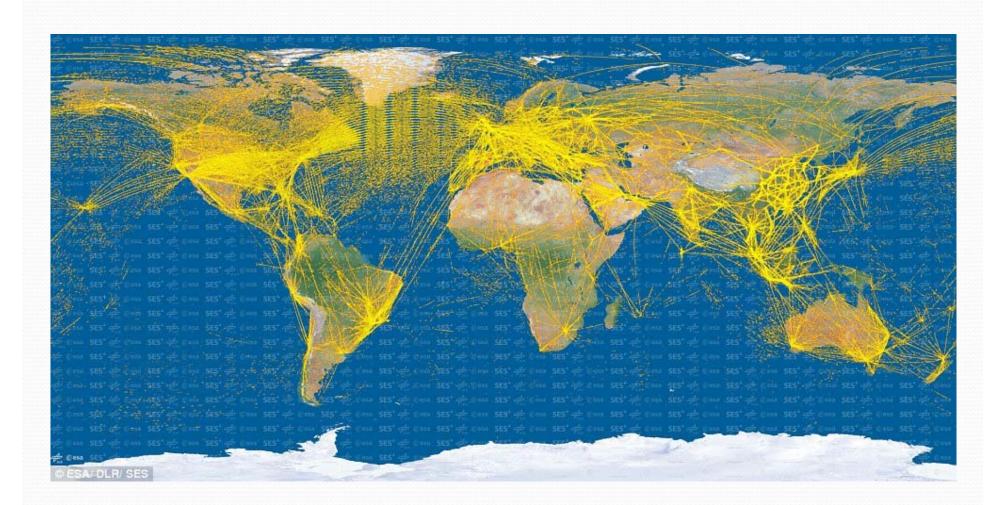
## **AmiEs 2016**

International Symposium on Ambient Intelligence and Embedded Systems

**Aircraft Identification using Machine Vision** 

Dimitrios Vidakis Dimitrios Kosmopoulos

22 - 24 September, 2016 Heraklion, Crete, Greece



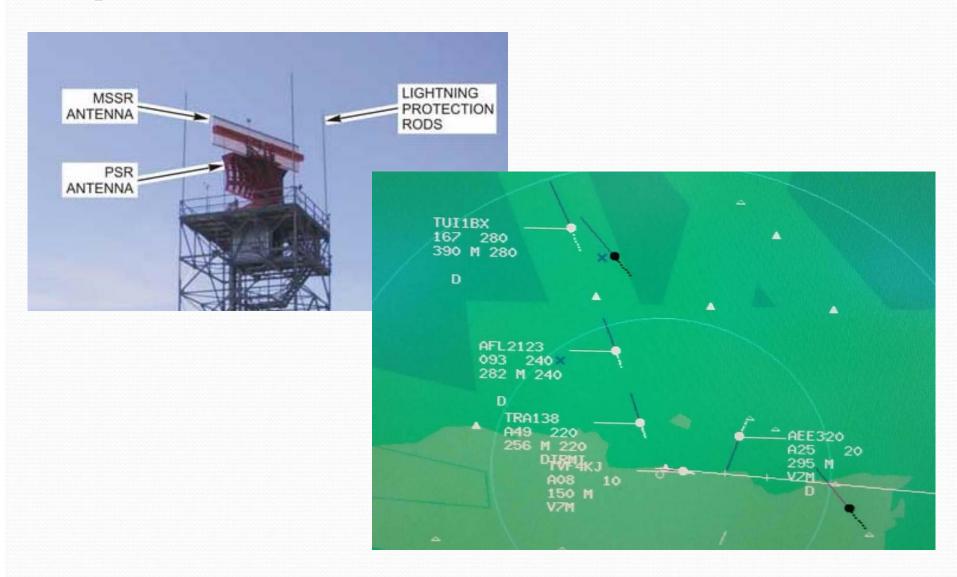
15,000 out of a total 102,000 worldwide daily flights!!!



Aircraft congestion on the ground - demand for technical support

Low Cost aircraft recognition in the airport using machine vision

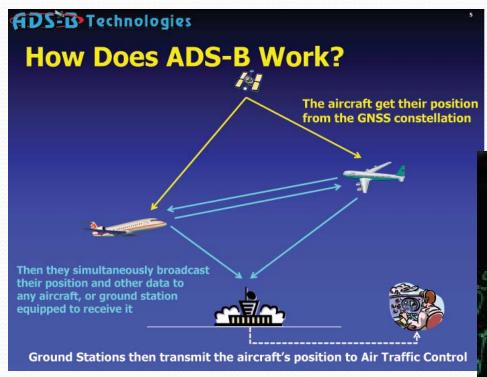
## Airport Surveillance Radar (ASR)

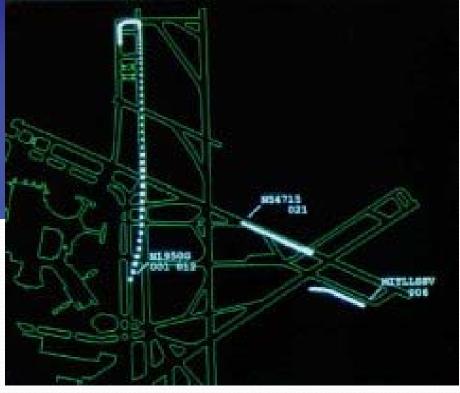


## **Surface Movement Radar (SMR) (Schwabn and Rost 1985)**



### **Automatic Dependent Surveillance Broadcast (ADS-B)**





Sensor	Active	Recognition	Targets	MET conditions
SMR	No	No	All	All
DGPS	Yes	Yes	Equipped	All
MS	Yes	Yes	Equipped	All
Camera	No	No	All	Clear

- **❖** SMR → Surface Movement Radar
- ❖ DGPS -> Differential GPS via digital data link
- **❖** MS → Multilateration systems

### Cameras - Passive sensors

Aircraft Identification using Machine Vision

### **Pros**

- > Zero radiation
- > Low cost
- Video availability
- High rate data refresh

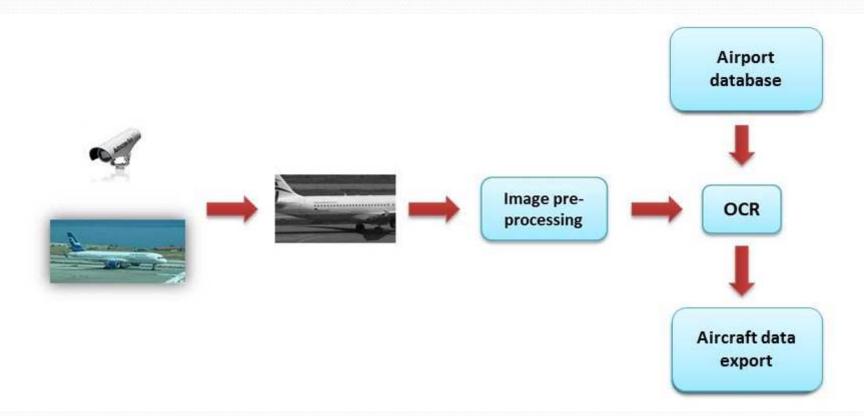
# Aircraft registration number

Aircraft Identification using Machine Vision

The aircraft registration number or tail number is an alphanumerical sequence that is unique for every aircraft flying worldwide.



**Registration numbers in various aircrafts** 



Video Recording and analyzing procedure

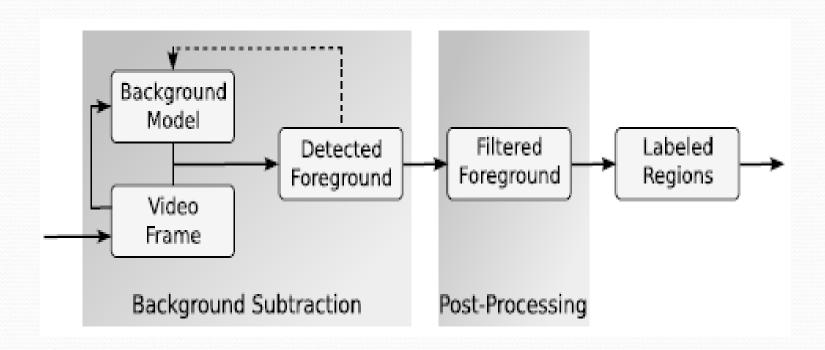
- Record and display aircraft movements in the aircraft moving area of an airport.
- Faster information delivery to everyone involved.
- Monitoring of aircrafts enlisted in 'blacklists'.
- Monitoring of aircrafts landed unauthorized eg when airports are closed
- Detection of aircraft suspected of any irregularities, such as "emitting noise above permissible limits."
- Detection of aircrafts involved in runway incursion incidents.

# **Optical Character Recognition**

Aircraft Identification using Machine Vision

### "ocr.m" - MATLAB 2014b +

- > Text in pictures
- > Text in Regions of interest (ROI)
- ➤ Words bounding boxes and confidence percentage



Background subtraction with post processing for surveillance applications

**International Symposium on** Aircraft Identification using **Algorithm description Ambient Intelligence and Machine Vision Embedded Systems** High definition video capture Setting background as first frame Variables initialization before movement detection Frame with the largest part of the Background Subtraction aircraft visible detection Connected pixels region detection and cropped >thresh MatLAB "mean2" function Pre-process of the frame cropped Sauvola with integral images <thresh Aircraft tail detection Top-hat filtering Sliding window size definition Matlab OCR word detection If rules: · 6 letters including "-" Only one solution Confidence over 60% accepted Only capital letters and numbers are satisfied Data base use for no requested data

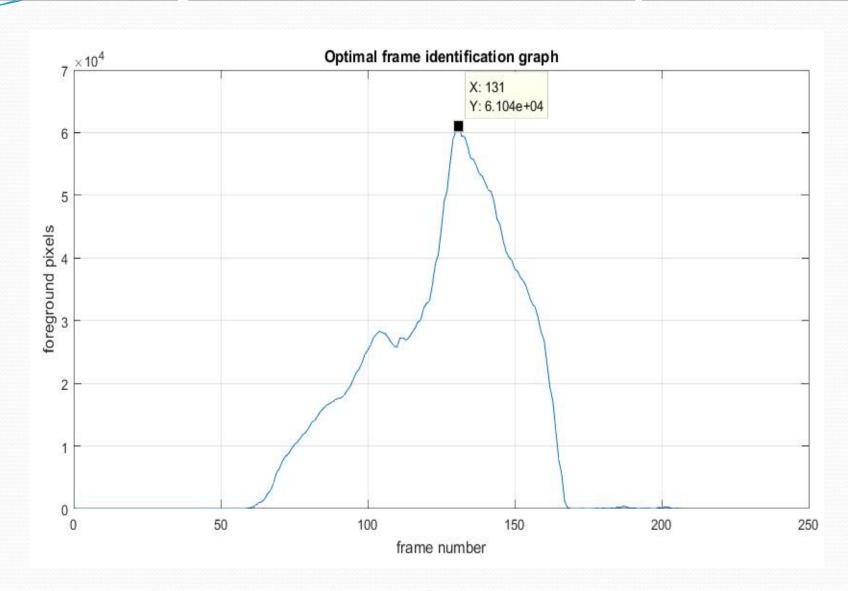
> No results answer produced

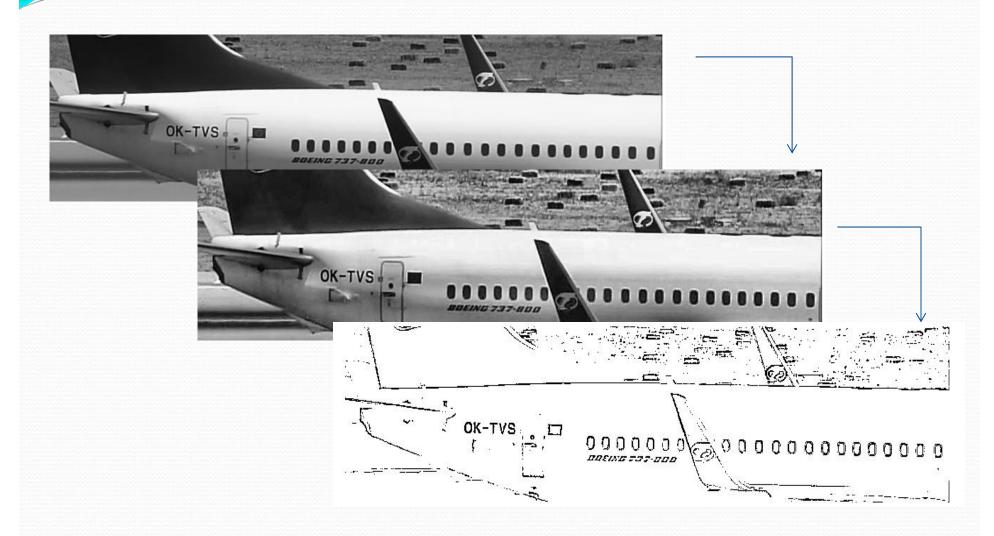
### Aircraft Tail in frame 131 Sum of values of pixels 43339306



Aircraft Tail in frame 131 Number of pixels "on" 83266 Frame Coverage Percentage 20.3127 %







#### Command Window

Possible Registrations and Confidence Percentages

index: 127, conf: 0.77303, words: OK-TII
index: 143, conf: 0.83222, words: OK-TVS
index: 158, conf: 0.81872, words: OK-TVS

#### Command Window

Possible Registrations and Confidence Percentages

index: 127, conf: 0.77303, words: OK-TII
index: 143, conf: 0.83222, words: OK-TVS
index: 158, conf: 0.81872, words: OK-TVS
Aircraft Registration and Origin: OK-TVS
 'Czech Republic'

fx >>

Aircraft Registration and Origin: SK-DVD

WARNING: There is a problem with the OCR of the Aircraft Regitration, so the origin country of the aircraft is not available

### Matlab time results



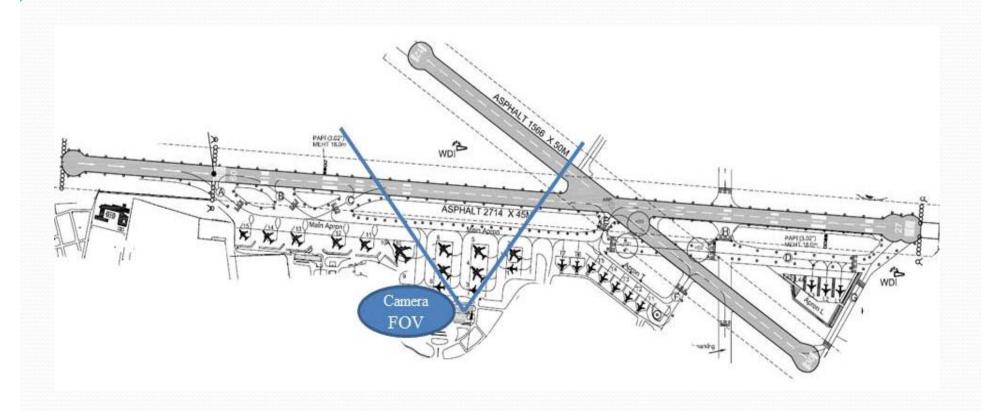
Start Profiling Run this code:

# **Profile Summary**

Generated 03-Jun-2015 22:47:25 using cpu time.

Calls	Total Time	Self Time*	Total Time Plot (dark band = self time)
1	61.384 s	9.401 s	
1	29.033 s	2.022 s	
672	24.787 s	13.915 s	
255	14.960 s	0.045 s	
255	14.270 s	13.985 s	
	1 1 672 255	1 61.384 s 1 29.033 s 672 24.787 s 255 14.960 s	1 61.384 s 9.401 s 1 29.033 s 2.022 s 672 24.787 s 13.915 s 255 14.960 s 0.045 s





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### Results

Aircraft Identification using Machine Vision

	video mp4)	slidi windov (pixe	v size	total time (sec)	ims (sec)	OCR (sec)	conf (%)	reg result	correct reg	succ (%)	country result	correct country	succ (yes/no)
		cols	rows	(360)									
A	EE_2	60	30	95	32	45	73.56	SX-DST	SX-DGT	83,33	Greece	Greece	yes
T	FL_1	110	40	187	26	45	84,00	PH-HZG	PH-HZG	100,00	The Nederlands	The Nederlands	yes
Т	VS_4	90	30	139	24	32	79,00	OK-TSJ	OK-TSJ	100,00	Czech Republic	Czech Republic	yes
Т	VS_4	90	50	336	68	101	84,00	OK-TSJ	OK-TSJ	100,00	Czech Republic	Czech Republic	yes
A	\FL_2	-	-	30	9,5	0,3	73,00	VP-BUS	VP-BOS	83,33	Bermuda	Bermuda	yes
A	EE_9	100	40	111	21	25	75,00	SX-DVV	SX-DVV	100,00	Greece	Greece	yes
A	EE_19	100	40	238	30	57	81,00	SX-DGI	SX-DGI	100,00	Greece	Greece	yes
E	BIE_1	90	40	208	28	51	75,00	F-GVAP	F-GVAP	100,00	France	France	yes
E	BIE_2	200	100	57	16	2,8	81,00	F-GVAD	F-GVAO	83,33	France	France	yes
Δ	LS_1	100	50	119	25	37	87,00	EC-LNC	EC-LNC	100,00	Spain	Spain	yes
В	ER_22	150	70	77	18	18	82,50	D-ABFZ	D-ABFZ	100,00	Germany	Germany	yes
E	BIE_8	150	70	55	11	16	76,00	F-HCOA	F-HCOA	100,00	France	France	yes
C	FG_8	150	70	42	11	7	75,00	D-AICL	D-AICL	100,00	Germany	Germany	yes
F	PO_6	150	50	71	15	22	80,00	F-GZTC	F-GZTC	100,00	France	France	yes
L	.LC_5	150	50	47	12	7,6	75,00	SP-HAF	SP-HAF	100,00	Poland	Poland	yes
L	.LC_7	100	40	112	25	30	72,00	SP-HAG	SP-HAG	100,00	Poland	Poland	yes
L	.LC_8	100	40	100	21	30	74,00	SP-HAD	SP-HAD	100,00	Poland	Poland	yes
M	IAV_3	150	50	58	12	18	83,00	SX-MAR	SX-MAR	100,00	Greece	Greece	yes
S	EH_1	150	50	36	8	5	83,00	SX-LDS	SX-LOS	83,33	Greece	Greece	yes
S۱	WG_2	150	50	82	15	24	88,00	OK-TVV	OK-TVV	100,00	Czech Republic	Czech Republic	yes

**Average Success: 91% - Average Confidence: =77%** 

To examine the effectiveness of the algorithm, high definition video of the actual movements of Nikos Kazantzakis airport of Iraklion was recorded. The time period was from May to August 2015.

The recording was made from the site of the airport control tower (Figure 14).

Table 1 presents the results of the algorithm. The Average Success of the algorithm was 91%.

- Use of a fixed position video recording system specialized for monitoring moving objects.
- Combination with airport collaborative decision making systems to confirm registration numbers and optimize results.
- Implementation of camera networks in an airport to achieve full area, real time registration scanning and consequently produce a low cost airport surface movement ground control system (SMGCS).
- Use as part of a hybrid radar camera surface surveillance system for optimum result

