

Implementation of AI methodologies for point cloud classification in the Spanish national LiDAR project (PNOA-LiDAR)

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GOBIERNO DE ESPAÑA

MINISTERIO DE TRANSPORTES Y MOVILIDAD SOSTENIBLE

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Season 1 recap



WINNER: AI-BOT

HUMAN: 8,85s

9,50s





- More classes
- Less budget
- Faster processing time

PNOA-LiDAR

THE BEAM THAT SEES ALL



ABOVE THE CLOUDS, IT PAINTS THE WORLD.

COMING SOON TO A SELF-DRIVING WORLD!

STARFORM: AERIAL SCANNING, ALSOFTIMES, AND GPS. WRITTEN BY ALI ASRISTEB.
PRODUCED BY SKYVIEW DYNAMICS.

SEASON 2

SPANISH NATIONAL LIDAR PROGRAM

PROYECTO PNOA-LIDAR: AÑOS DE VUELO 3º COBERTURA



- 2022-2025
- 505.000 km²
(2023-2025 160.000 km² per year)
- 5 p/m²
- RMSEz: 10 cm
- RGBI medium format
- GSD 25cm
- Galaxy T2000
(except Catalunya Terrain Mapper)

IA IMPLEMENTATION

A thick, hand-drawn style orange line that underlines the text "IA IMPLEMENTATION". It starts under the first letter 'I' and ends under the last letter 'N', following the baseline of the text.



Levels of classification

- NPC0: traditional classification (geometric, macros...)
-

- NPC1:
 - Methodology: AI (inference)
 - Basic Classes: ground, vegetation, buildings,
 - Advanced classes: powerlines, poles, antennas, wind turbines, solar panels, vehicles, water
 - Classes under testing: walls, dams, buildings (inner courtyards, facades, types of roofs (flat, pitched, green))
 - Reliability: not defined



Levels of classification

- **NPC2:**
 - Methodology: semi-automatic and manual review
 - Classes: NPC1
 - Reliability: in accordance with what is established in the specifications



Levels of classification

The following F-score accuracy thresholds are established depending on the classification level and class

- Basic classification: general F-Score $\geq (0,97/0,96/0,90)$ in rural / urban / mountainous terrain (rocky or scrubland areas)
- Ground class: $\geq 0,97/0,97/0,85$ in rural / urban / mountainous terrain (rocky or scrubland areas)
- Vegetation class: $\geq 0,97/0,90/0,60$ in rural / urban / mountainous terrain (rocky or scrubland areas)
- Buildings: $\geq 0,95/0,90$ in rural / urban
- Bridges: $\geq 0,95$

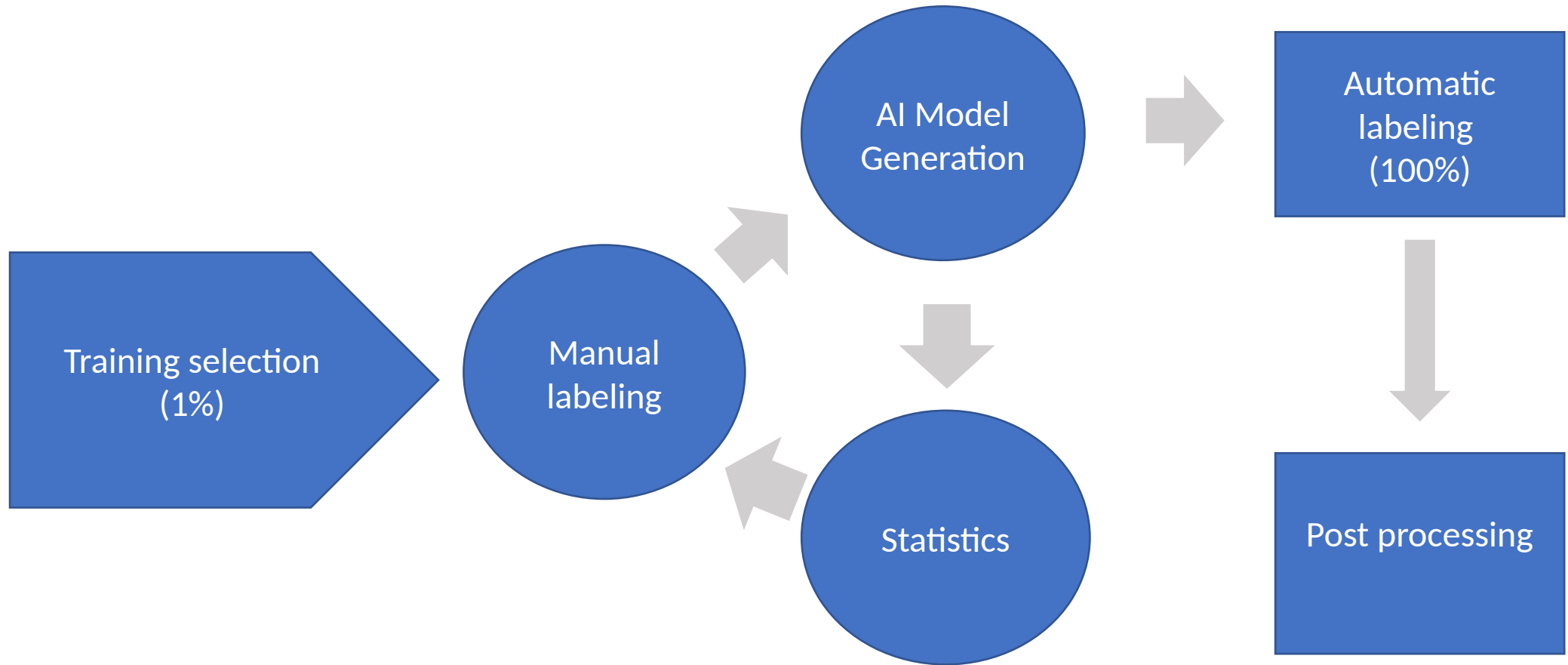
- Advanced classes : $\geq 0,85$

- For advanced classes, the specification have been conservative in some classes and higher accuracies are being achieved.

METHODOLOGY



METHODOLOGY AI



Training

Type	Training		Validation
Number of files	166		40
1 - Default	32,489,109	3.08%	7,076,152
2 - Ground	654,795,407	62.11%	160,995,975
3 - Low vegetation	69,325,684	6.58%	14,733,267
4 - Medium vegetation	243,008,648	23.05%	52,737,137
6 - Building	28,085,221	2.66%	7,094,186
7 - Noise	135,264	0.01%	31,292
9 - Water	13,165,557	1.25%	4,456,189
12 - Overlap	361,299,501		95,846,486
14 - Power lines (electric cables)	952,235	0.09%	189,870
15 - Power pylons	170,231	0.02%	36,555
16 - Bridges (structural elements)	50,070	0.00%	14,094
17 - Bridges (deck)	713,389	0.07%	104,751
64 - Vehicles (trains_cars_boats)	1,075,978	0.10%	372,550
65 - Construction walls	796,976	0.08%	107,828
66 - Signs_various_advertising	12,156	0.00%	2,611
67 - Solar panels	6,369,639	0.60%	1,732,608
68 - Wind turbines	65,162	0.01%	14,746
69 - Dam / Dike	141,328	0.01%	1,139
70 - Antenna	3,021	0.00%	349
76 - Fixed urban furniture	159,008	0.02%	14,744
77 - Mobile urban furniture	34,034	0.00%	4,572
78 - Irrigation structure (pivot)	17,297	0.00%	91
79 - Bales	24,225	0.00%	11,094
80 - Shade nets	76,941	0.01%	7,286
81 - Greenhouses	415,606	0.13%	74,928
82 - Vegetation walls	583,078	0.06%	206,933
83 - Metal fences	78,347	0.01%	13,640
84 - Aerial pipelines	18,840	0.00%	10,473
85 - Electrical transformer station	42,745	0.00%	5,974
86 - Ruins	10,571	0.00%	4,232
87 - Ramparts / city walls	345	0.00%	0
88 - Hoppers and conveyor belts	11,537	0.00%	4,116
89 - Railway catenary gantry	17,691	0.00%	2,002
90 - Material accumulation	327,208	0.03%	

RESULTS



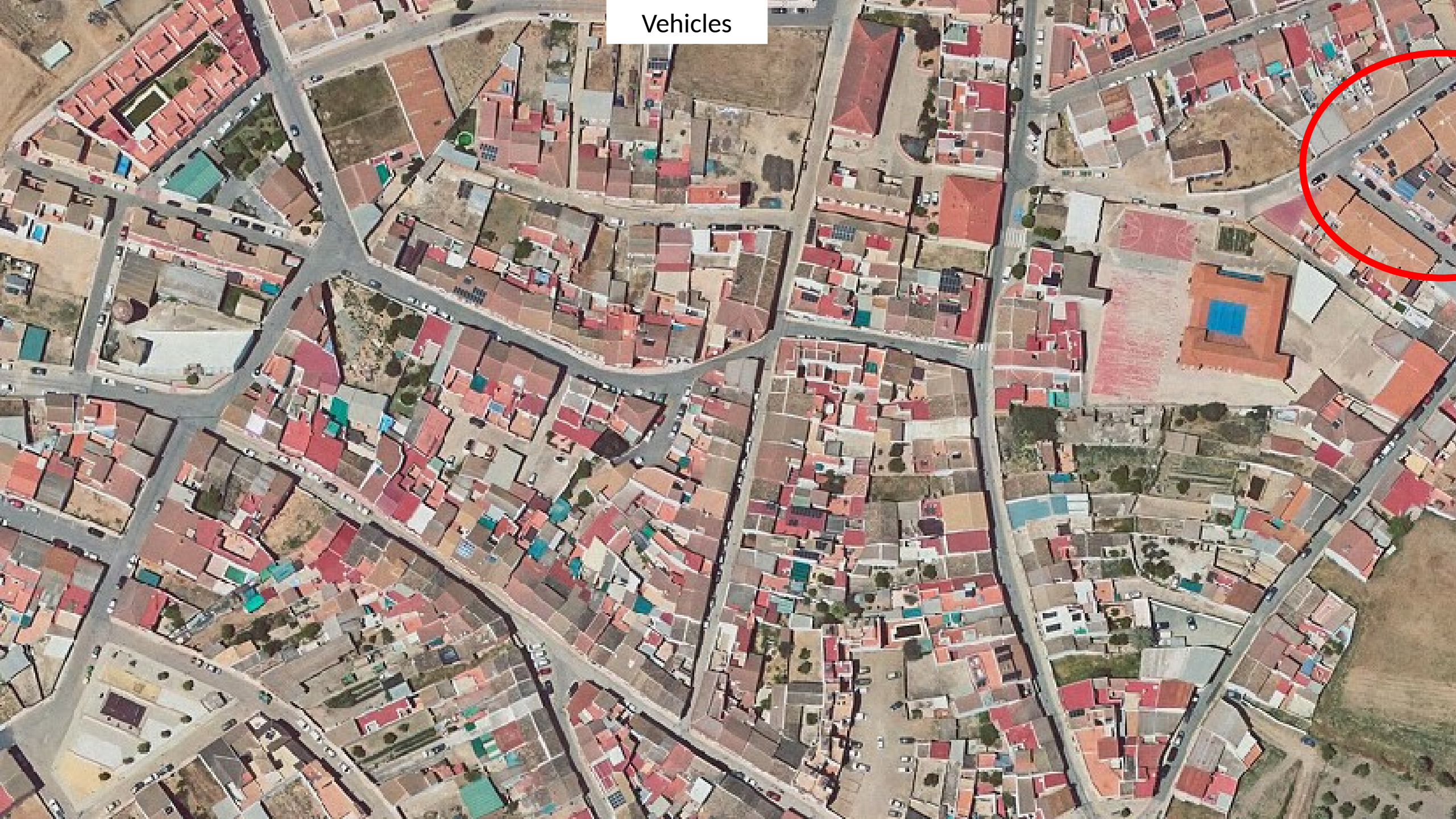
Solar panes



Solar panels



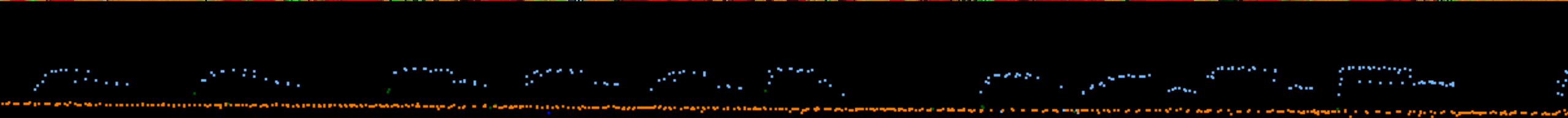
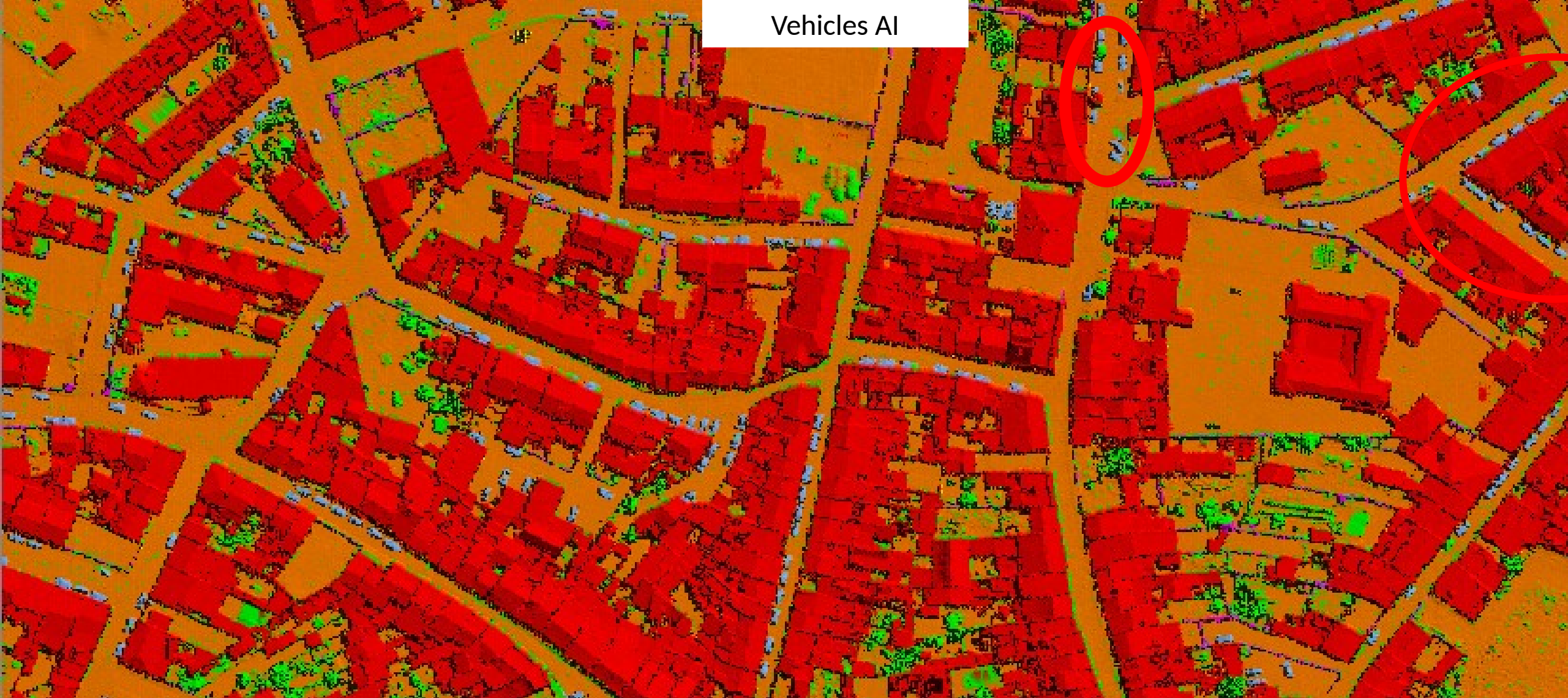
Vehicles



Vehicles traditional



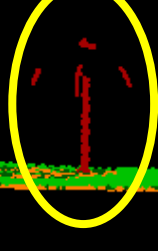
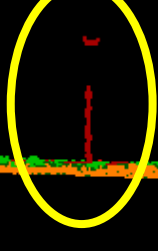
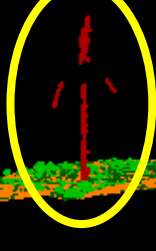
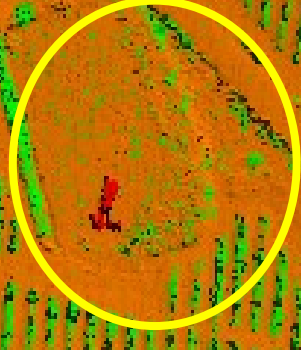
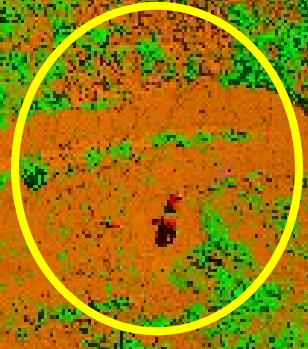
Vehicles AI

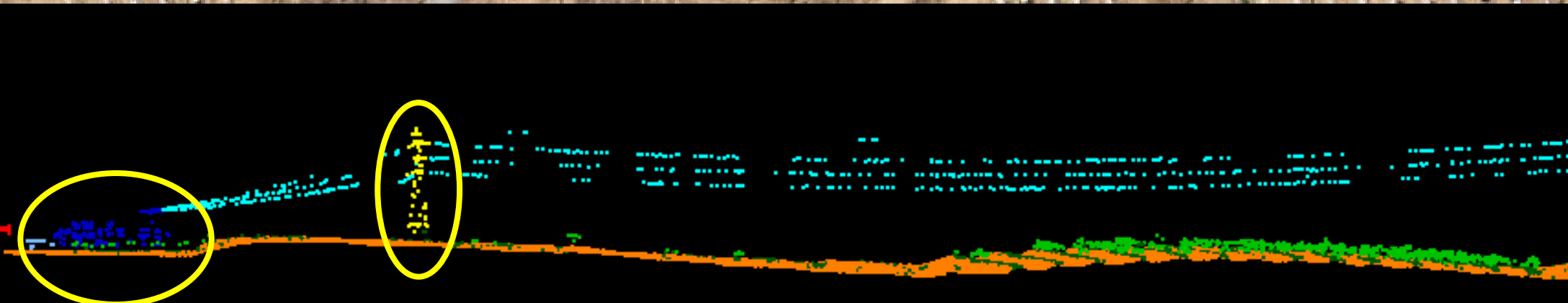


Wind turbines



Wind turbines





Electrical transformer station

Pole

Walls



Walls





ERRORS POSTPROCESSING



ERRORS POSTPROCESSING

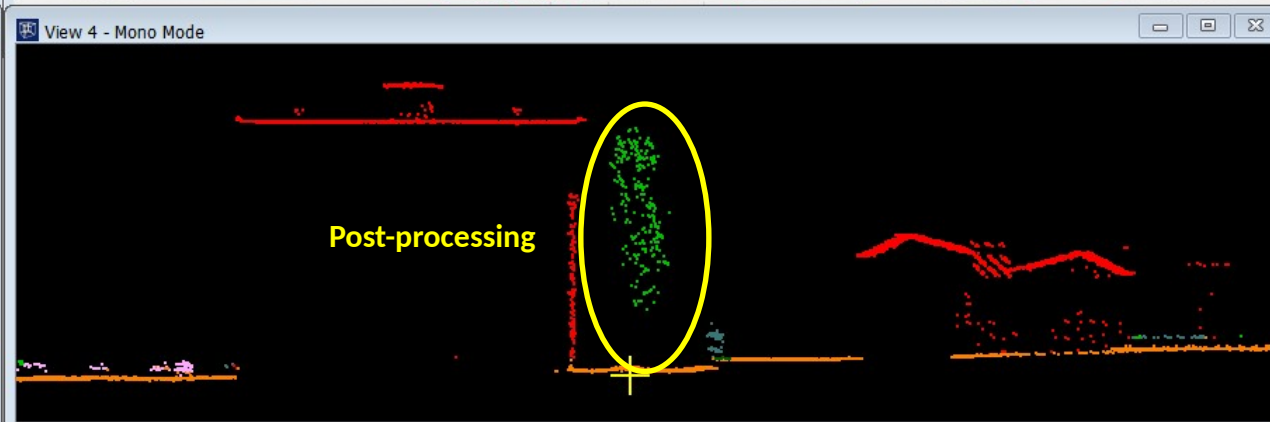
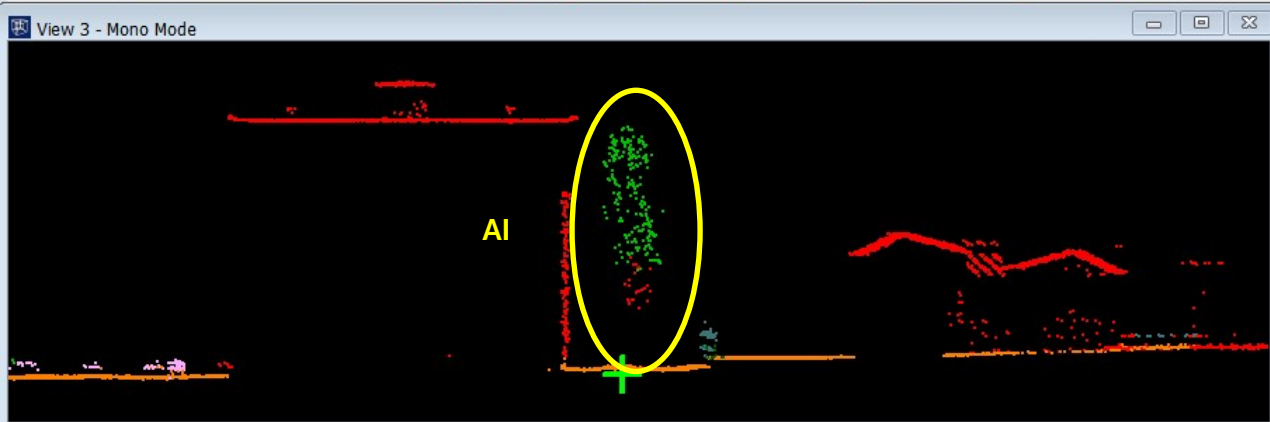
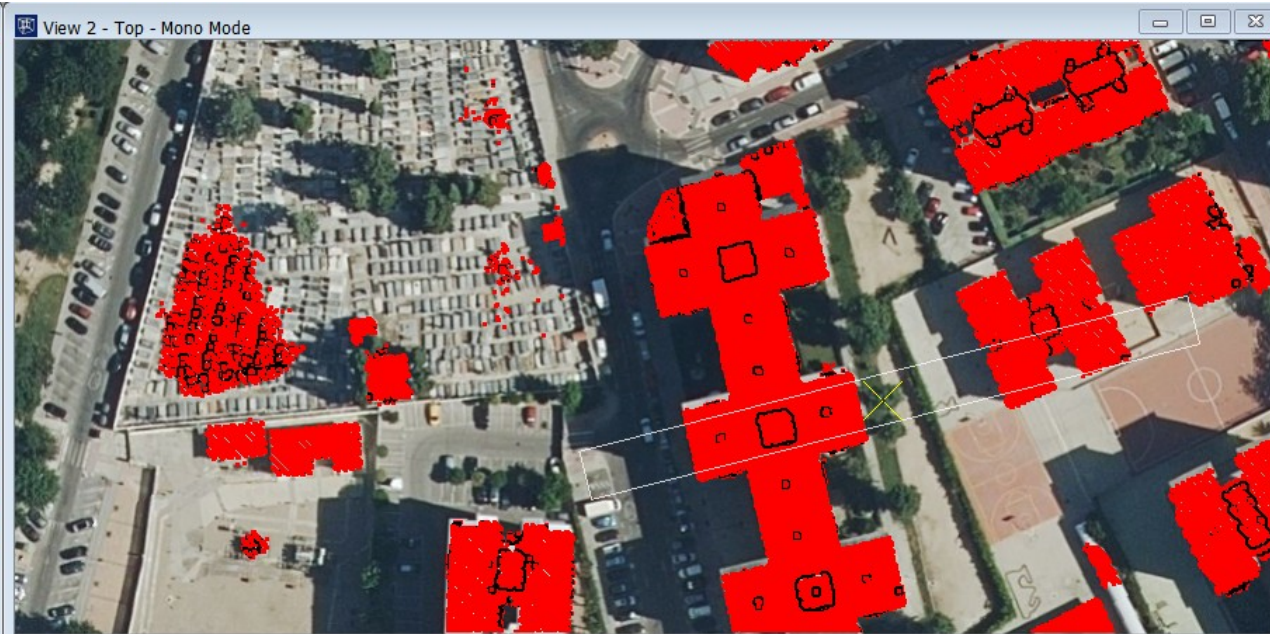
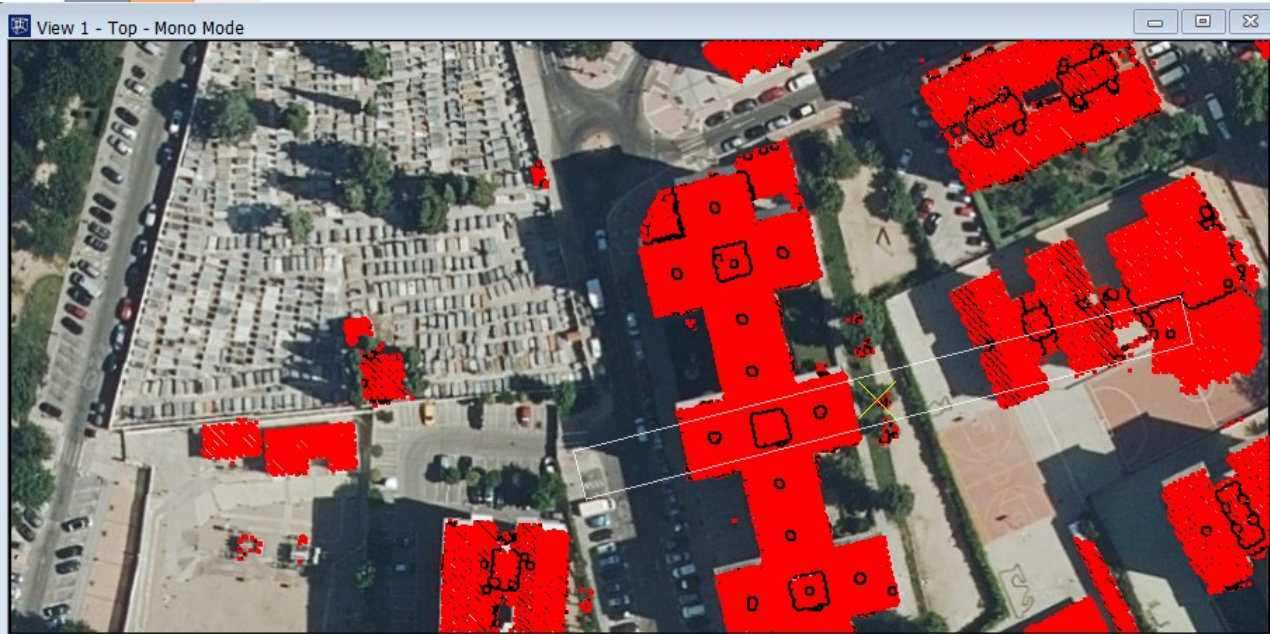


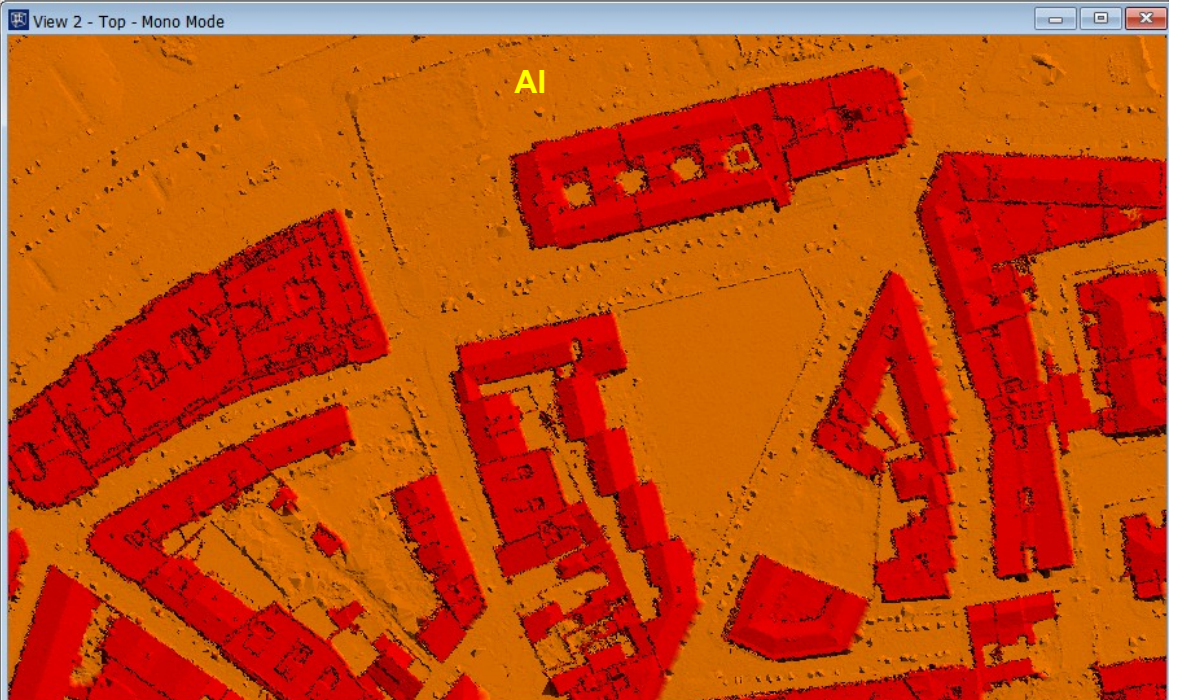
AI



Post-processing









		suelo	veg baja	veg media	veg alta	edif	agua	cables	torres	puentes	vehic	muros	pan sol	aerog	antenas
	Acc	f1_class_2	f1_class_3	f1_class_4	f1_class_5	f1_class_6	f1_class_9	f1_class_14	f1_class_15	f1_class_17	f1_class_64	f1_class_65	f1_class_67	f1_class_68	f1_class_70
V_5(100)	96,4191	97,6079	66,1368	98,7722		97,5220		94,5259	74,6996	80,4914	90,8084	56,1533	97,7593	81,4997	
V_5(100_post) 14 clases	95,9651	97,4580	66,0315	95,5041	99,1024	97,7937	54,2850	95,4492	75,5967	83,7060	90,9746	54,2632	97,9348	95,5578	75,3014

Class 2: Ground	Class 9: Water	Class 65: Walls
Class 3: Low vegetation	Class 14: Power lines	Class 67: Solar panels
Class 4: Medium veg	Class 15: Poles	Class 68: Wind generator
Class 5: High vegetation	Class 17: Bridges	Class 70: Antennas
Class 6: Building	Class 64: Vehicles	

CONCLUSIONS





	Classes	Costs
Traditional	Basics	10€/km ²
AI	Basics, advanced, under testing	7€/km ²

Potential future cost saving

- Reuse training data in other areas
- More mature methodology



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GRACIAS
POR SU ATENCIÓN

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