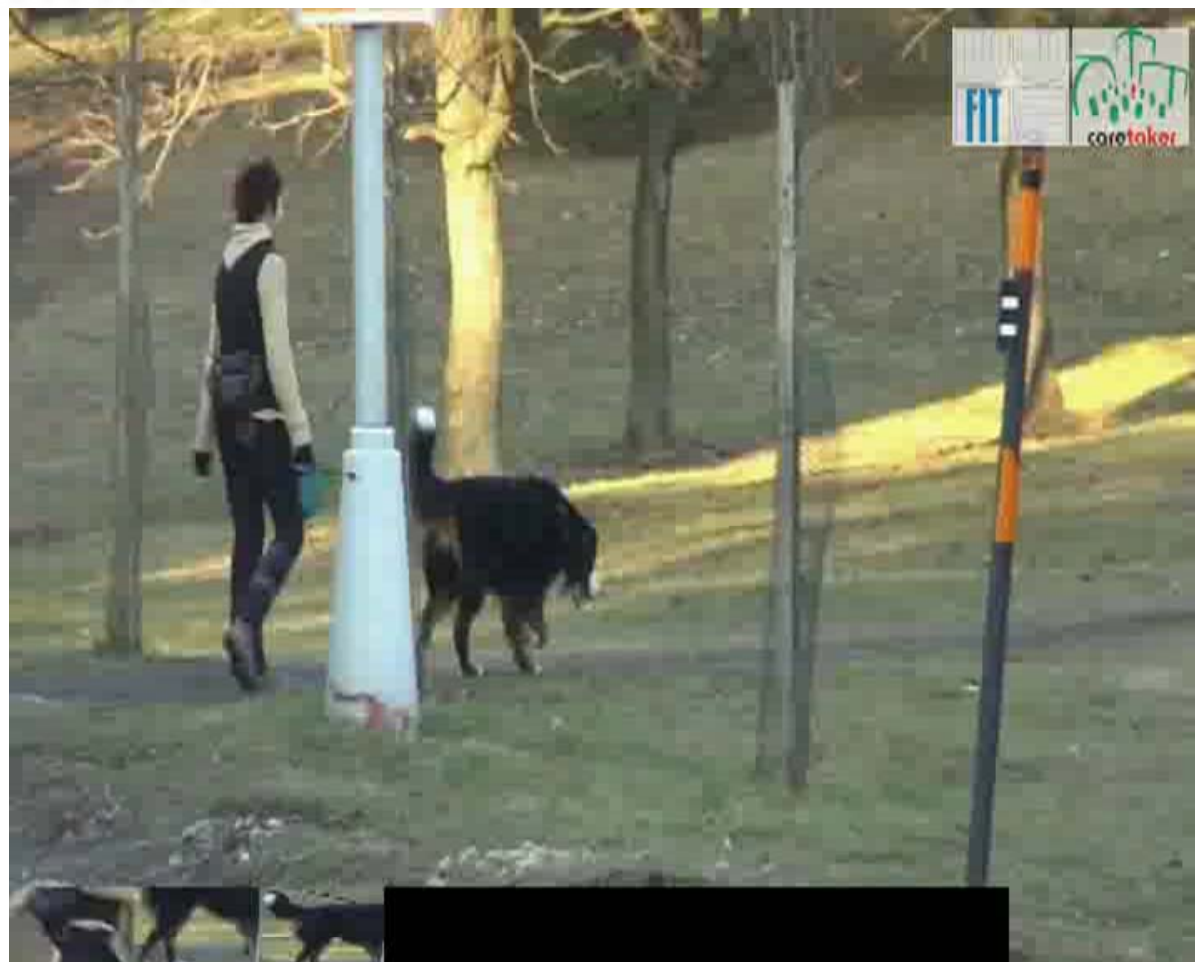
Results

- classifiers trained on data that differ from detection application
- classifier trained for dark and light dogs together



ROC and PRC of dog detection classifier ($\alpha=0.05$ – 5% False Negative Rate)

Examples



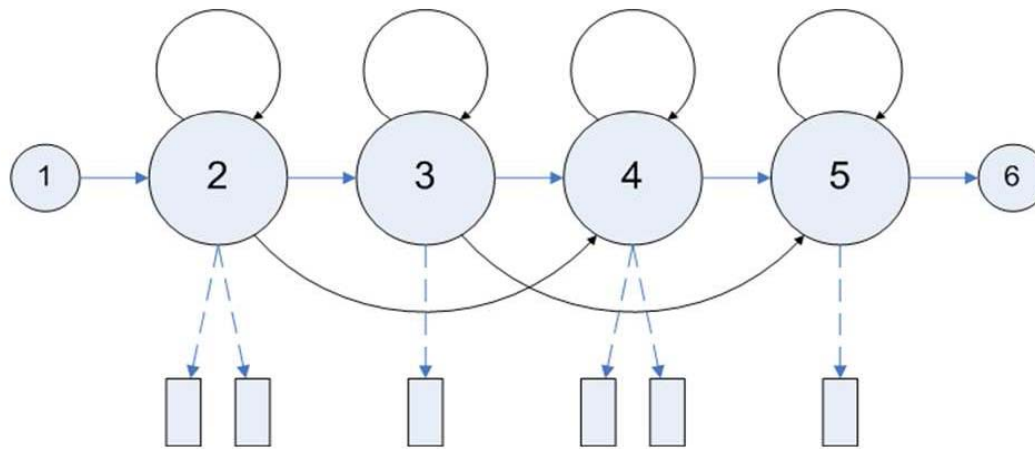
Abnormal Behaviour Detection



Trajectory

- spatial information about object in time
- object detection
 - simple activity detection based on foreground detection
- quadruple
 - position x, y
 - velocities dx, dy

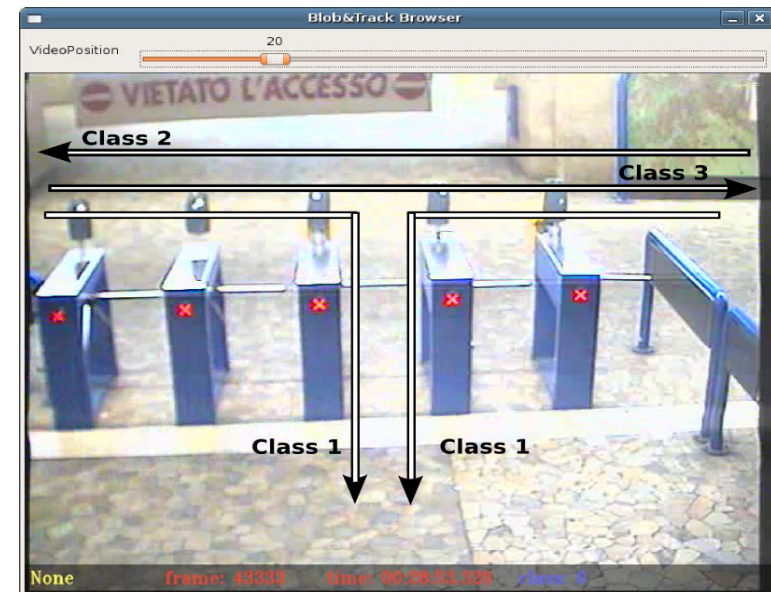
Hidden Markov Models



- hidden states
- transition probabilities between states
- train the model by changing transition weights
- model evaluation – find *best way* through the model
- usage – walk classification, gesture recognition, etc.

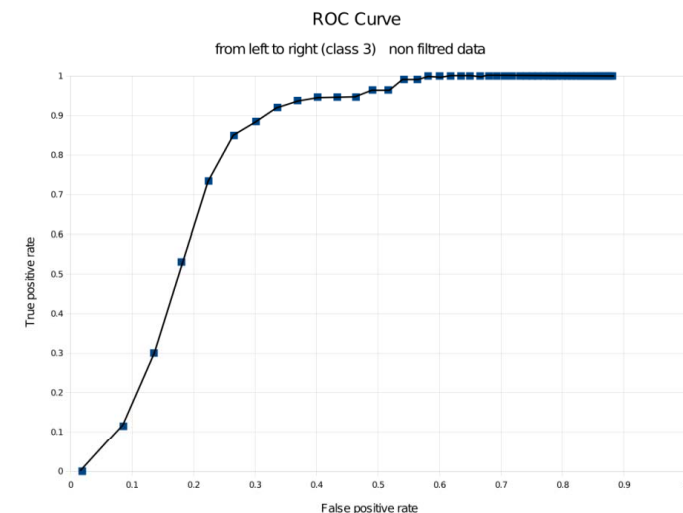
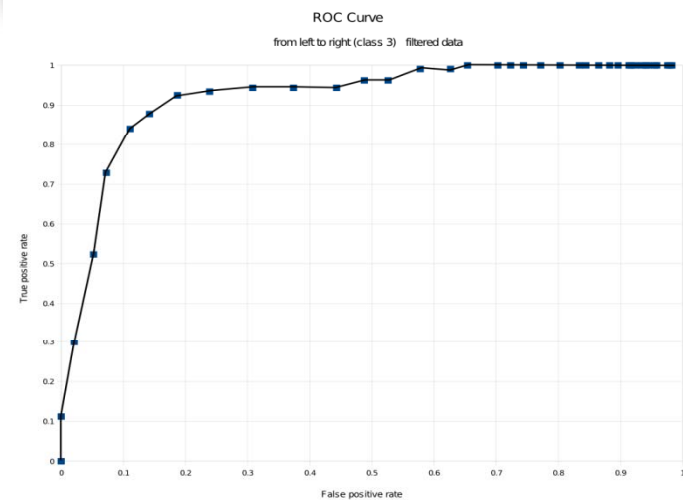
Trajectory Classification

- each scene needs particular configuration
- definition of all normal trajectories in scene
- data annotation
- model estimation
 - each trajectory class is modelled by HMM
- unusual trajectory
 - none of known class



Trajectory classification

- classifier behaviour described by ROC
 - example of class 3
- results on filtered data
 - containing only well defined trajectories
 - classifier accuracy 92%
- results on non-filtered data
 - containing also merged trajectories and ambiguous trajectories
 - classifier accuracy 37%



Example

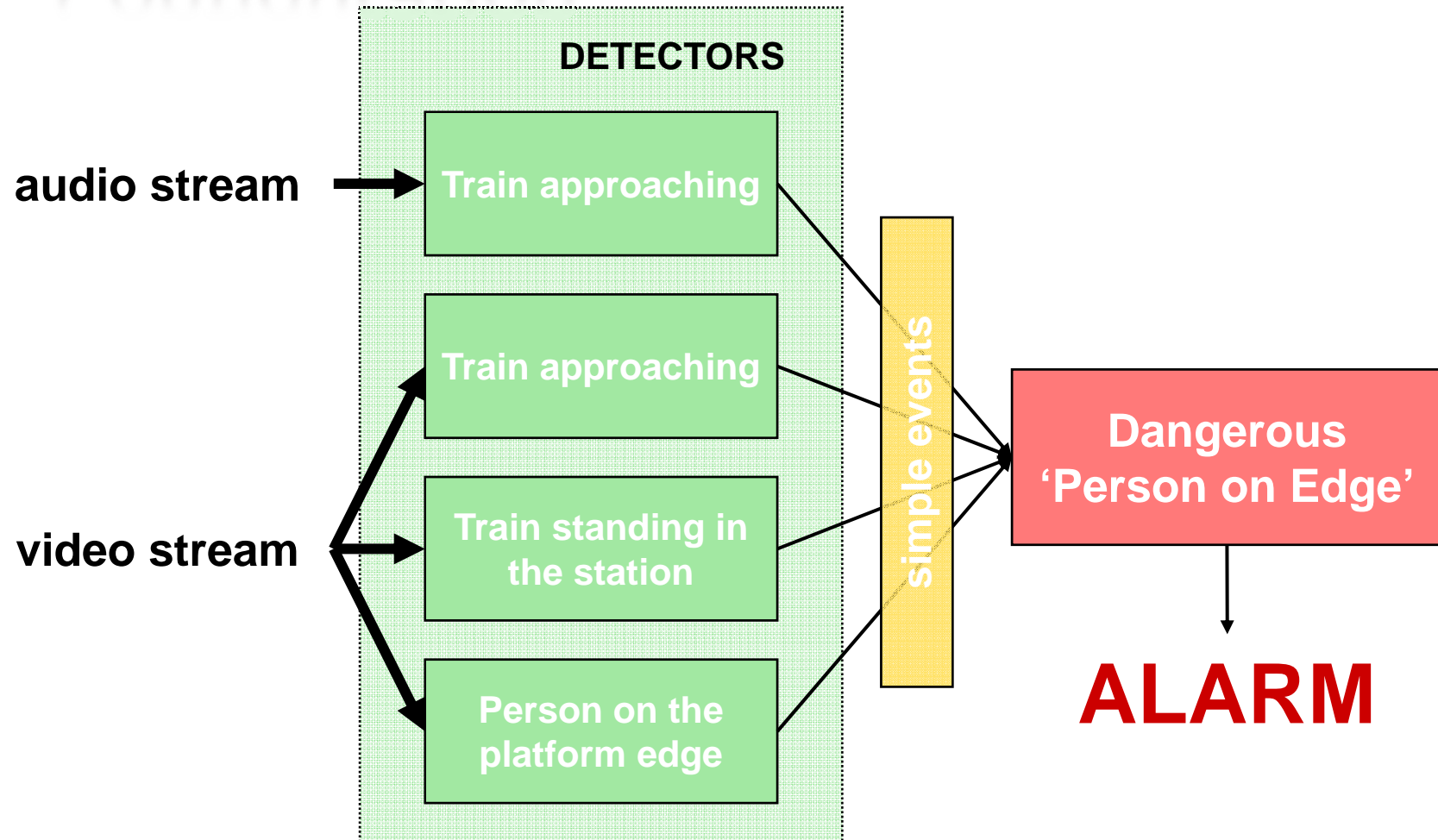


On-line Compound Event

- detection of complex audio/video events
- temporal and spatial relations

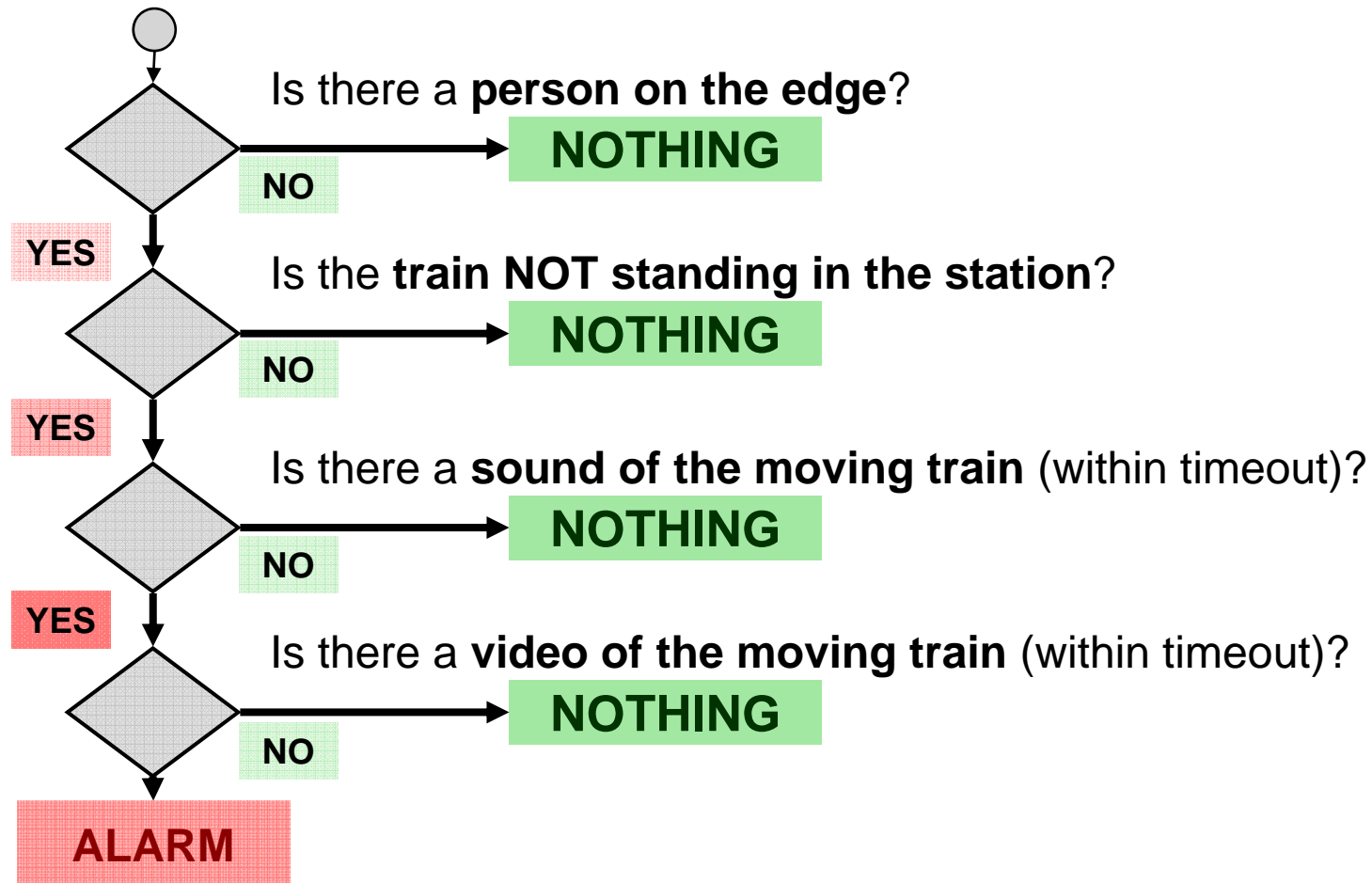
“dangerous occurrence of a person on an edge of a platform”

Configuration

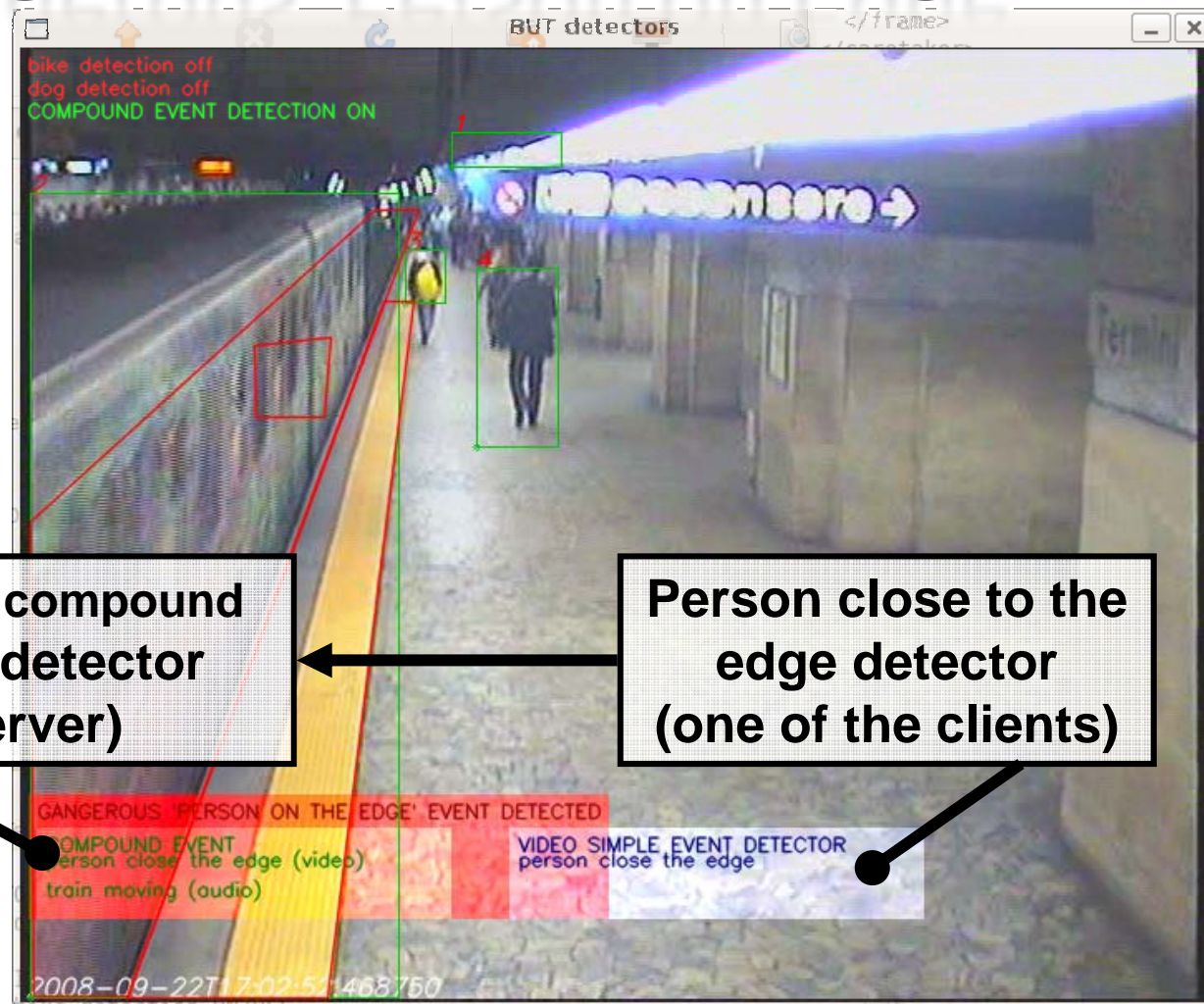


Rules

- Rule based compound audio/video event evaluation



Dangerous 'Person on Edge'



**On-line compound
event detector
(server)**

**Person close to the
edge detector
(one of the clients)**

Conclusion

- methods are implemented in the running surveillance system (CARETAKER project)
 - underground stations in Roma and Torino
- methods was proven to be usable in such scenarios
- results are also used to off-line events processing (data mining, statistics, etc.)

Thank you for your attention.