

Detecting solar panels and skylights on roofs using machine learning

SDS Business Presentation

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Hail Damage

Schadensbilanz für 2021

Schweiz verzeichnet grösste Unwetterschäden seit 2007

2 Milliarden Franken wegen Sturm und Hagel: 2021 wird für die Schweiz eines der teuersten Schadenjahre aller Zeiten

Rekord: Sommer-Unwetter richten im Kanton Luzern Schäden von 400 Millionen Franken an

Kanton Di 29.06.2021 - 14:30

Hagel sorgt für grösste Gebäudeschäden seit 15 Jahren



Hail Damage on Buildings



Elementar: Übersicht 2002 - 2021 – Elementarschäden an Gebäuden

	Elementar Total		Sturm		Hagel		Überschwemmung		Erdrutsch Steinschlag		Lawinen Schneedruck	
	Schadenssumme in Mio. CHF	Anzahl Schäden	Schadenssumme in Mio. CHF	Anzahl Schäden	Schadenssumme in Mio. CHF	Anzahl Schäden	Schadenssumme in Mio. CHF	Anzahl Schäden	Schadenssumme in Mio. CHF	Anzahl Schäden	Schadenssumme in Mio. CHF	Anzahl Schäden
2002	243.1	52'968	30.4	20'247	129.3	23'359	76.5	8'722	6.0	348	0.9	292
2003	139.3	47'757	57.6	31'331	39.7	10'388	37.0	4'546	1.2	239	3.8	1'253
2004	146.2	50'461	28.4	17'632	94.3	26'266	19.9	5'102	1.2	243	2.4	1'218
2005	880.6	58'930	21.2	10'846	142.9	24'525	695.4	18'405	7.9	405	13.2	4'749
2006	128.9	39'357	19.3	11'742	44.2	13'114	33.8	6'296	3.6	431	28.0	7'774
2007	414.9	51'939	35.4	18'837	81.5	14'701	294.2	17'746	2.7	232	1.0	423
2008	68.0	19'622	21.7	10'678	15.7	3'982	23.6	2'587	0.7	103	6.2	2'272
2009	414.8	83'357	47.5	19'576	318.9	55'368	22.5	2'915	1.4	132	24.5	5'366
2010	76.2	21'981	31.5	13'251	16.1	4'506	25.0	3'020	0.8	140	2.9	1'064
2011	307.2	64'451	70.8	25'933	206.5	35'011	27.9	3'060	0.7	79	1.2	368
2012	157.1	39'200	41.1	16'052	72.6	16'225	20.7	2'163	1.0	154	21.8	4'606
2013	250.9	39'354	30.5	10'309	166.4	23'220	46.2	4'022	2.4	225	5.4	1'578
2014	111.2	23'554	33.8	14'988	9.8	2'746	54.5	4'251	5.1	145	7.9	1'424
2015	130.2	28'894	28.2	14'056	31.0	7'153	62.9	5'070	0.9	129	7.2	2'486
2016	98.7	21'299	19.8	9'877	20.0	5'716	54.6	5'222	3.5	166	0.9	318
2017	217.0	40'935	48.6	18'622	74.9	16'363	90.6	5'463	1.7	108	1.2	379
2018	300.6	91'041	187.1	75'906	27.6	6'267	77.4	7'548	2.5	214	5.9	1'106
2019	116.5	33'446	46.9	22'500	21.8	5'510	29.5	2'186	1.1	117	17.2	3'133
2020	150.9	67'031	115.5	60'773	12.2	3'236	20.8	2'455	0.8	119	1.6	448
2021	1'085.0	115'465	69.8	17'234	779.8	74'988	210.2	18'079	3.7	298	21.5	4'866

Source: Schadenstatistik der VKG

→ Where do those damages occur?

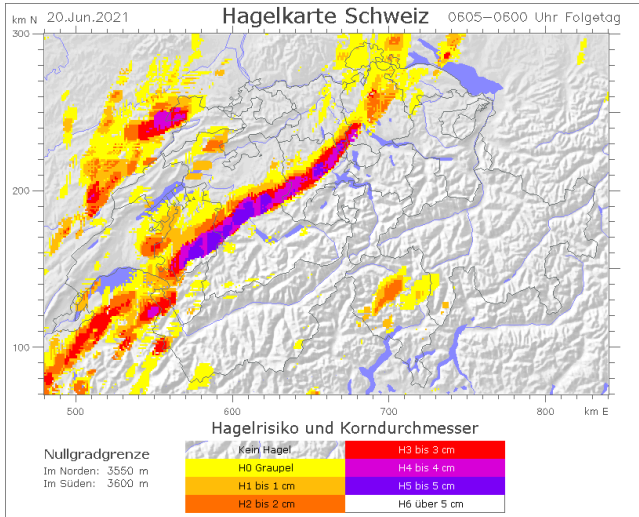
Hail Damage on Buildings



→ For a successful risk assessment, information on **hail** and **roofs** is needed.



Hail Damage



© meteoradar gmbh, Data: MeteoSchweiz



We know where it hails, ... but where are the vulnerable roofs?

→ Use machine learning to get information on roofs

Project for the building insurance of the canton Zug



1. Dataset of solar panels
2. Dataset of skylights

Content



- About Metetest
- Project goals for the *Gebäudeversicherung Zug* (GVZG)
- Data and workflow (for detecting solar panels and skylights)
 - Pre-processing
 - Training data
 - Algorithm training
 - Object detection
- Example of other applications

«Meteo-climate-environment-web Swiss army knife»



Weather Forecasts



Air Quality



Energy & Climate



Measurements



Wind & Ice



Digital Services



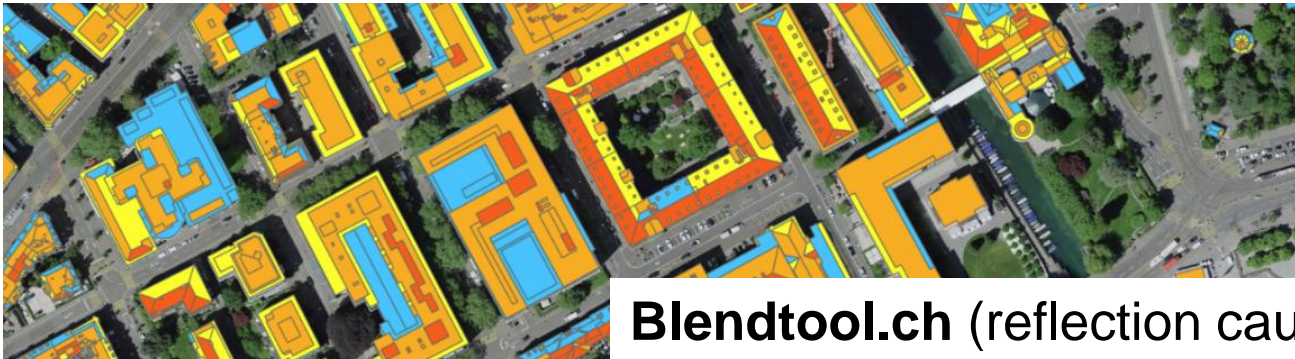
Geoinformatics



About Meteotest



Sonnendach.ch (roof dataset with PV production)



Blendtool.ch (reflection caused by solar panels)

Emission | **Immission**

Latitude [°] Longitude [°]
47.36550 8.5376

Azimuth [°]
180

Inclination [°]
30

Height above ground [m]
40

Terrain height [m]
407.9

Length [m] Width [m]
5 5

Bundle expansion [°]
5

Minimum relevance range [°] Maximum relevance range [°]
-30 30

Year
2023

Reset Calculate

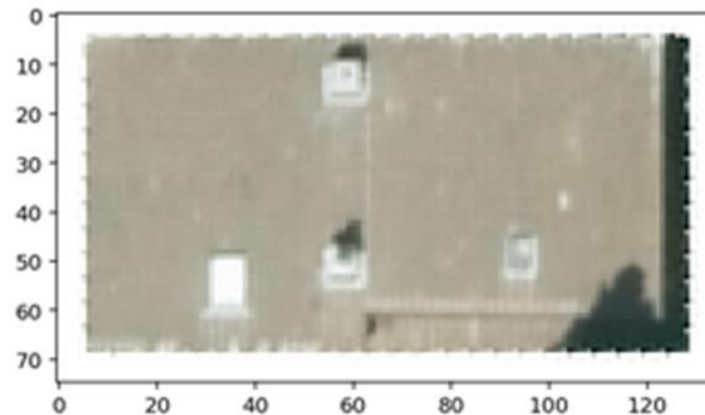
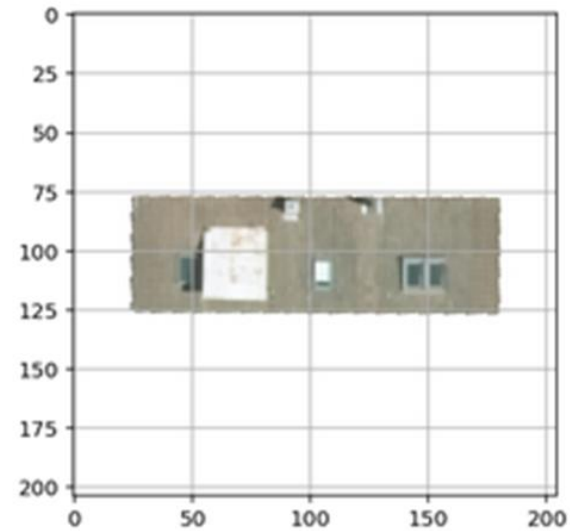
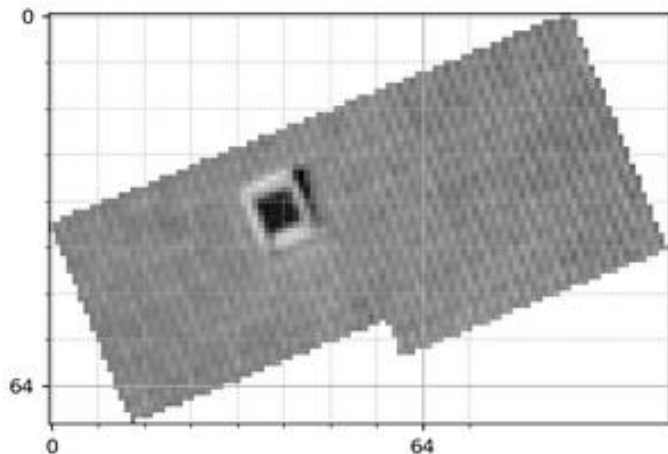
Map Sun position Reflection Table Analysis Export Help Legal notice

Legend:
- With reflections in the selected relevance area. Analysis of places of immission recommended.
- Without reflections in the selected relevance area.

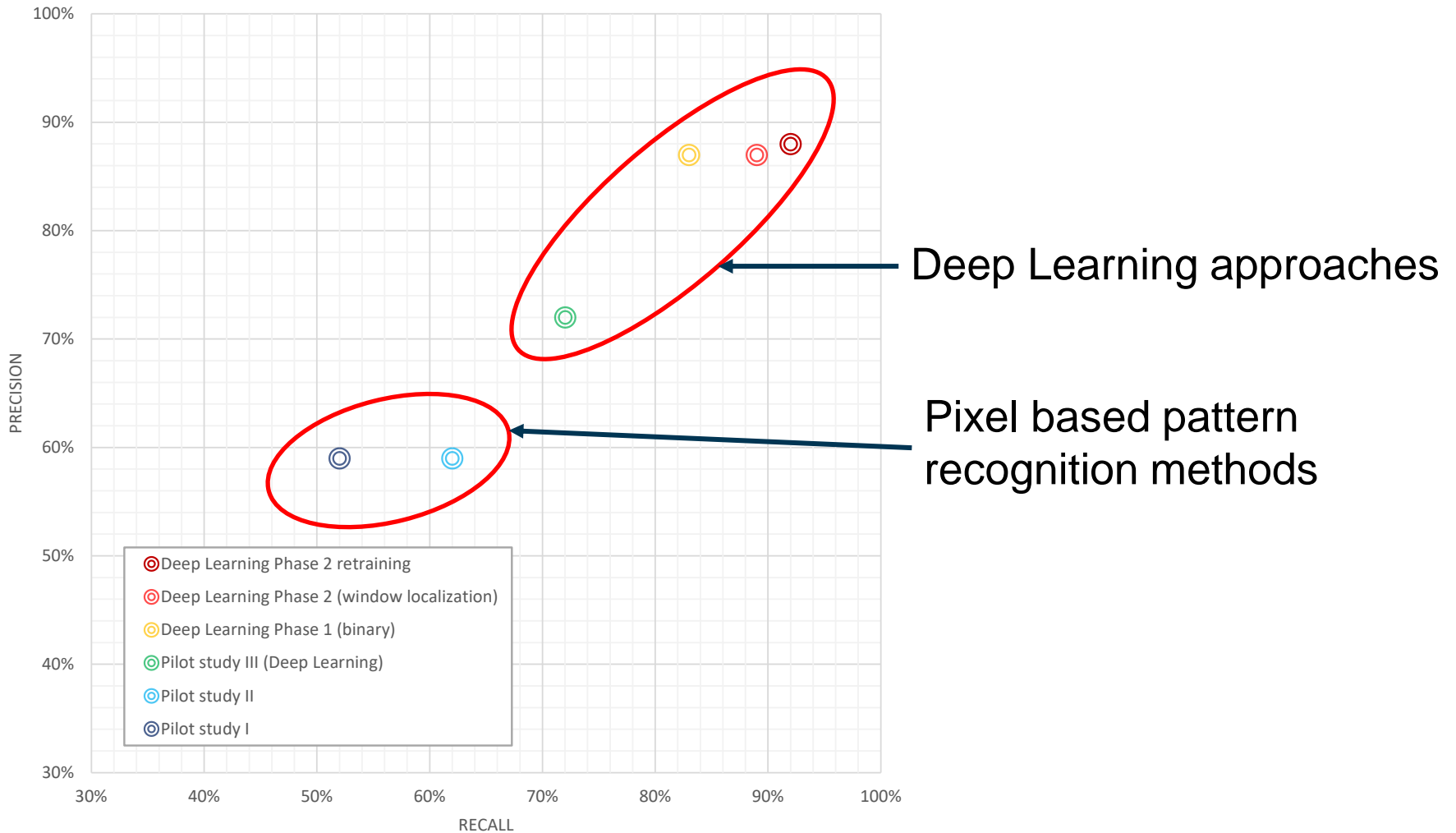
Meteotest and object detection

Detecting roof windows for a client
in the construction industry

→ What do you see?



Meteotest and object detection



Project Goals

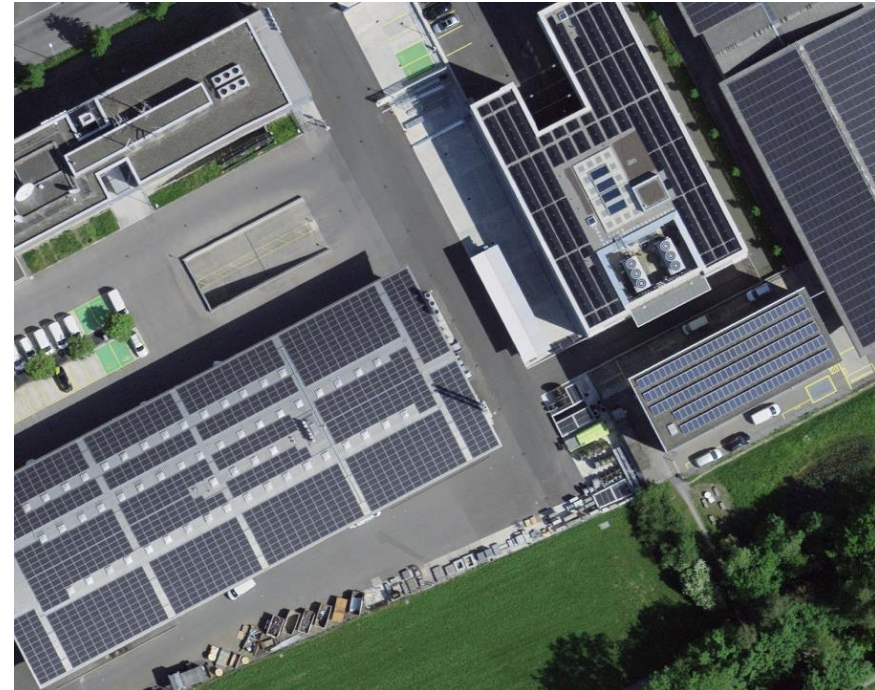


Dataset of solar panels in the canton of Zug

→ Detection of solar panels on roofs with area $\geq 10 \text{ m}^2$



© Christoffer Riemer/AS Solar



Project Goals

Dataset of skylights in the canton of Zug

→ Detection of skylights on flat roofs (slope $\leq 3^\circ$) with area $\geq 10 \text{ m}^2$



Data and Workflow



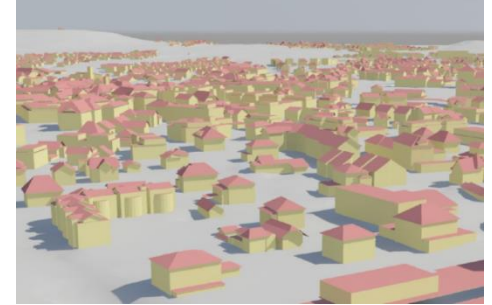
SWISSIMAGE 10 cm



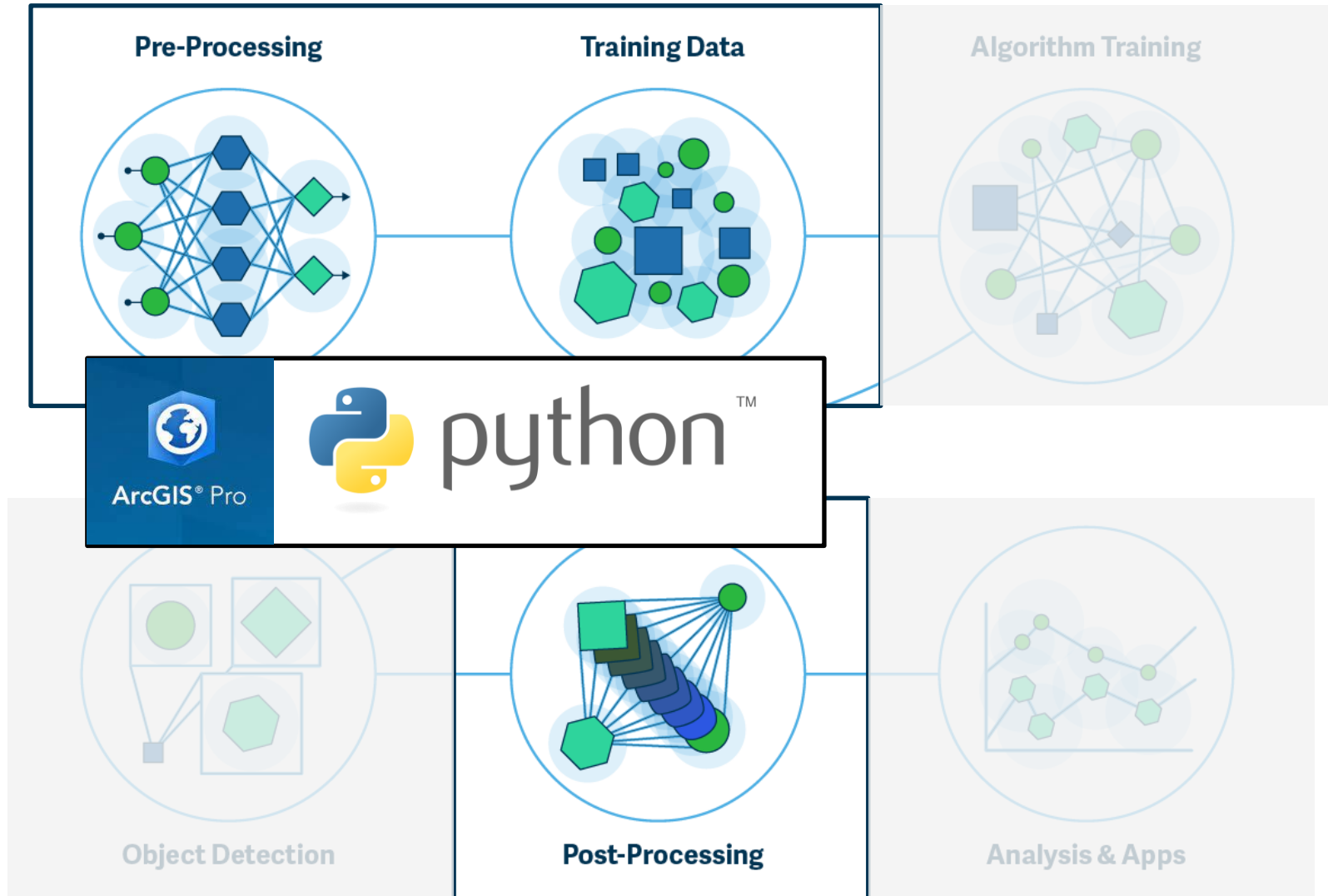
Manual annotating



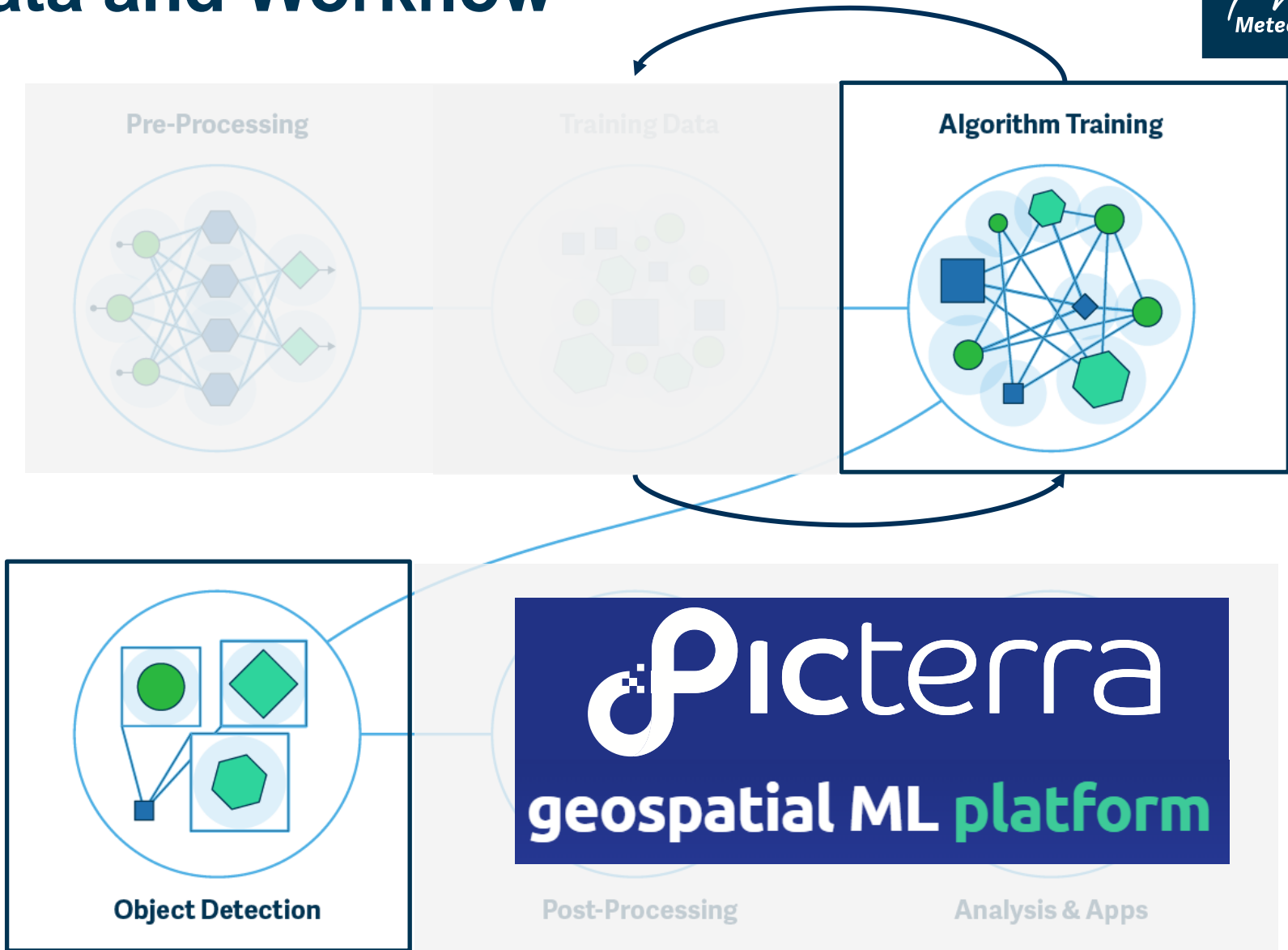
swissBUILDINGS3D 2.0



Data and Workflow

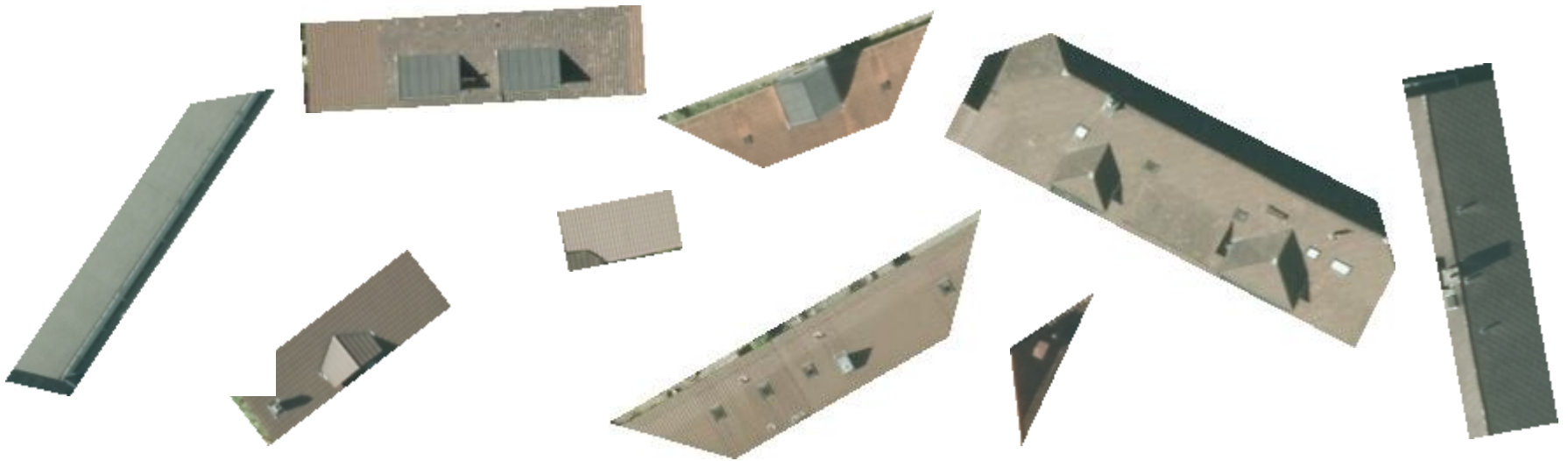
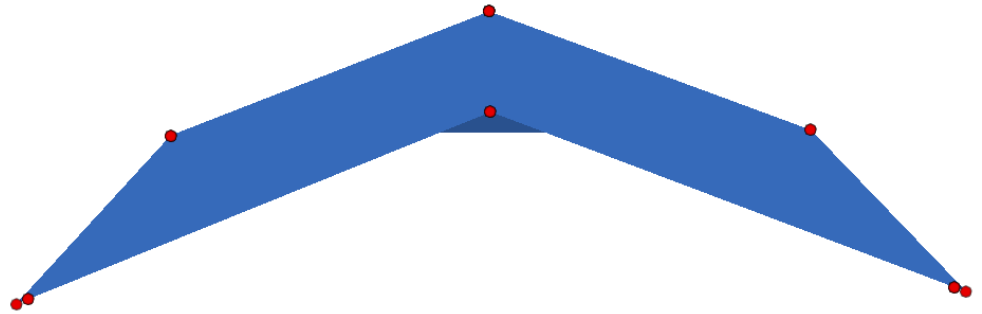


Data and Workflow



Pre-Processing

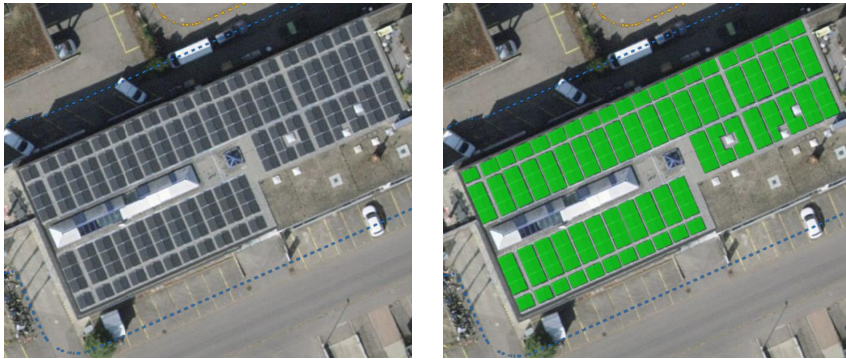
- Calculation of roof properties (area, slope)
- Selection of roofs with
 - Area $\geq 10 \text{ m}^2$
 - Slope $< 3^\circ$ (flat roofs for skylights)



Training Data

Solar panels

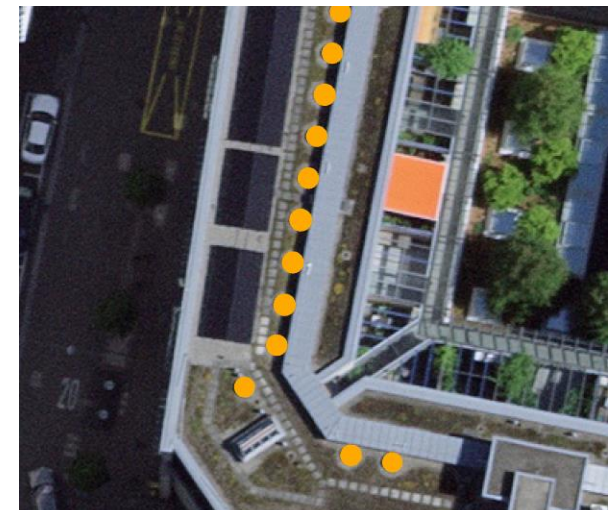
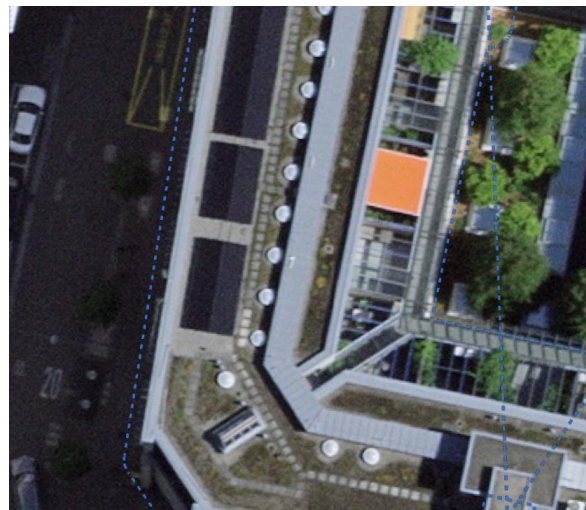
- Annotated 1'400 solar panel objects on 450 roof segments



Skylights

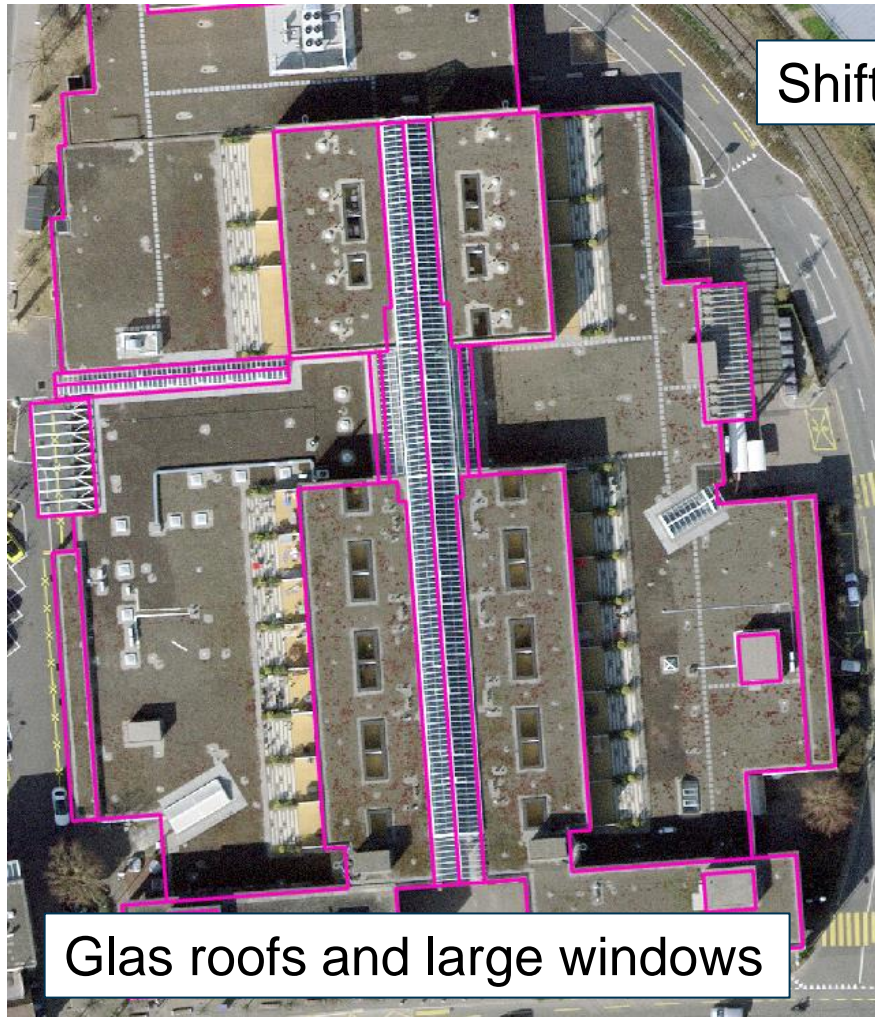
- Annotated 2'100 skylights on 600 roof segments
- Separated into 2 classes
 - Skylights bright (1'300)
 - Skylights dark (700)

Split annotation data into:
a) Training (75%)
b) Validation (25%)

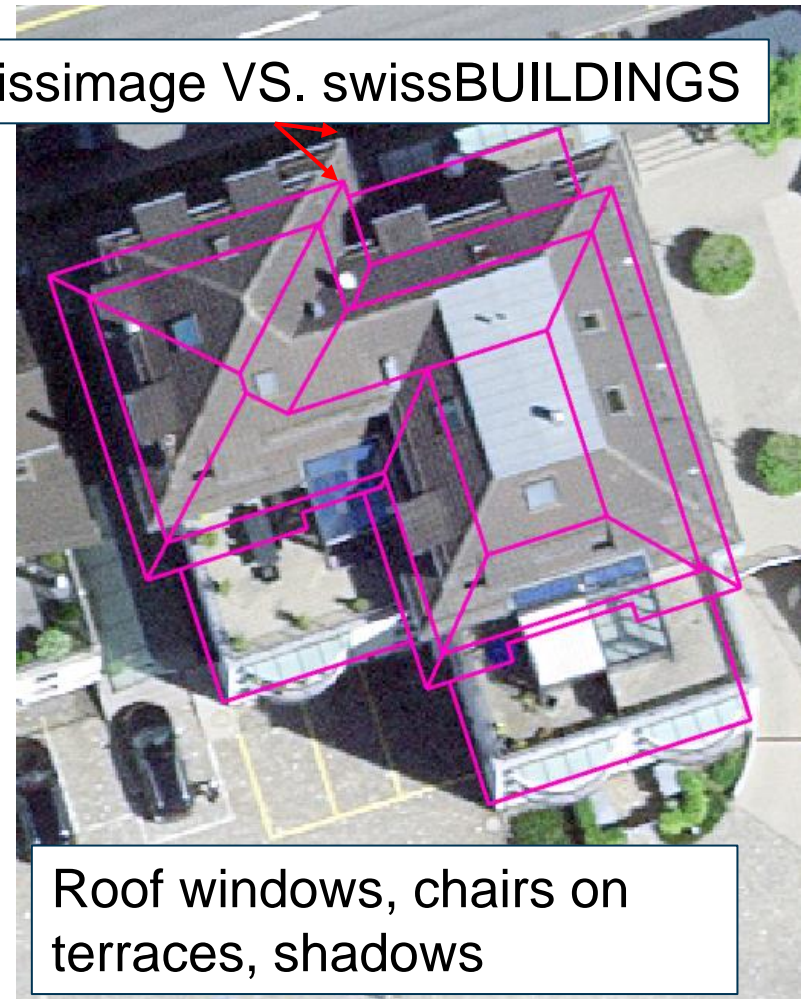


Training Data

Show the algorithm what is NOT a solar panel («negative examples»)



Shift Swissimage VS. swissBUILDINGS



Algorithm Training – Example solar panels

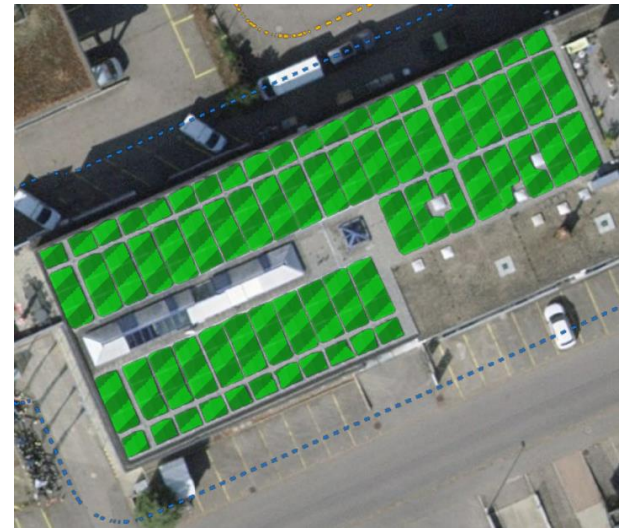
Aerial image



Annotation

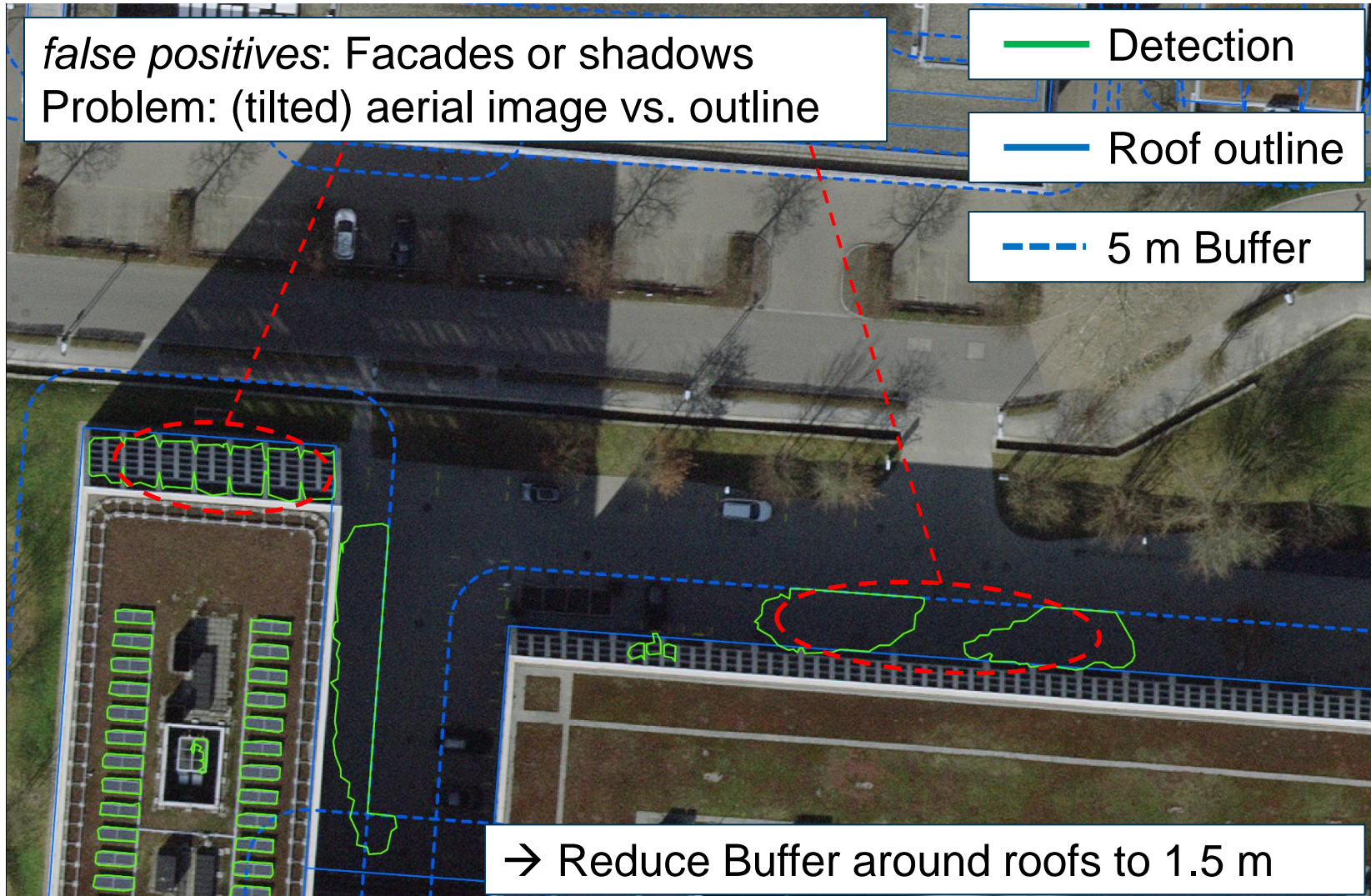


Detection

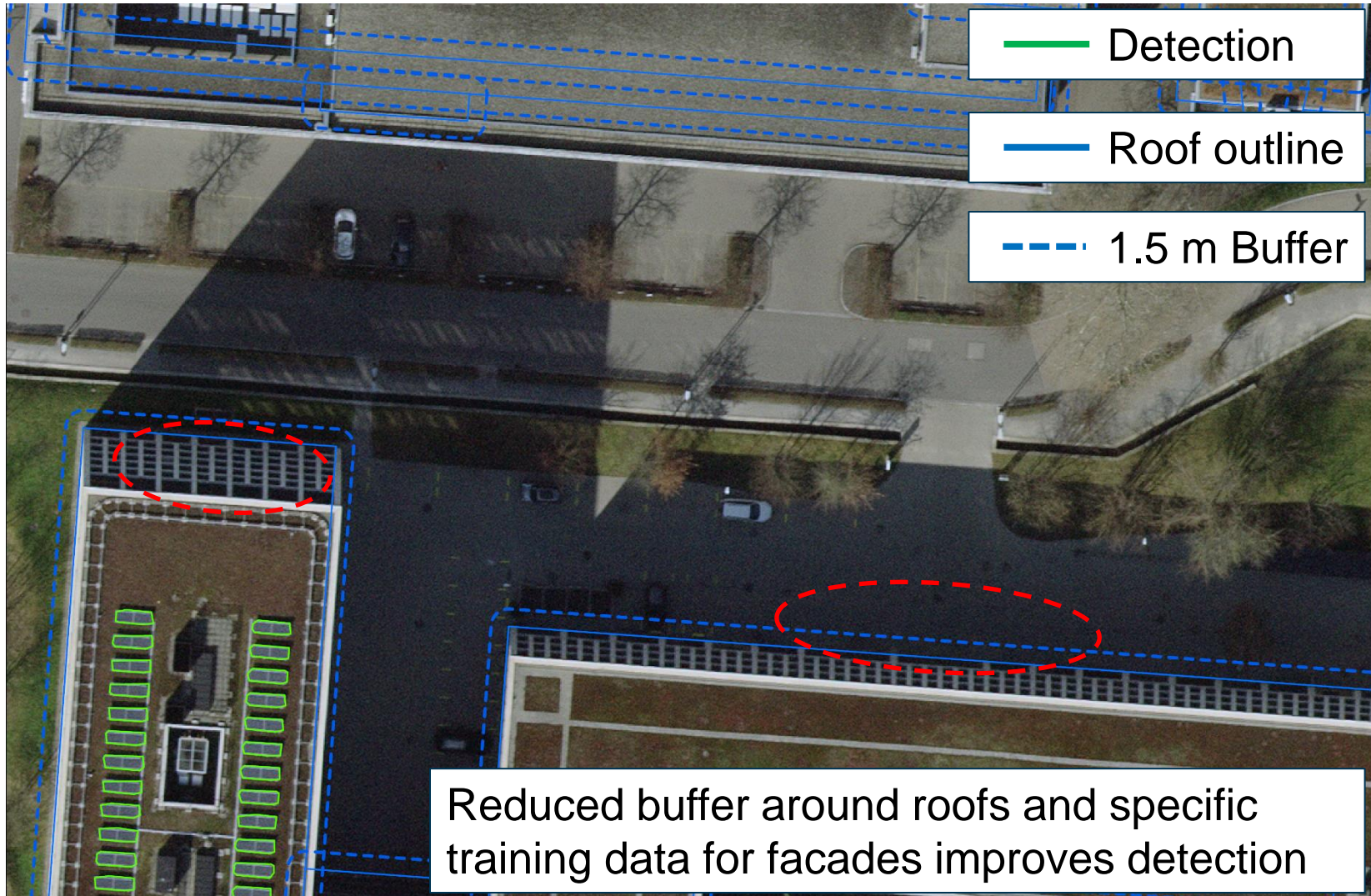


→ Good results early on. Solar panels are recognized and other objects differentiated.

Algorithm Training – Challenges solar panels

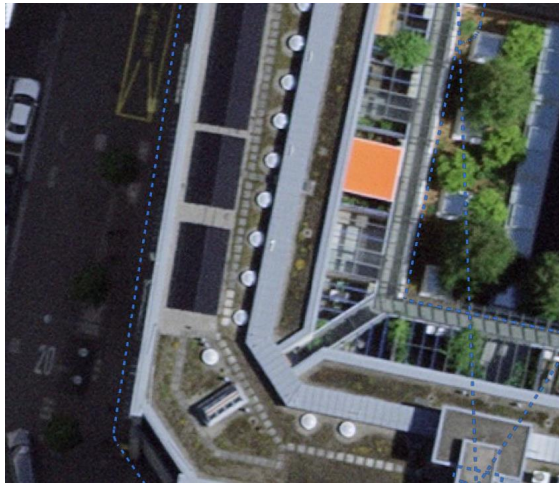


Algorithm Training – Challenges solar panels

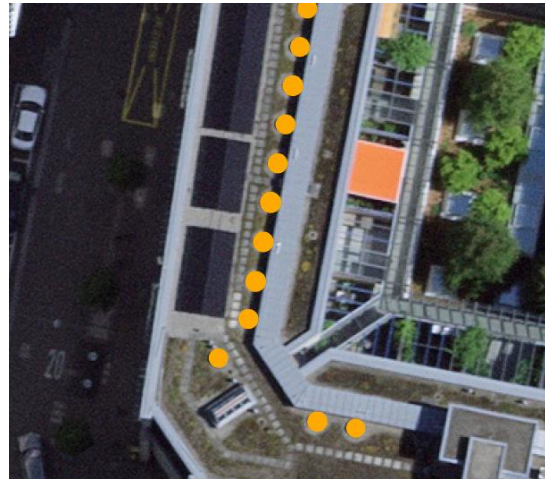


Algorithm Training – Examples skylights

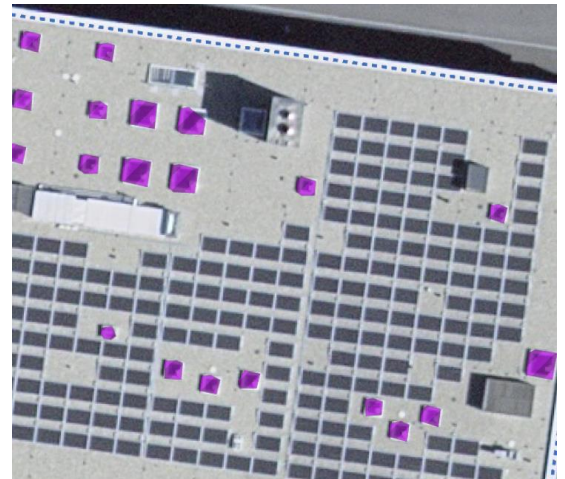
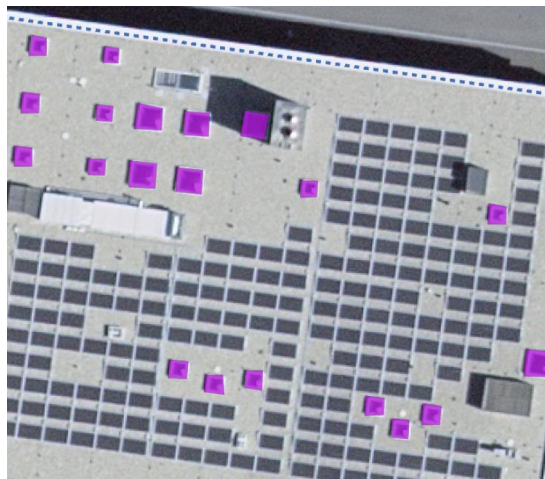
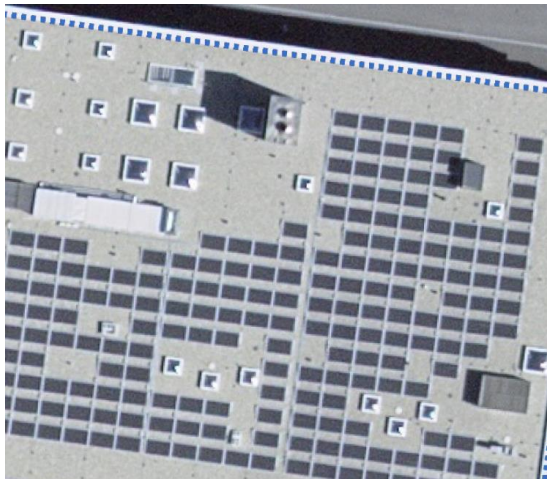
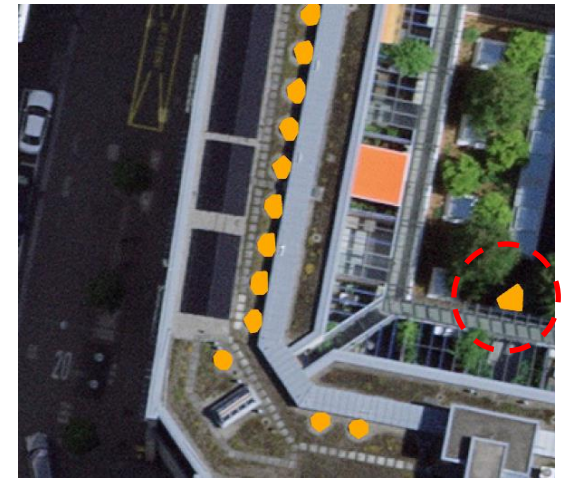
Aerial image



Annotation



Detection

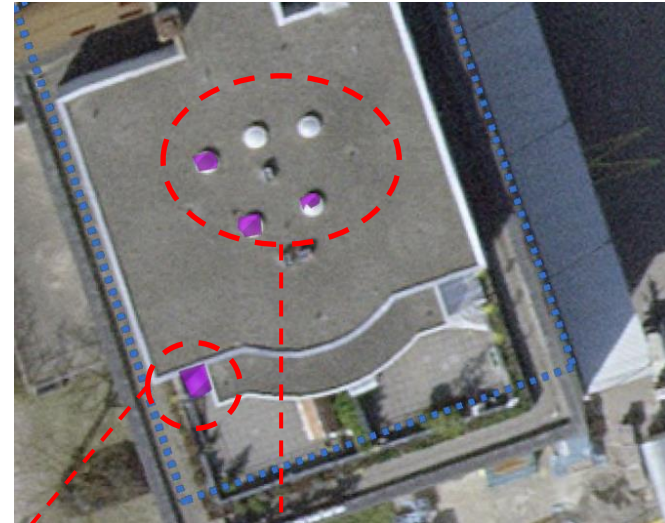


Algorithm Training – Challenges skylights dark

Aerial image



Detection



- Other dark objects on roofs (small roof windows, antennas, air vents) can be hard to differentiate
- «*false positives*» on bright skylights lowers precision (only 76%)
43% of «*false positives*» are bright skylights. These should not be considered actual false positives. Overlaps removed in post-processing.

Algorithm Training – Precision and Recall

Object Type	Precision	Recall
Solar panels	89%	93%
Skylights light	95%	95%
Skylights dark	76%	92%

Percentage of detections that
matched an outline
(key word: *false positives*)

Percentage of outlined
objects, which are
detected

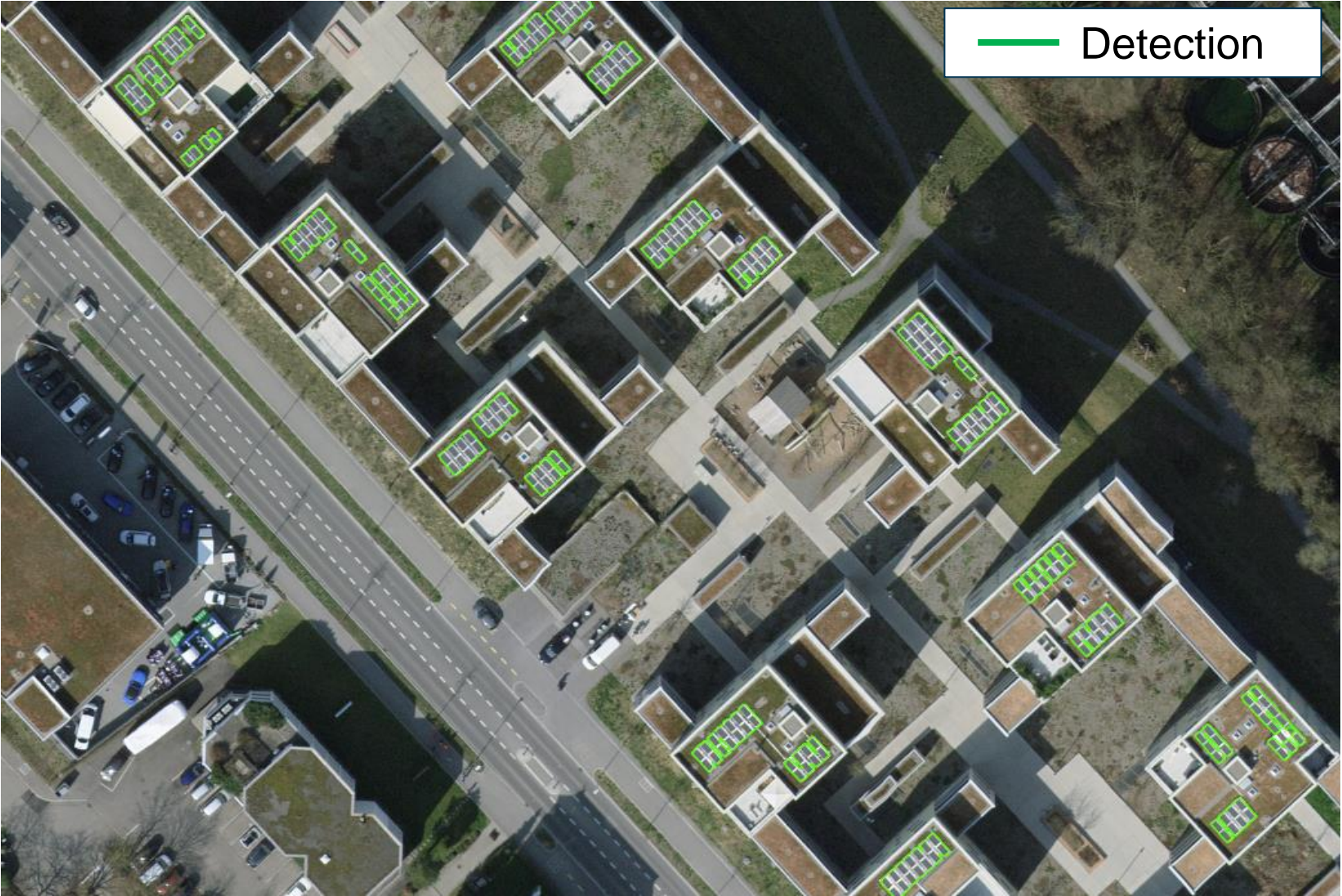
Object Detection – Solar panels



Object Detection – Solar panels



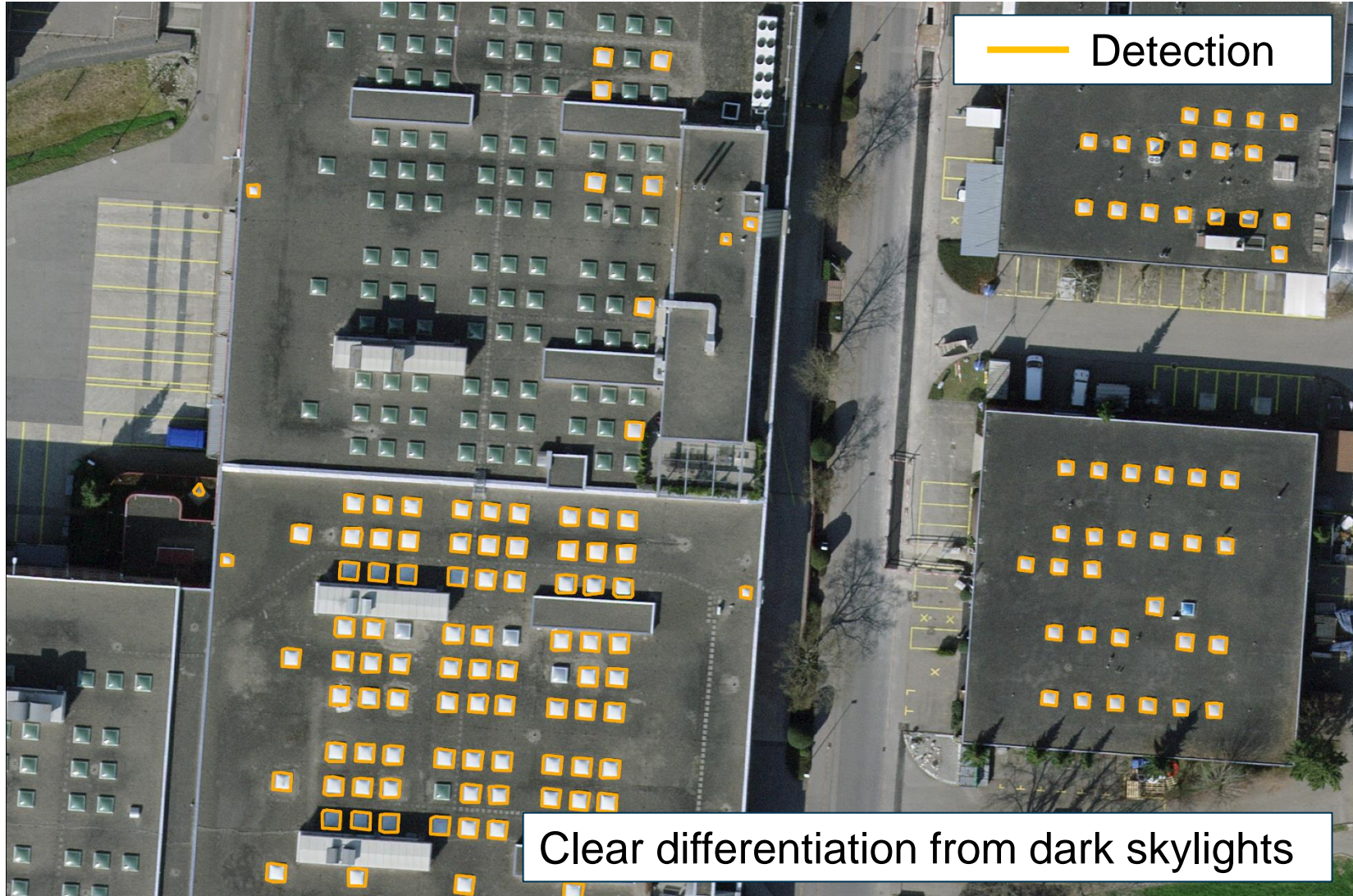
Object Detection – Solar panels



Object Detection – Skylights bright



Object Detection – Skylights bright



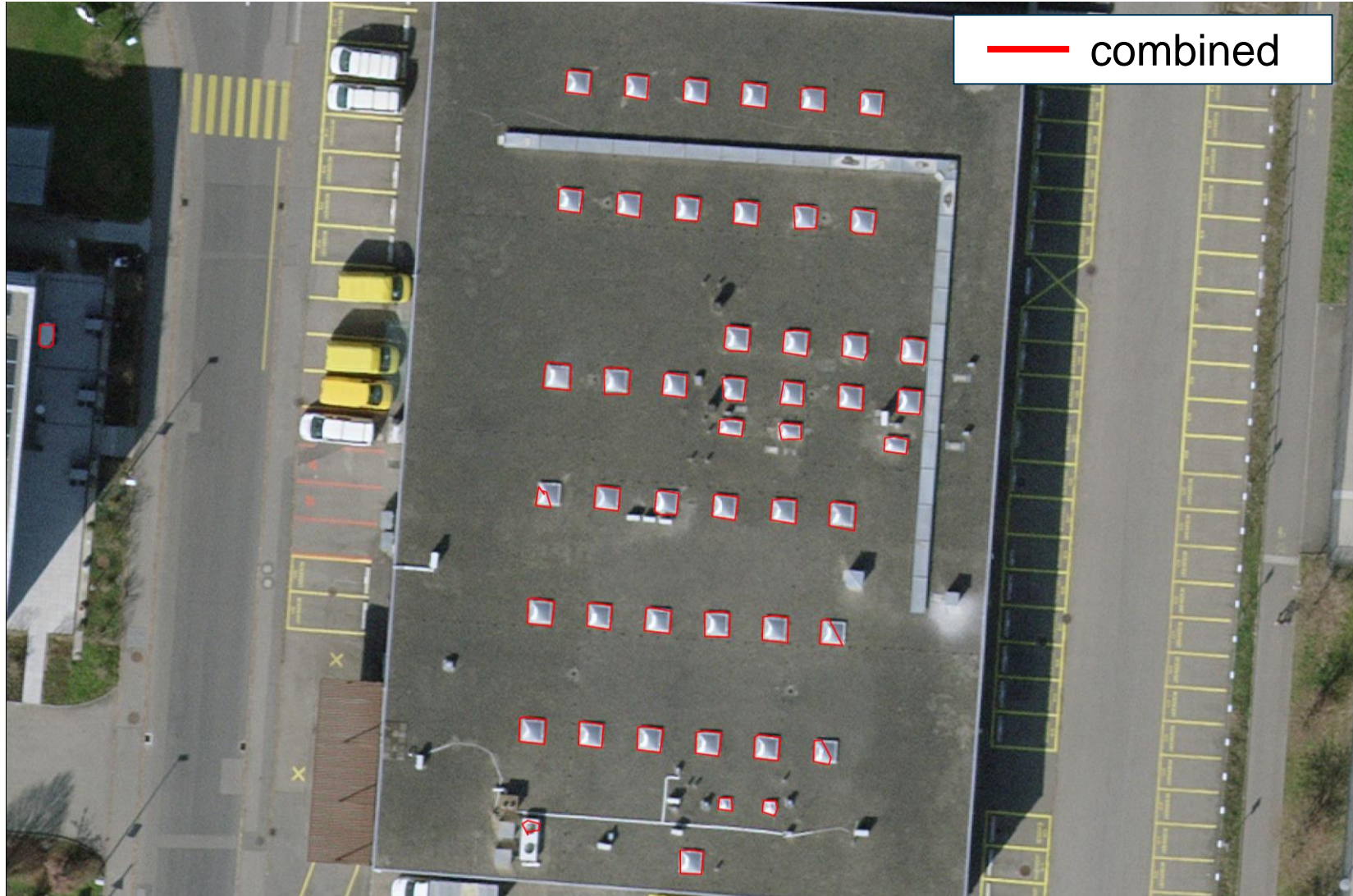
Object Detection – Skylights dark



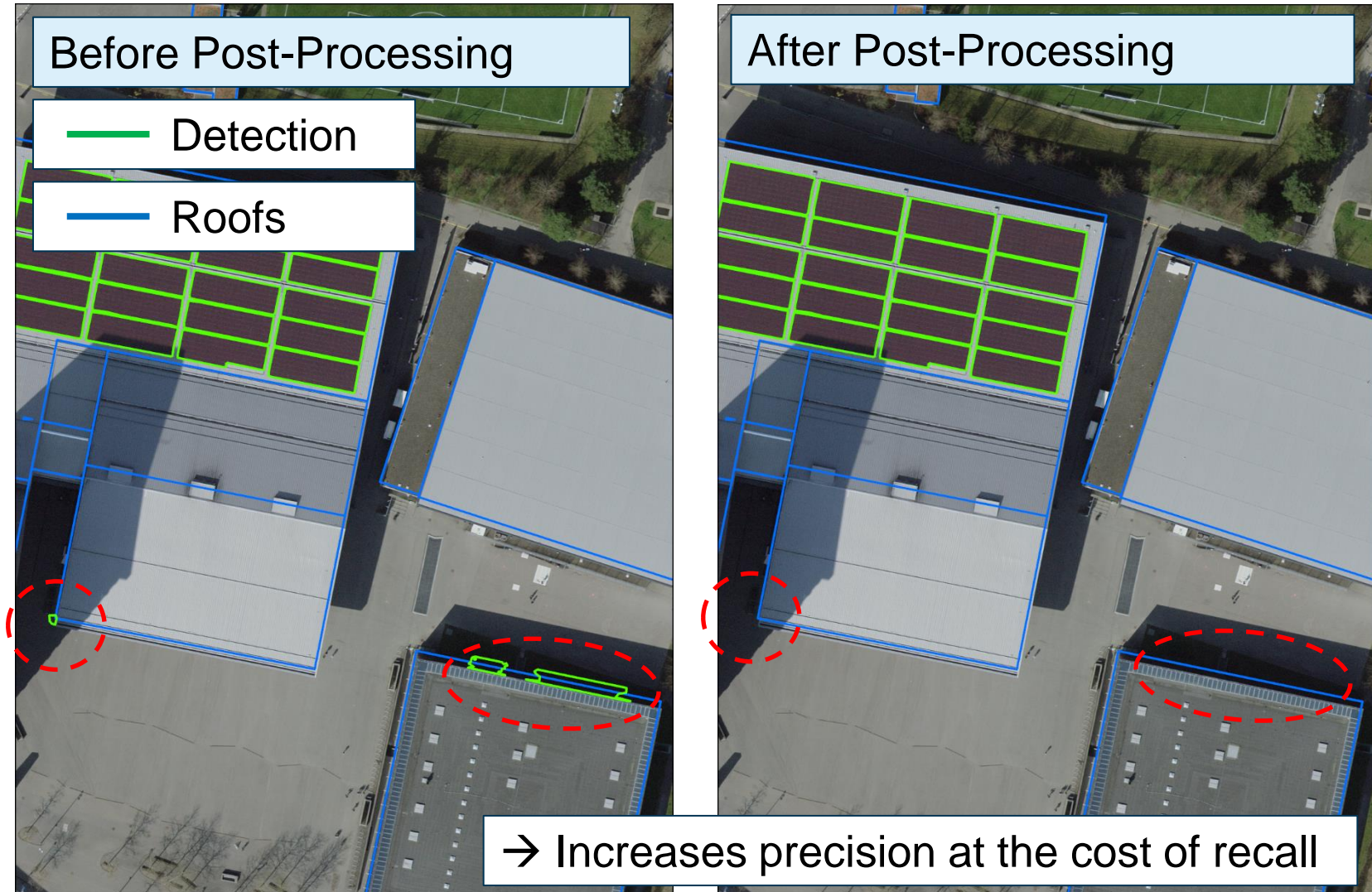
Post-Processing – Overlaps between Skylight types



Post-Processing – Overlaps between Skylight types



Post-Processing – Solar panels on facades due to buffer



Summary

- Successful detection of solar panels and skylights in the whole canton of Zug
 - 6'900 solar panels
 - 14'900 skylights (13'000 type light; 1'900 type dark)
- Recall and precision of around 90%
- ~ 4 weeks of work within 6 months
- Cost effective, flexible, and fast approach combining GIS methods and machine learning approaches
- Trained detectors can be scaled to new and much larger areas
- Updates over time are easy and inexpensive

Further Applications – Detecting drainage ditches in bogs

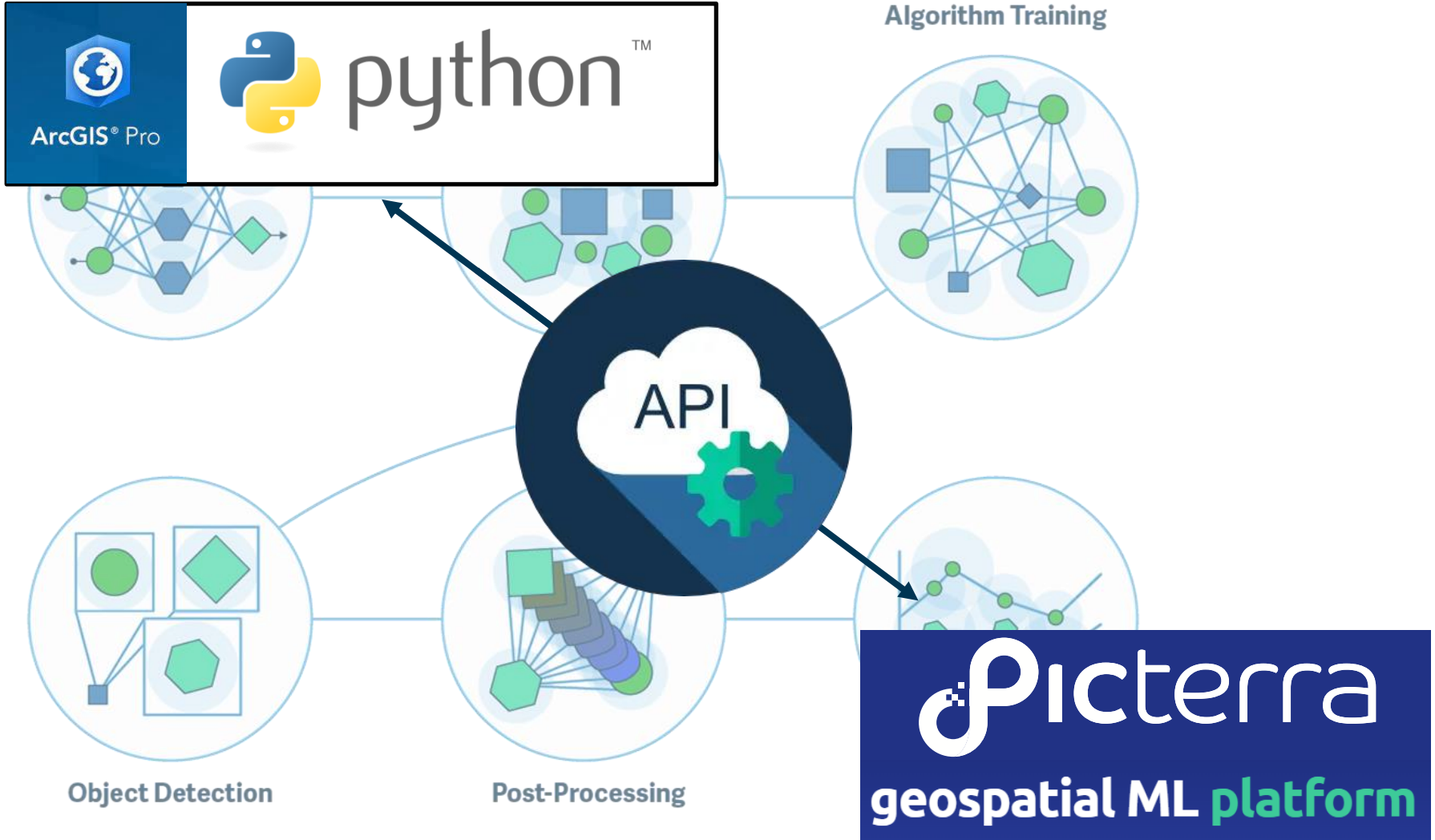


Aerial image



Detection

Further Applications



Questions & Feedback?



www.meteotest.ch