

# The PERIGEO project: Inertial and Imaging sensors Processing, Integration and Validation on UAV platforms for Space Navigation

P. Molina <sup>a</sup>, E. Angelats <sup>a</sup>, I. Colomina <sup>a</sup>, A. Latorre <sup>b</sup>, J. Montaño <sup>b</sup>, M. Wis <sup>b</sup>



# Summary

## 1. PERIGEO project

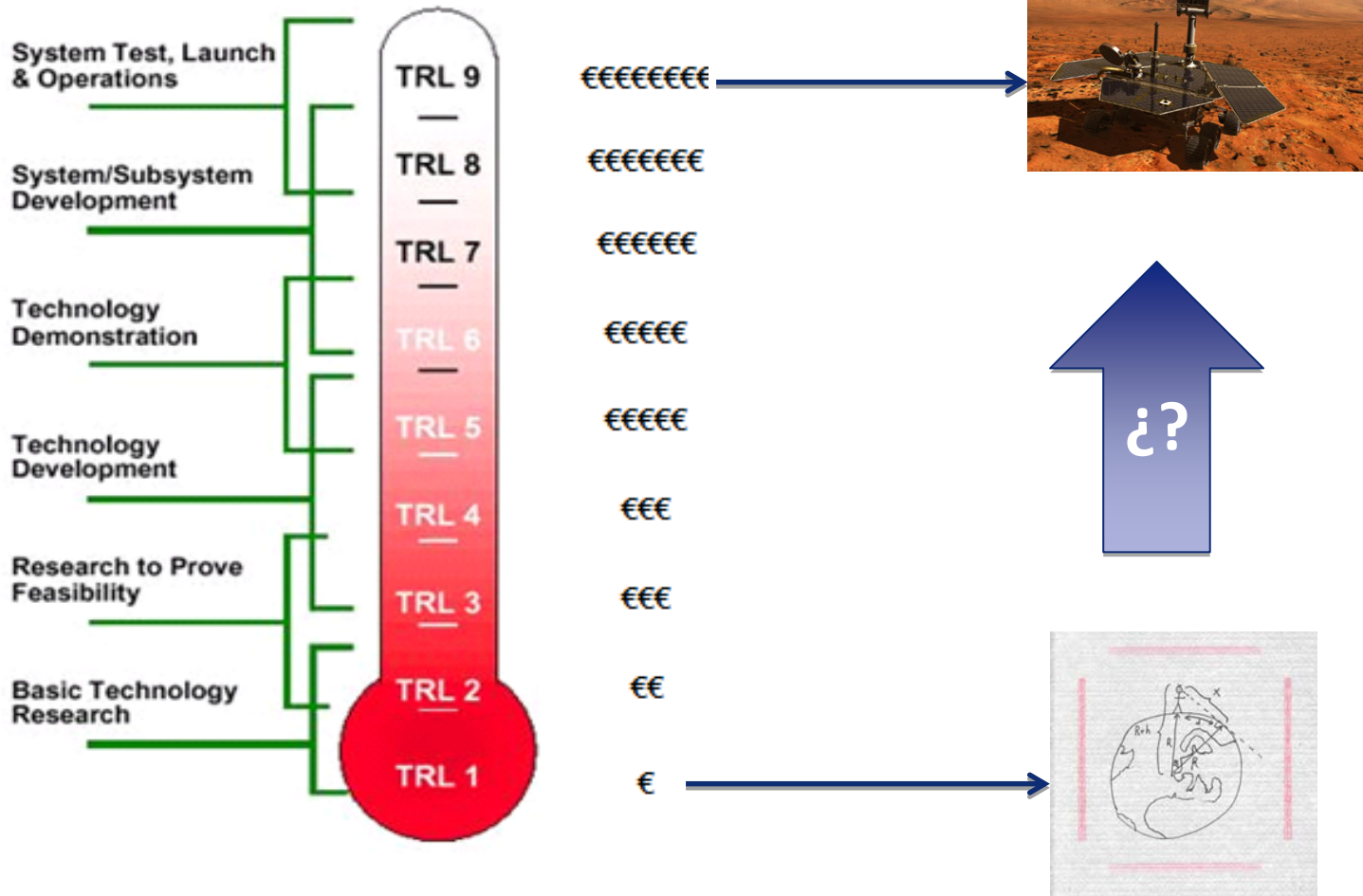
- Goals and Scenarios
- Validation and testing: from simulation to UAVs

## 2. Inertial- and Imaging-based Navigation

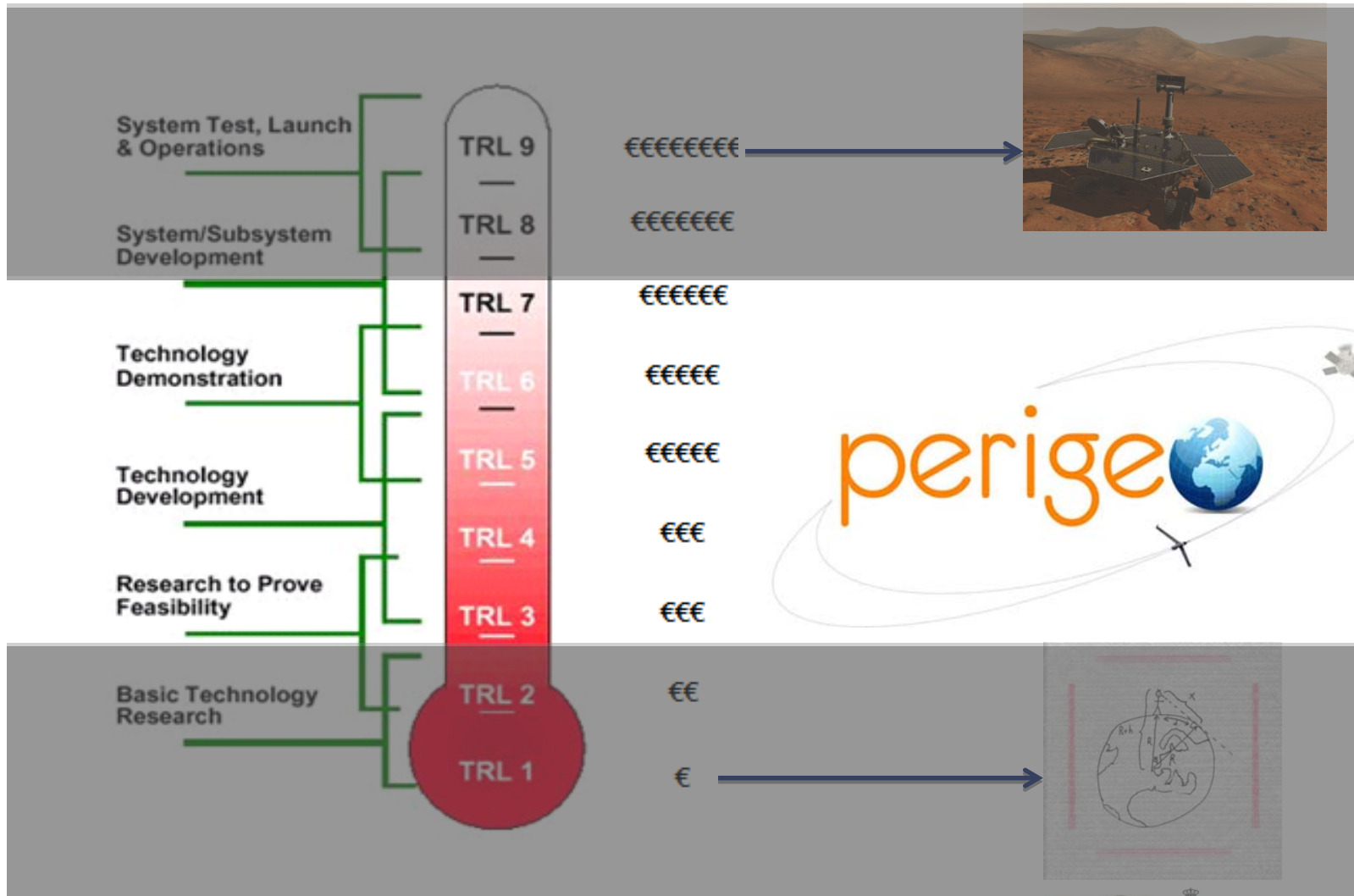
- Relative navigation based on features extraction and matching
- Absolute navigation based on craters detection and matching

## 3. Results

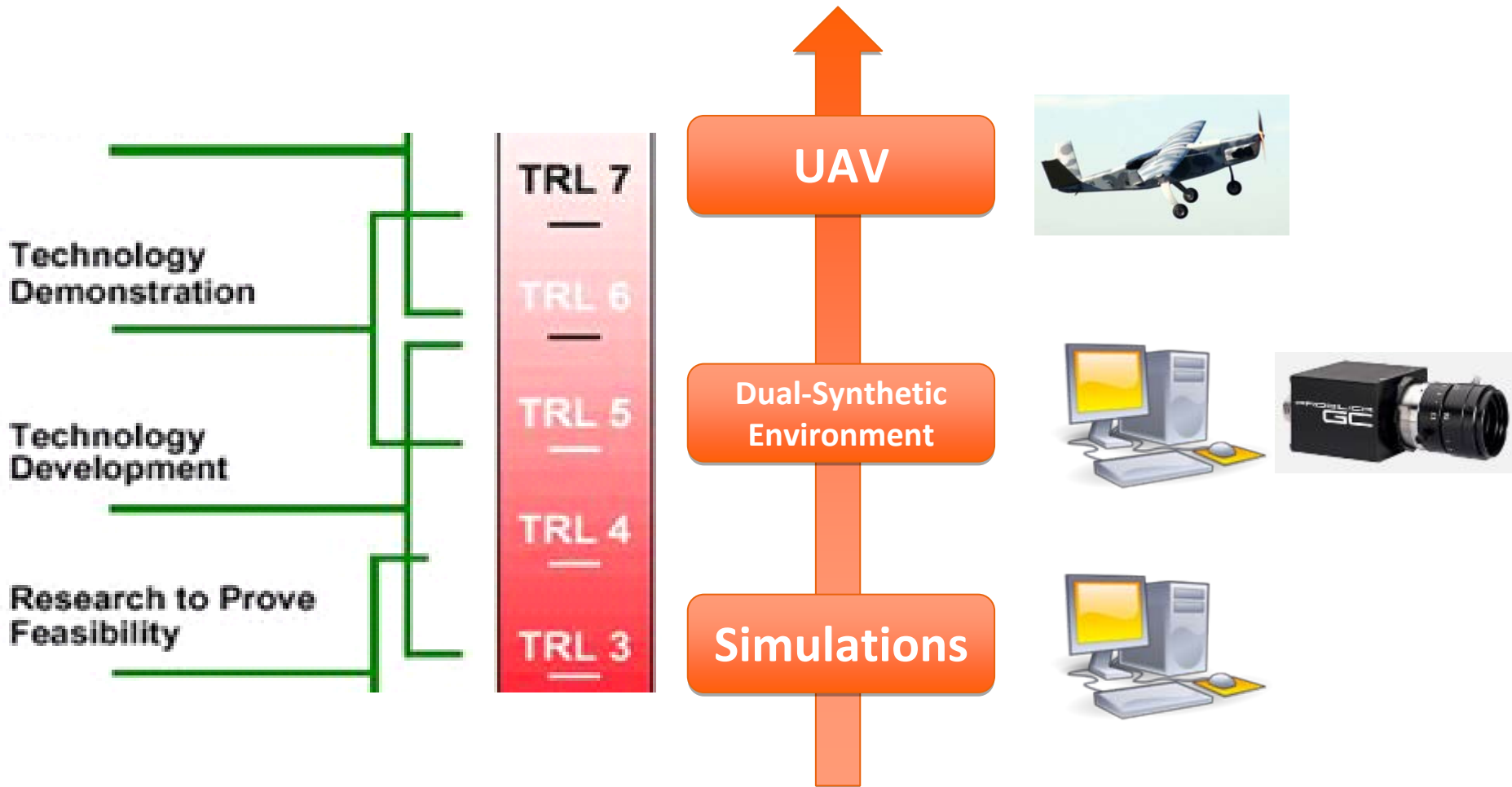
# Why PERIGEEO?



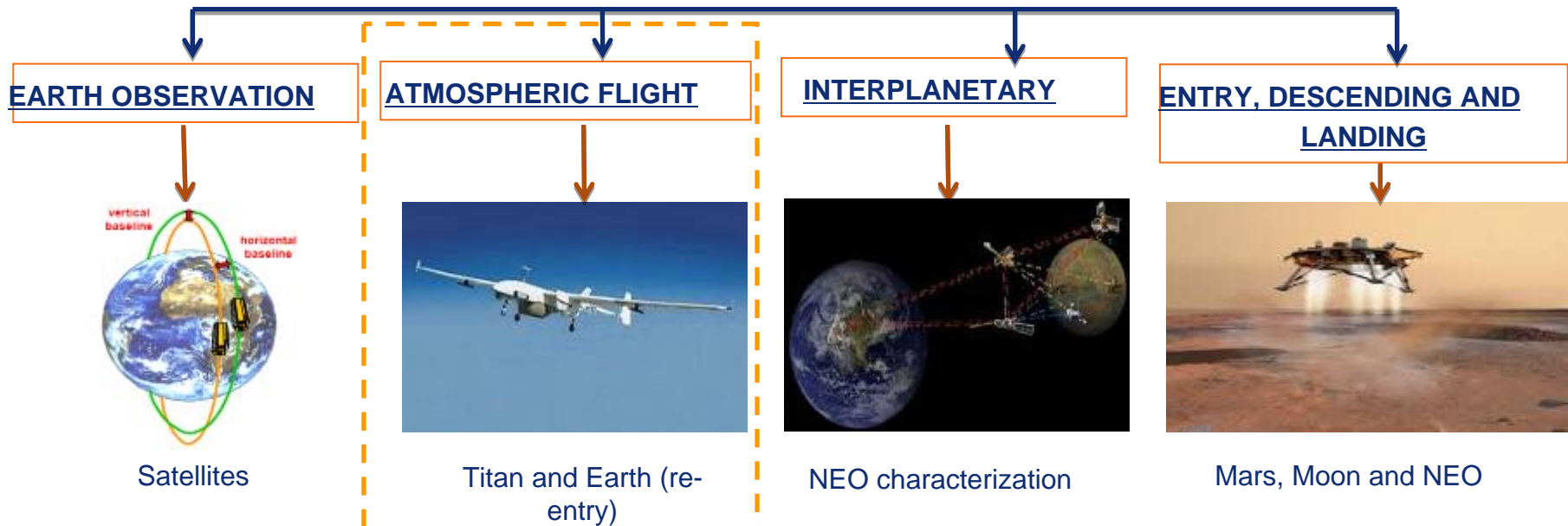
# Why PERIGEO?



# Why PERIGEEO?



## Scenarios and Missions:



## Technologies:

OPTIMAL DESIGN

ADVANCED CONTROL

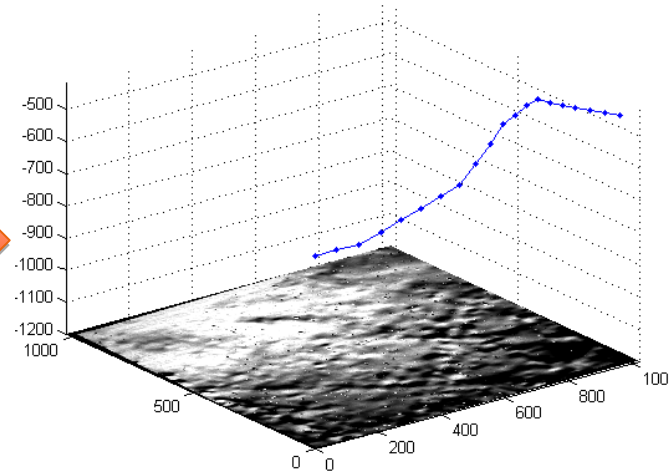
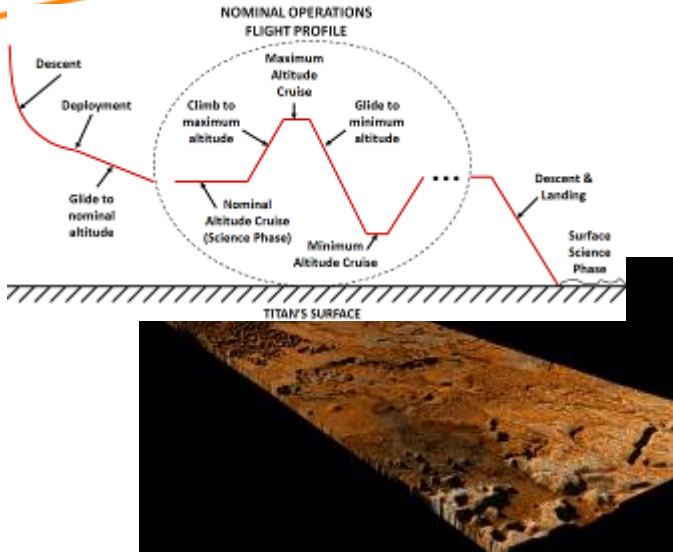
AUTONOMOUS NAVIGATION

RESEARCH AND DESIGN INTEGRATED PROCESS

# Atmospheric Flight on Titan

## SPACE MISSION

## PERIGEO

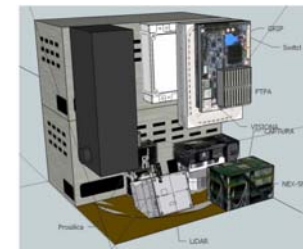
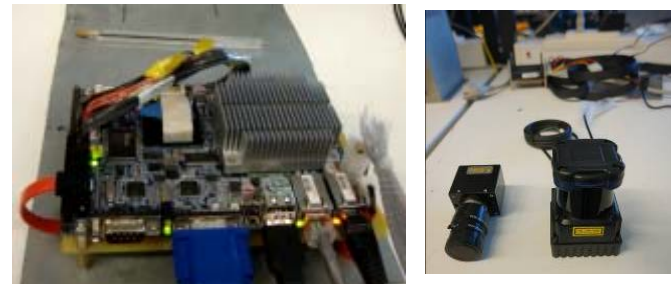
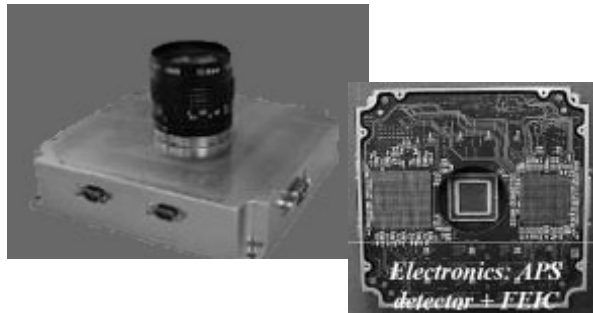


Tests are representative and scalable

Simulations

Dual-Synthetic Environment

UAV



# Key requirements

## Reqs Sensors

## Flight Platform Reqs

## Mission Reqs

Entorno Espacial

Camera: FOV, resolution, fps, pixel-size, etc.



Trayectoria: manoeuvre, payload, consumo, profile, etc.



Environment: morphological characteristics, illumination, etc.



Technology Key requirements

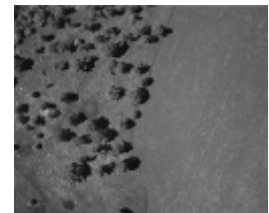
Pixel-size

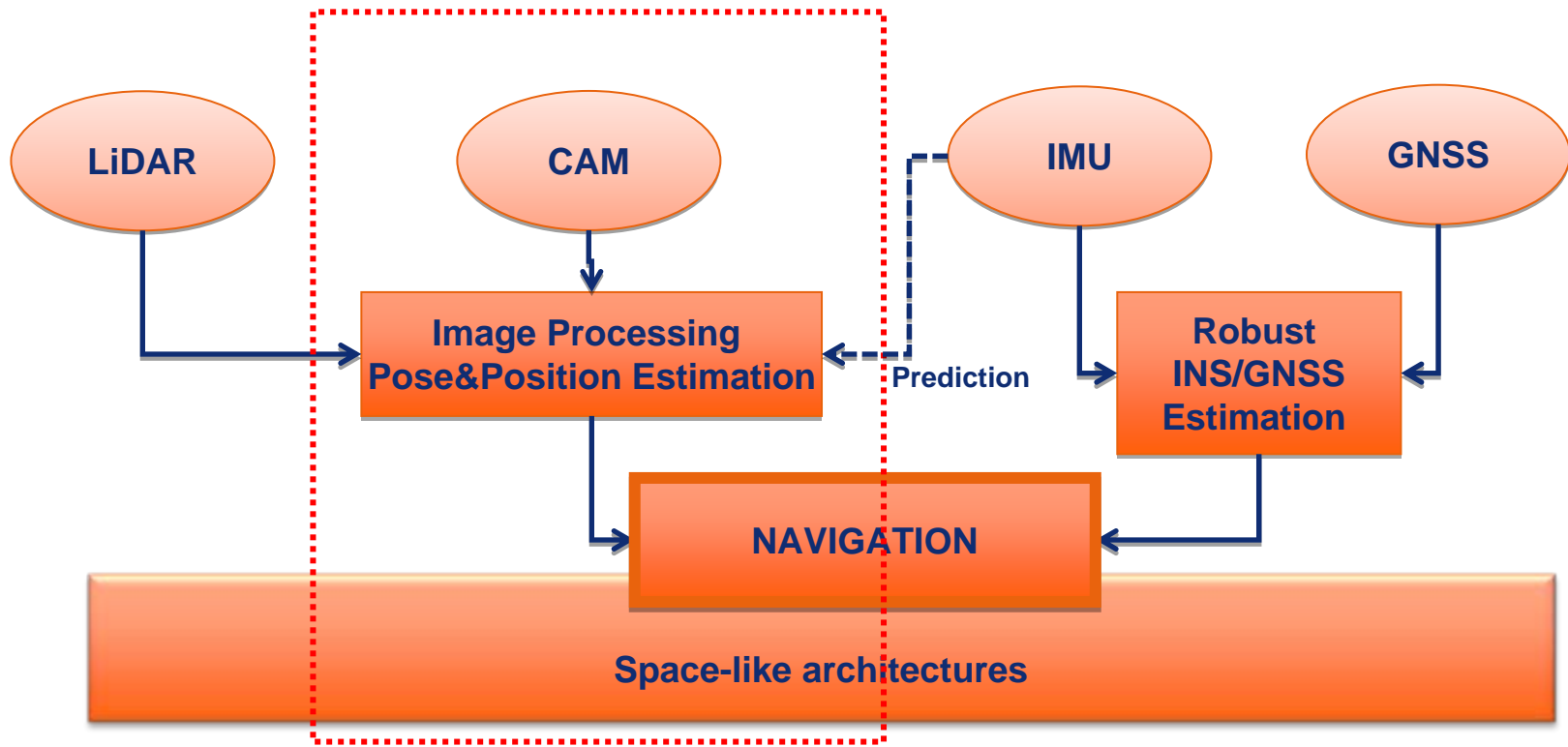
Profile

Characteristics

ESD

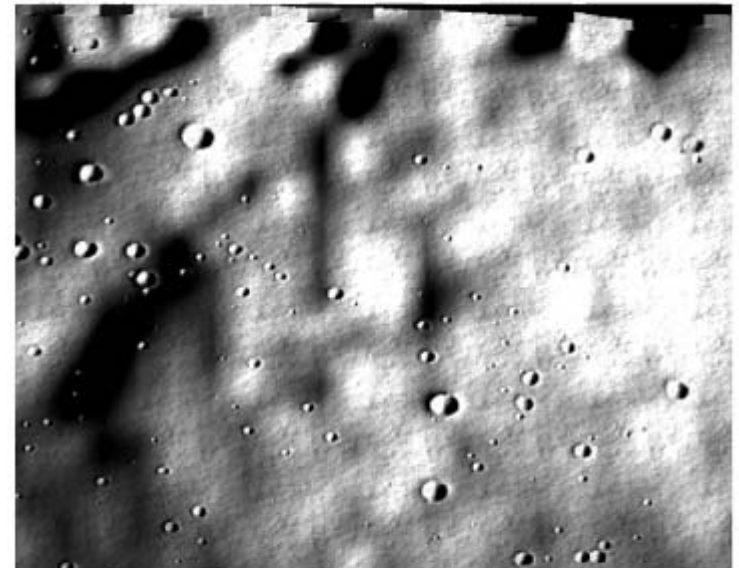
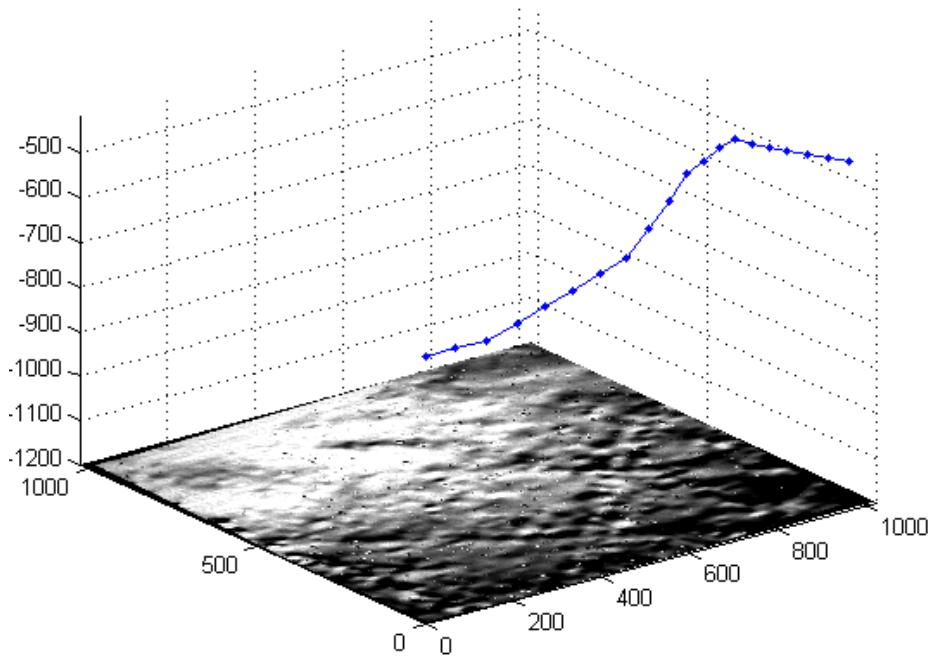
Entorno Terrestre

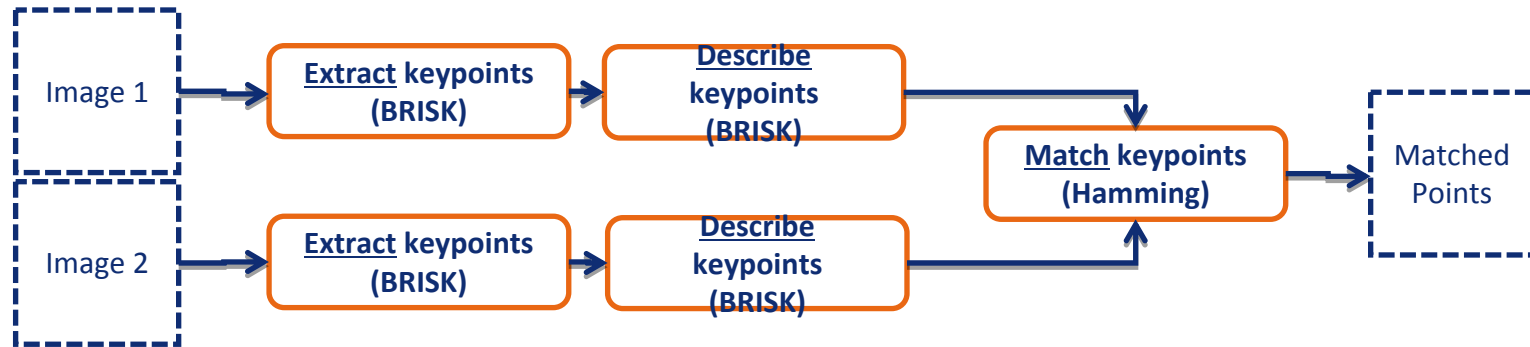




## Optical Absolute and Relative Navigation based on Camera Images

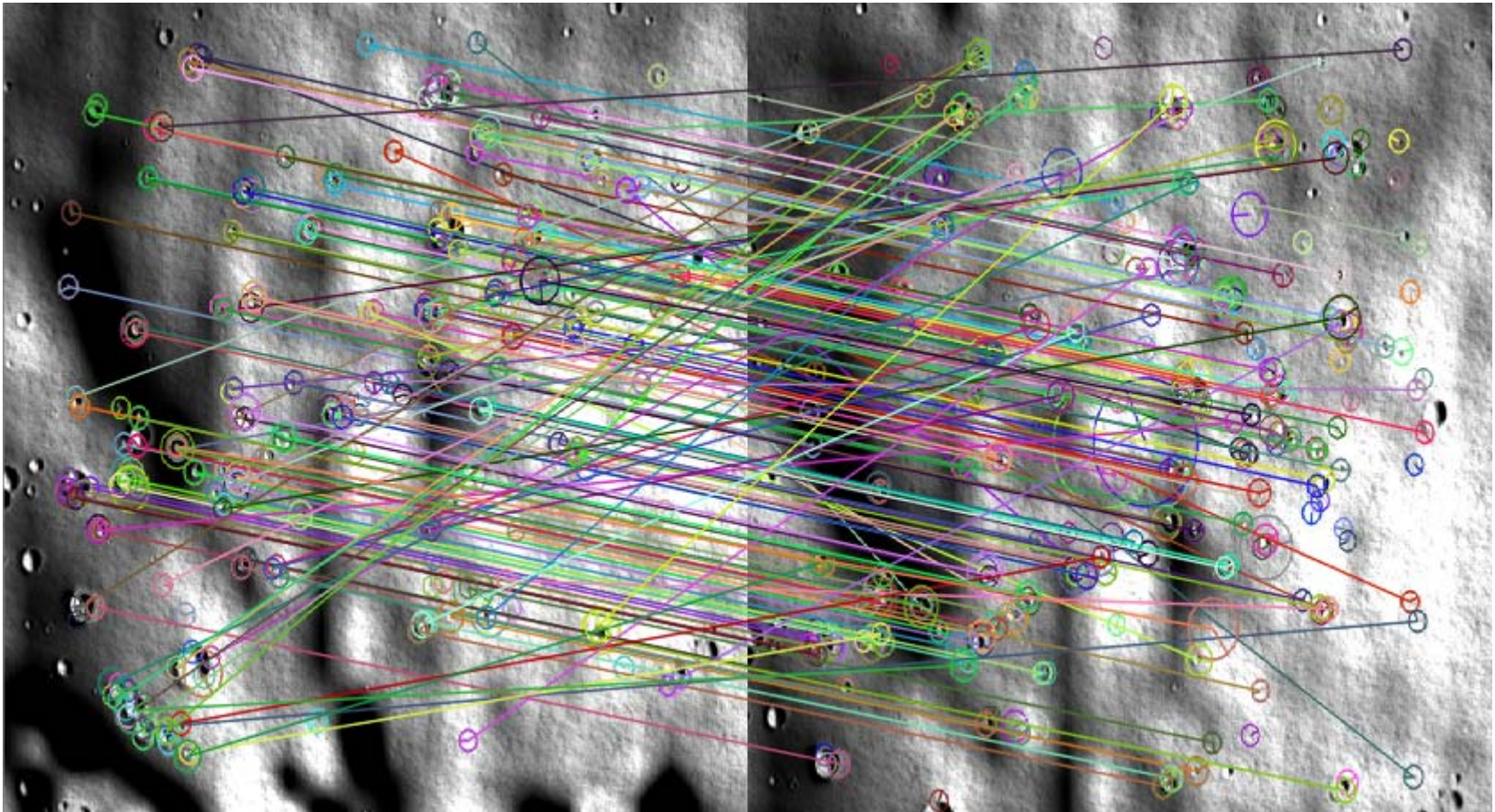
# Trajectory Example





- Point extraction and matching
  - BRISK (fast extraction; binary descriptor = fast matching)
  - 30% inliers nominally in space images
    - Robustness is mandatory!
- Parallax-based outlier removal
  - propagate orientation (naive model; IMU; ...)
  - reject pairs with large parallax
- Estimate  $t = (t_x, t_y, t_z)$  the translation vector (up to scale), and  $\Delta\Omega = (\Delta\omega, \Delta\phi, \Delta\kappa)$ , the relative attitude between two images using the ‘coplanarity’ condition in a standard least-squares estimation approach.

# Mixed inliers/outliers

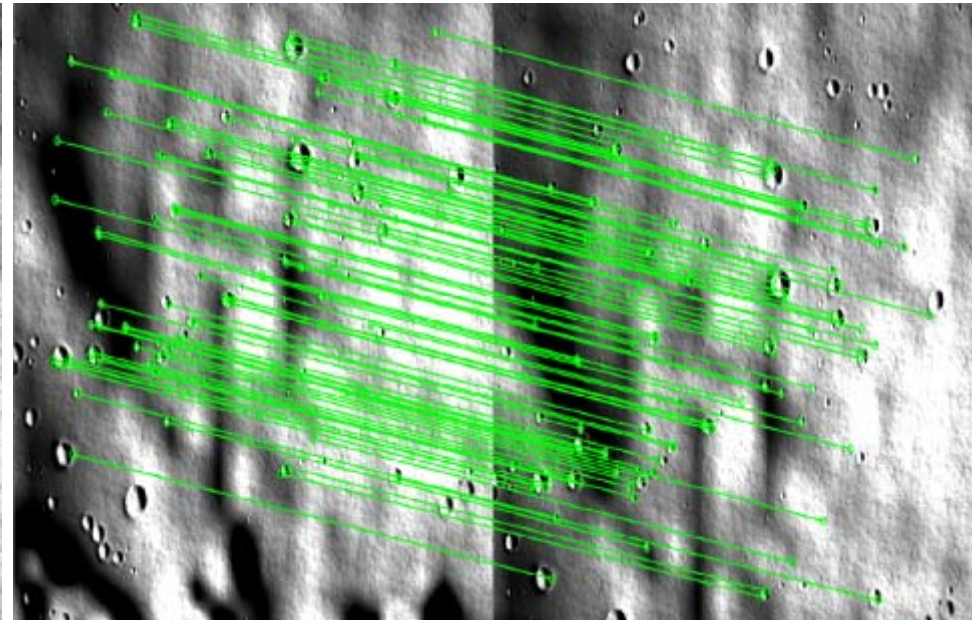
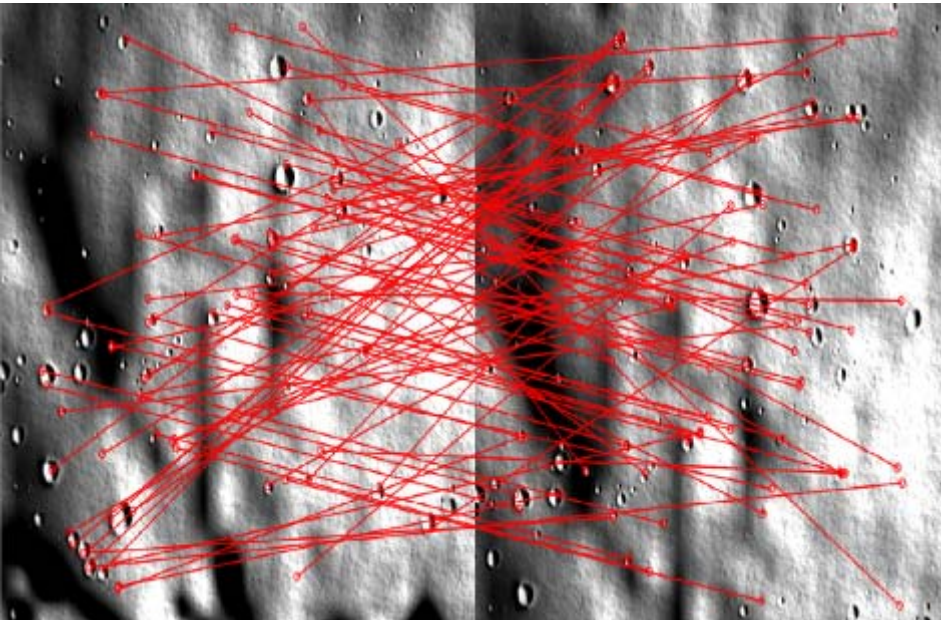


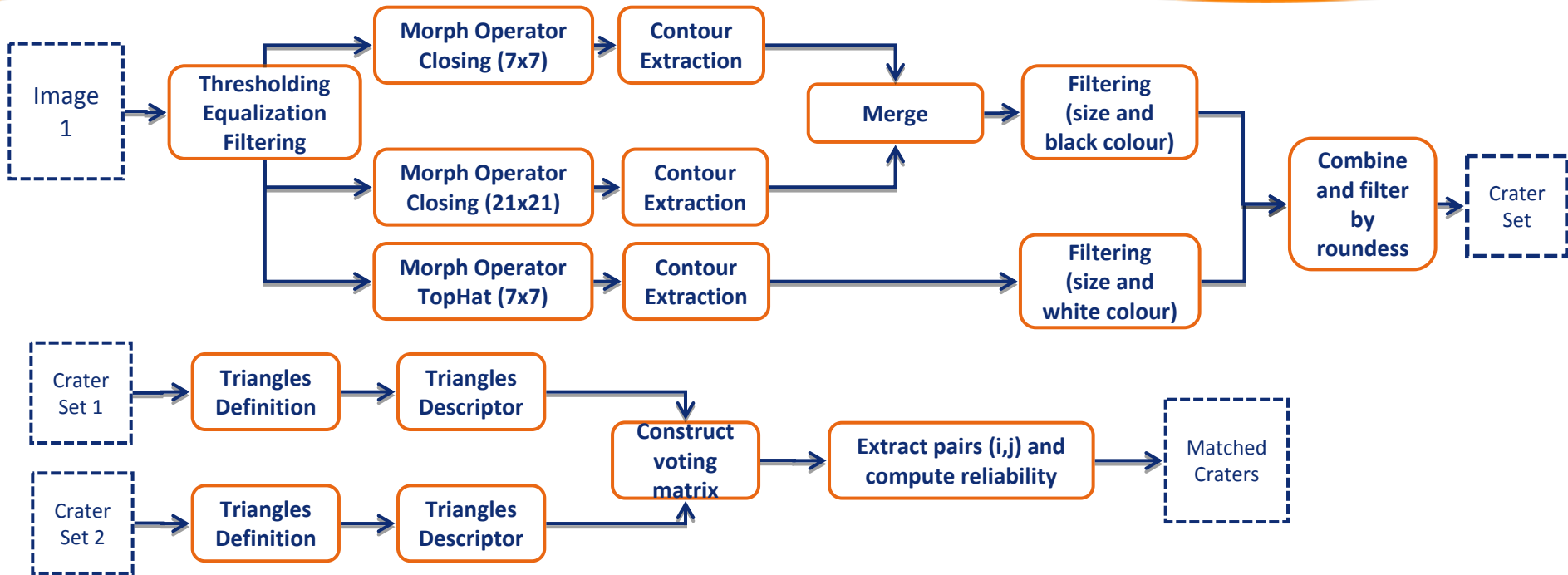
**BRISK operator: 144 'correct' matchings over 386 (37%) [correct = parallax < 0.5 m]**

# Inlier/outlier discrimination

Using orientation prediction and parallax computation, we distinguish **inliers** and **outliers**.

Still some wrong matches escape → projecting rays intersect ‘between camera and terrain’



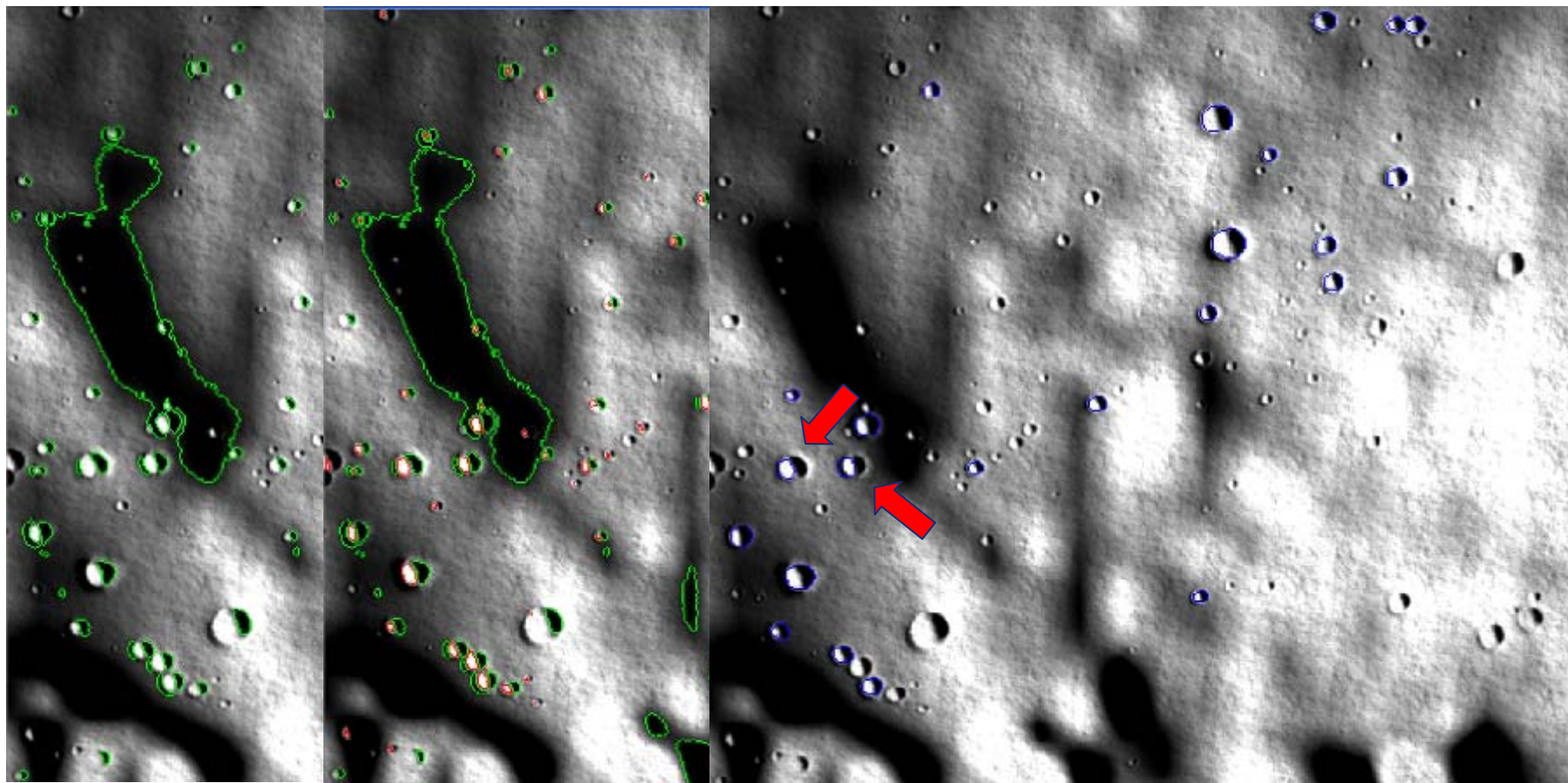


- Crater extraction:
  - Image processing (morphological operators, thresholding)
    - contour extraction
- Crater description and matching:
  - normalized distances and areas of crater groups
  - matching to geo-referenced imagery
    - 2D-3D crater centroid correspondences

Estimate  $(p, \Omega)$  the camera exterior orientation using the 'colinearity' condition.

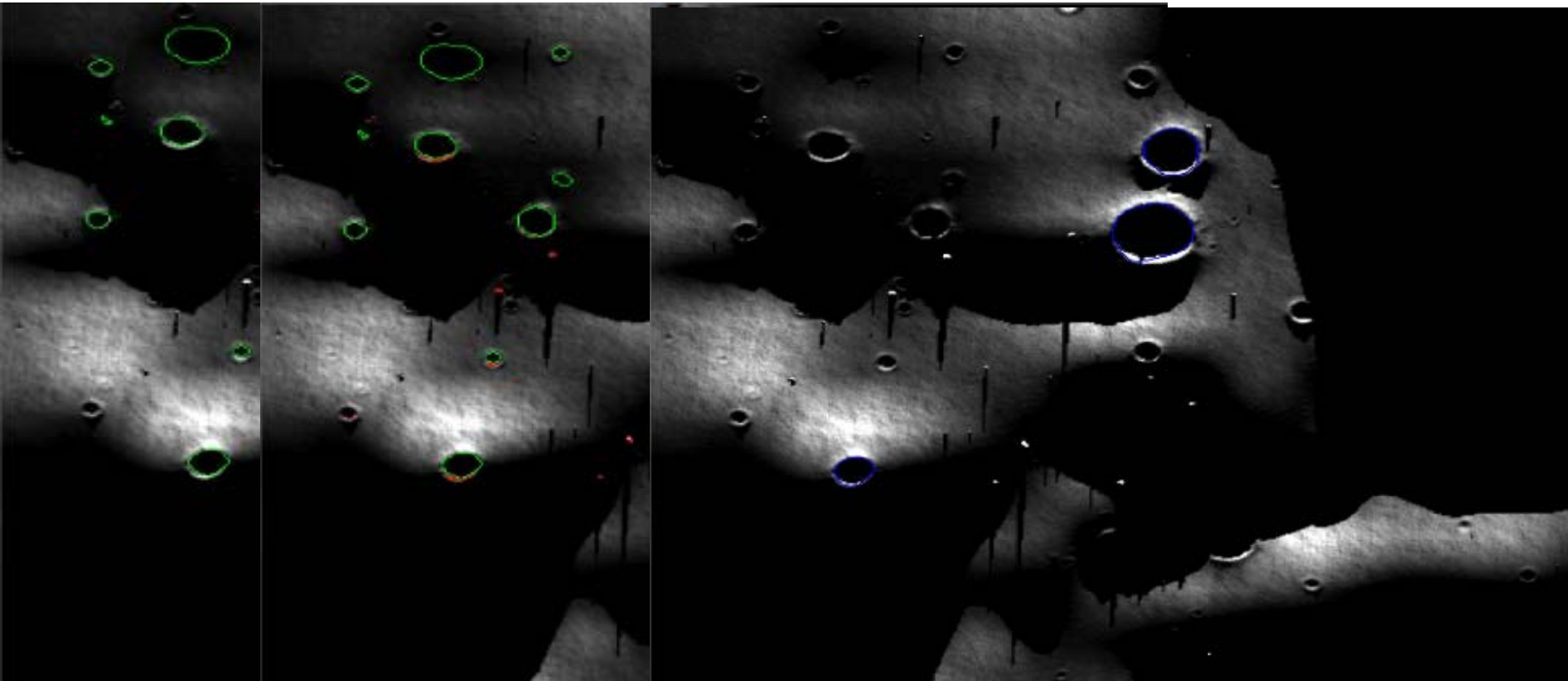
# Illumination-based incremental contour extraction (45° Sun elevation)

- Results:**
- 23 detections over more than 50 craters
  - 2 miss-associations (black/white areas switched)
  - towards illumination independency (next slide)



# Illumination-based incremental contour extraction ( $< 1^\circ$ Sun elevation)

- Results:**
- 3 detections over around 15 craters
  - Illuminated areas (white) areas are poor  
→ algorithm tuning is key



# Preliminary results with Simulated Images

## Synthetic trajectory

- forward motion towards North, 'descend-ascend', roll & heading turns (Titan's case)
- 45° Sun elevation
- PANGU simulator (simulation of camera and digital elevation models)
  - see .fig and pics

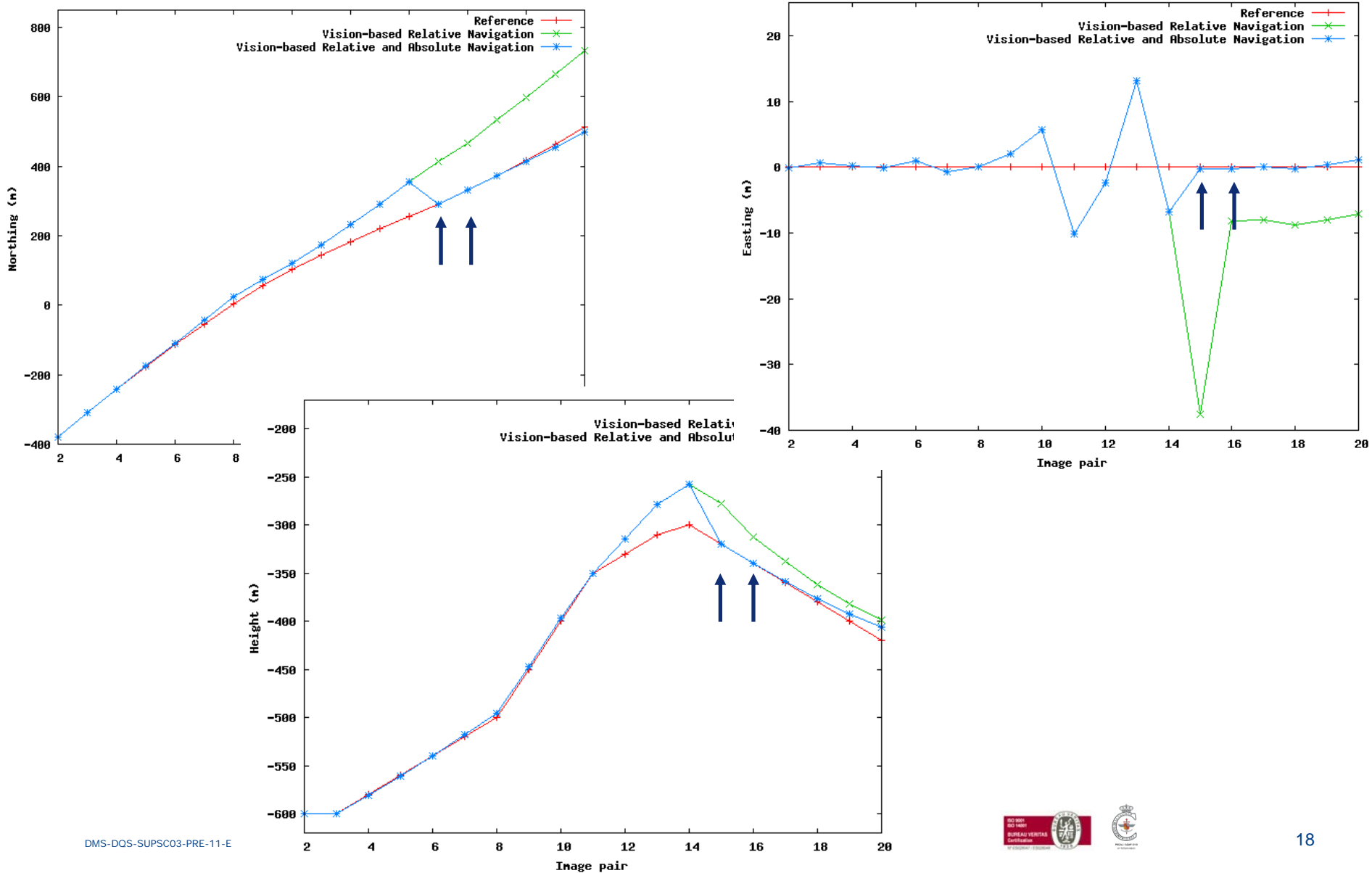
## For the 19 image pairs,

- Points extracted and matched using BRISK
- All outliers removed (equivalent to using a perfect propagation model, e.g. perfect IMU)

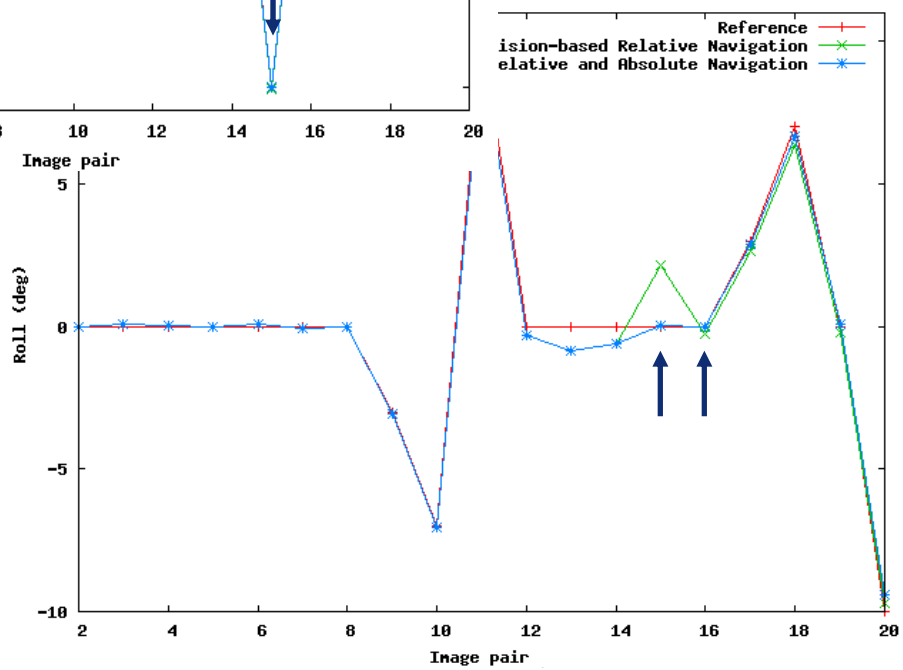
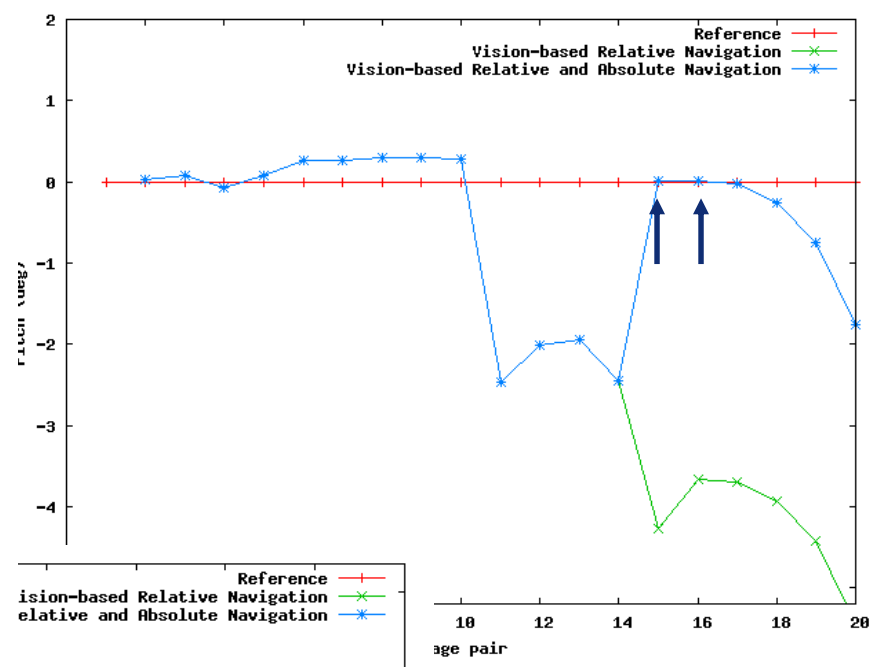
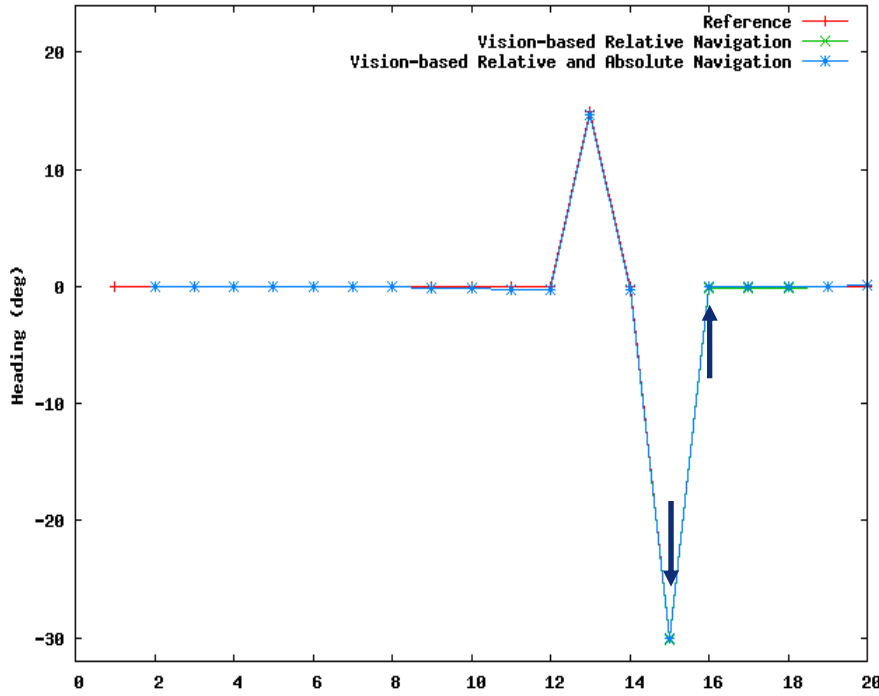
## → We compare camera-only {relative} vs {relative/absolute} navigation

- In images 15 and 16, absolute navigation was performed by extracting 5 craters and matching those to geo-referenced imagery.

# VISION-BASED RELATIVE NAVIGATION: POSITION ESTIMATION RESULTS

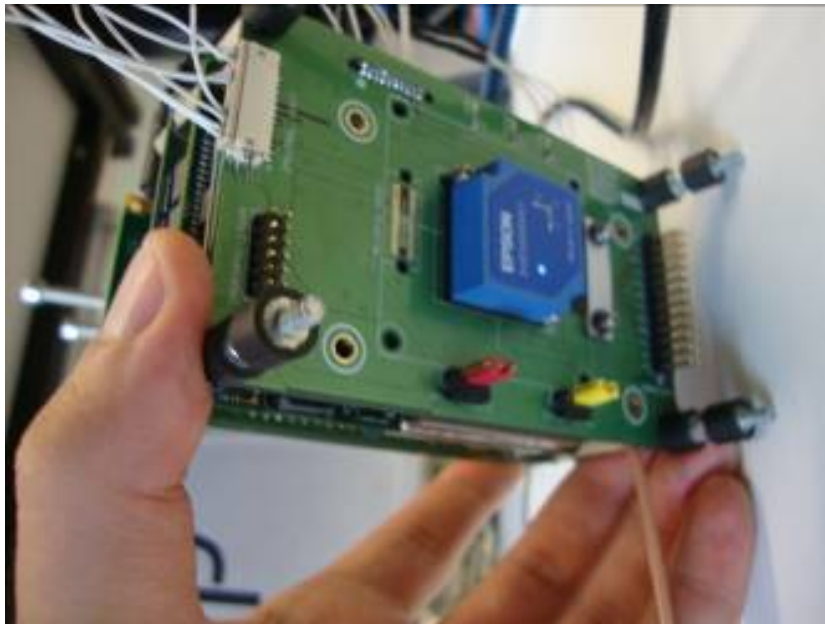


# VISION-BASED RELATIVE NAVIGATION: ATTITUDE ESTIMATION RESULTS



# REPRESENTATIVE ENVIRONMENT: FLYING ON THE UAV

Subsystem or sensors	Functionalities
GRIP	In-house multi-frequency GNSS receiver, featuring Galileo's AltBOC signal reception
VISIONA	Camera- and LiDAR-based navigation system, and input provider for HDA tasks
CAPTURA	Miniaturized INS/GNSS acquisition system and time-reference server
Fault-tolerant Processing Architecture	LEON3-based processing architecture including INS/GNSS/image processing software
Prosilica GC2450C	Camera for online image processing tasks
Hokuyo UTM-30LX	LiDAR for online digital elevation model generation
Sony NEX-5N	Camera for high-resolution observation
EPSON M-G350-PD11	Low-cost miniaturized IMU
Javad TR-G3T	Geodetic-grade, multi-frequency GNSS receiver

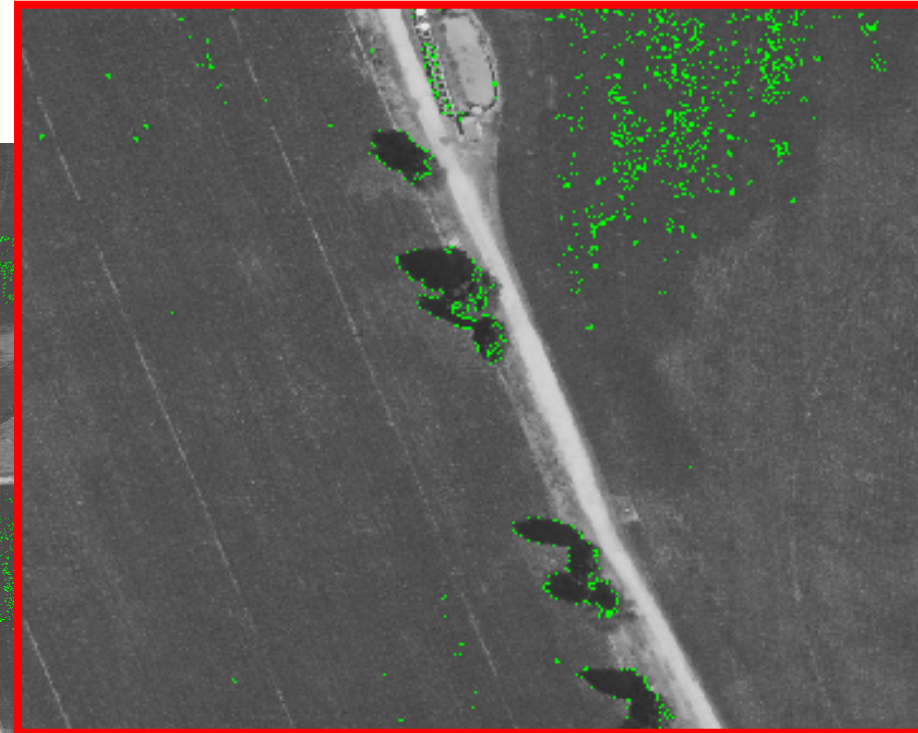
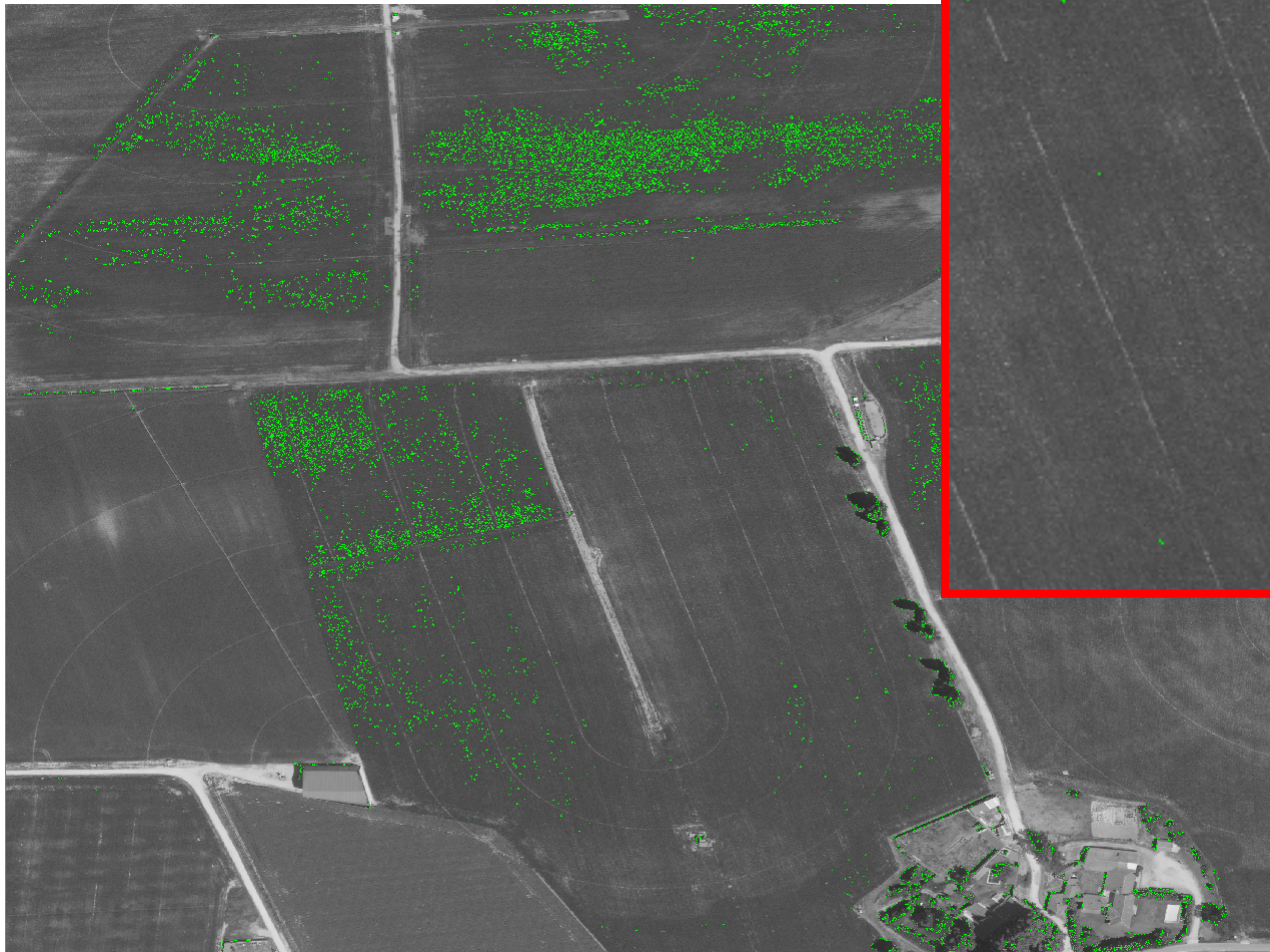


# PREPARING THE REPRESENTATIVE ENVIRONMENT

Adaptation of the algorithms is needed (mainly, for crater detection)

- Trees as craters ... ?

- Adaptation of environment is requirement for testing campaign



# ELEMENTS IN THE REPRESENTATIVE ENVIRONMENT

