
Absolute map-based localization for a planetary rover

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Work developed within
the ESA founded Startiger activity “Seeker”

Outline

- On the importance of localization
- Absolute map-based localization
- Outlook

On the importance of localization

“Reach that goal”, “Map this area”...

→ Missions are defined in terms of localization

Environment models are required

→ Spatial consistency ensured by localization

Safe execution of the planned trajectories

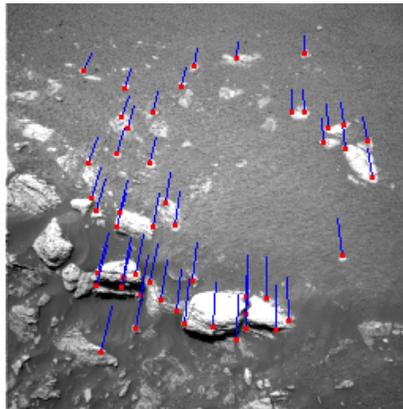
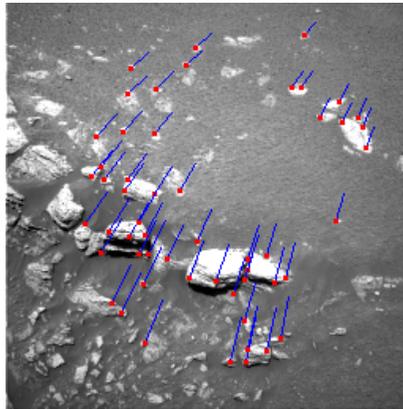
→ Robust locomotion control and diagnosis ensured by localization

“If you are not localized, you are lost !”

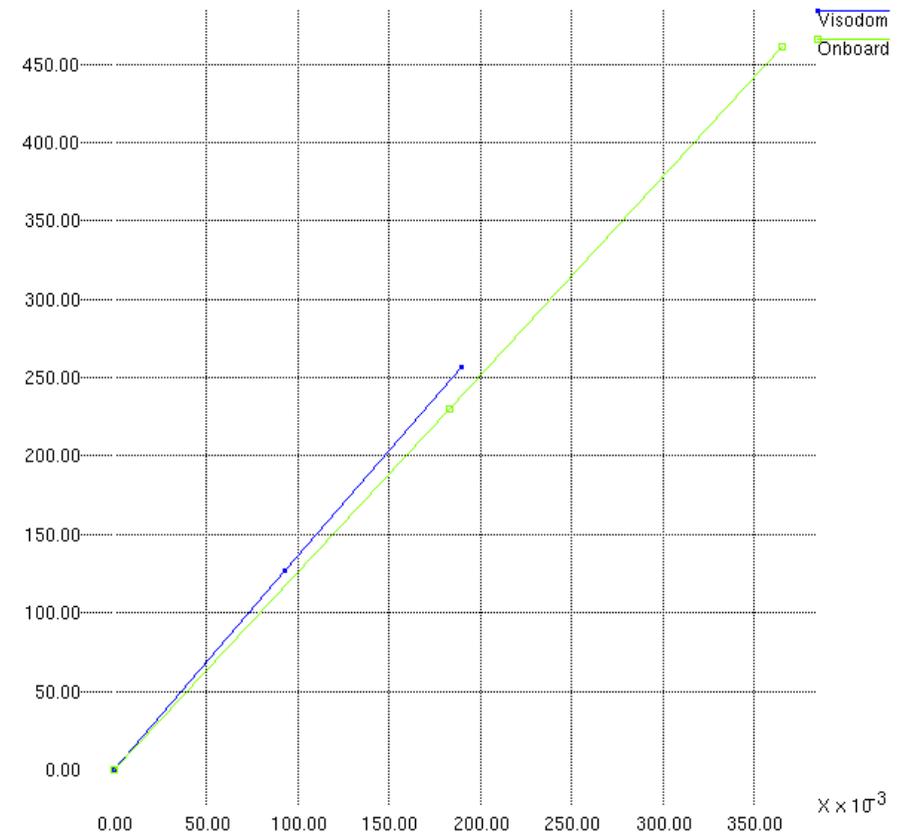
Trajectory following

Trajectories are defined by geometric primitives : a precise localization estimate is required

50% slip
→ visual
odometry
(Spirit in
2004)



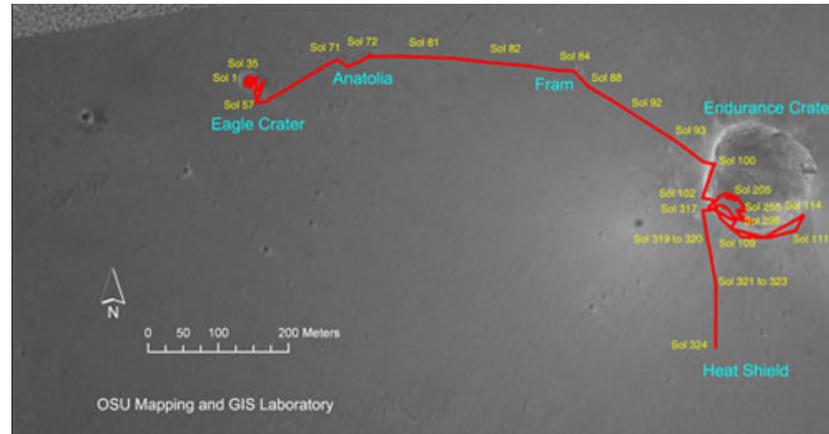
X Graph



Required localization : *cm* accuracy, local frame, frequency function of speed

Locomotion diagnosis

Opportunity traverse



April 26th, 2005



Rear left wheel

Required localization : *cm* accuracy, local frame, high frequency

Locomotion diagnosis

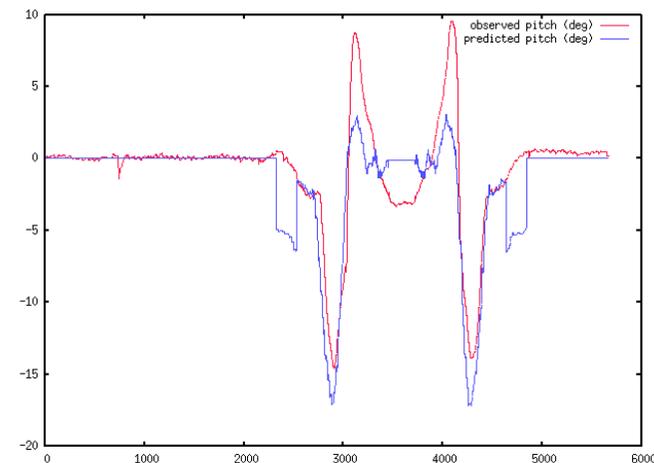
Monitoring the terrain profile



predicted

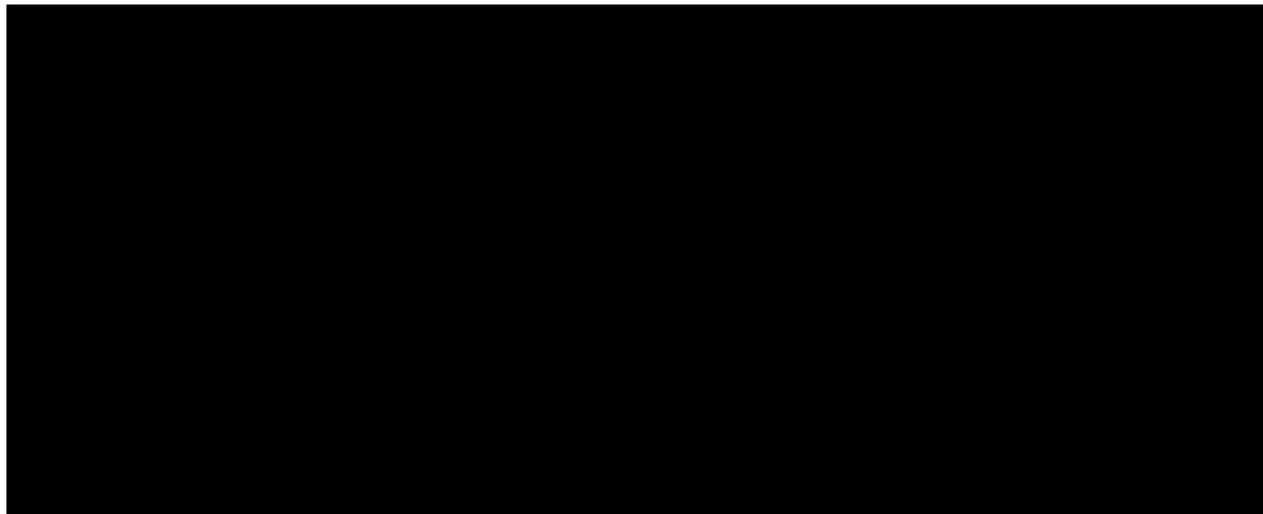
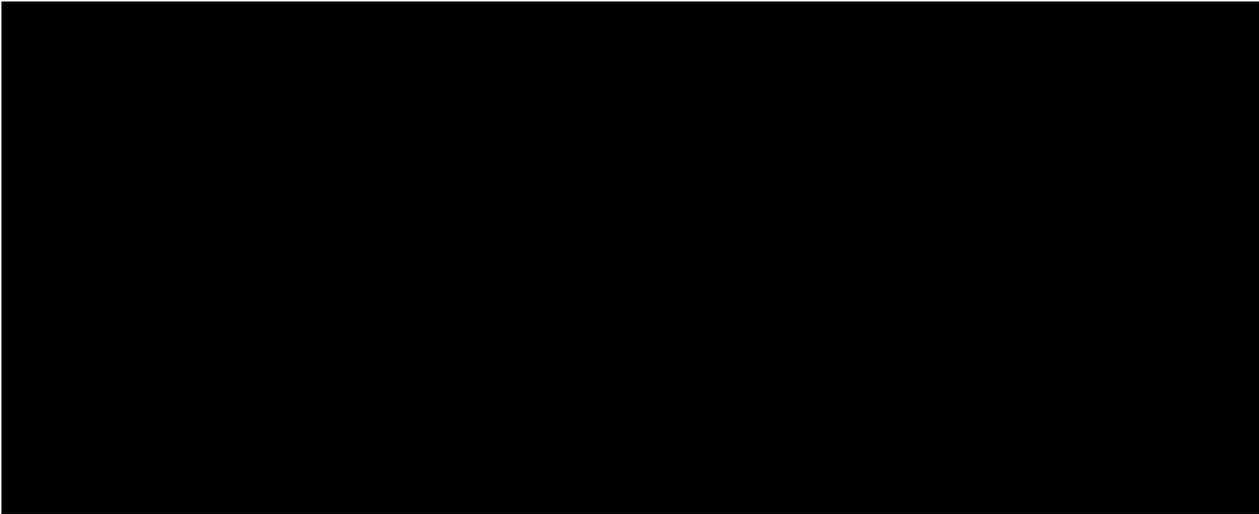
measured

Predicted vs. observed
robot pitch angle



Required localization : *cm* accuracy, local frame, high frequency

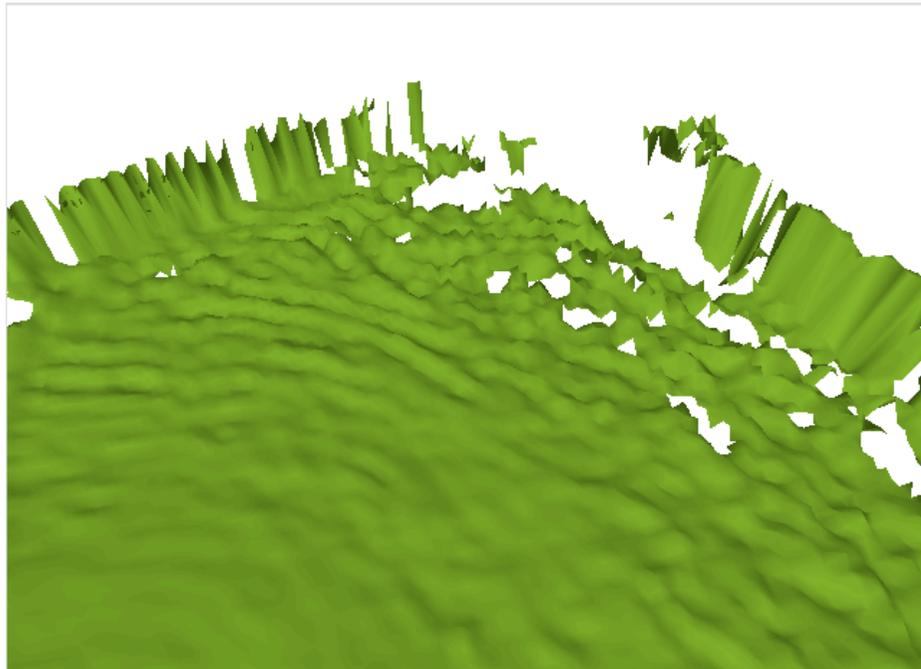
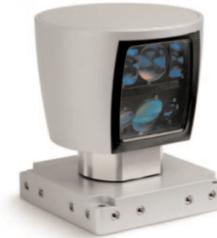
(From visual odometry to INS/visual SLAM)



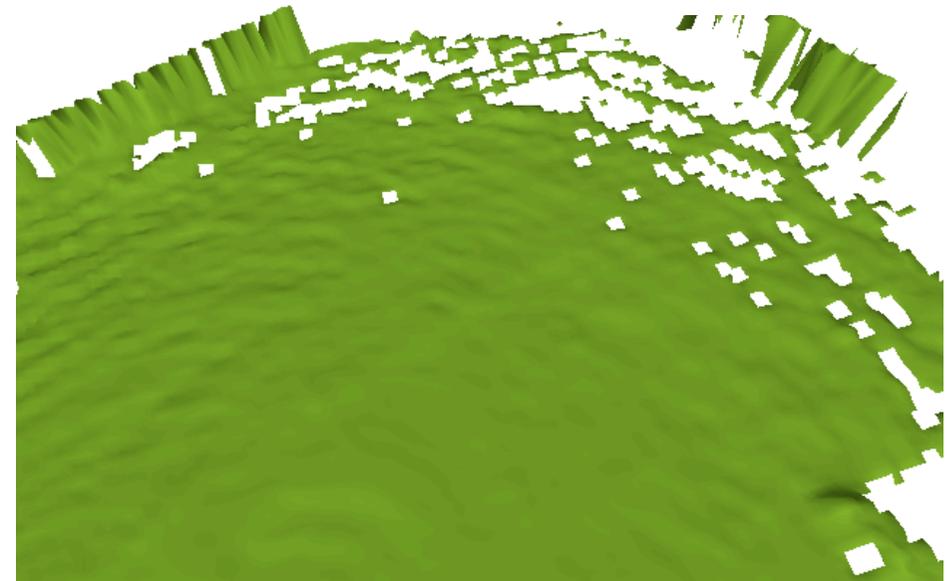
State of the art (<http://rtslam.openrobots.org>) : 12 state parameters @ 200 Hz, errors of the order of 1% over hundreds of meters

Environment models spatial consistency

- Velodyne lidar provides chunks of 64 points @ 3.5 kHz:
1° error on pitch yields a 17cm elevation error @ 10m



2m/s, GPS RTK @ 20Hz
+ Xsens AHRS @ 50Hz
+ FOG gyro @ 50Hz

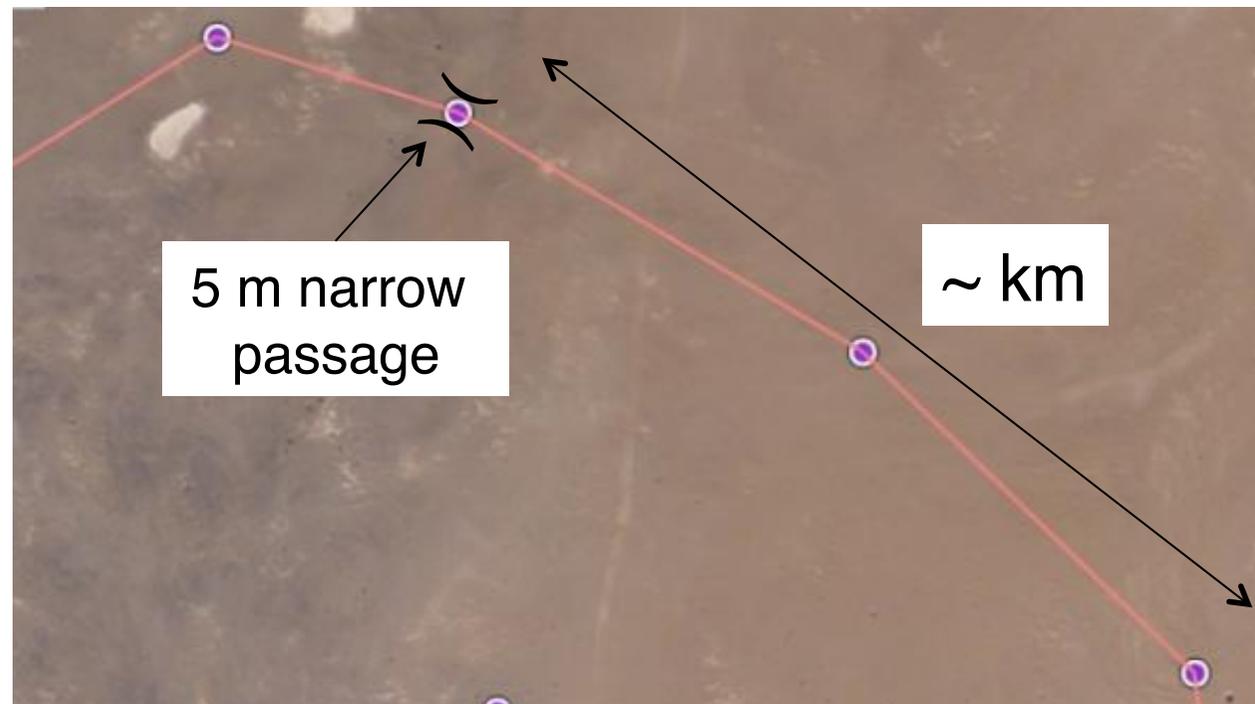


2m/s, RT-SLAM @ 100Hz

Required localization : *cm* accuracy, local (or global) frame, frequency defined by the range data acquisition rate (often high)

Ensuring long range missions

- Odometry, VO, INS, SLAM (without loop closures) localisation solutions eventually drift
 - ⇒ No way to ensure the achievement of global mission plans



⇒ Need to rely on “absolutely localized” features : geo-referenced maps

Required localization : m accuracy, global *absolute* frame, very low frequency

Which localization?

Essential questions to answer to define a localization process :

1. With what precision ? From *cm* to *meters*
2. In which frame ? Absolute vs. local
3. At what frequency? From *kHz* to “sometimes”

cm accuracy,
10 to 100 *Hz*,
local frame

- Ensure the lowest level (locomotion) controls
- Ensure the proper execution of paths / trajectories
- Ensure the spatial consistency of the built models

~*m* accuracy,
“sometimes”,
global frame

- Ensure the achievement of the missions, most often defined in localization terms (“goto (*goal*)”, “explore / monitor (*area*)”, ...)

- On the importance of localization
- **Absolute map-based localization**
 - Related work
 - Approach
 - Results
- Outlook

The essence of the problem

- Given
 - A map
(*various maps*)



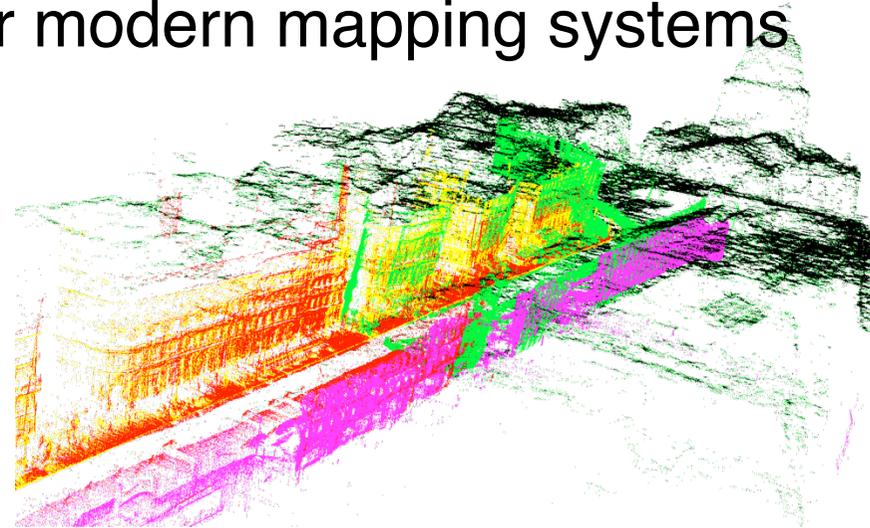
- Local data



⇒ Where am I ?

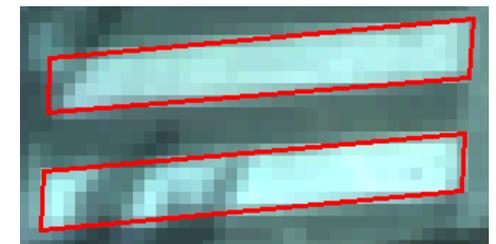
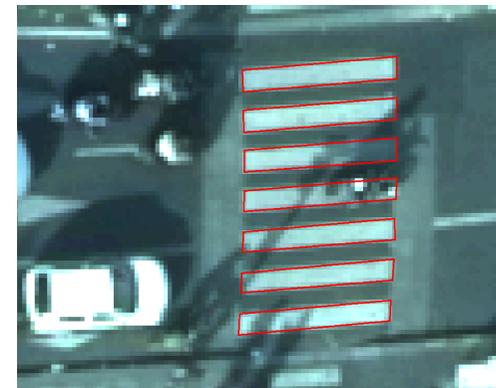
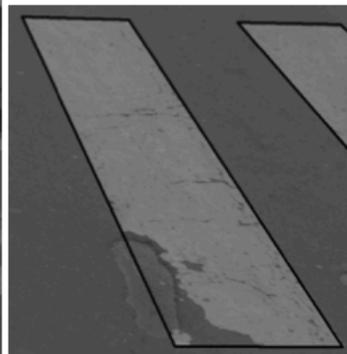
(Matching aerial and terrestrial data on earth)

- A very relevant problem for modern mapping systems



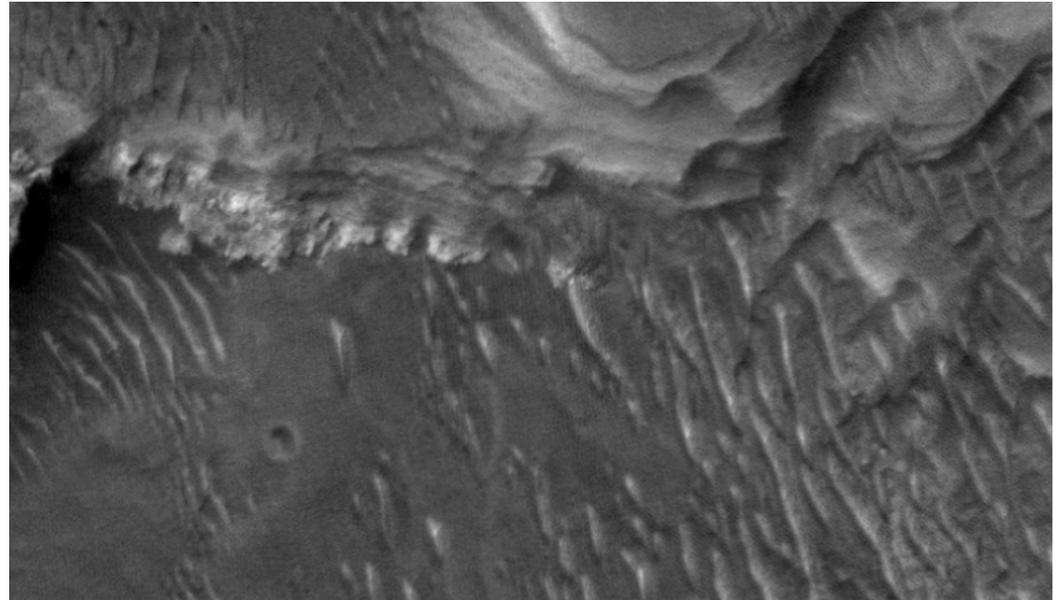
- 3D points registration

- Using landmarks

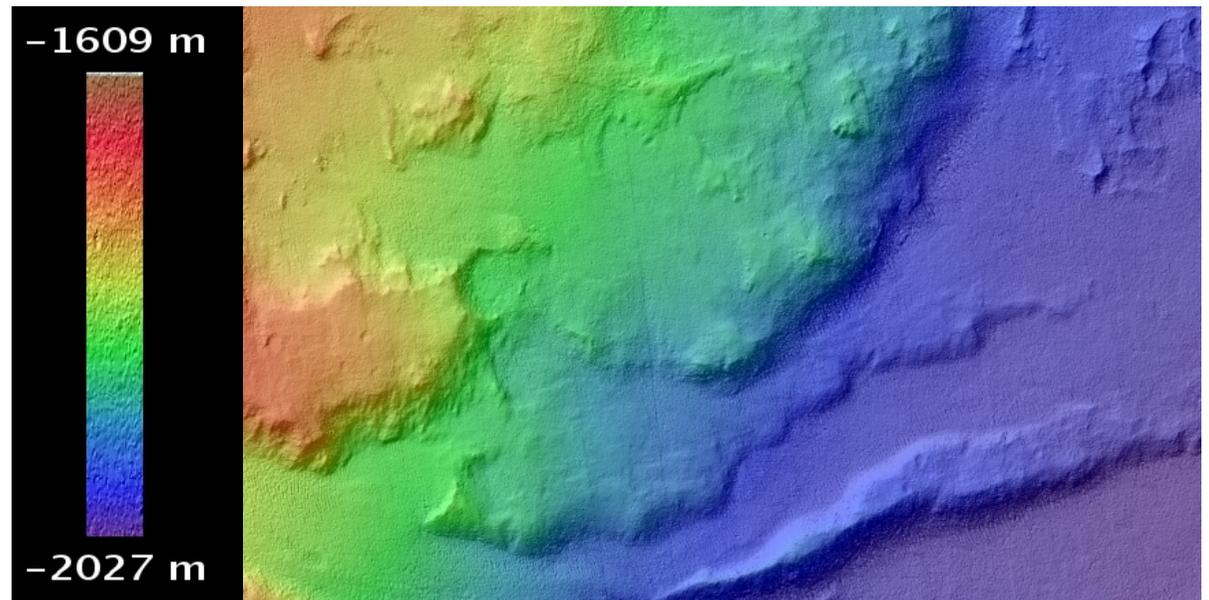


Available maps on Mars

HIRISE Mars orbiter images
0.25m/pixel (here 0.5/pixel)



DTM of Holden Crater
(1m/pixel)

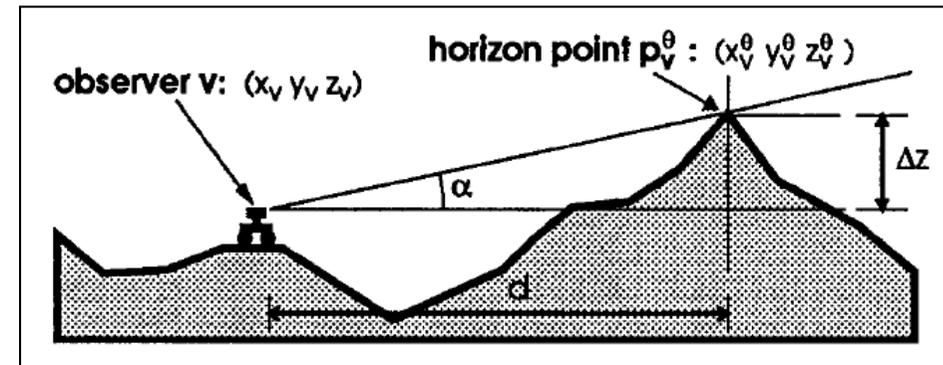
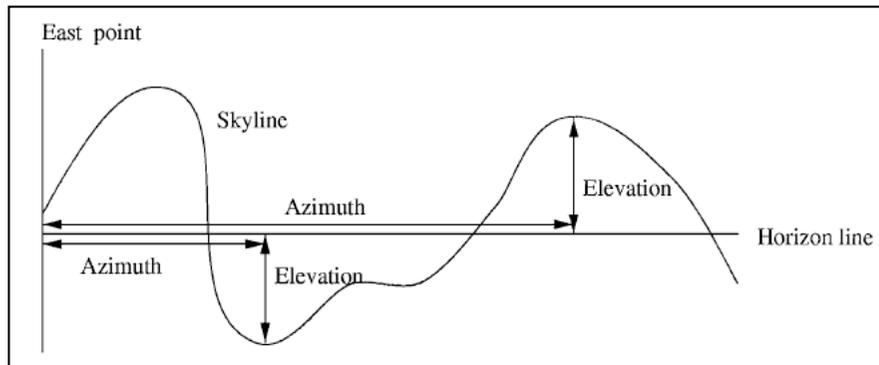


+ descent imagery
(cf MSL)

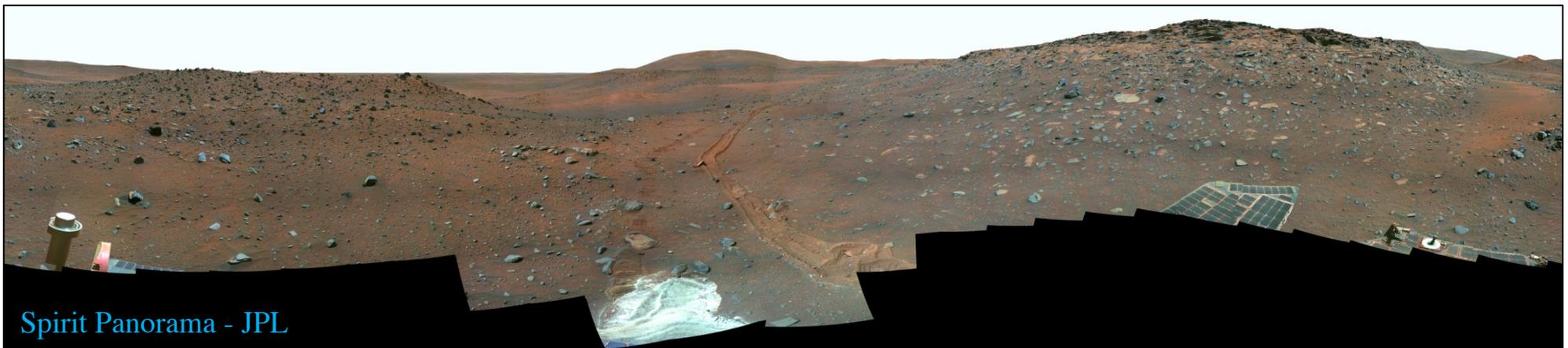
Map-based localization solutions (1)

- Skyline matching [Cozman 2000]

1. Extract Skyline Signature for each orbiter DTM cell



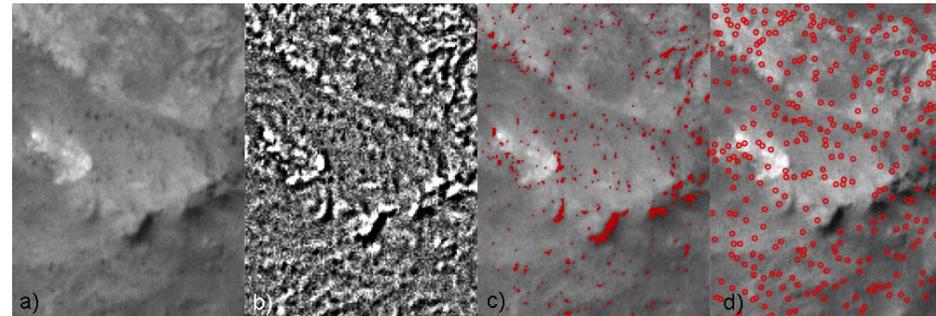
2. Compare robot perceived skyline with DTM skylines



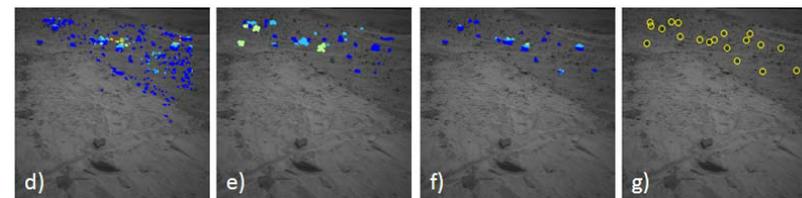
Map-based localization solutions (2)

- Surface feature matching [Hwangbo 2009]

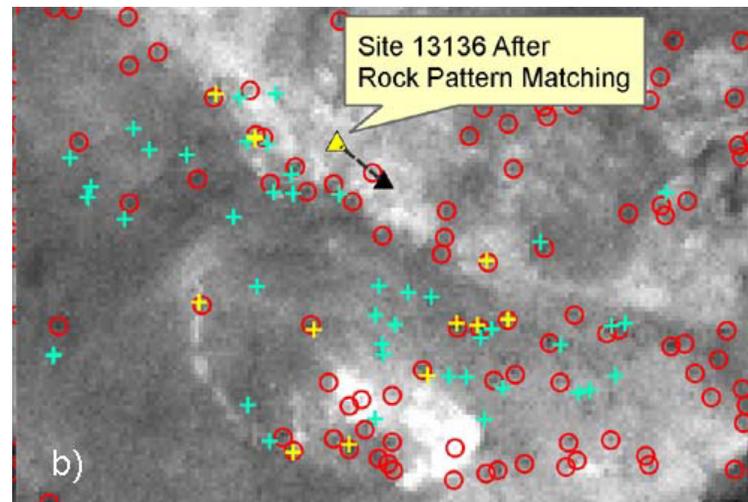
1. Extract rocks from orbital data (e.g. using shadows)



2. Extract rocks from rover stereo data

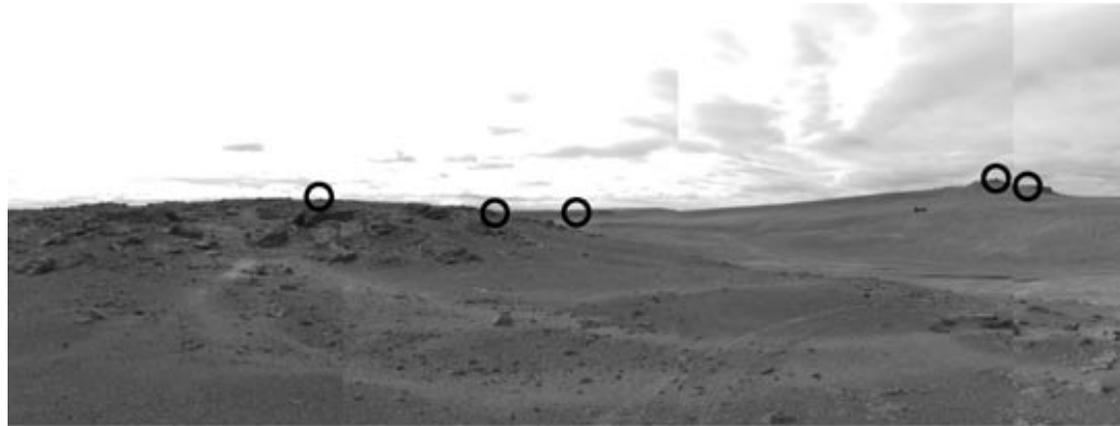
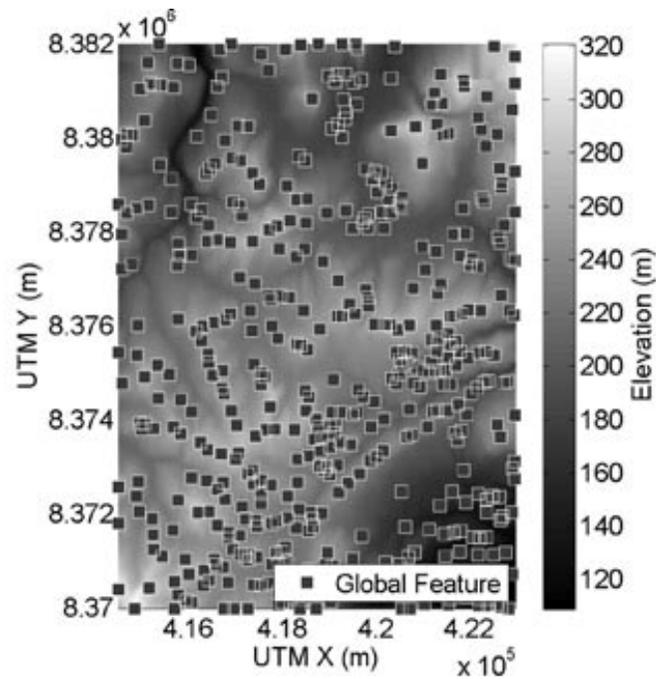


3. Match rock patterns between rover and orbital data



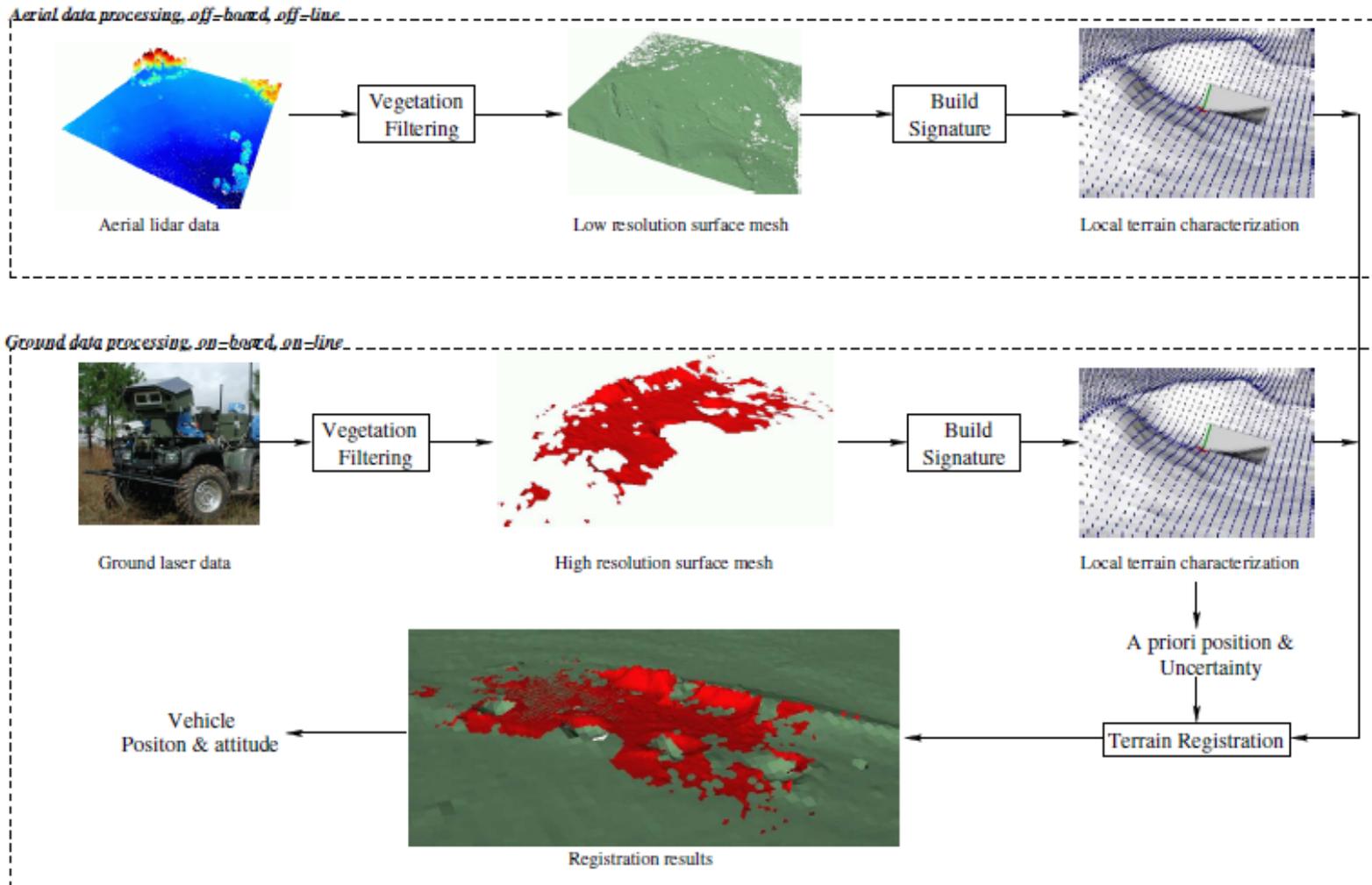
Map-based localization solutions (3)

- DTM peaks matching [Carle 2010]



Map-based localization solutions (4)

- DTM feature matching [Vandapel 2006]



Map-based localization solutions: summary

- Two processes involved:
 1. Matching rover and orbiter data (“perception” process)
 2. Assessing the position (“estimation” process)

- Skyline matching
- Surface feature matching
- DTM peaks matching
- DTM feature matching



All the matching processes involve additional computations

⇒ Proposal: exploit for the matching process the rover DTM build for path planning

Proposed solution: overview

- Two processes involved:

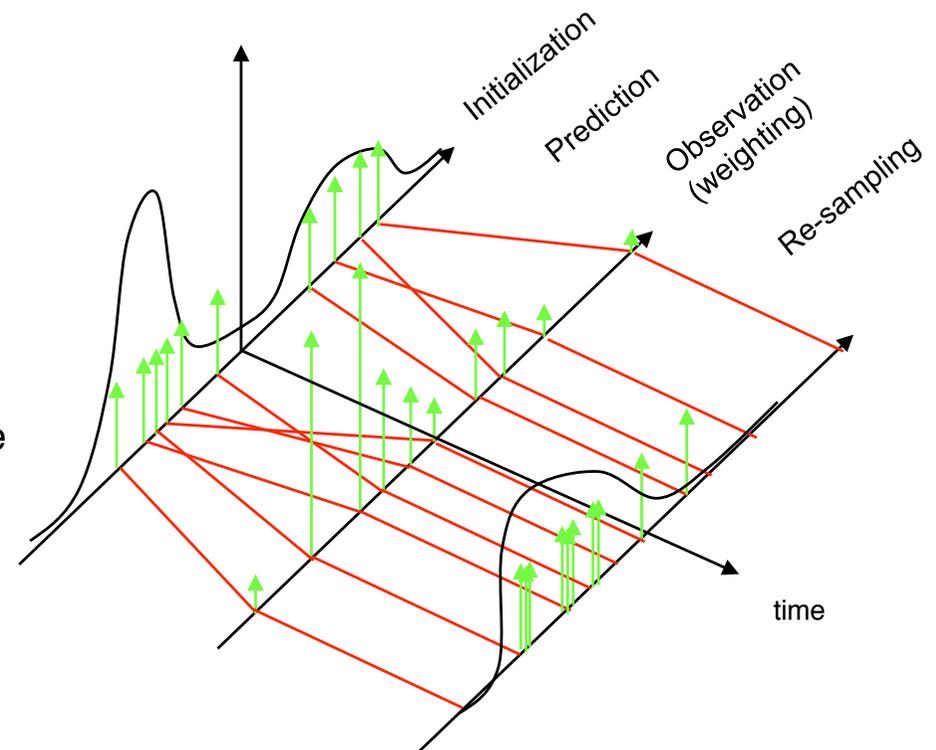
1. Matching rover and orbiter data (“perception” data association process)

⇒ Comparison between the rover DTM and the orbiter DTM (fast to compute)

2. Assessing the position (“estimation” process)

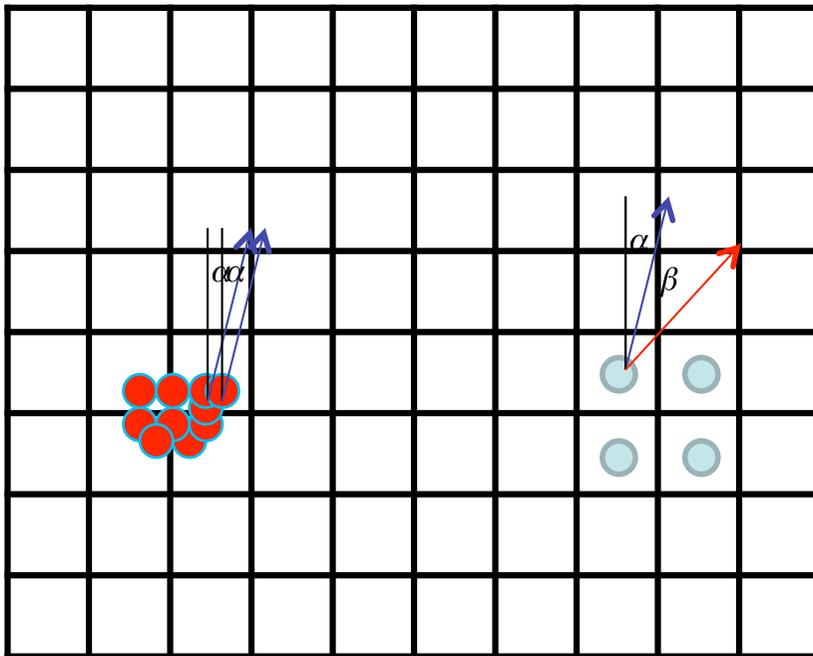
⇒ Particle filter

- The robot pos uncertainty is represented by a population of position hypotheses (the particles)
- Prediction step: the particle positions evolved according to the motion estimation (with associated error model)
- Observation step: the likelihood of each particle to represent the true position is estimated (with the perception process), particles are weighted accordingly
- Resampling: “light” particles are killed, “heavy” ones are “multiplied”



Proposed solution: optimizations

- Particles management
⇒ “discretized particles” on the orbital DTM



- One (x,y) particle per DTM cell
- Orientation (heading) angle is also discretized: $\beta = k\alpha$
- Particle uncertainty is assumed to be Gaussian

-  Conventional particles
-  Discretized particles

Getting the orbiter map

With a UAV, using a camera to build a high resolution DTM
Up-to-date commercial bundle adjustment techniques



- Down-sampled to 1m/cell

Getting the orbiter map

With a UAV, using a camera to build a high resolution DTM
Up-to-date commercial bundle adjustment techniques



Proposed solution: illustration

- On the quarry test field
- “Full” initial position uncertainty (worst possible case)



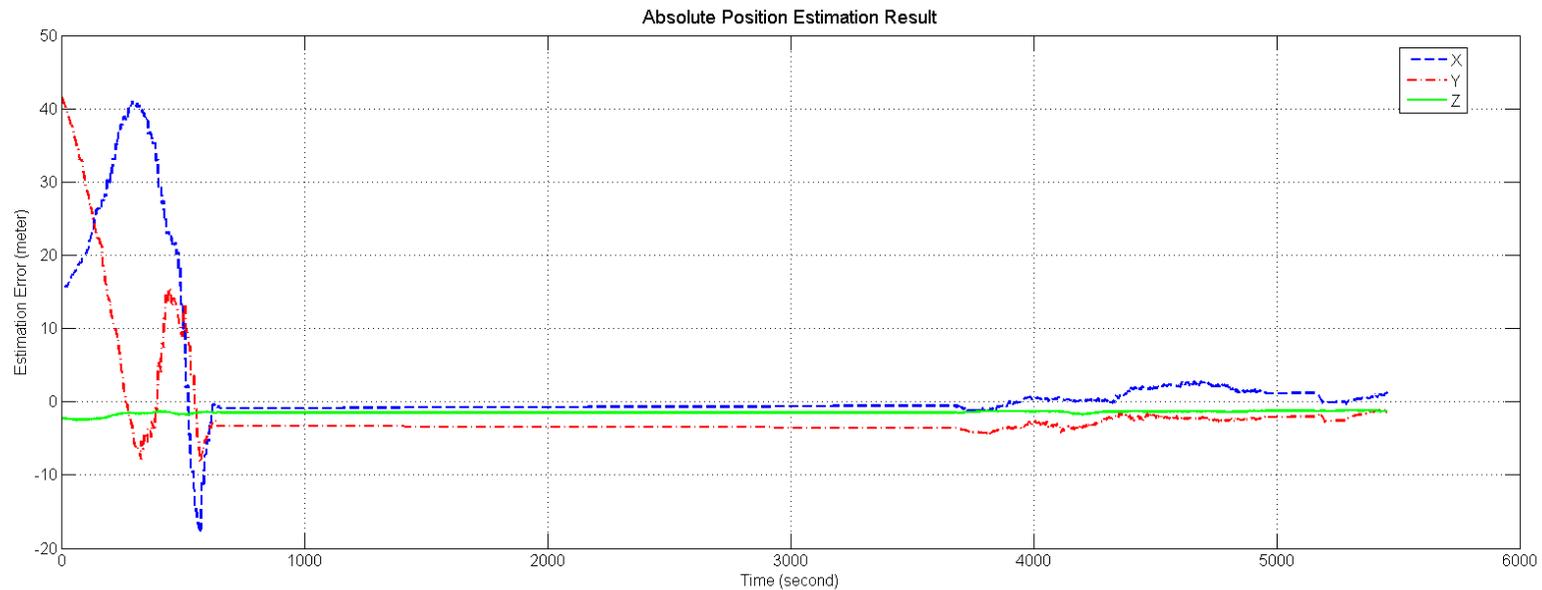
Proposed solution: illustration

- On the quarry test field
- “Full” initial position uncertainty (worst possible case)

Full search with heading information

Proposed solution: illustration

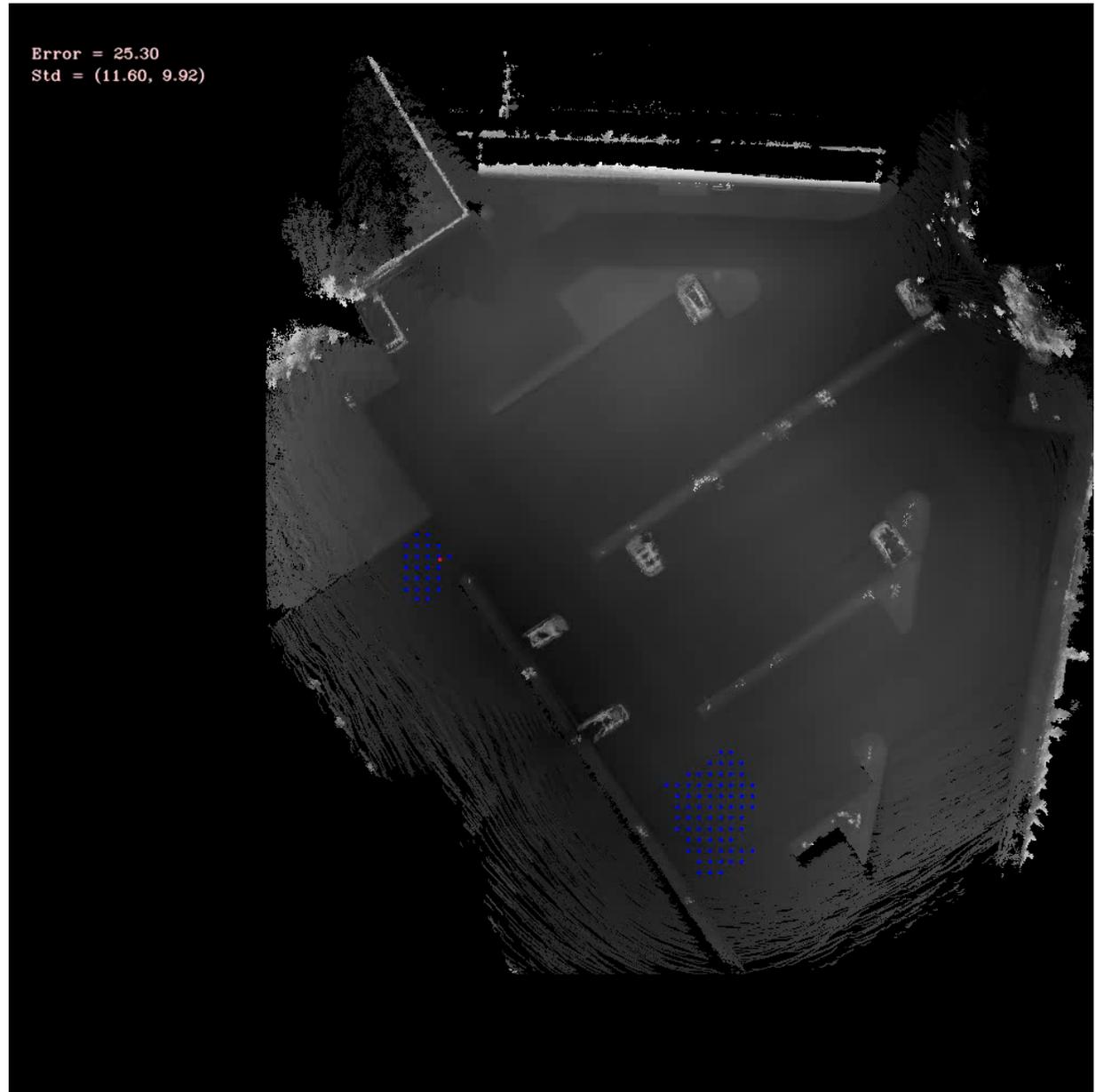
- On the quarry test field
- “Full” initial position uncertainty (worst possible case)



Evolution of the error of the (x,y,θ) position

Proposed solution: illustration

- On a parking lot
- “Full” initial position uncertainty (worst possible case)



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Summary and outlook (1/2)

- Showed good performances on the quarry test site
- Need for “signal” in the orbiter DTM



No terrain relief: no absolute localization possible



Sufficient terrain relief: possible absolute localization

Further extensions:

- Need to assess “localisability” in initial orbiter maps
- Exploit this information to plan long range traverses
- Integrate other absolute localisation processes

(this work is pursued at LAAS, and extended to terrestrial areas)

Outlook (2/2)

- More generally on localization:
 - A variety of needs
 - A variety of solutions
- Need to develop an overall framework that integrates the solutions and allows their control (in a distributed multi-robot context)

SLAM processes
are at the heart
of such a
framework

