

# Identifying foreground objects from IP cameras



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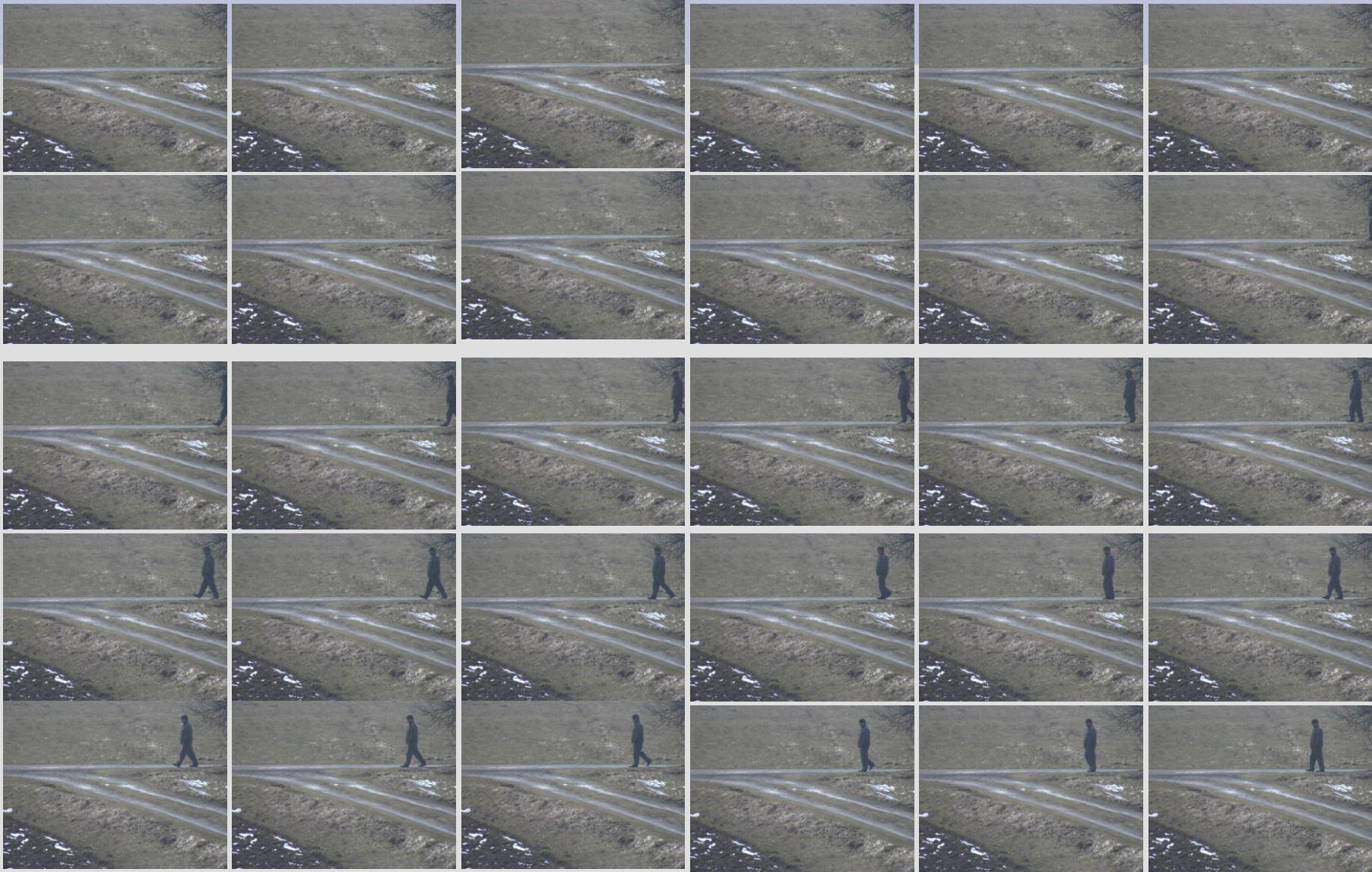
University of Ostrava

Faculty of science

## Typical records example

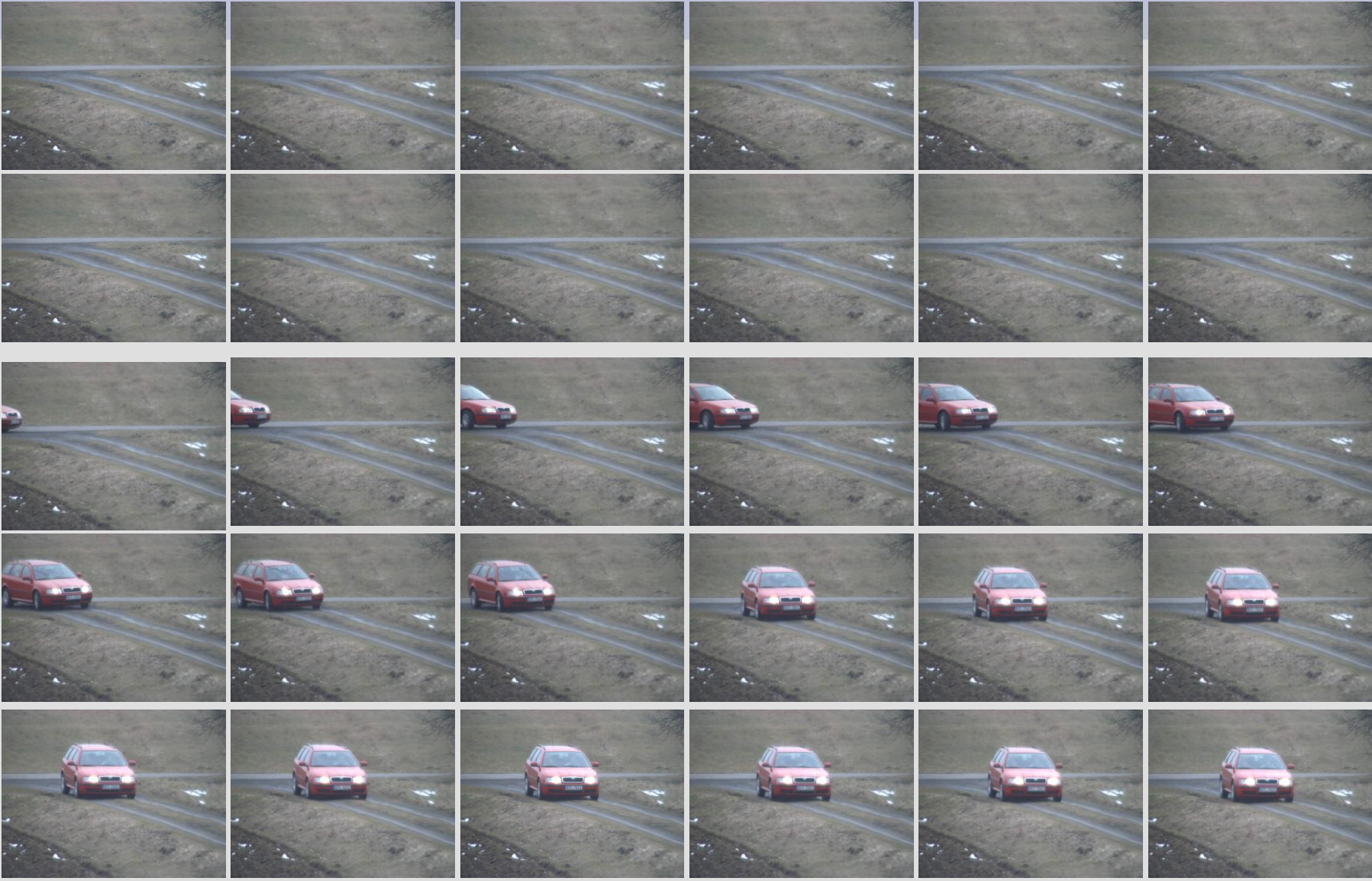


As these series of images look like?





## Example with the car



## Aim

- Recognizing foreground object in series of recorded images from IP cameras
- Verification using Gaussian mixture to separate the foreground with a small number of static background images
- Selection and testing classification methods

## The conventional methods

- HOG –histogram of oriented gradient
- HAAR cascades
- PCA principal component analysis

### Benefits

- detection from a single still image

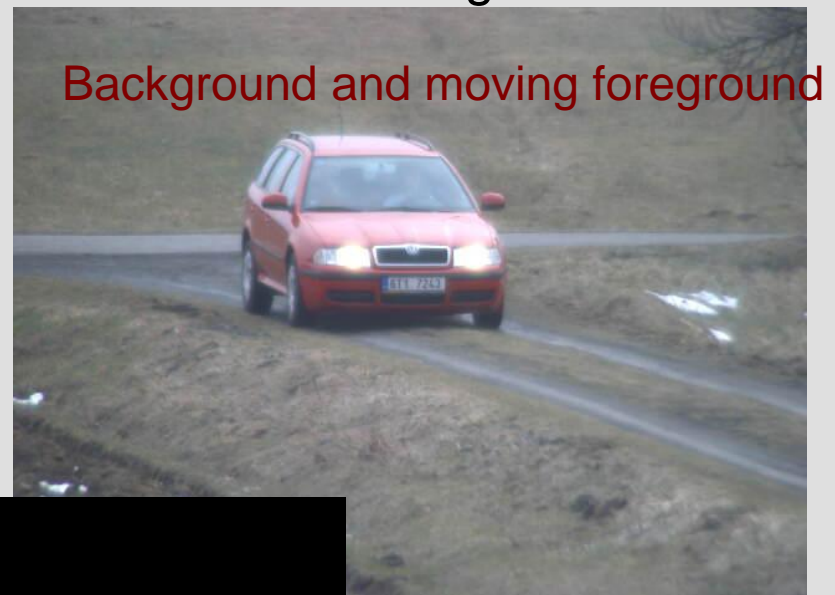
### Disadvantages

- high number of sample images - training data (in thousands)

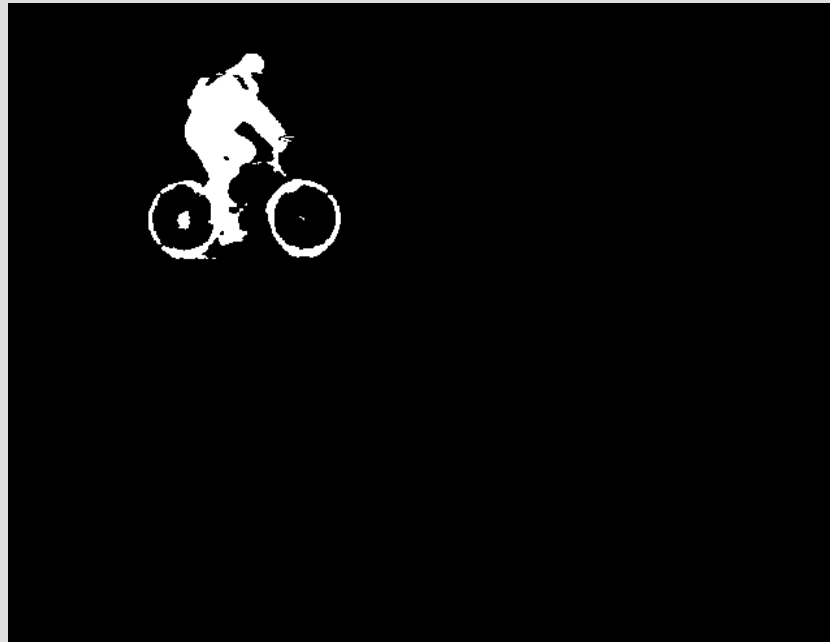
# Solution

Segmentation of the scene in the foreground and background

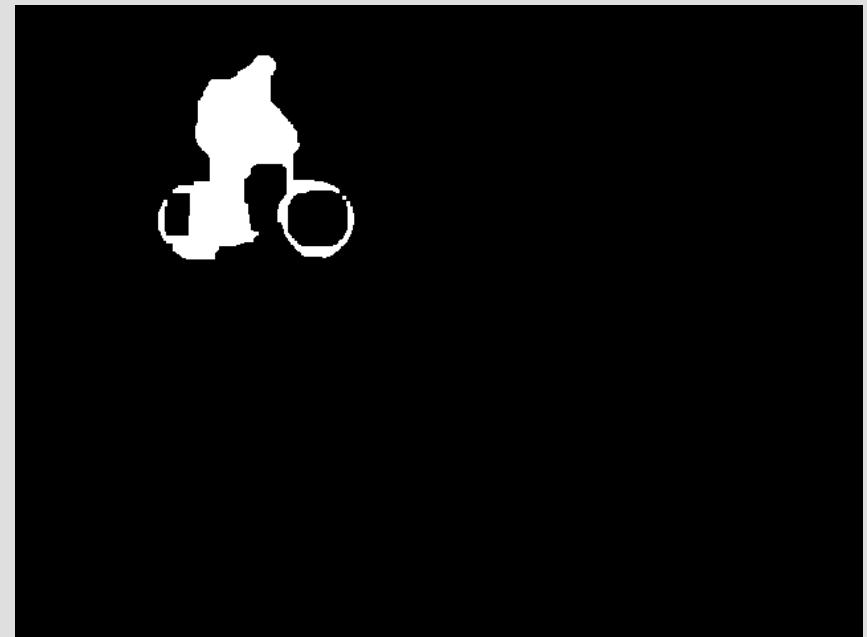
- Mixture of Gaussians with little numbers of statical background



# Removal of noise suppression splitting of one object to more objects



Erosion and  
dilatation





# Mathematical parametrization

## Normalized central moments

Geometric moments

$$m_{pq} = \sum_{x=0}^{X_{max}} \sum_{y=0}^{Y_{max}} x^p y^q f(x, y)$$

Center of gravity

$$\bar{x} = \frac{m_{10}}{m_{00}}$$

$$\bar{y} = \frac{m_{01}}{m_{00}}$$

Central moments

$$\mu_{pq} = \sum_{x=0}^{X_{max}} \sum_{y=0}^{Y_{max}} (x - \bar{x})^p (y - \bar{y})^q f(x, y)$$

Normalized central moments

$$\eta_{pq} = \frac{\mu_{pq}}{\mu_{pq}^{(0)}} \quad \eta' = \frac{p+q}{2} + 1 \quad \text{pro } p+q=2,3, \dots$$

# HU invariant moments

Non linear combination normalized central moments

$$\Phi_1 = \eta_{20} + \eta_{02}$$

$$\Phi_2 = (\eta_{20} - \eta_{02})^2 + 4\eta_{11}^2$$

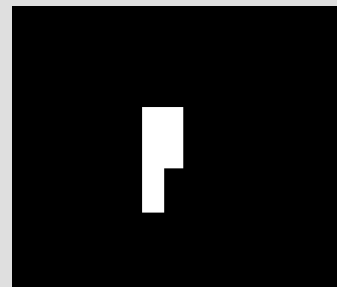
$$\Phi_3 = (\eta_{30} - 3\eta_{12})^2 + (3\eta_{21} - \eta_{03})^2$$

$$\Phi_4 = (\eta_{30} + \eta_{12})^2 + (\eta_{21} + \eta_{03})^2$$

$$\Phi_5 = (\eta_{30} - 3\eta_{12})(\eta_{30} + \eta_{12})[(\eta_{30} + \eta_{12})^2 - 3(\eta_{21} + \eta_{03})^2] \\ + (3\eta_{21} - \eta_{03})(\eta_{21} - \eta_{03})[3(\eta_{30} + \eta_{12})^2 - (\eta_{21} + \eta_{03})^2]$$

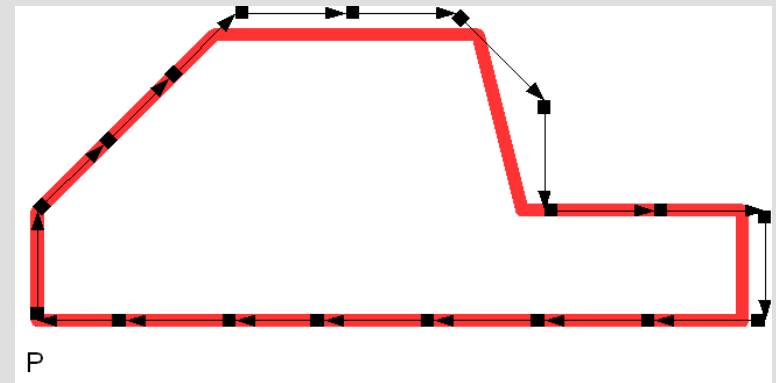
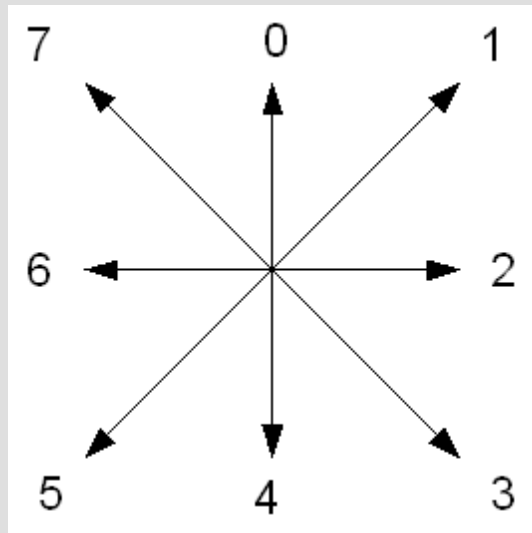
$$\Phi_6 = (\eta_{20} - \eta_{02})[(\eta_{30} - \eta_{12})^2 - (\eta_{21} + \eta_{03})^2] + 4\eta_{11}(\eta_{30} + \eta_{12})(\eta_{21} + \eta_{03})^2$$

$$\Phi_7 = (3\eta_{21} - \eta_{03})(\eta_{30} + \eta_{12})[(\eta_{30} + \eta_{12})^2 - 3(\eta_{21} + \eta_{03})^2] \\ + (3\eta_{21} - \eta_{03})(\eta_{21} - \eta_{03})[3(\eta_{30} + \eta_{12})^2 - (\eta_{21} + \eta_{03})^2]$$



# Freeman chain code

Edge presentation



[0,1,1,1,2,2,3,4,2,2,4,6,6,6,6,6,6,6]

## The percentage results of the tested parametrization

Class	Normované centrální momenty	HU invariantn moments	Histogram Freeman code
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CAR	98,2	95,7	60,6
SKEWCAR	89,9	91,5	98
CYCLIST	57,6	33,9	79,7
PERSONDOG	90	44	42,6
DOG	100	57,7	0
PERSON	94,6	91,5	25,1
GROUP	70,1	53,7	72

Totally correctly classified	<b>91,7</b>	<b>85,3</b>	<b>43,2</b>
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## The percentage results of the classifiers

<b>Classifier</b> <b>Class</b>	<b>Neurall networks</b>	<b>Normal Bayes classifier</b>	<b>Support vector machine</b>	<b>Linear discriminant analyse</b>
CAR	98,1	100	48,3	99,6
SKEWCAR	90,5	69,1	80,8	86,8
CYCLIST	58,6	27,6	0	58,6
PERSON-DOG	89,9	30,2	0	89,2
DOG	100	96,2	10,9	100
PERSON	94,7	96,7	81,7	74,1
GROUP	69,9	43,3	50,9	85,3

Totally correctly classified	<b>91,7</b>	<b>87,8</b>	<b>72,3</b>	<b>80</b>
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# Typical classification errors

Class	Wrong		Correctly	
PERSON				
GROUP OF PERSONS				
PERSON				

Thank you for your attention