

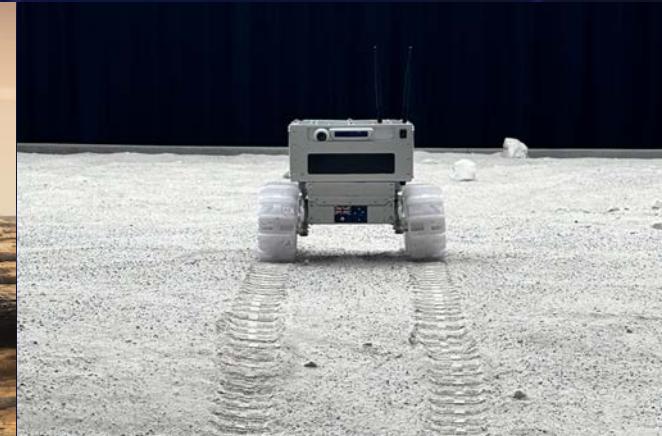
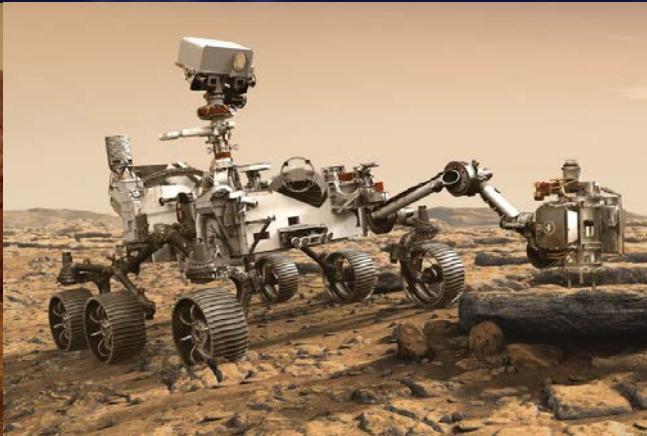


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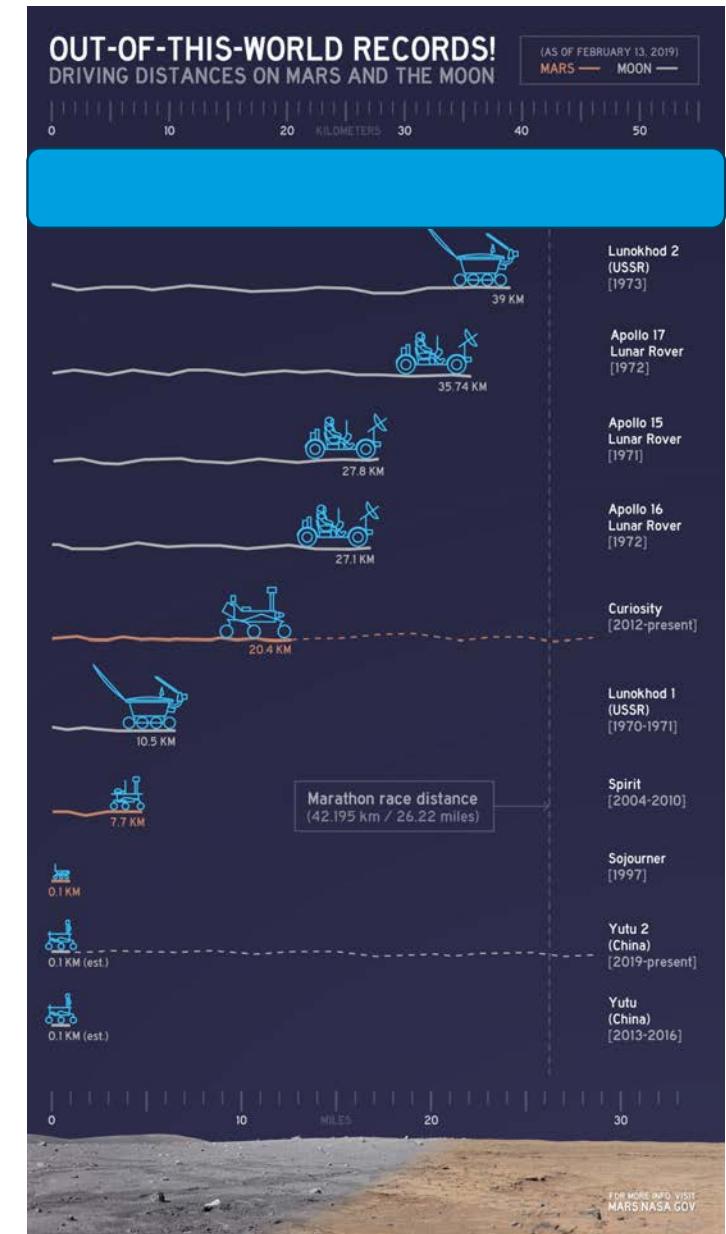
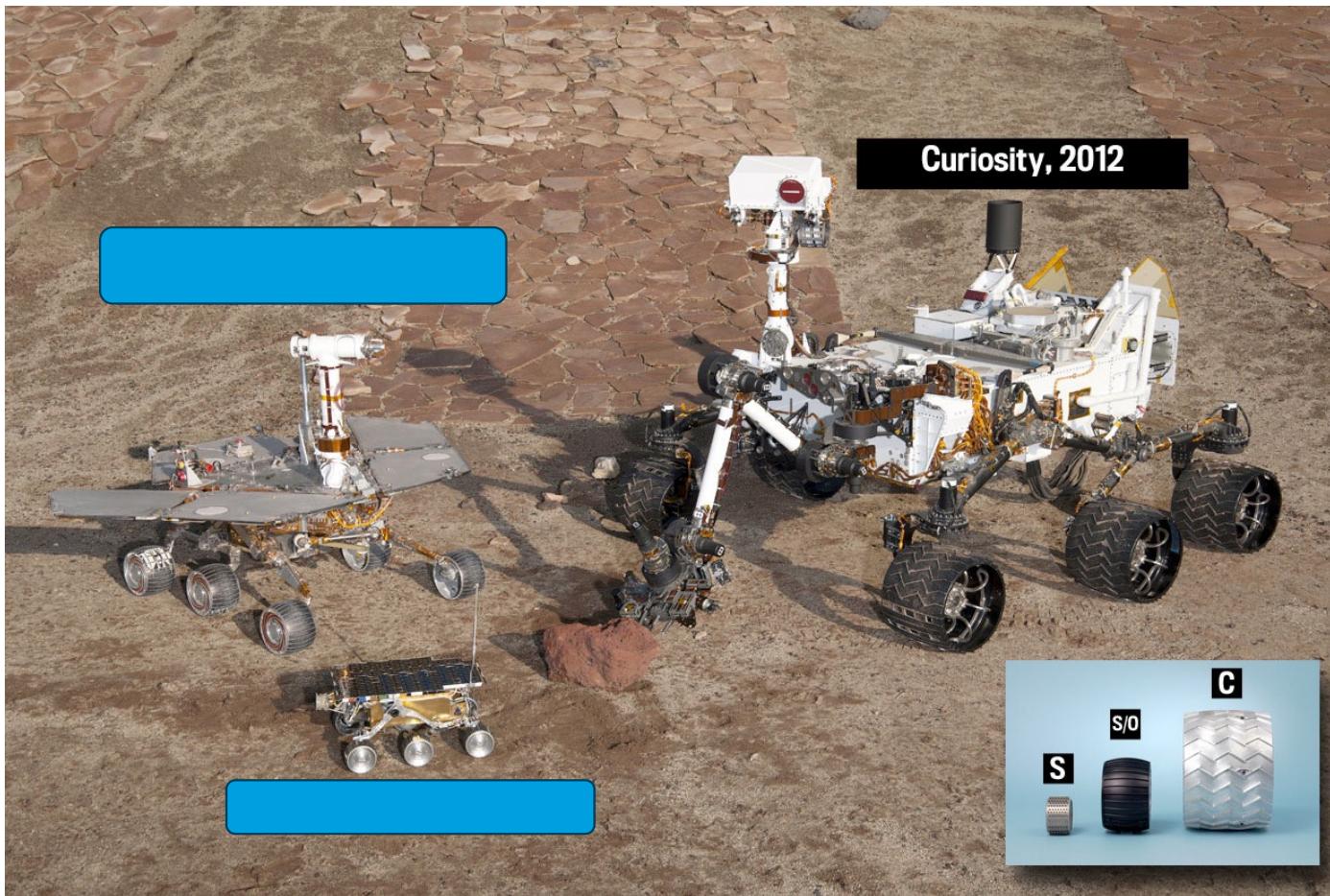
Planetary Rovers - Autonomous Navigation in Extreme Environments

A/Prof. Thierry Peynot

QUT Centre for Robotics

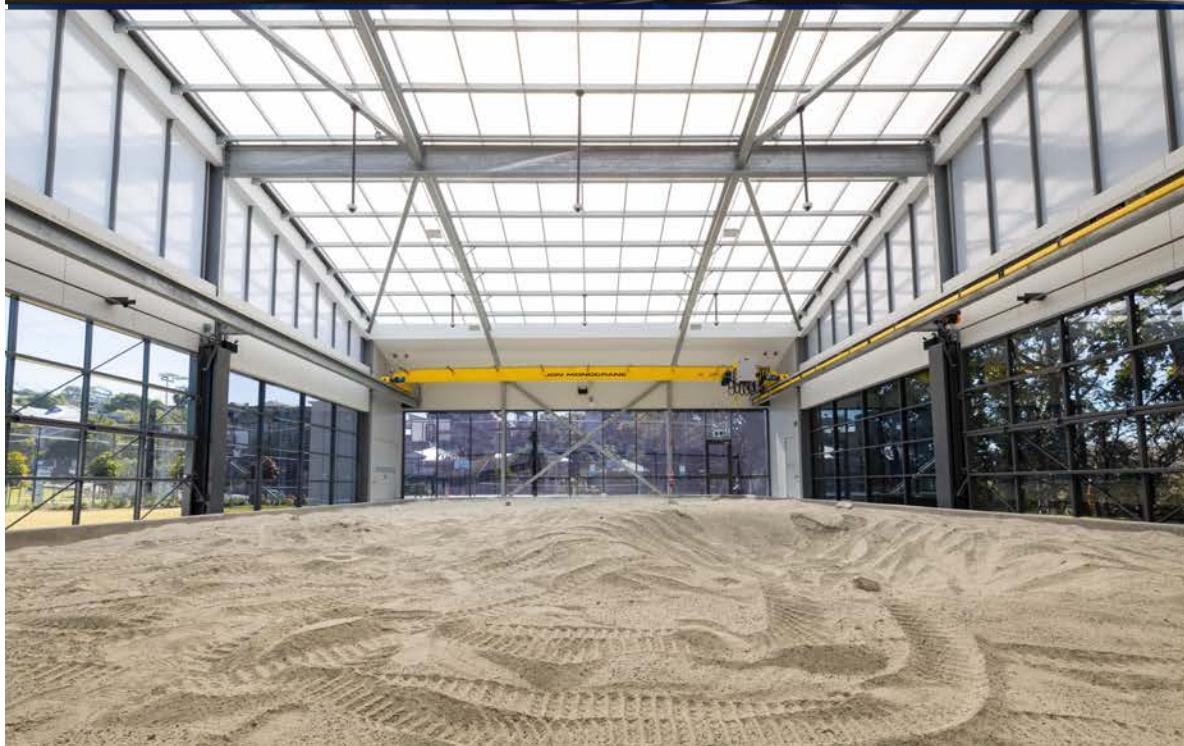


Planetary Rovers: Endurance race





Yandiwanba: QUT Lunar Testbed & Space Lab

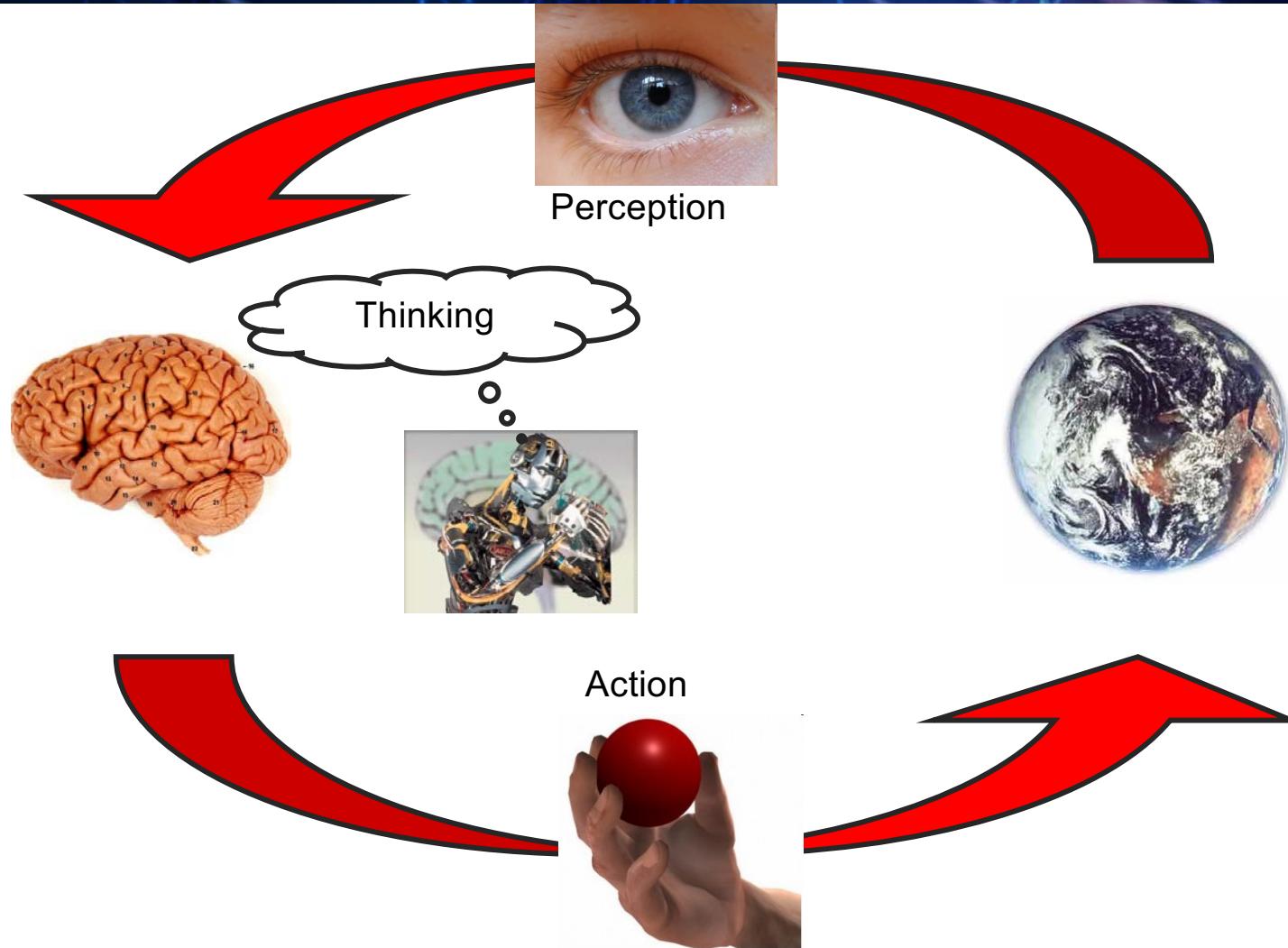




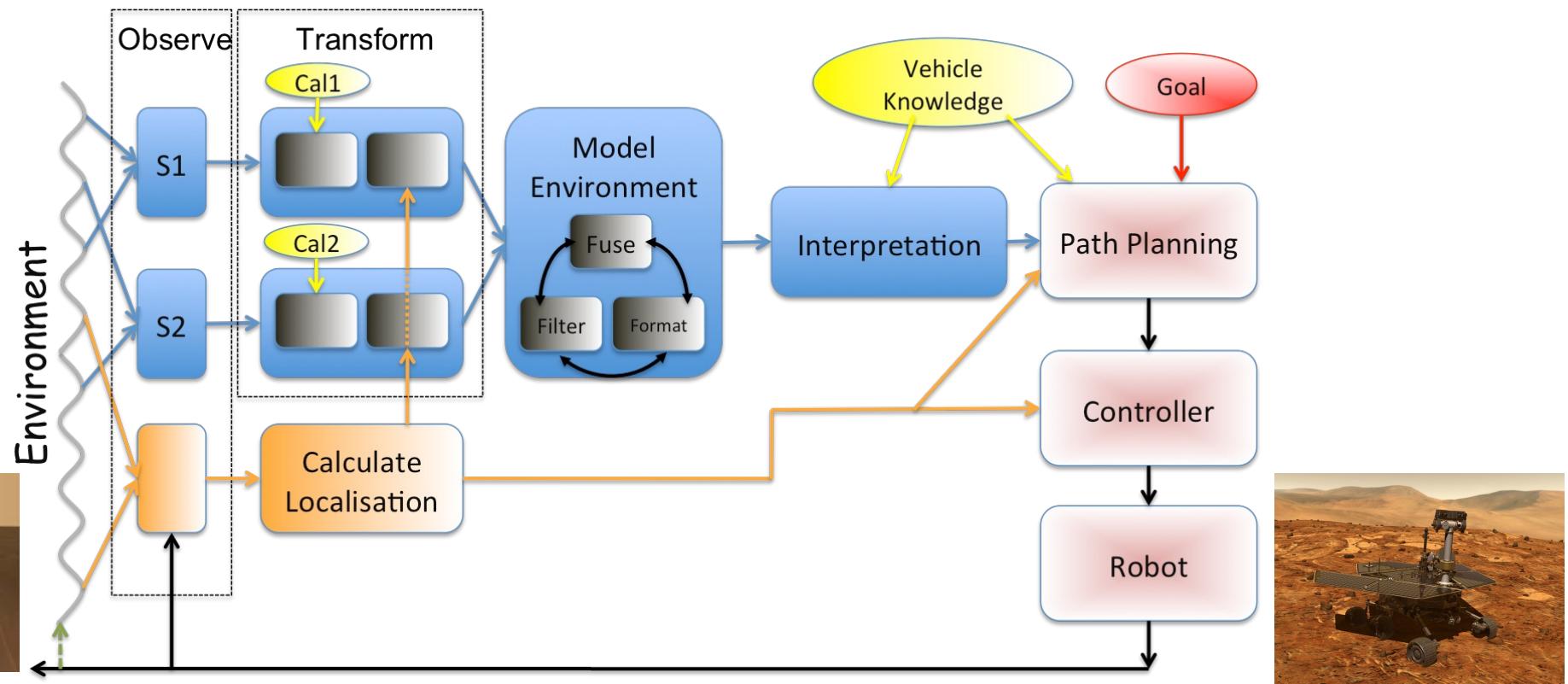
Australian
Space Agency



Autonomous Robot



Autonomous Navigation (Architecture)



Environmental Constraints

And other external constraints



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Robotics

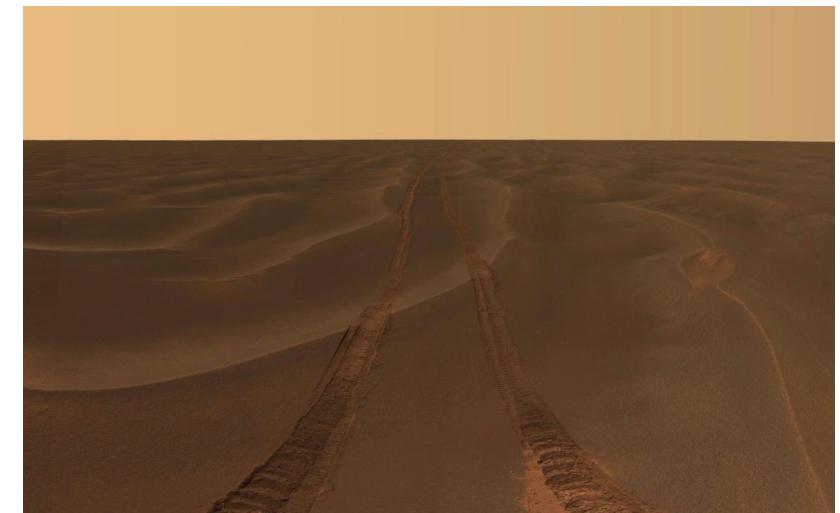
Environmental Constraints (Moon, “Near Side”, South Pole)

- Vacuum (no atmosphere)
- Radiations
- Temperature range
- Lower gravity (about 1/6 of Earth gravity)
- **Lunar regolith**
- **Distance to Earth** (Comm. delays: ~2.5s)
- Strong **Shadows**



Environmental Constraints (Mars)

- Lighter atmosphere
- Radiations (less than on the moon, but 40-50 times more than on Earth)
- Temperatures
- Gravity (~2.6 times lower than Earth)
- **Mars Dust / Soil**
- **Great distance to Earth** (=> Large Communication delays)



Localisation & Mapping



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ORB-SLAM2: an Open-Source SLAM for Monocular, Stereo and RGB-D Cameras
<https://www.youtube.com/watch?v=ufvPS5wJAx0>

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Next Challenges in Localisation

- Extreme lighting conditions
- Textureless environments
- Edge cases
- **Introspection**

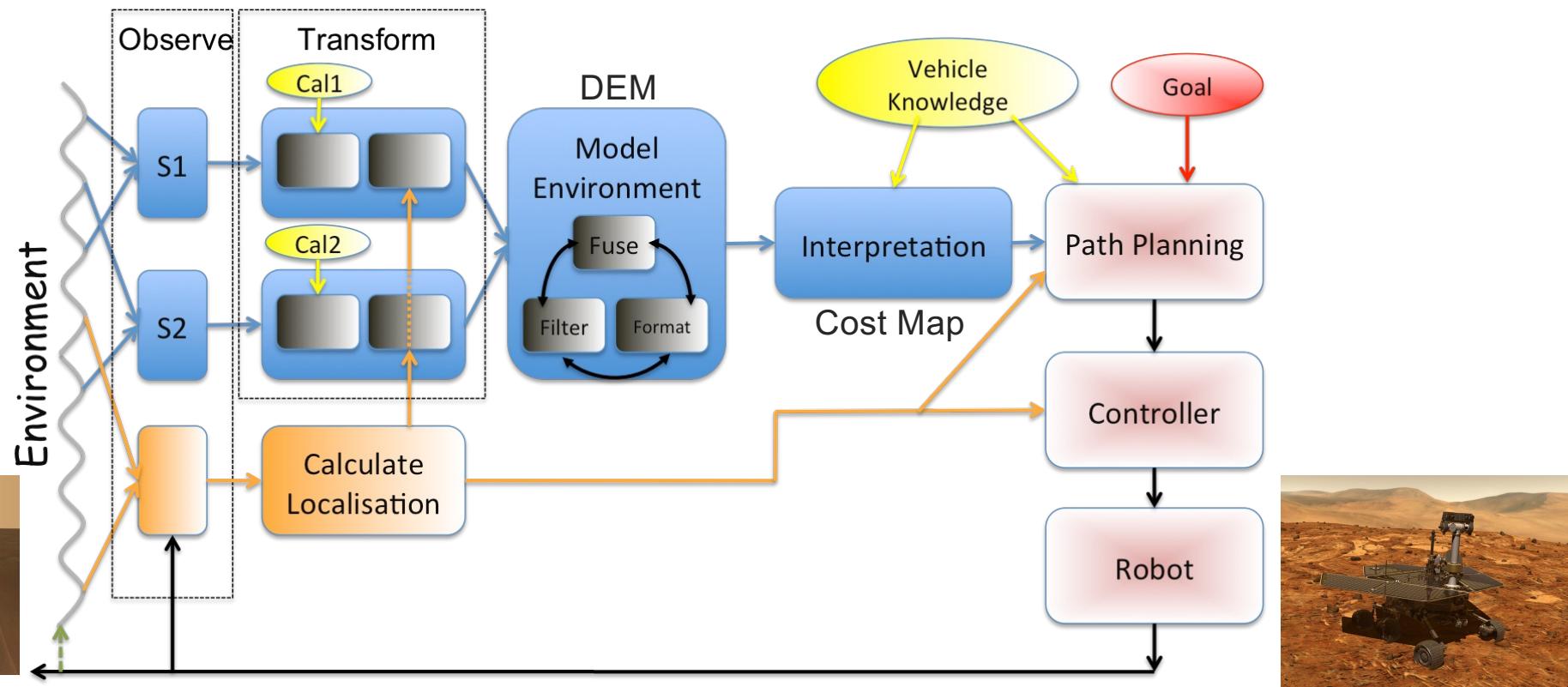


Terrain Traversability Estimation / Cost Maps

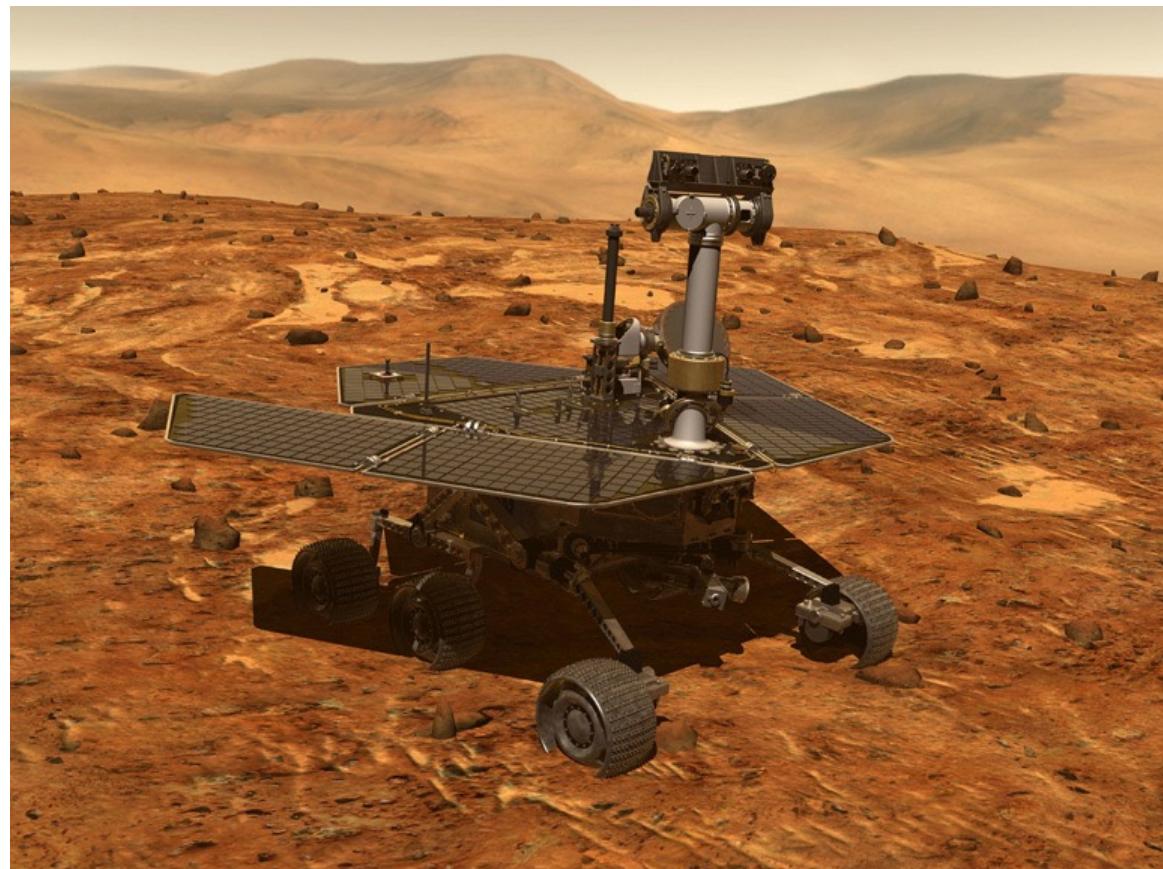
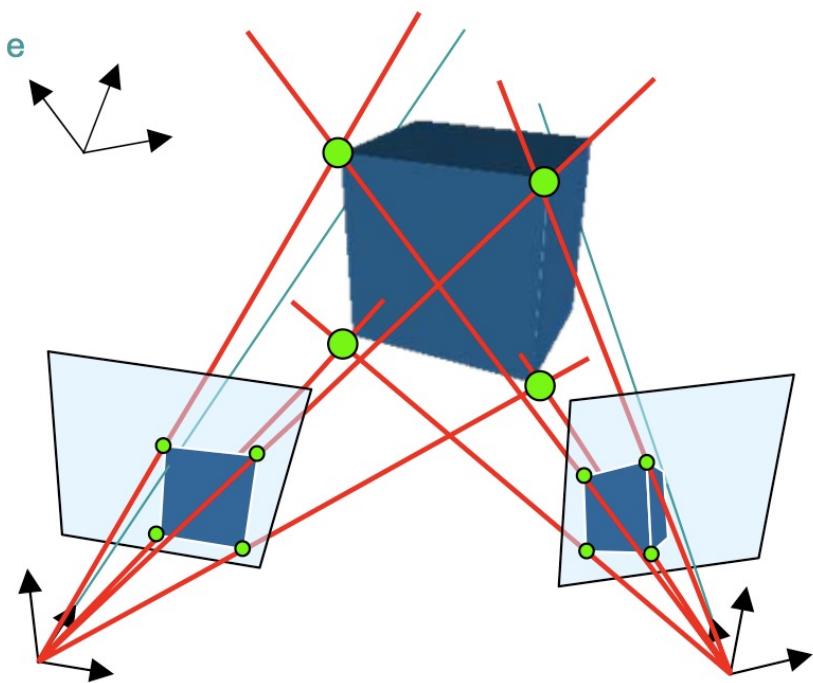


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Autonomous Navigation (Architecture)

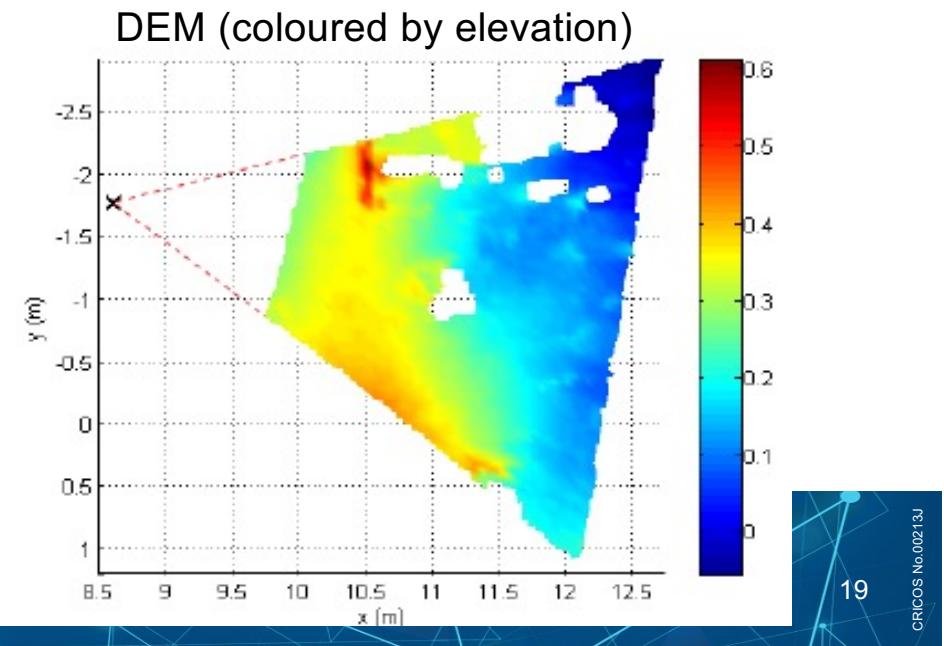


Stereovision

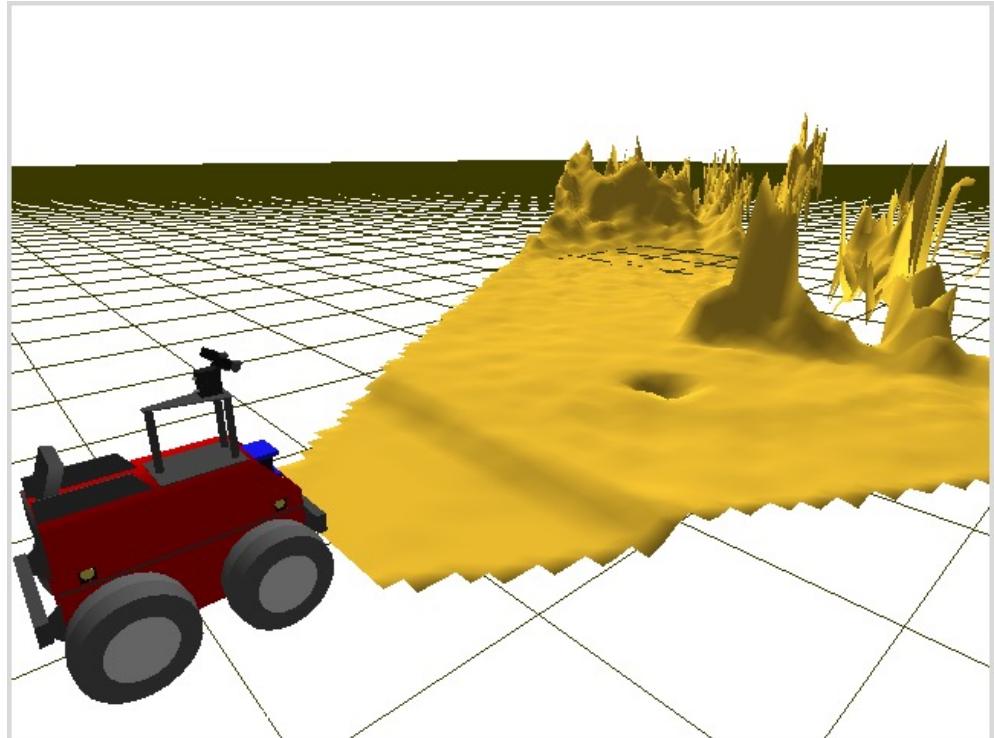
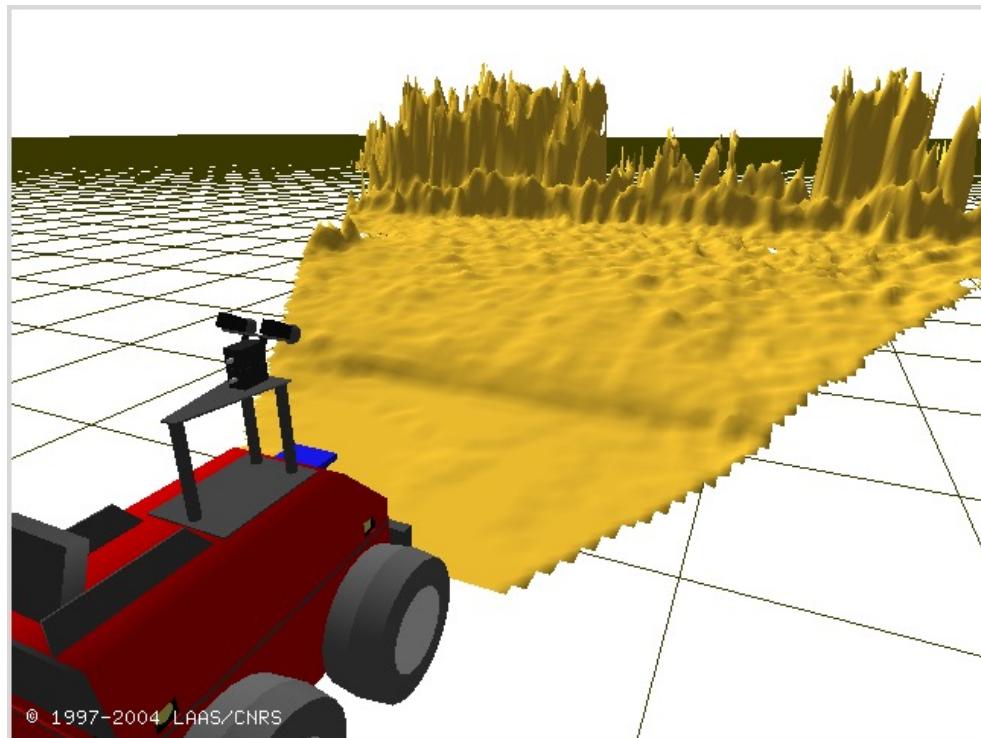


Digital Elevation Map (DEM) / 2.5D Map

- For each position (x,y) in a 2D map/grid, the DEM contains **one** elevation value $z=f(x,y)$
- Distinct from a 3D Map, which can contain multiple elevation values for each position (x,y) (e.g. voxel map)



Digital Elevation Map (DEM)



From Occupancy maps to traversability/cost maps

- **Occupancy map/grid:**
 - Usually discrete (grid) but can also be continuous (map)
 - each grid cell or point in the map contains either:
 - A binary, e.g. occupied (1) / free (0)
 - A probability that this cell is occupied
- **Cost map** for navigation (a.k.a. difficulty map, (terrain) traversability map):
 - Usually discrete but can also be continuous
 - each grid cell or point usually contains:
 - a (normalised) value of cost/difficulty*, from 0 to 1
 - Other characteristics such as the features that led to the cost computation (e.g. slope, rugosity etc.)

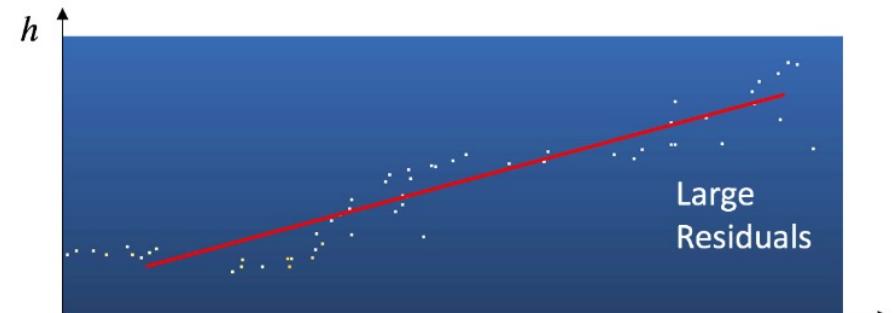
Terrain Traversability Estimation

Traversability metrics:

- Slope
- Roughness



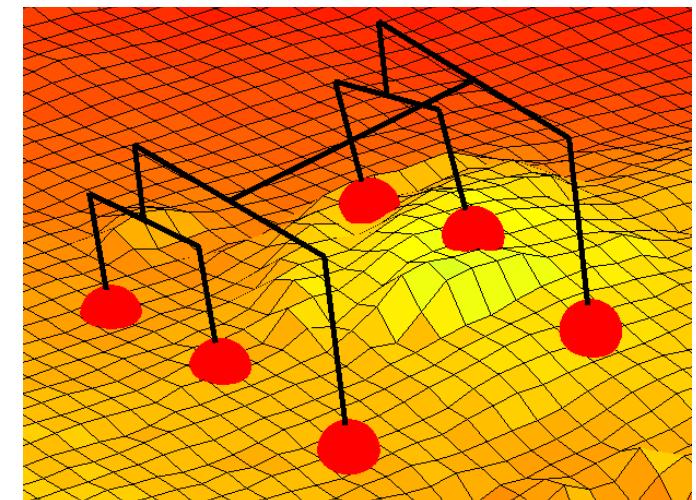
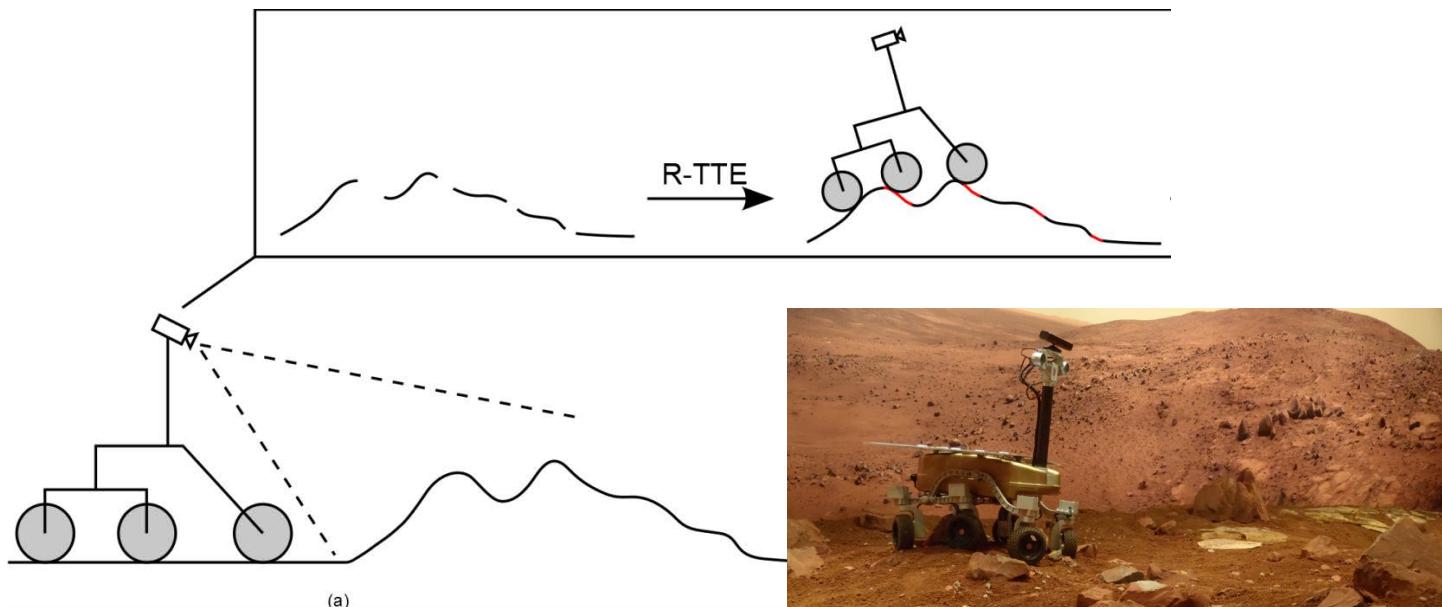
Independent from platform?



P. Borges, T. Peynot, et al. A Survey on Terrain Traversability Analysis for Autonomous Ground Vehicles: Methods, Sensors, and Challenges. In Field Robotics, Vol. 2, July 2022.

Digital Elevation Map and Terrain Traversability Estimation

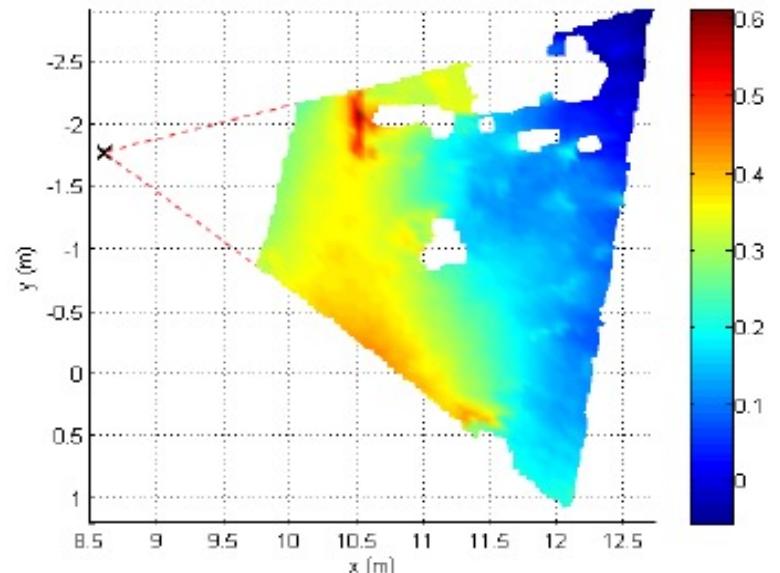
- Evaluation of the configuration of the rover (roll, pitch, chassis configuration) when ‘placed’ on the terrain



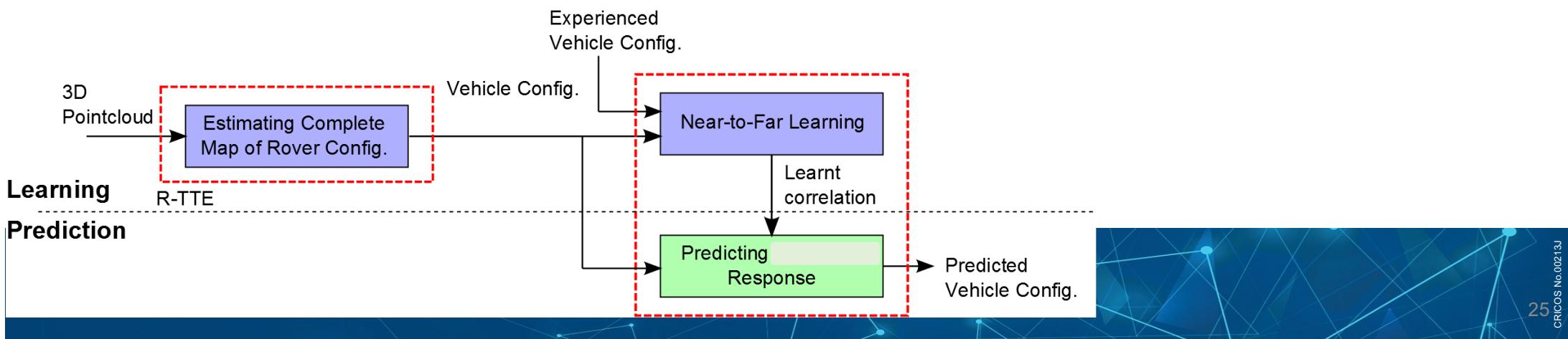
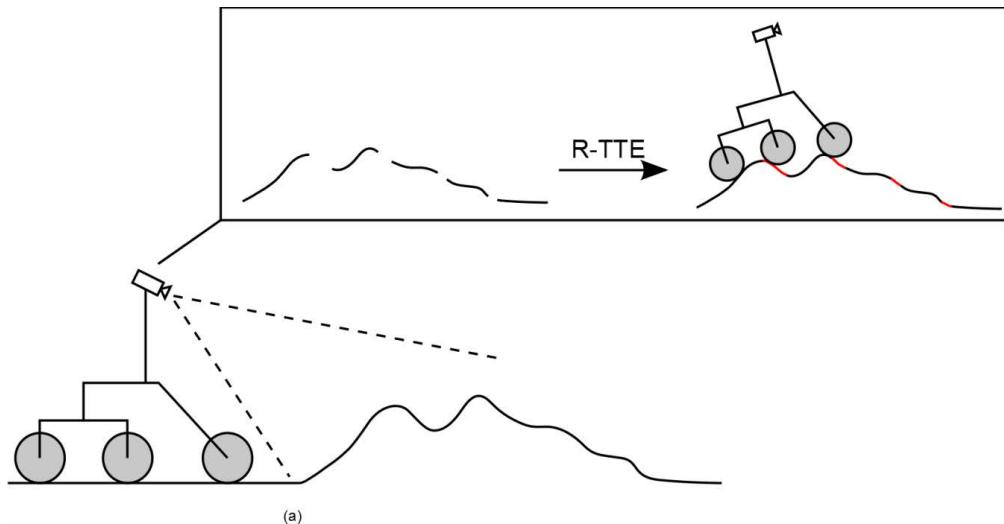
Challenge 1



DEM (coloured by elevation)



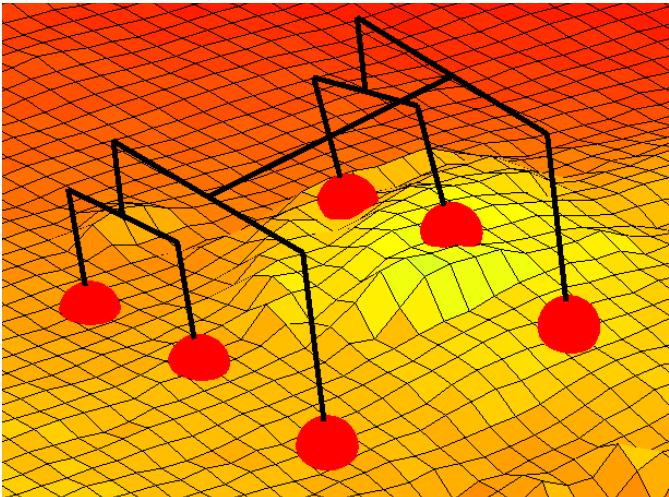
Near-to-Far Learning



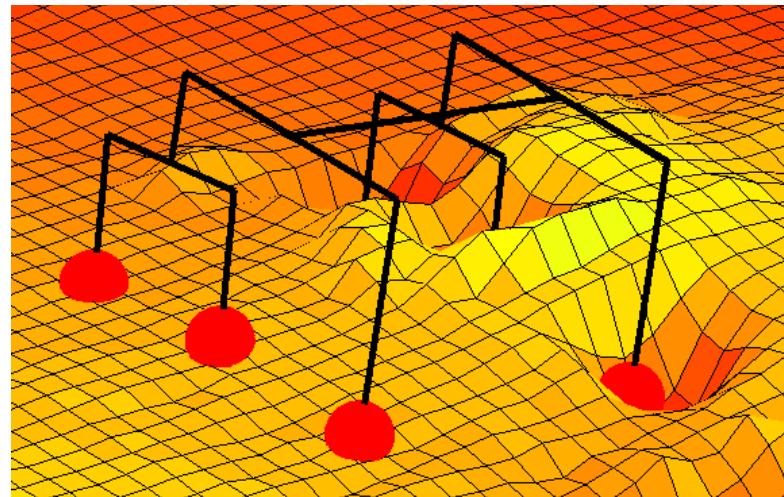
Challenge 2



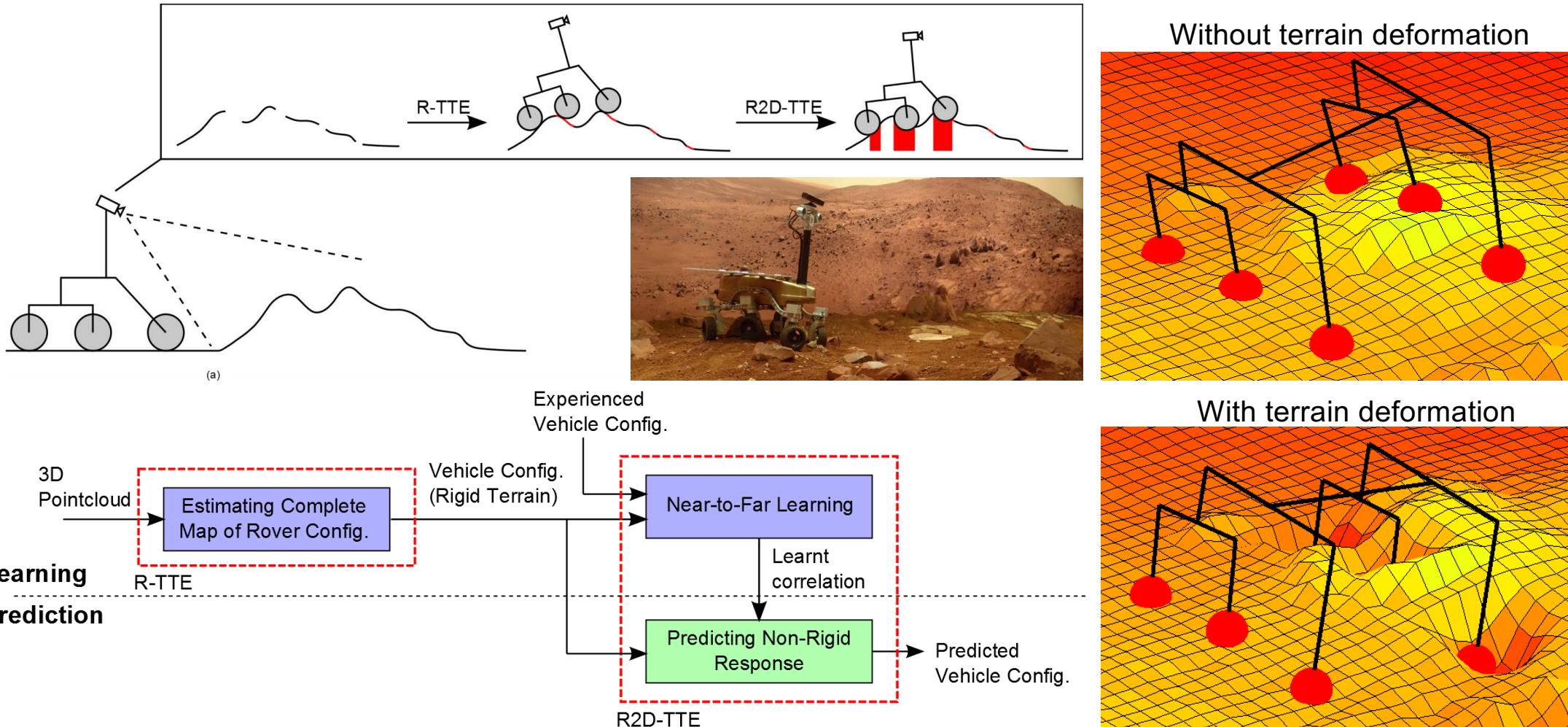
Without terrain deformation



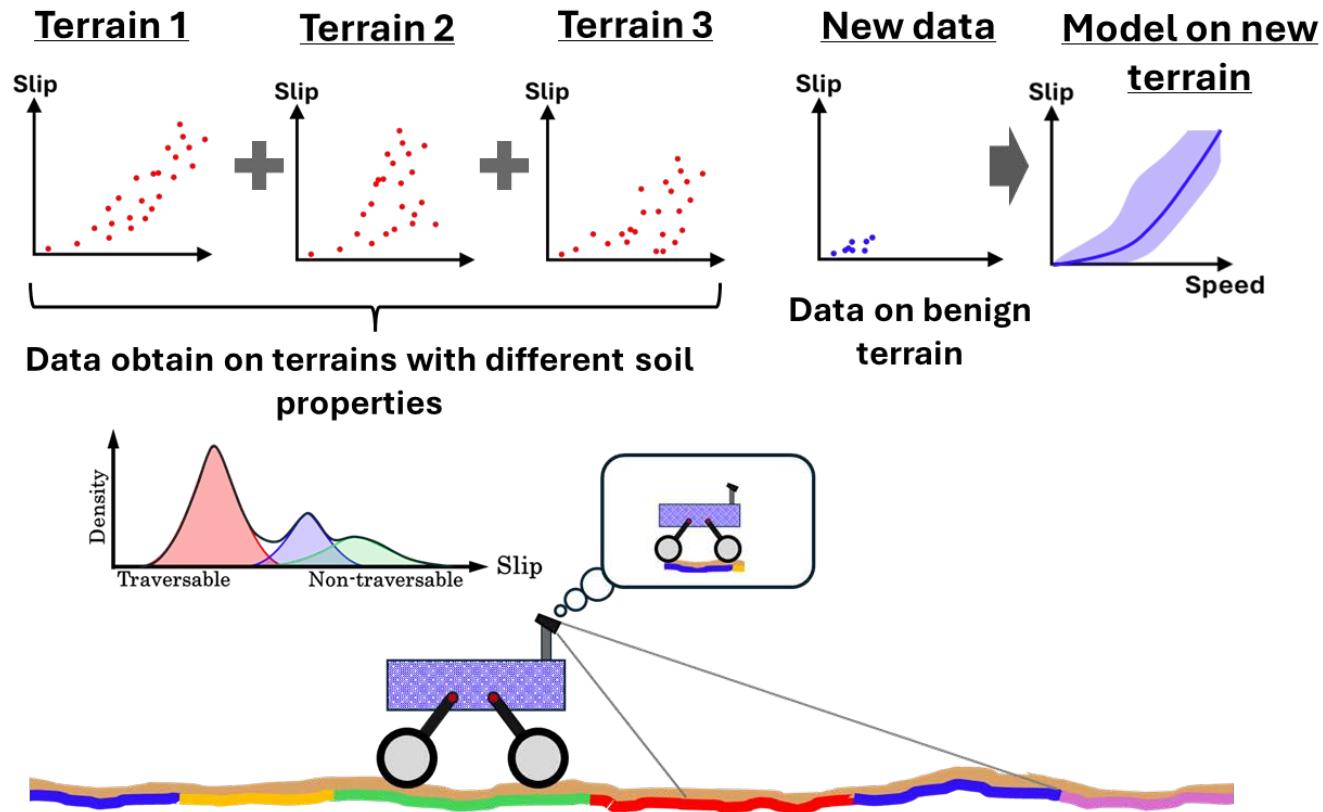
With terrain deformation



Rigid-to-Deformable Terrain Traversability Estimation

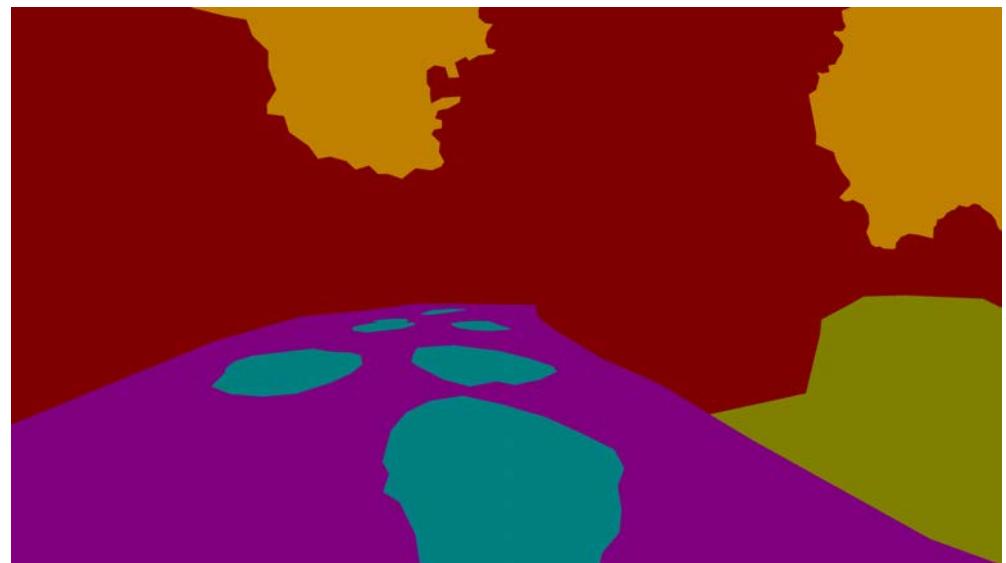


Slip



SlipNet: Enhancing Slip Cost Mapping for Autonomous Navigation on Heterogeneous and Deformable Terrains
<https://arxiv.org/html/2409.02273v2>

Terrain Classes





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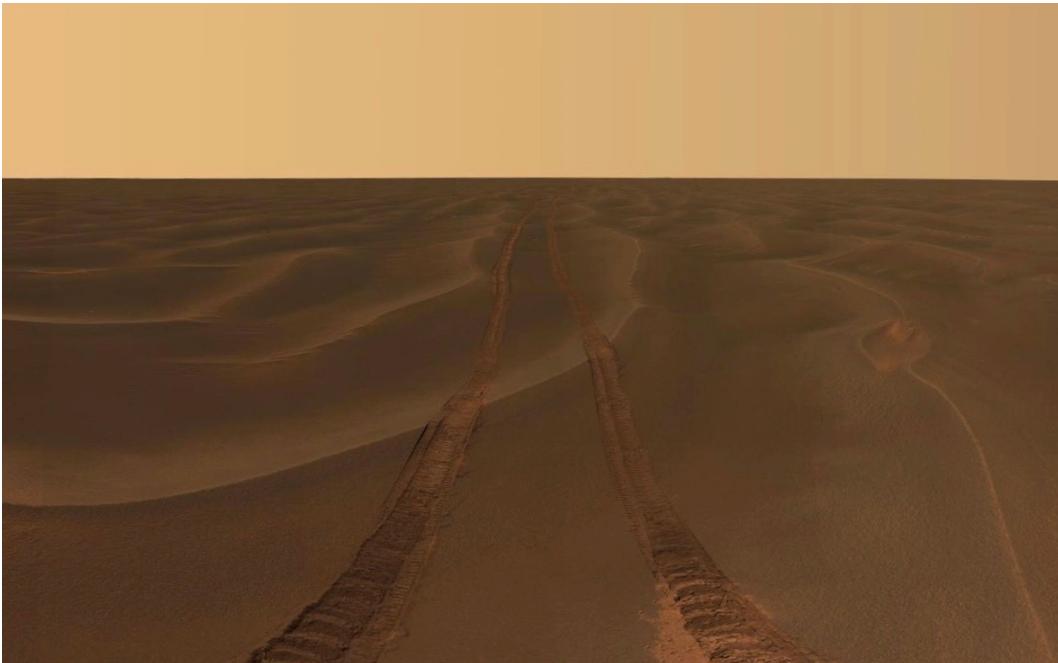


On the moon



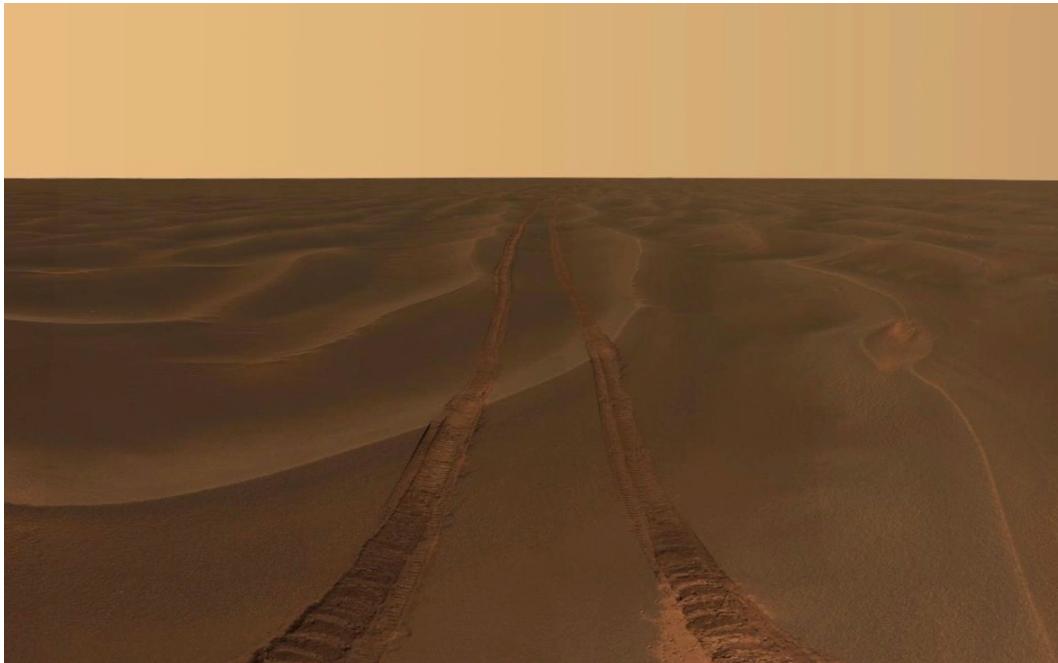
On Mars

- Terrain traversability estimation



On Mars

- Terrain traversability estimation

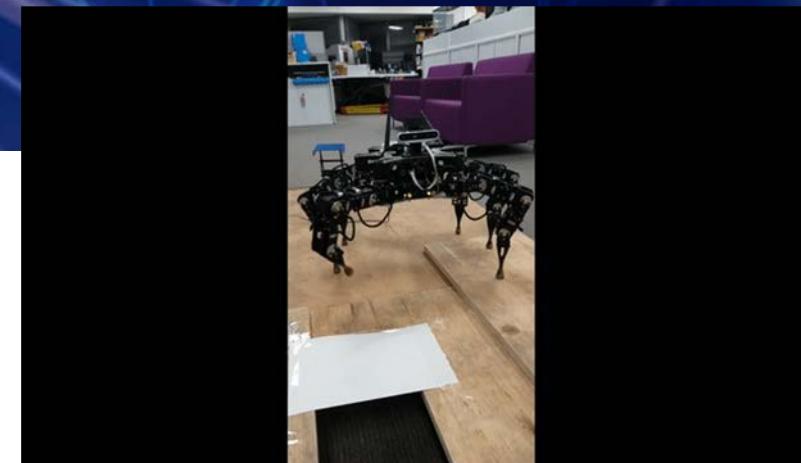


On Mars

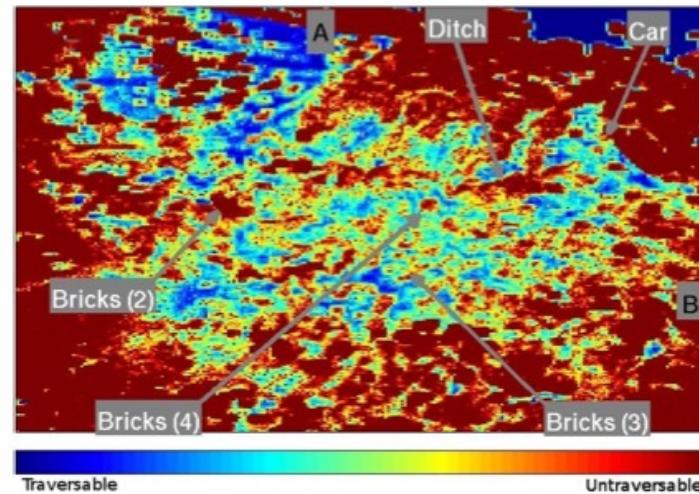


Current Challenges in Terrain Traversability Estimation

- Predict Rover-Terrain Interaction when there is no geometric or visual distinction (e.g. lunar regolith)
- Learn with limited negative examples
- **Introspection**
 - Known Unknowns
 - Unknown Unknowns



Terrain Traversability Estimation in Densely-Vegetated Environments (Earth)



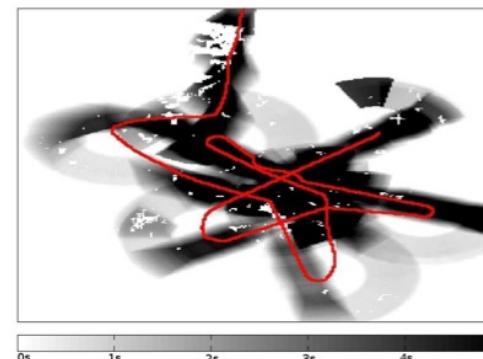
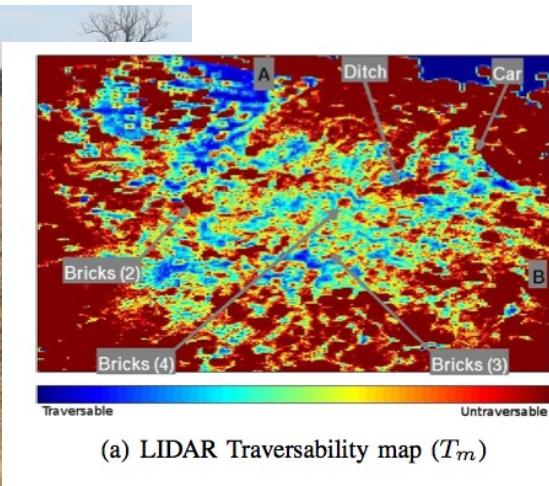
(a) LIDAR Traversability map (T_m)

[Ahtiainen-IROS-2013]

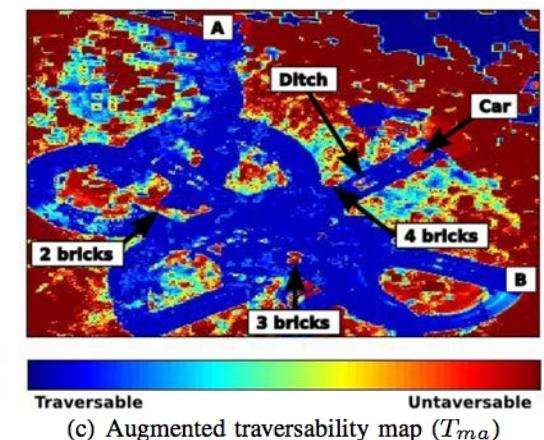
[Ahtiainen-FUSION-2015]



Augmenting Traversability Maps with UWB Radar to Enhance Obstacle Detection in Vegetated Environments



(b) Updated cells in radar FOV

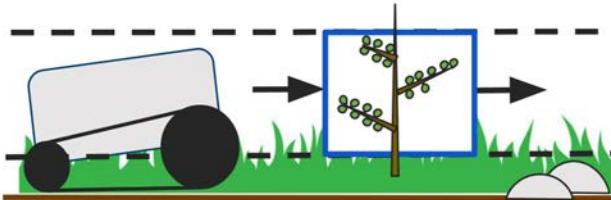


[Ahtiainen-IROS-2013]
[Ahtiainen-FUSION-2015]

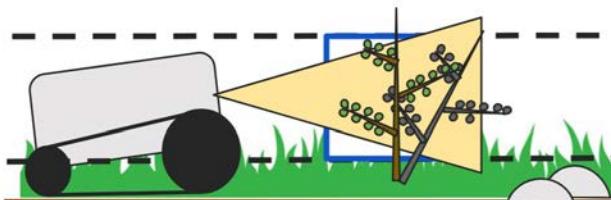




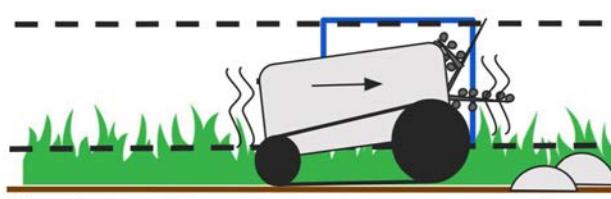
ForestTrav: Accurate, Efficient and Deployable Forest Traversability Estimation for Autonomous Ground Vehicles



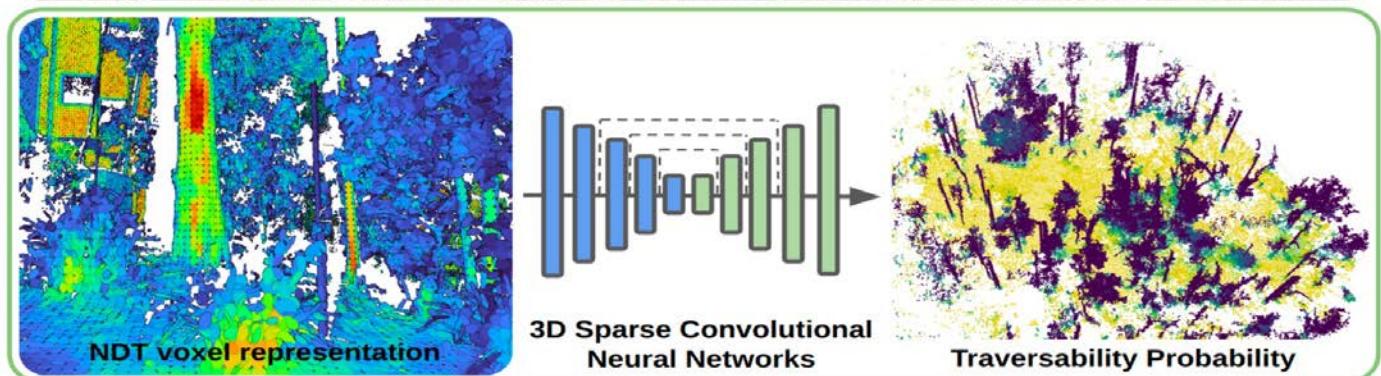
Volume or instance for which to estimate "interaction cost"



Estimating "interaction cost" from afar using exteroceptive sensors



Attempting to push through and sensing the current interaction

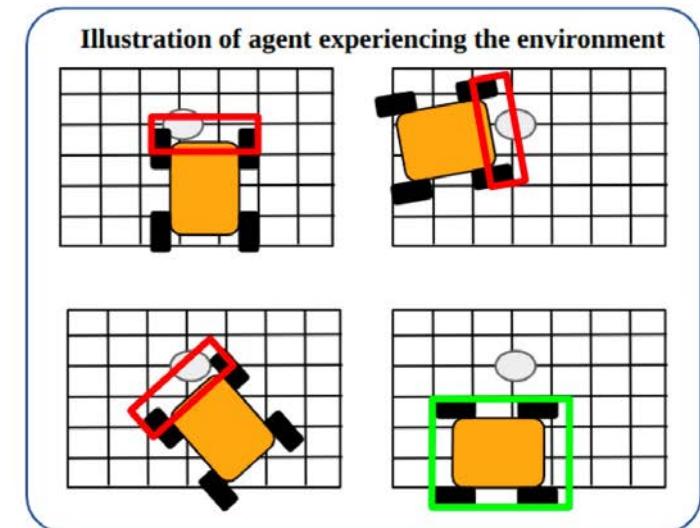


[Ruetz et al, IROS 2022 & IEEE Access 2024]

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Cell size for Occupancy Grid / DEM / Cost Map

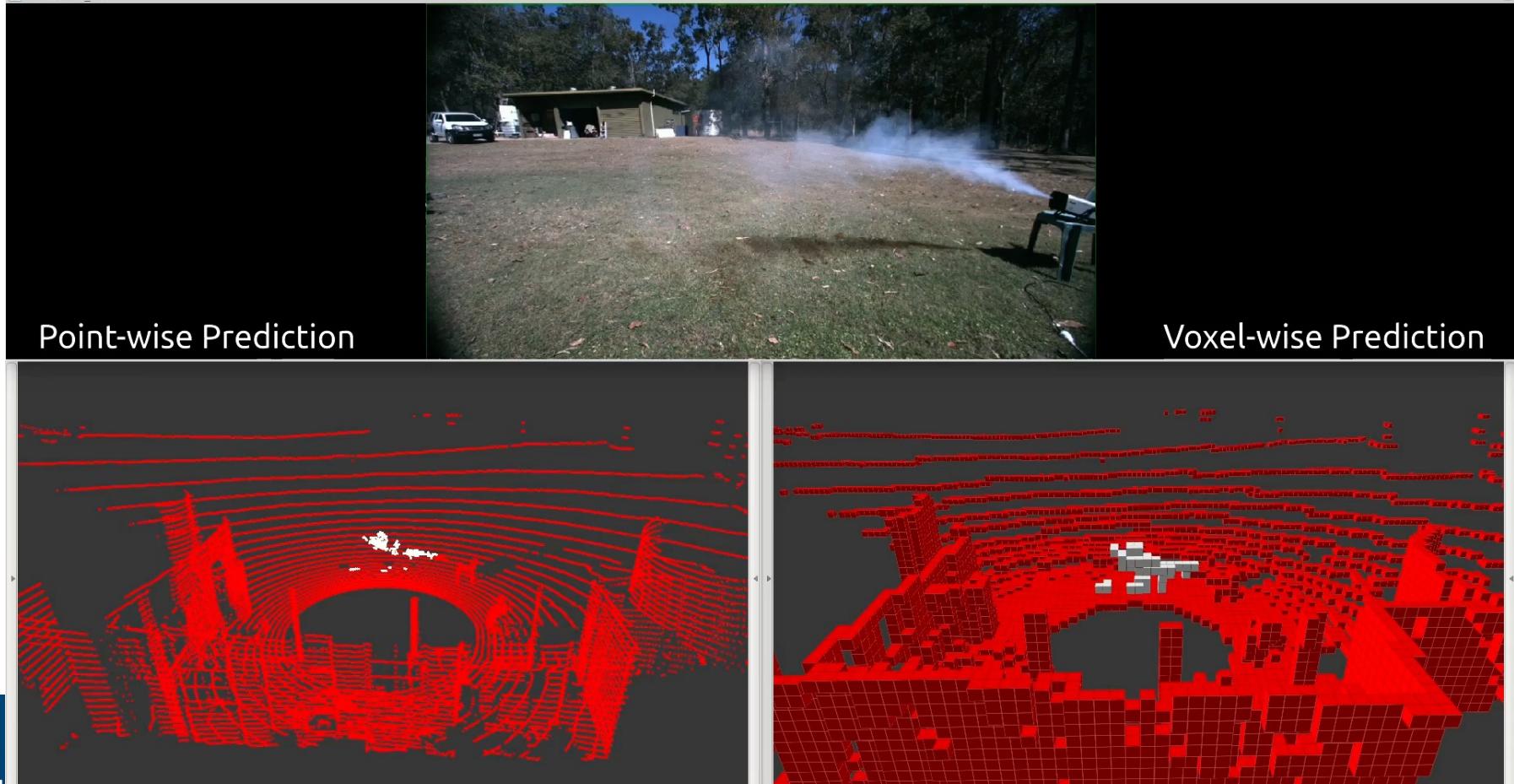
- To select the cell size, consider:
 - Size (footprint) of the robot
 - Size of the environment to be mapped
 - Computation resources, e.g. memory
- We usually would like the highest resolution possible, computation constraints



F. A. Ruetz, N. Lawrence, E. Hernández, P. V. K. Borges and T. Peynot,
"ForestTrav: 3D LiDAR-Only Forest Traversability Estimation for
Autonomous Ground Vehicles," in IEEE Access, vol. 12, 2024.

Occupancy Grid in 3D: Voxel map

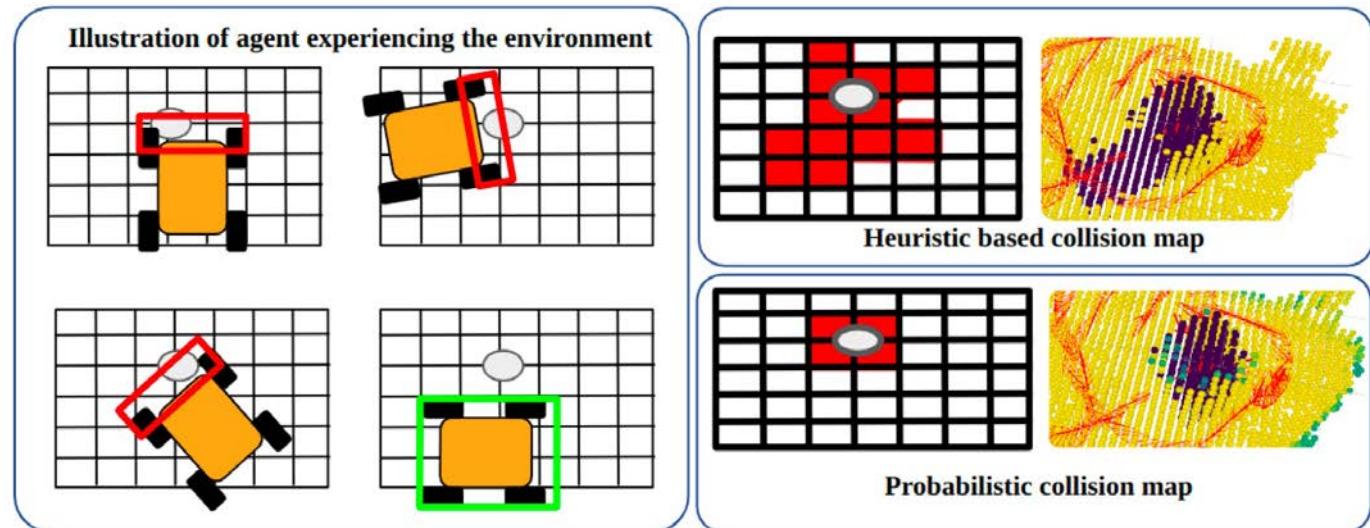
- Voxels





ForestTrav: Accurate, Efficient and Deployable Forest Traversability Estimation for Autonomous Ground Vehicles

Traversability Labelling

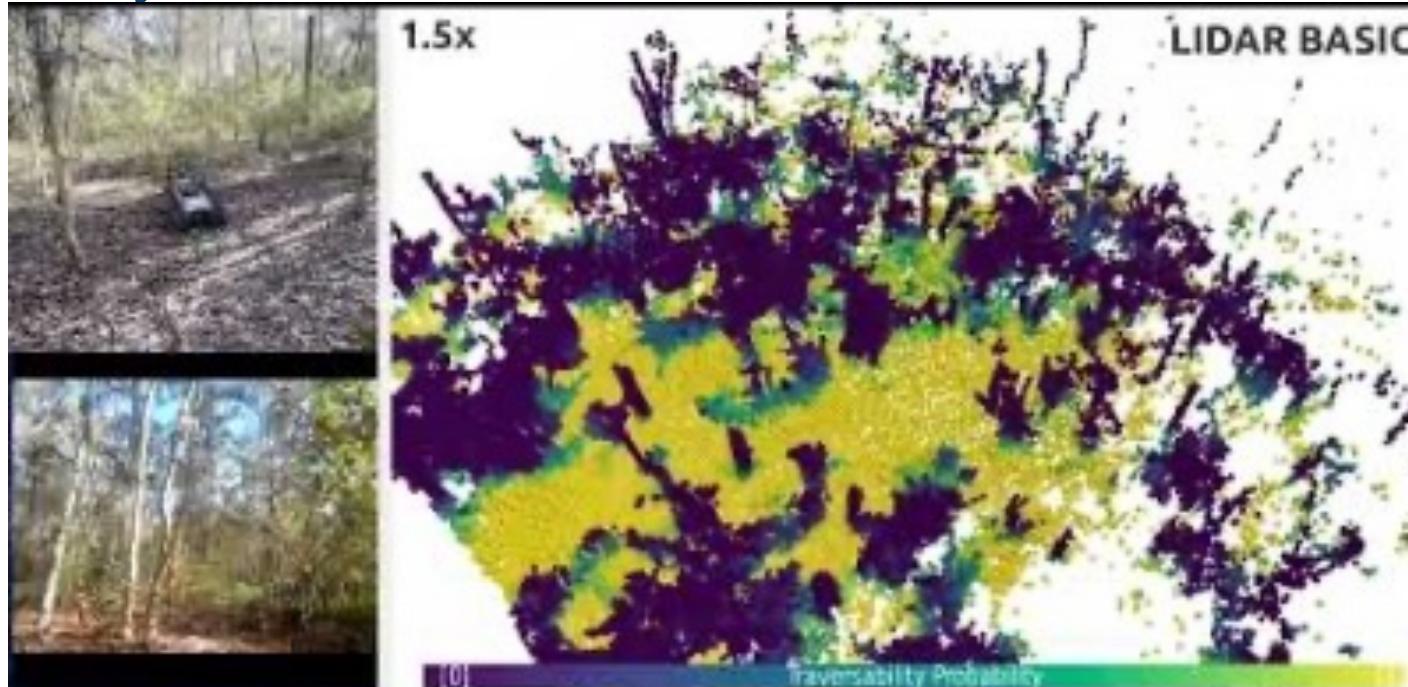


[Ruetz et al, arXiv & IEEE Access 2024]

Data Set: <https://data.csiro.au/collection/csiro:58941>

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ForestTrav: Accurate, Efficient and Deployable Forest Traversability Estimation for Autonomous Ground Vehicles



A variation using only five lidar features is now shown. This variation is also capable of assessing the environment accurately.

- <https://www.youtube.com/watch?v=Kw8easF89Zg>

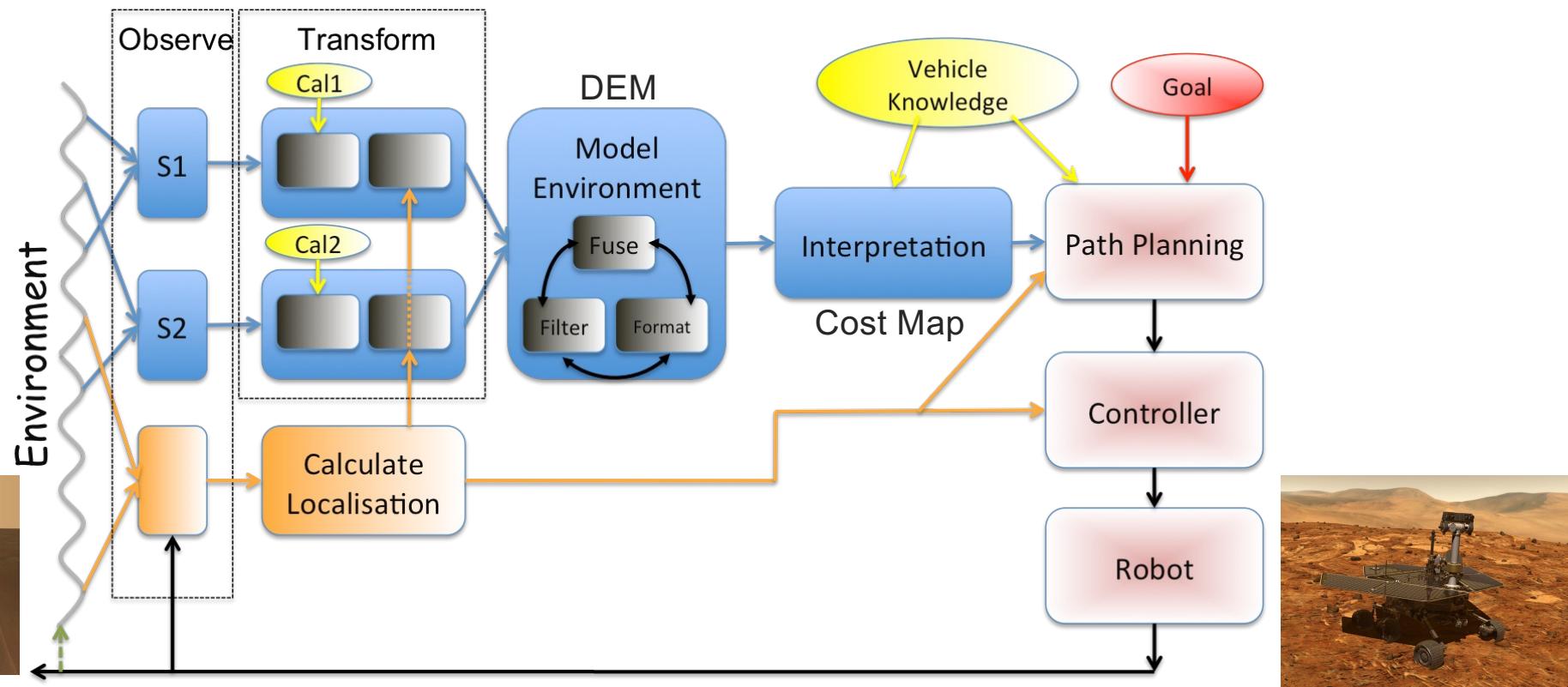
Self-Supervised or ‘end-to-end’ Traversability/Navigation

BADGR: An Autonomous Self-Supervised Learning-Based Navigation System

Gregory Kahn, Pieter Abbeel, Sergey Levine



Autonomous Navigation (Architecture)



Planning (with Uncertainty)



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Learned Stochastic Mobility Prediction for Planning with Control Uncertainty on Unstructured Terrain



**Learned Stochastic Mobility
Prediction for Planning with Control
Uncertainty on Unstructured Terrain**

Thierry Peynot, Angela Lui, Rowan McAllister,
Robert Fitch and Salah Sukkarieh





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Planetary Rovers - Autonomous Navigation in Extreme Environments

A/Prof. Thierry Peynot

QUT Centre for Robotics

