

Search and Rescue UAV missions with object recognition using convolutional neural networks

Fabio Andrade

Research Scientist, Norut

Assistant Professor, CEFET/RJ

PhD Candidate, NTNU

Supervisors: Rune Storvold and Tor Arne Johansen



Agenda

Search and Rescue

Convolutional Neural Networks

Examples

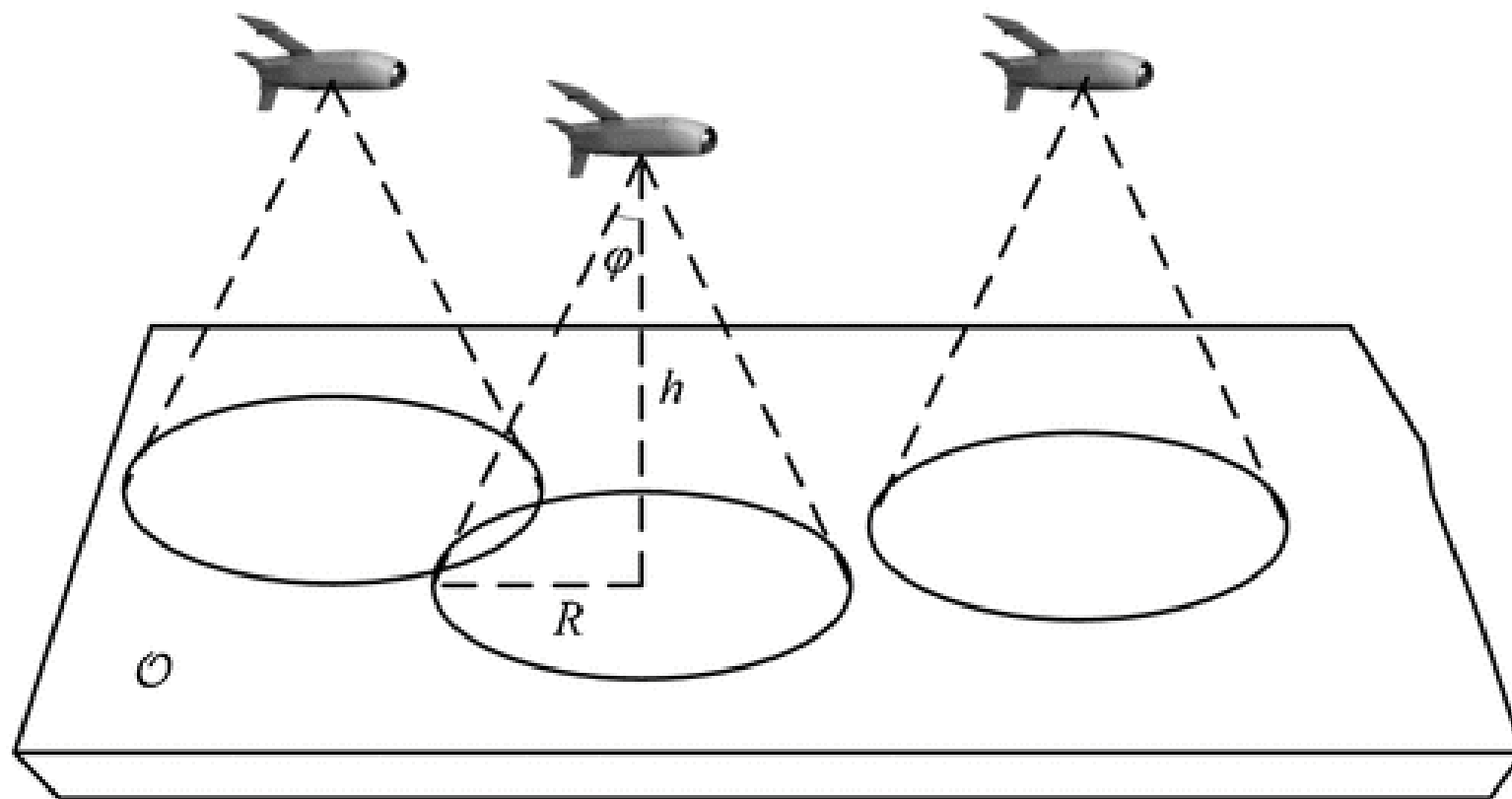
Our dataset



Source: Norut

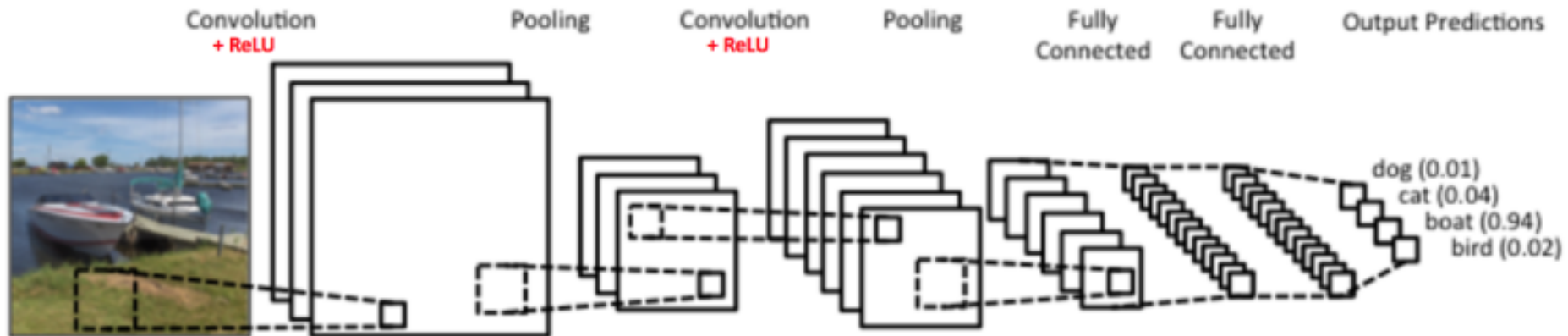


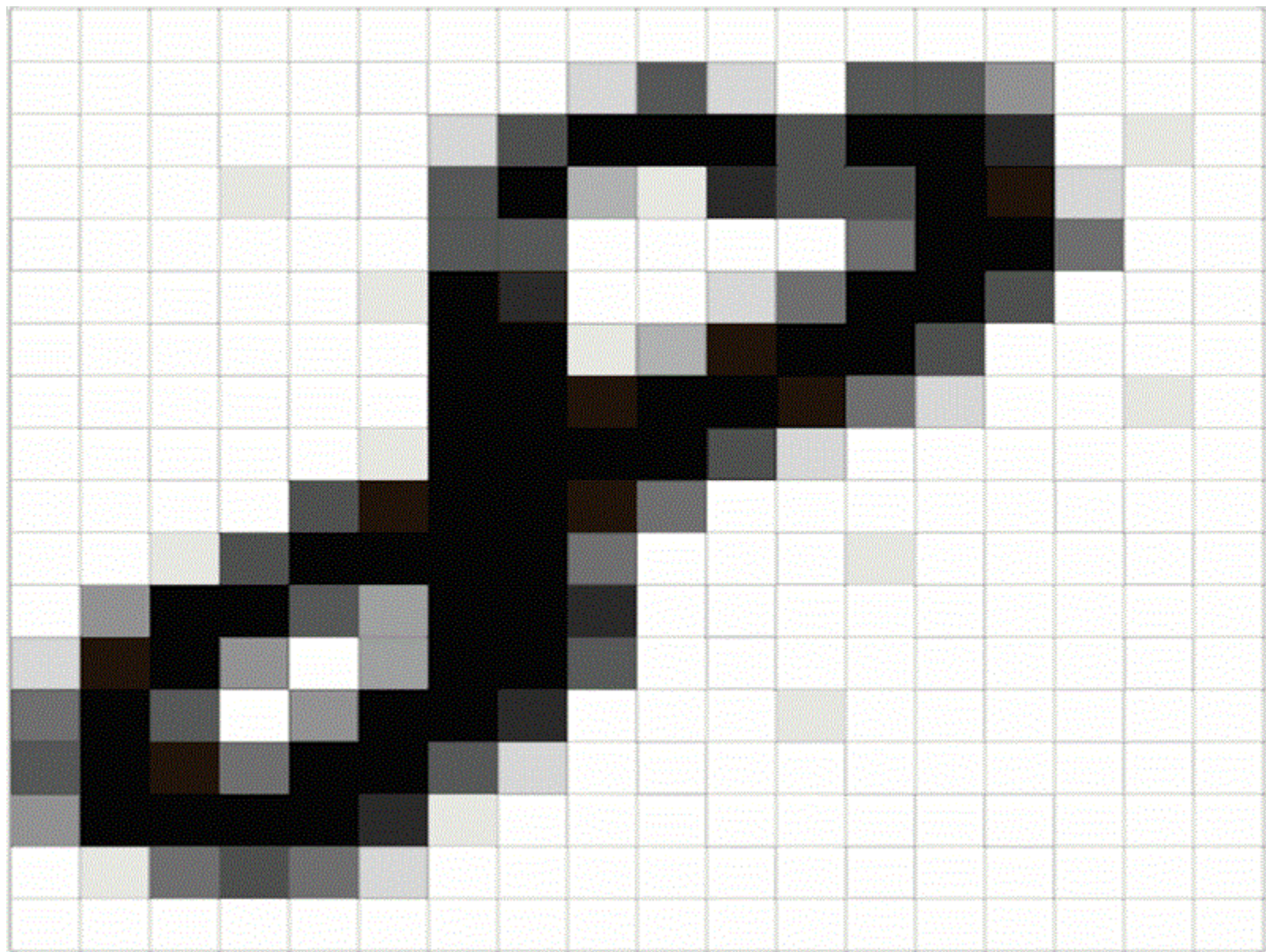
Source: Norut



Source: Minqiang Zhang

Convolutional Neural Networks?





The Convolution Step

1	1	1	0	0
0	1	1	1	0
0	0	1	1	1
0	0	1	1	0
0	1	1	0	0

 $*$

1	0	1
0	1	0
1	0	1



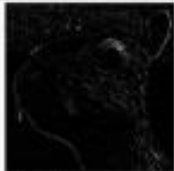




1 <small>$\times 1$</small>	1 <small>$\times 0$</small>	1 <small>$\times 1$</small>	0	0
0 <small>$\times 0$</small>	1 <small>$\times 1$</small>	1 <small>$\times 0$</small>	1	0
0 <small>$\times 1$</small>	0 <small>$\times 0$</small>	1 <small>$\times 1$</small>	1	1
0	0	1	1	0
0	1	1	0	0

Image

4		

Convolved
Feature



Operation	Filter	Convolved Image
Identity	$\begin{bmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$	
Edge detection	$\begin{bmatrix} 1 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{bmatrix}$	
	$\begin{bmatrix} 0 & 1 & 0 \\ 1 & -4 & 1 \\ 0 & 1 & 0 \end{bmatrix}$	
	$\begin{bmatrix} -1 & -1 & -1 \\ -1 & 8 & -1 \\ -1 & -1 & -1 \end{bmatrix}$	
Sharpen	$\begin{bmatrix} 0 & -1 & 0 \\ -1 & 5 & -1 \\ 0 & -1 & 0 \end{bmatrix}$	
Box blur (normalized)	$\frac{1}{9} \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 1 \\ 1 & 1 & 1 \end{bmatrix}$	
Gaussian blur (approximation)	$\frac{1}{16} \begin{bmatrix} 1 & 2 & 1 \\ 2 & 4 & 2 \\ 1 & 2 & 1 \end{bmatrix}$	



Input

Input Feature Map

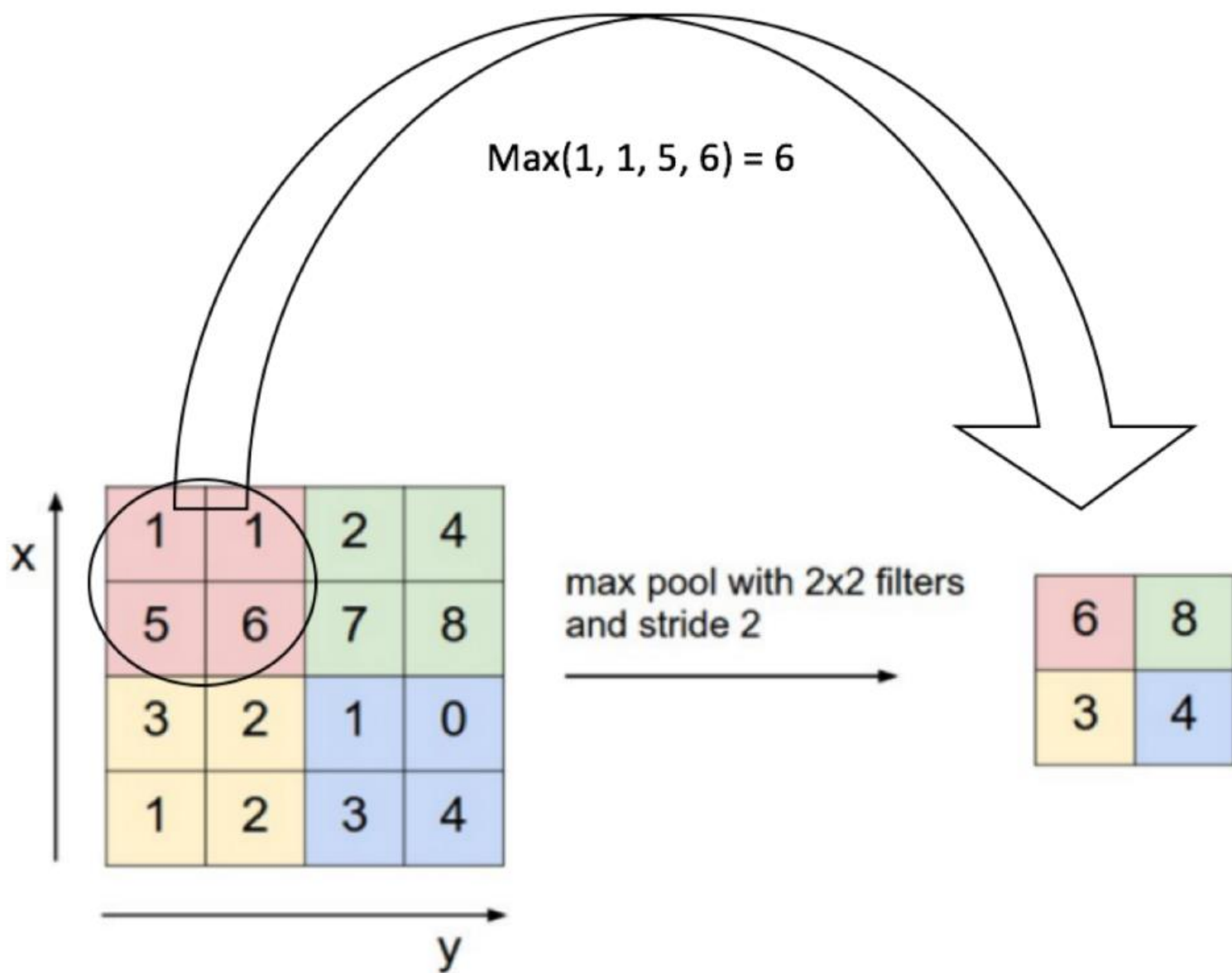


ReLU



Rectified Feature Map





Rectified Feature Map

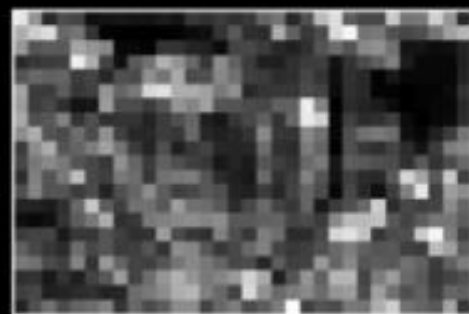


Rectified Feature Map

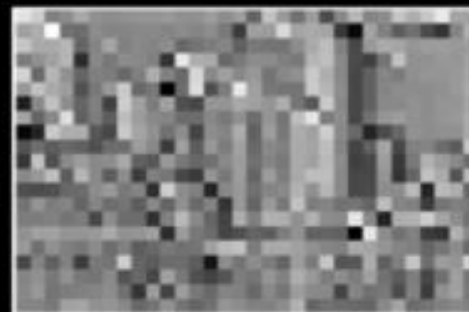
Pooling

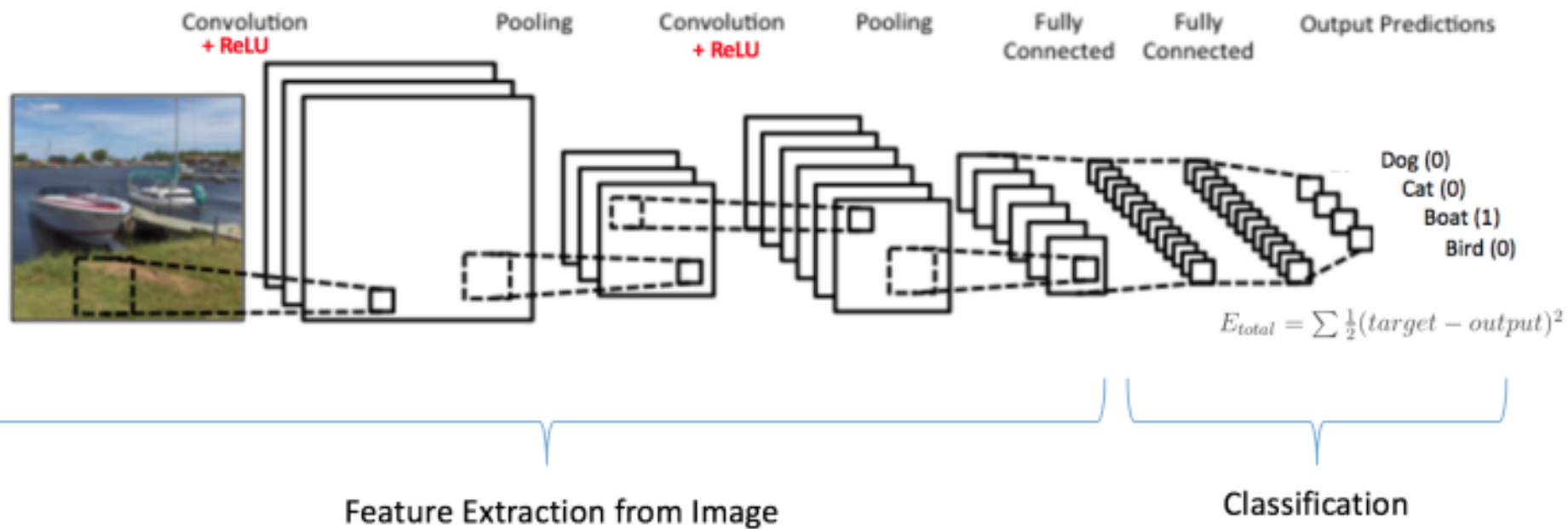


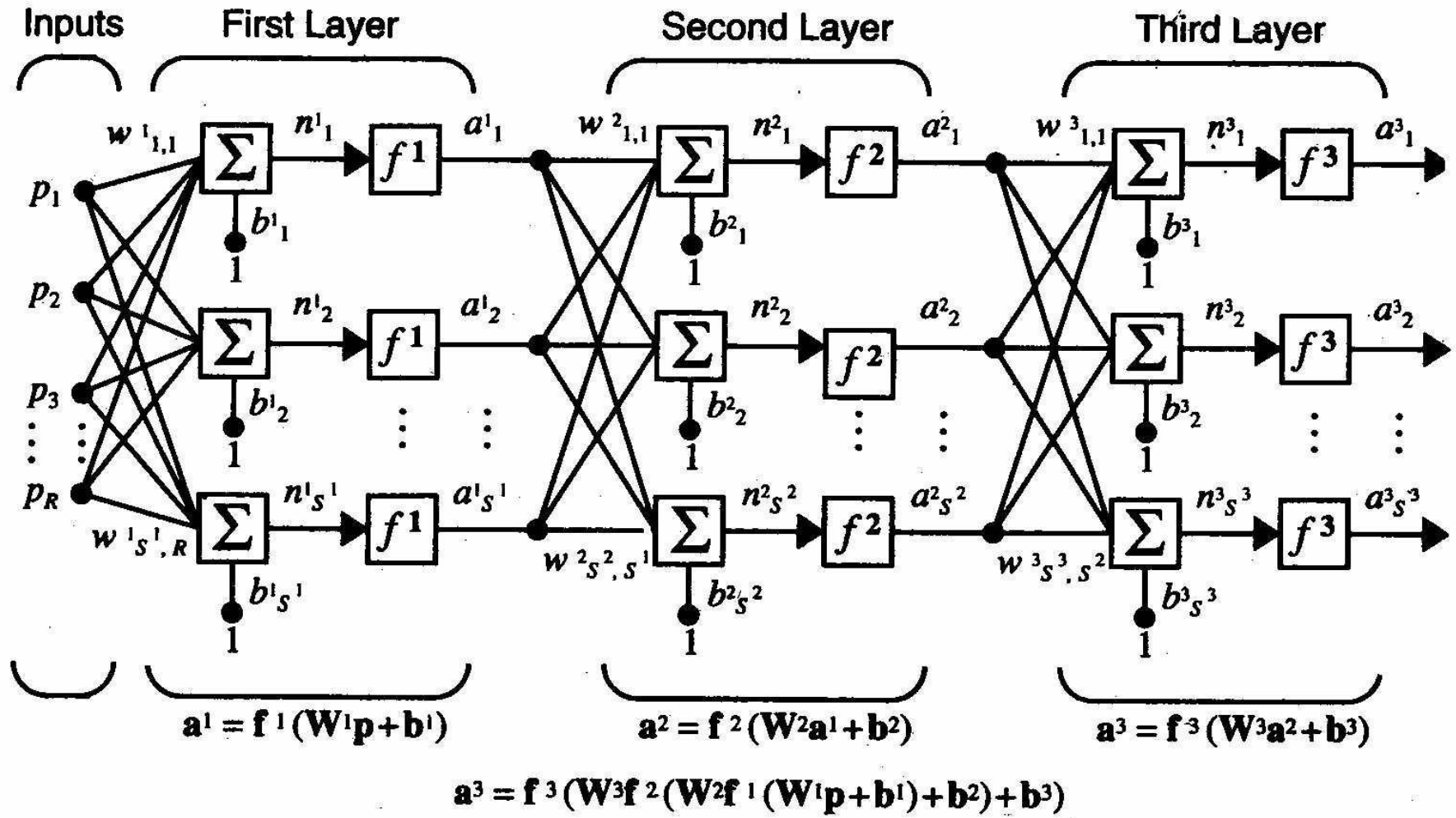
Max

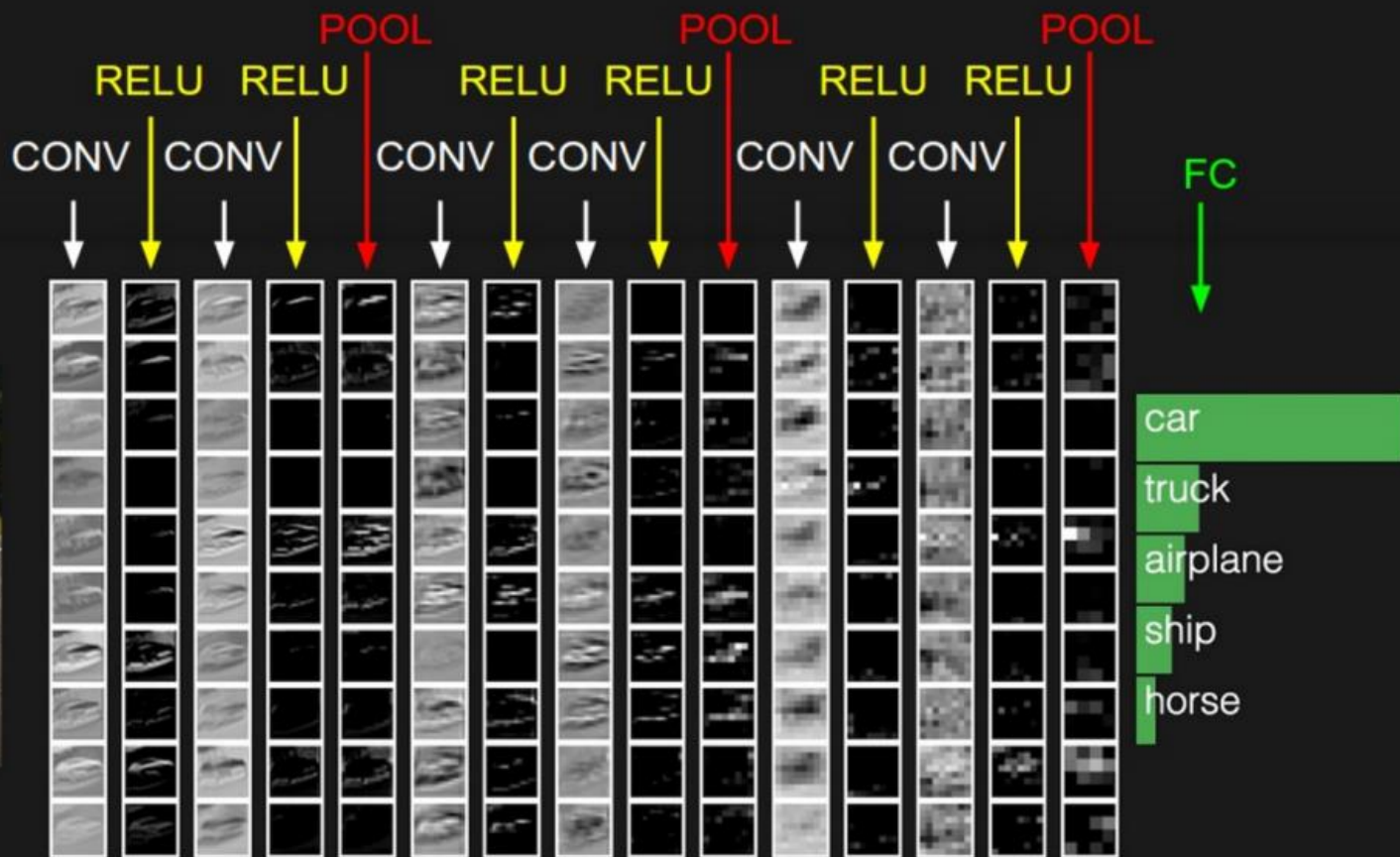


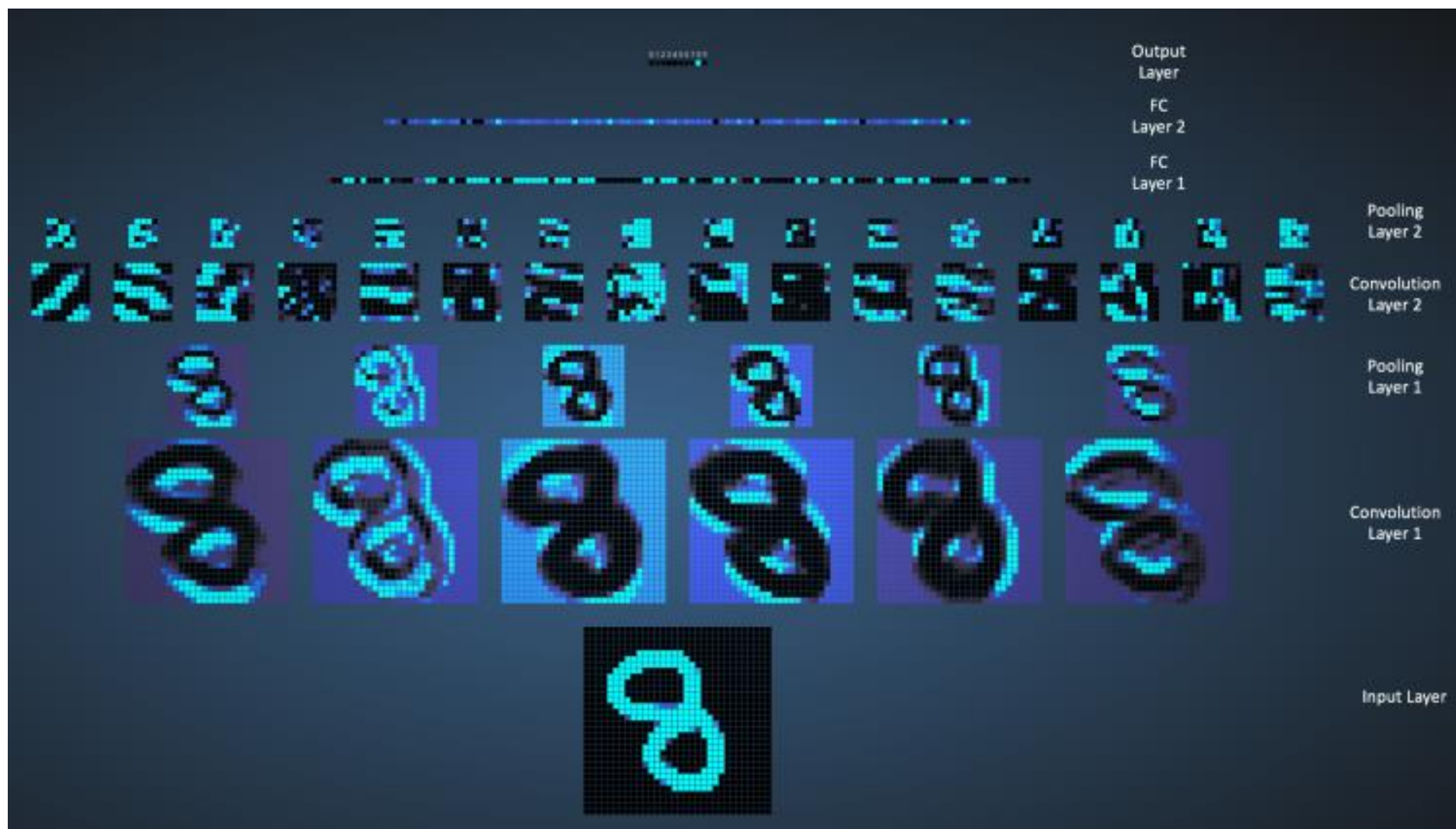
Sum











Vision-based Vehicle Detecting and Counting for Traffic Flow Analysis

Zhang et al



Count vehicles

Fast R-CNN

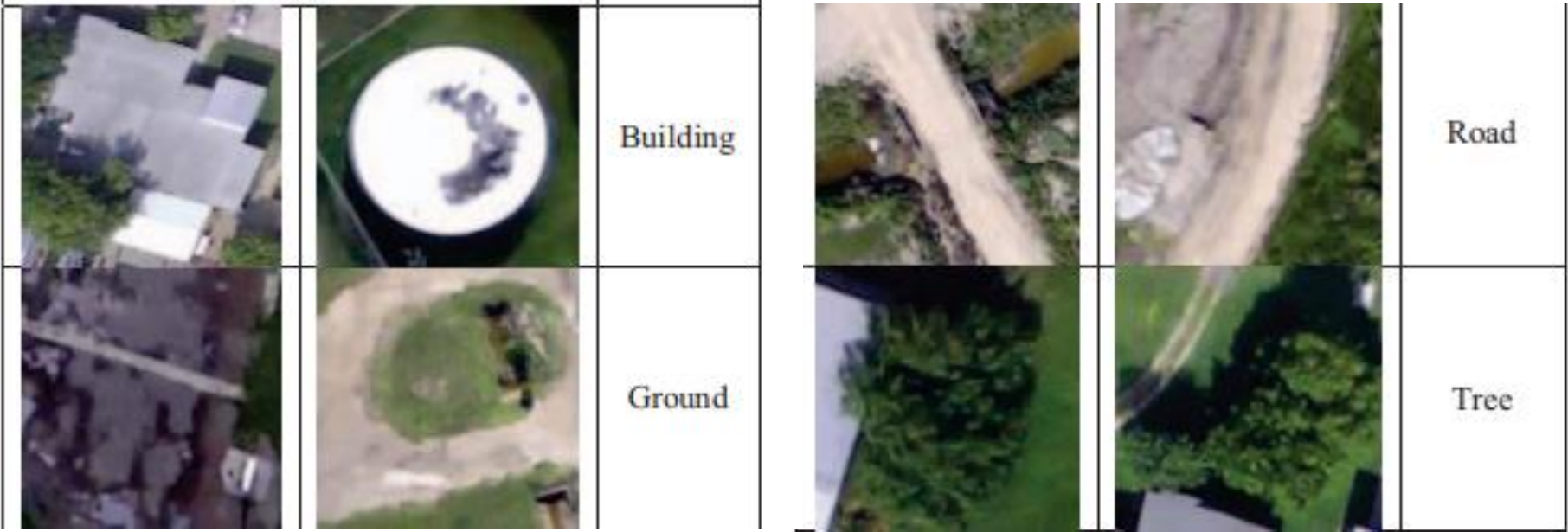
Set of object proposals as input

Produces bounding boxes

Fine-tune a pre-trained Fast R-CNN with traffic videos

Real-time Scene Understanding for UAV Imagery based on Deep Convolutional Neural Networks

Clay Sheppard and Maryam Rahnemoonfar



UAV vertical images
High altitude
3864 images
4 classes: building, ground, road, tree

Generating binary tags for fast medical image retrieval based on convolutional nets and radon transform

Xinran Liu, H.R.Tizhoosh, J.Kofman

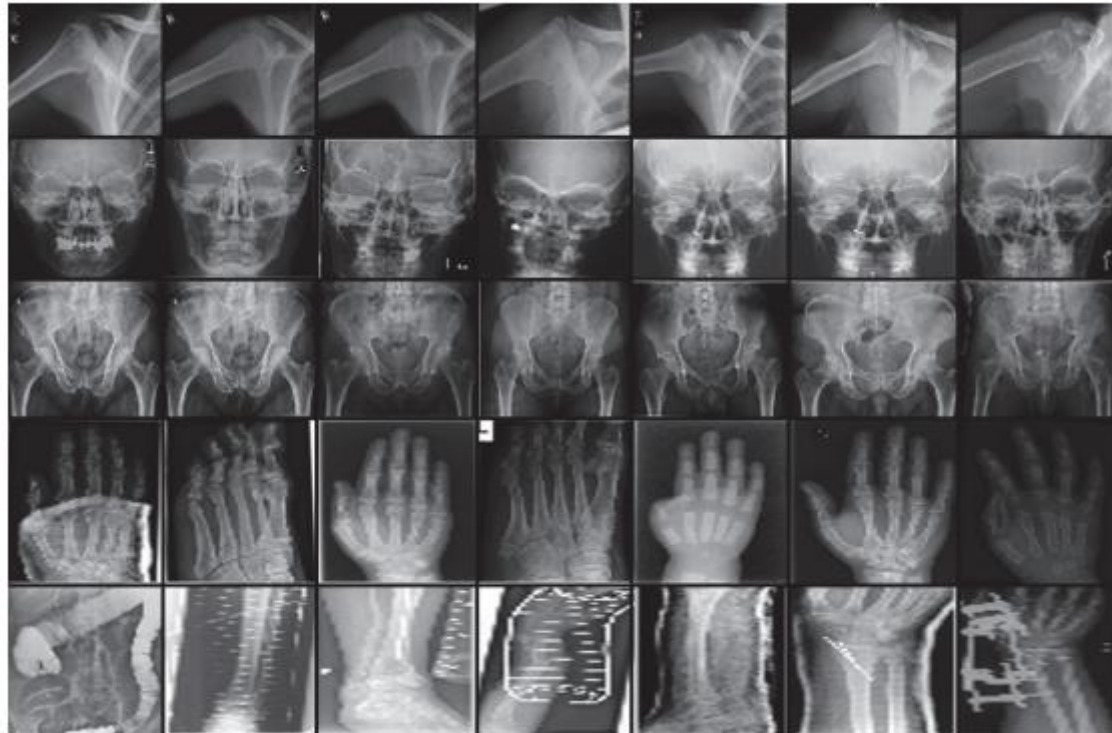


Image retrieval
Gray level (no color)

Fast animal detection in uav images using convolutional neural networks

Benjamin Kellenberger, Michele Volpi, Devis Tuia

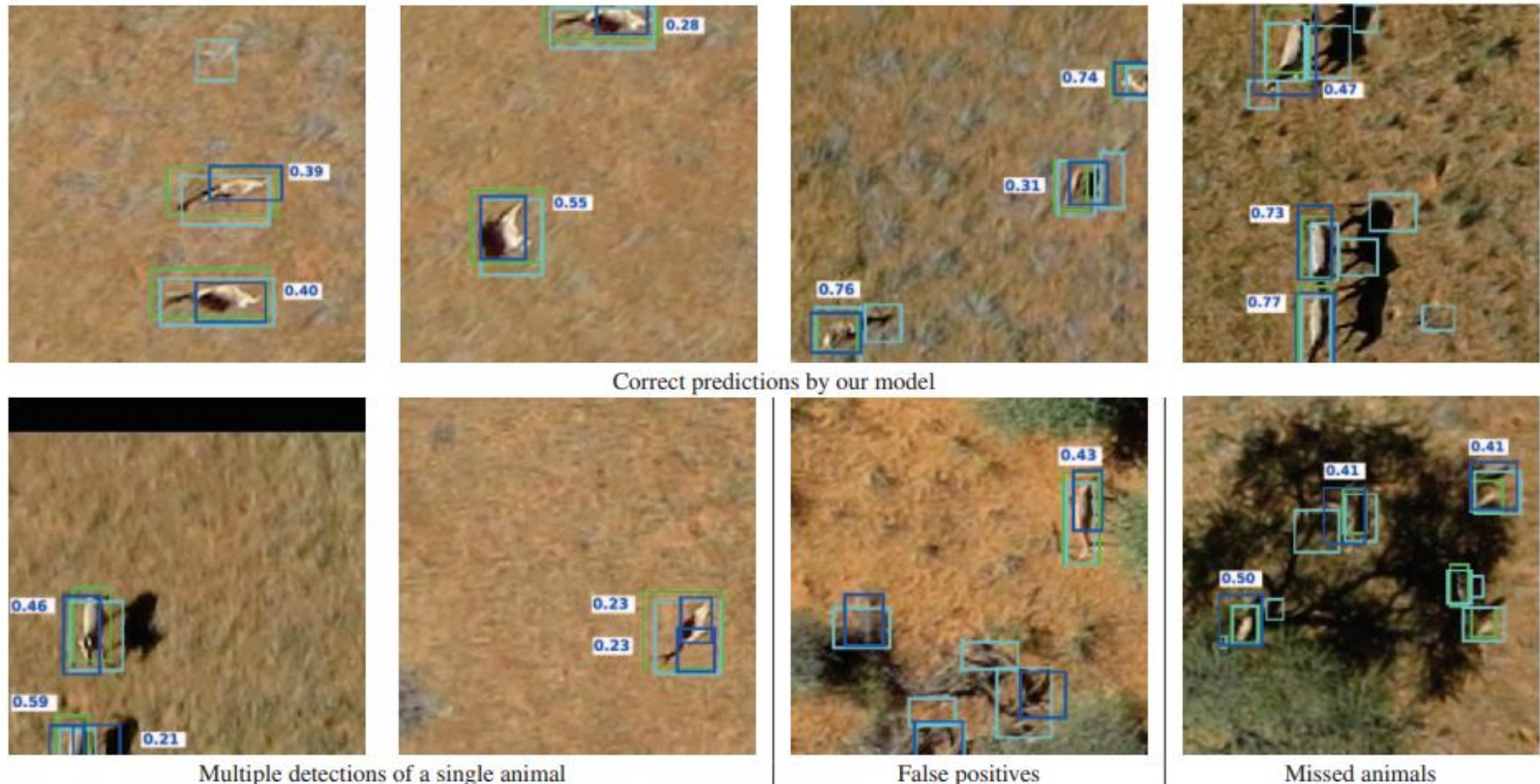


Fig. 3. Detection examples on the test set (blue; with IoU scores for predictions) and Fast R-CNN baseline (cyan). Ground truth is in green. Top row shows correct detections for our model, while bottom row show failure cases.

Large animals in Namibia

Real time

654 RGB images, 1196 animals after fine tuning

Convolutional Neural Networks For Near Real-time Object Detection from Uav Imagery In Avalanche Search And Rescue Operations

Mecnu Relete Reina, Abdallah Zennadi, and Farid Melanni



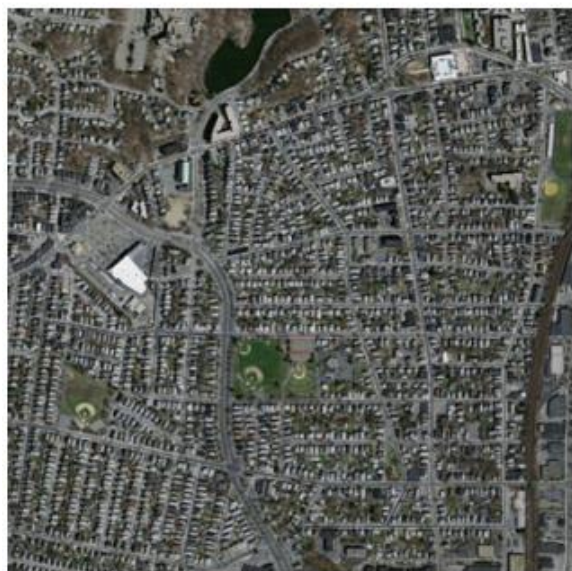
Fig. 3. Example of correctly classified negative (top left) and positive (top right) images, false positive object in yellow rectangle (bottom left), and false negative or undetected object (bottom right). Objects of interest are indicated by yellow arrow.

Pre-trained CNN to detect objects

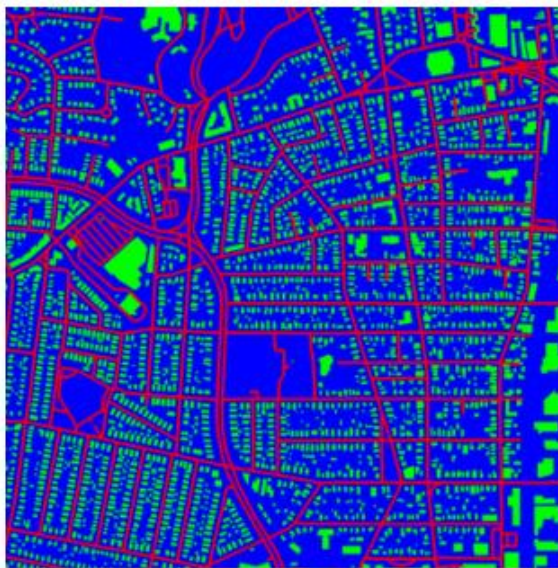
SVM to classify

Multiple Object Extraction from Aerial Imagery with Convolutional Neural Networks

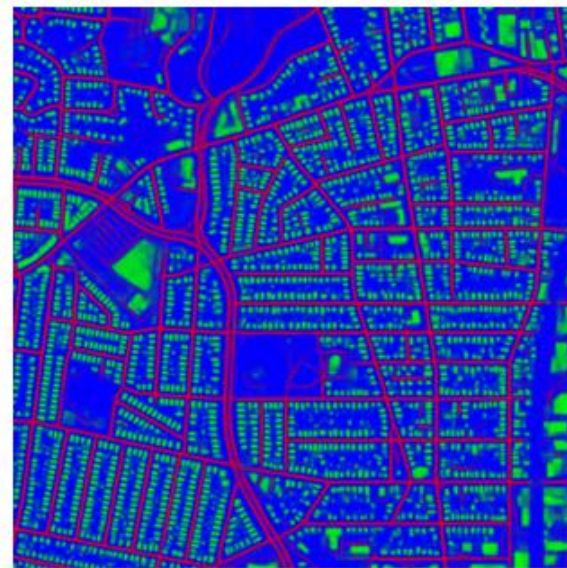
Shunta Saito, Takayoshi Yamashita, and Yoshimitsu Aoki



(a) Aerial imagery S



(b) Ground truth \tilde{M}



(c) Predicted label image \hat{M}

Figure 2. An example of the resulting predicted label image.

Aerial images

3 classes: road, building, background

A Convolutional Neural Network for Automatic Analysis of Aerial Imagery

Frederic Maire, Luis Mejias and Amanda Hodgson



Marine species

Bboxes using blobs (confidence of the pixel being the center of a window containing a mammal)

Deep Learning for Infrared Thermal Image Based Machine Health Monitoring Olivier Janssens et al

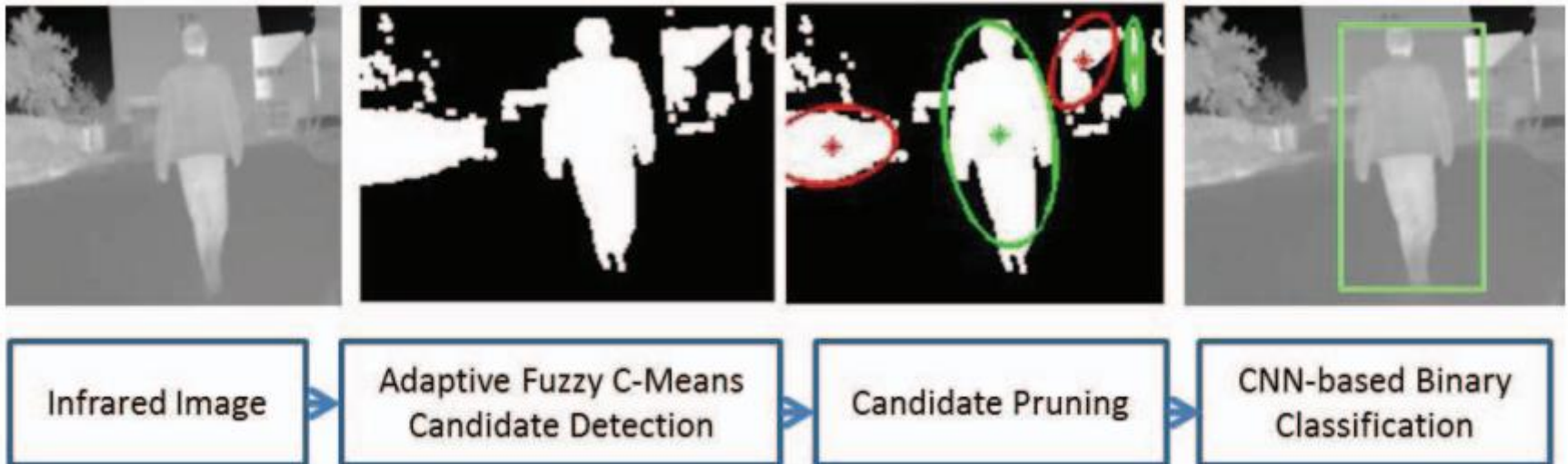


Fig. 6: Regions that influence the CNNs output for a healthy

Infrared thermal video
Determine condition of the machine
Machine fault detection and oil level prediction

Pedestrian Detection in Thermal Images Using Adaptive Fuzzy C-Means Clustering and Convolutional Neural Networks

Vijay John et al



Infrared images

Candidate bbox is a pedestrian or not?

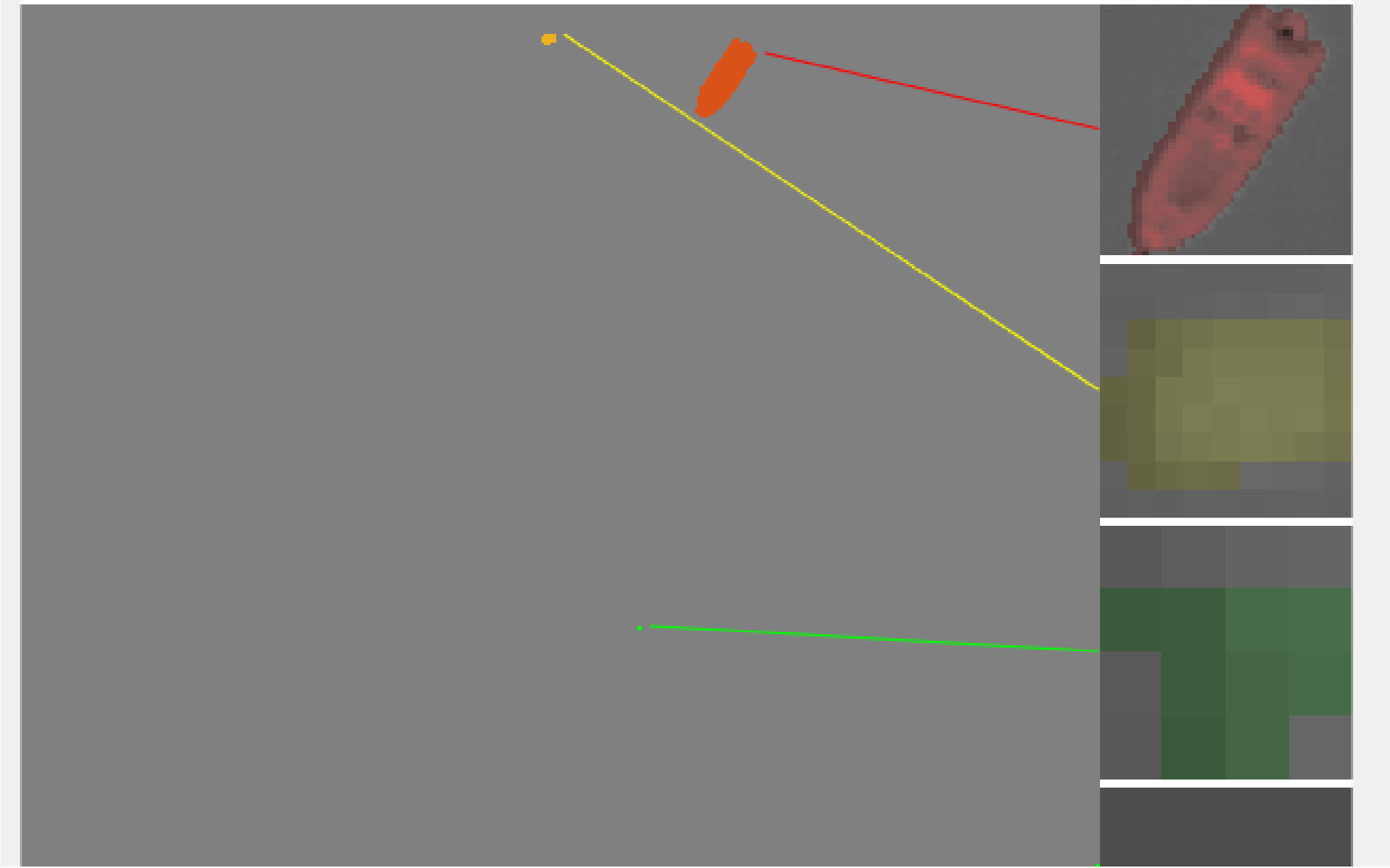
LSI Public dataset (8000 pedestrians and 8000 negative samples)

RGB

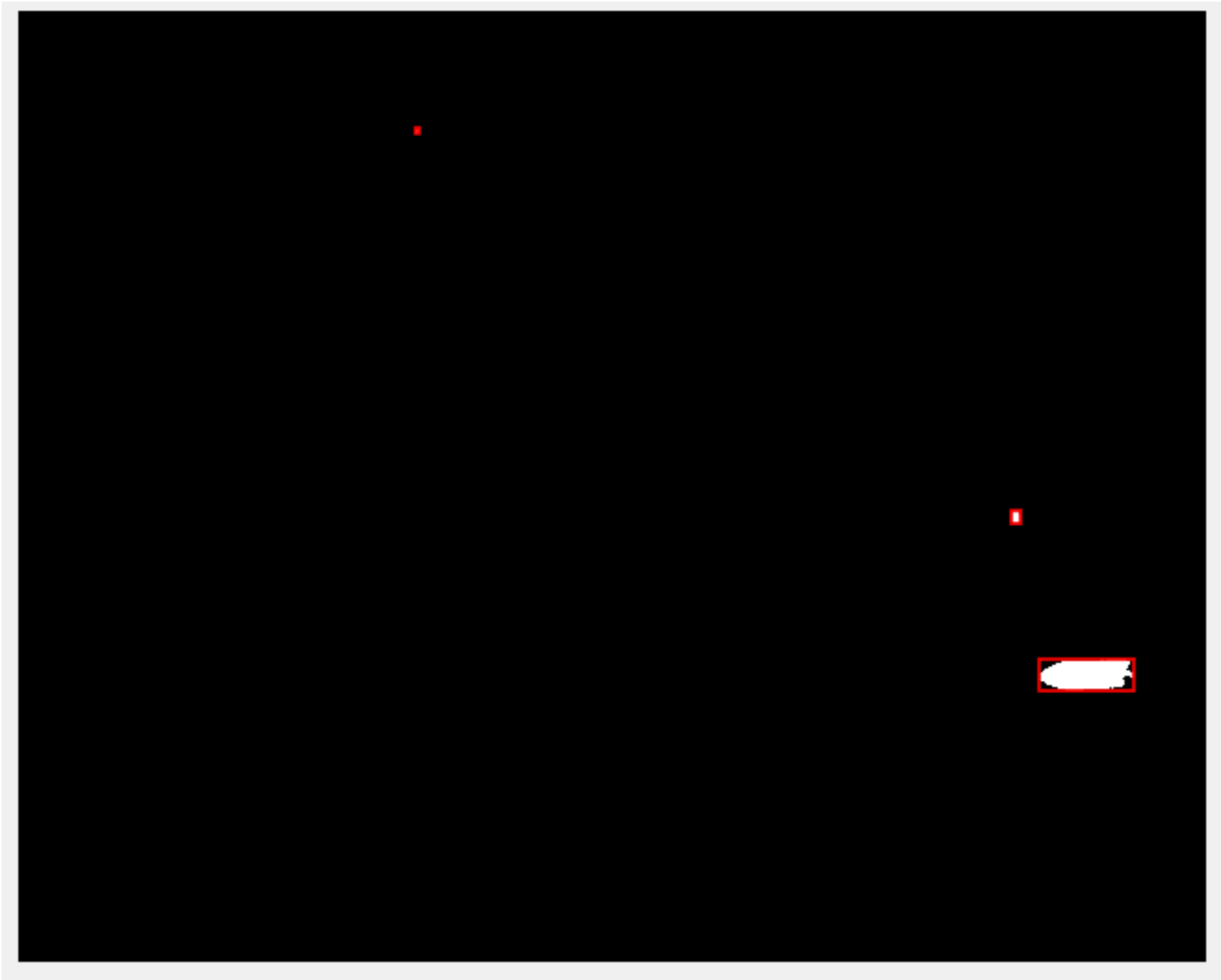


IR



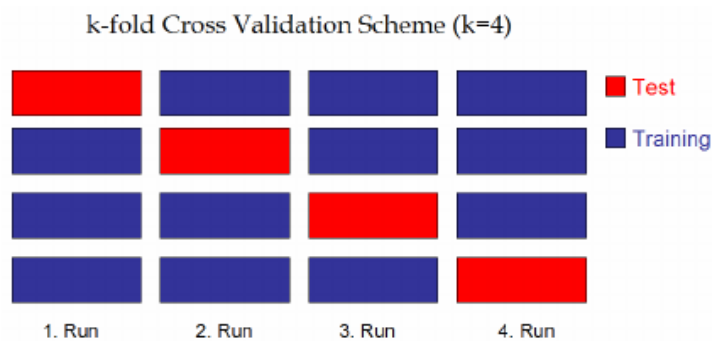


Bounding boxes

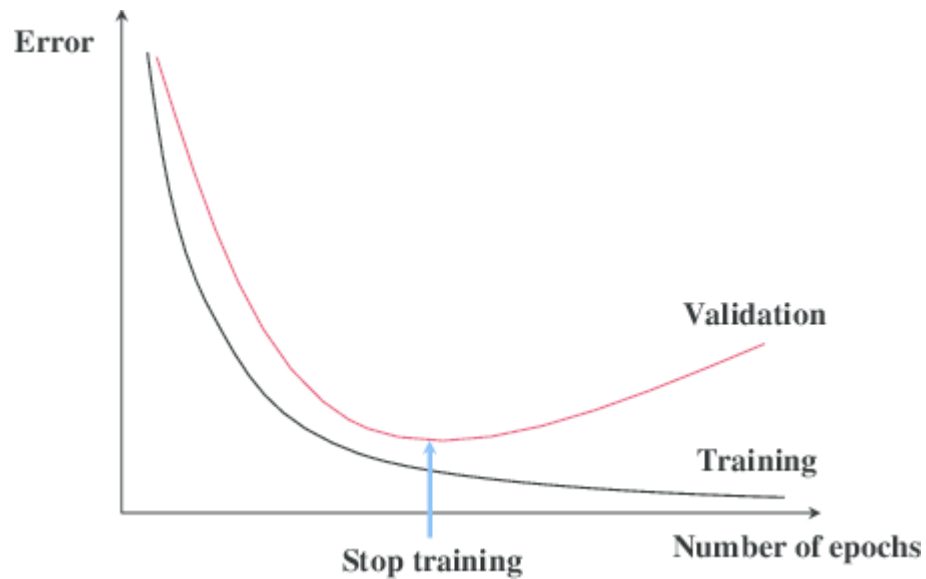


4 classes: boat, pallet, buoy, person

K-fold cross validation



Early stop by validation curve



Thank you!

fabio@ieee.org