

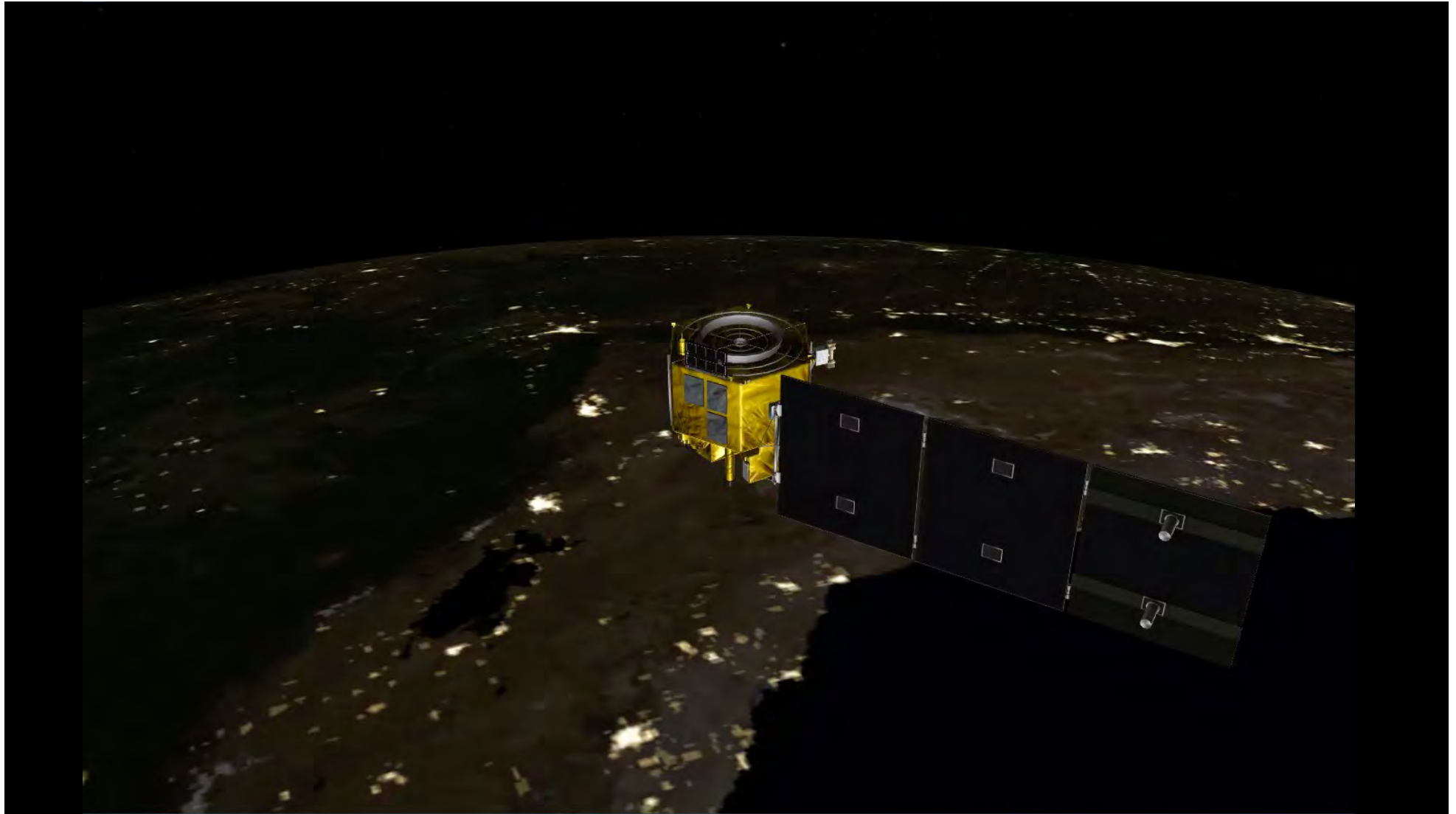
Artificial Intelligence in Space and the Hunt for life beyond Earth!

Steve Chien
Jet Propulsion Laboratory
California Institute of Technology

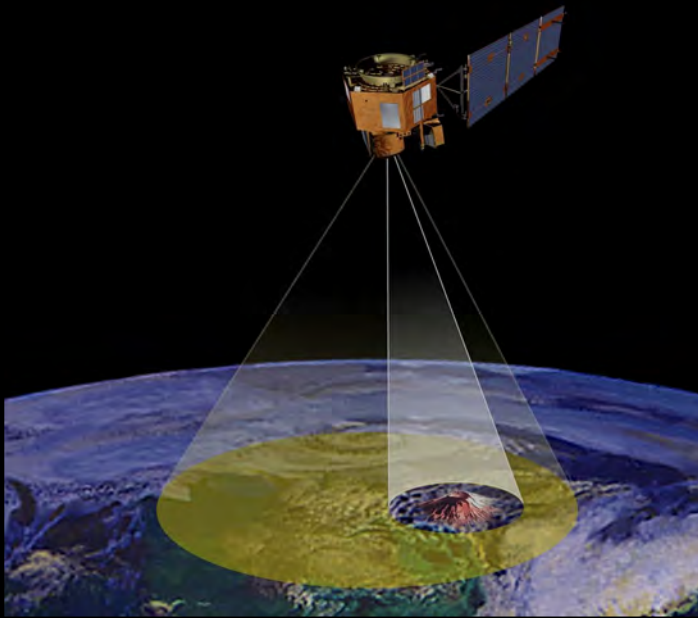
ai.jpl.nasa.gov ml.jpl.nasa.gov dus.jpl.nasa.gov

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Portions of this work were carried out at the Jet Propulsion Laboratory, California Institute of
Technology, under a contract with the National Aeronautics and Space Administration.
JPL Clearance: CL 23-4639.

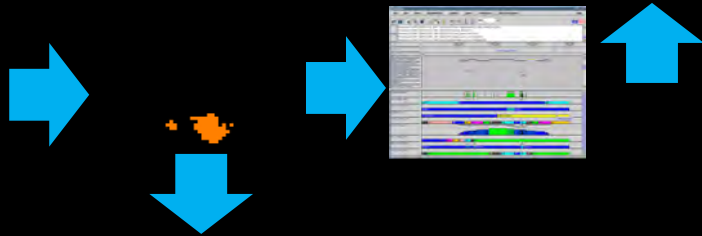
Predecisional, for planning and discussion only.
Any cost information contained in this document is of a
budgetary and planning nature and is intended for
informational purposes only. Any such information
does not constitute a commitment on the part of JPL
and/or Caltech.



The Autonomous Sciencecraft on Earth Observing One



- 2004-2017 Primary Operations
- 60000+ autonomous data collects over 12+ years
 - ~ 3 M commands!
- Operations cost reduction > \$1M/yr
- Reduce re-planning time to respond to anomalies from: days → hours
- (2009) R5 upgrade enabled significant increase in scene acquisition rate of ~ +33%
 - Estimated added value \$800K+ / year
- ASE co-winner NASA Software of the Year 2005

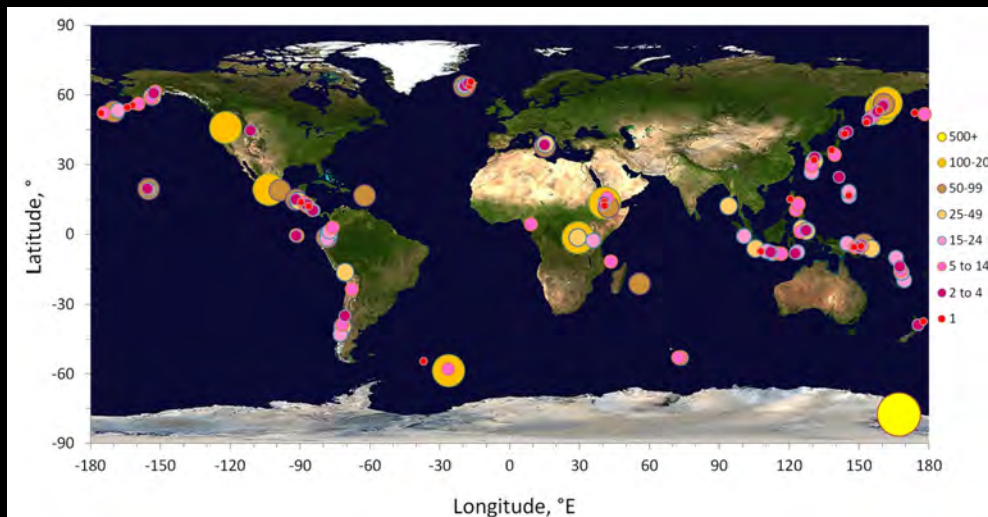


POC: Steve Chien/JPL

<https://ai.jpl.nasa.gov/public/projects/ase/> **JPL**

Example: NASA ASE/EO-1 Volcanoes

- Automated tasking: Volcano Sensorweb
 - Links together scores of space, ground, other assets
 - Automated Data analysis, triage to generate prioritized requests → ASE/EO-1 service → products delivered to stakeholders.
 - Over 100,000 alerts/triggers
End Result, - Thousands of volcanic scenes 2008-2017,
35%+ of said scenes with thermal signatures!
Compare to MODIS background < 1% of scenes with active thermal signature.



Partners (incomplete list):

MODVOLC

GOESVOLC

VIIRS (NPOES)

AFWA

VAAC

Iceland/MEVO

Etna VO (U. Firenze)

MEVO (NM Tech)

HVO (Kilauea)

IEGPN (Ecuador)

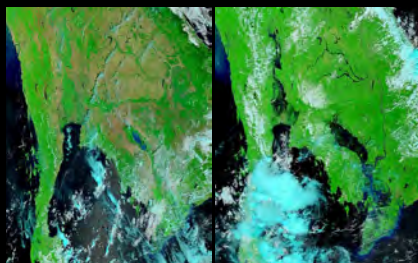
CVO (Mount St. Helens)

See [Chien et al. 2020 JAIS]

<https://ai.jpl.nasa.gov/public/projects/sensorweb/>

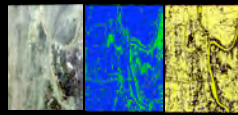
Example: NASA ASE/EO-1 Flooding

- Automated tasking: Thailand Flood Sensorweb
 - Links together space, ground assets
 - Automated Data analysis, triage to generate prioritized requests
 - ASE/EO-1 observation service and others
 - products to stakeholders
 - Fuse data from satellite, ground sensor, and model sources
- +100% temporal coverage for 2010-2011, 2011-2012 Flooding Seasons**

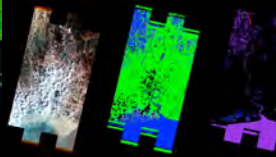


Dry: March 6, 2011

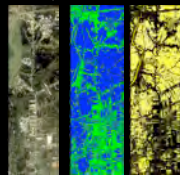
Flooded: October 27, 2011



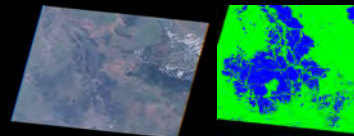
GeoEye-1



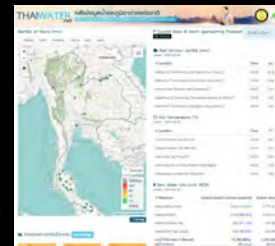
EO-1/ALI



Ikonos



Landsat-7 ETM



In-situ data and Model

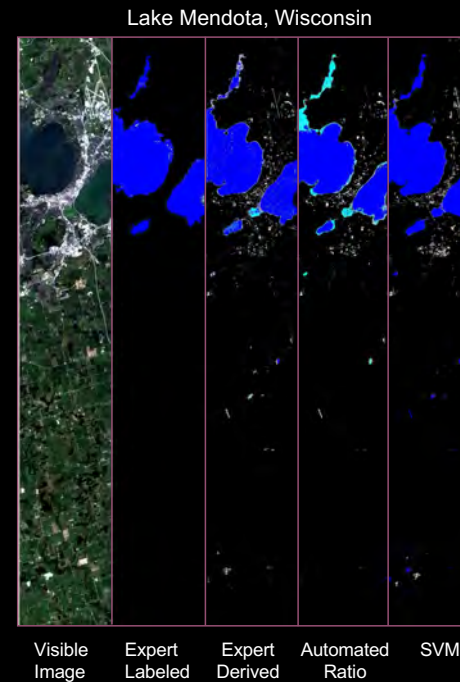
Partners:
 HAI (Thailand)
 Digital Globe (Worldview)
 Geo-Eye
 Radarsat
 Landsat
 LANCE-MODIS

See [Chien et al. 2011 IGARSS, 2013 JSTARS].
 See also Wildfires [Chien et al. 2011 JSTARS, Chien et al. JAIS 2018]

Land, Ice, Water, Snow Detection using Support Vector Machines

- Primary Purpose
 - Identify areas of land cover (land, ice, water, snow) in a scene
- Three algorithms:
 - Scientist manually derived
 - Automatic best ratio
 - Support Vector Machine (SVM)

Classifier	Expert Derived	Automated Ratio	SVM
cloud	45.7%	43.7%	58.5%
ice	60.1%	34.3%	80.4%
land	93.6%	94.7%	94.0%
snow	63.5%	90.4%	71.6%
water	84.2%	74.3%	89.1%
unclassified	45.7%		

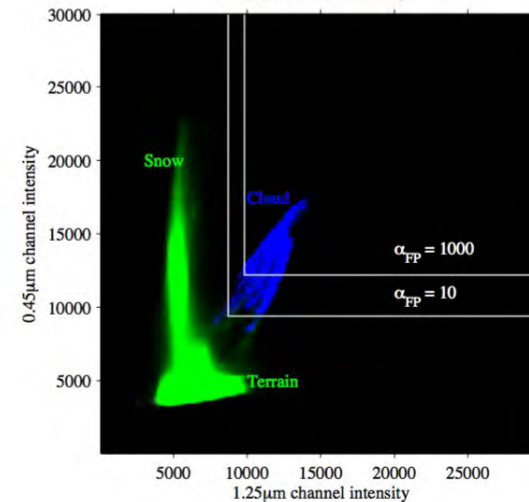
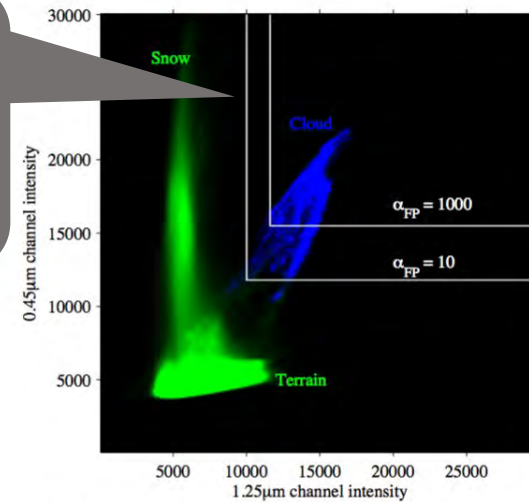


Bayesian Thresholding

Bayesian thresholding exploits the natural division between dark surface materials and bright cloudy regions at particular wavelengths.

- While the RDF method examines a window of values around the pixel to be classified, BT classifies each pixel independently.
- BT was previously employed to analyze data collected by the AVIRIS-C airborne sensor [Thompson et al. 2014].
- For EO- 1, BT used Hyperion bands at 447, 1245, and 1658 nm to span the range from blue to short-wave infrared.

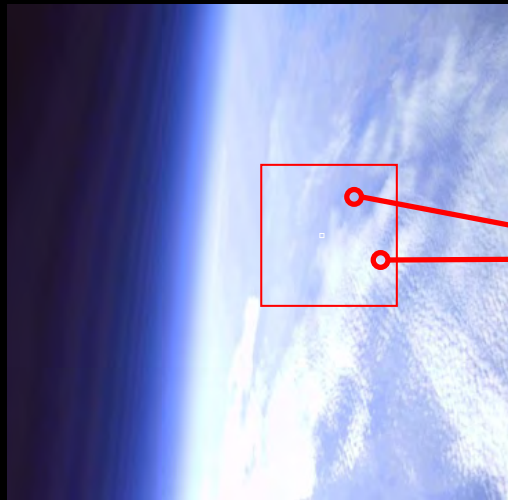
Image courtesy Thompson et al. 2014
TGARS



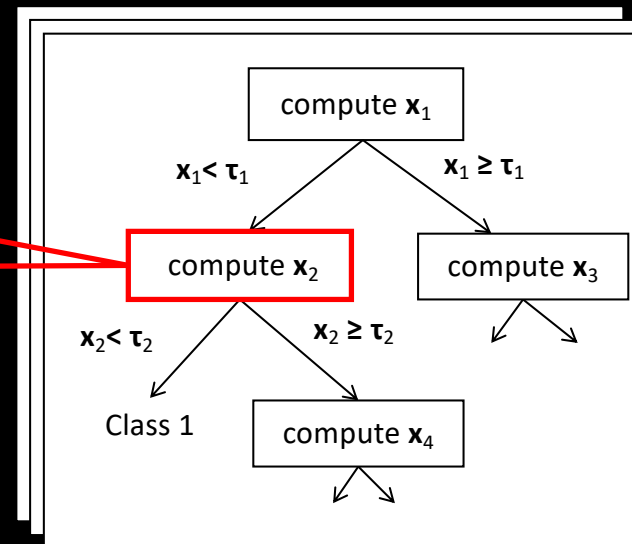
TextureCam – Random Decision Forests

Pixel classification for cloud screening,

Pixel to be classified



Random forest classifier



Onboard Hyperspectral Analysis



Superpixel segmentation

+

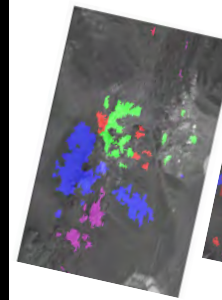
The sequential maximum angle convex cone
(SMACC)
endmember extraction

Results from onboard EO-1 (9/2011)

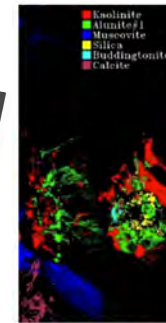
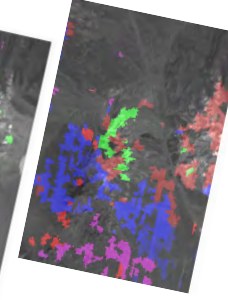
D. Thompson et al. 2012 TGARS

Repeatability: maps

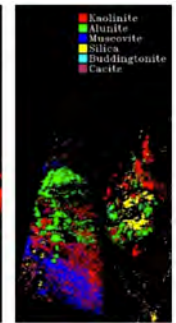
EO-1 Onboard
Sept. 21, 2011



EO-1 Onboard
Sept. 27, 2011

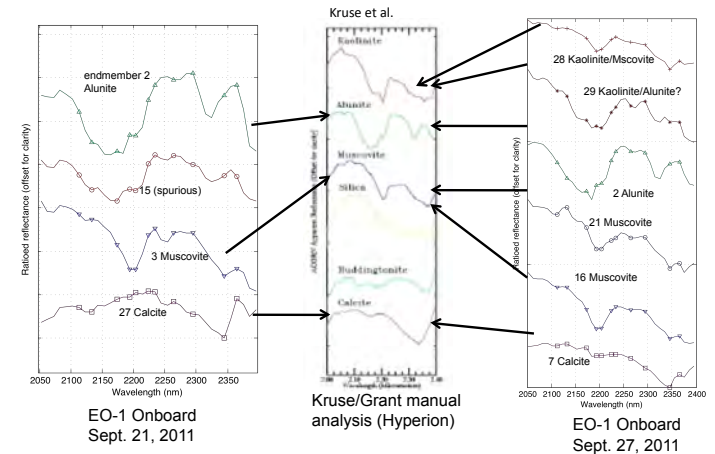


Kruse/Grant
manual analysis
(AVIRIS)

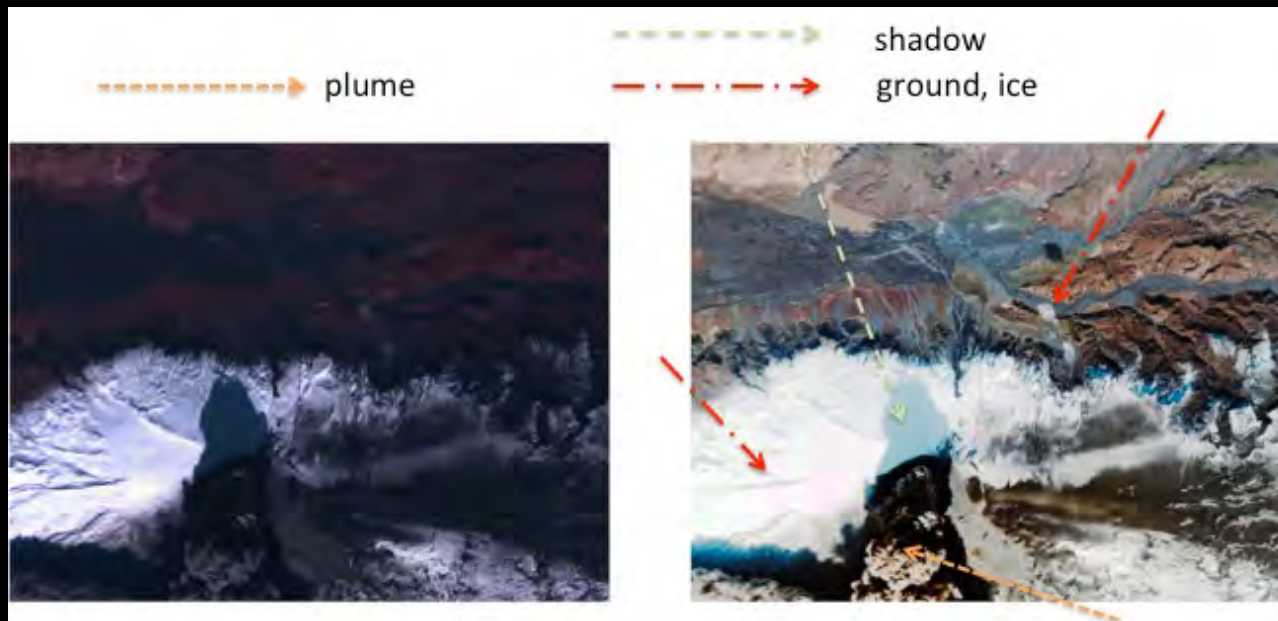


Kruse/Grant
manual analysis
(Hyperion)

Repeatability: detections



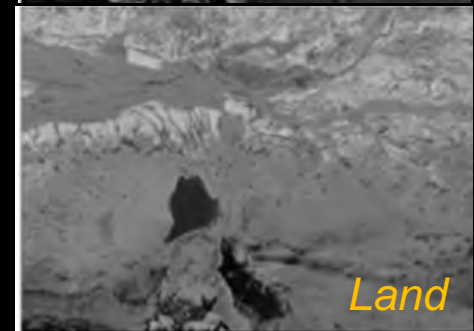
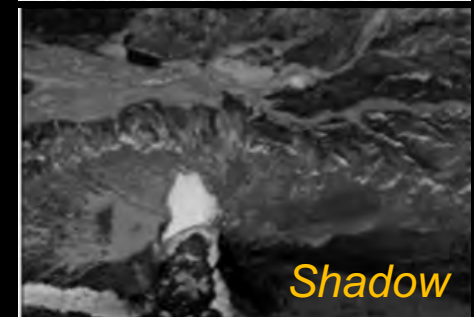
WorldView-2 Data



WorldView-2 Image of Eyjafjallajökull eruption, acquired April 17, 2010

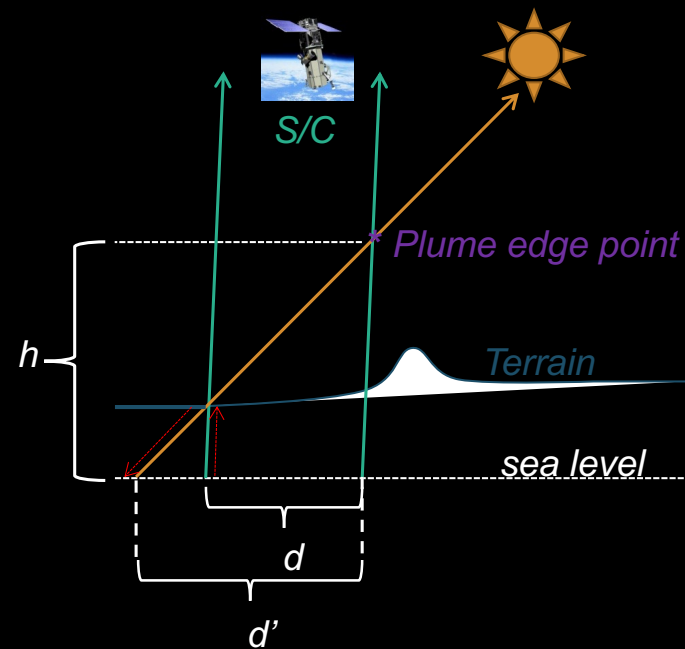
Histogram-equalized image

Mclaren et al. 2012, SPIE



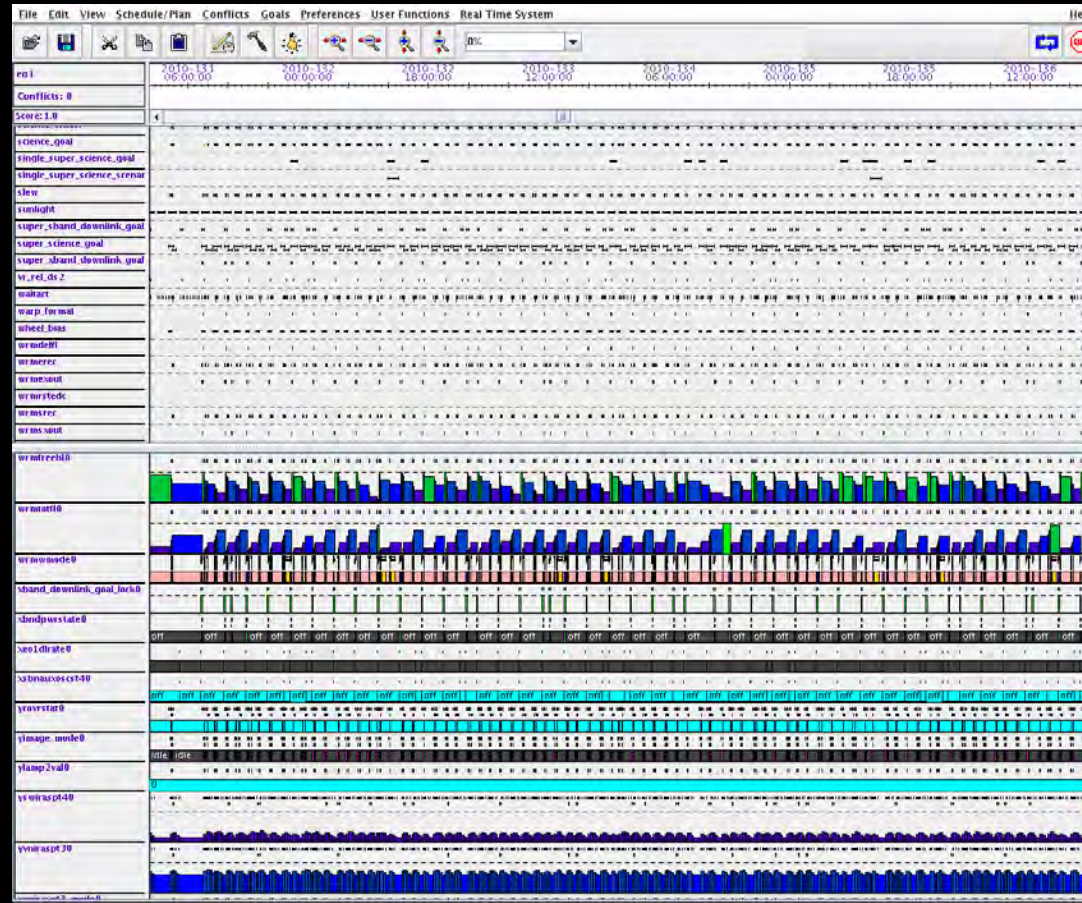
Height Estimation

- Estimate plume height from shadows
- Followed calculations derived in A. J. Prata and I. F. Grant, "Determination of mass loadings and plume heights of volcanic ash clouds from satellite data"
- Rotated classification maps so sun rays are coming from $-Y$ axis (bottom of the image)
- Collected shadow line segments which have a neighboring plume region in sunward direction
- Corrected shadow lengths for:
 - Sun and spacecraft azimuth, elevation
 - Ground elevation at shadow edge
 - ASTER GDEM2 DEM
 - 30m horiz. spacing, 1m vert.



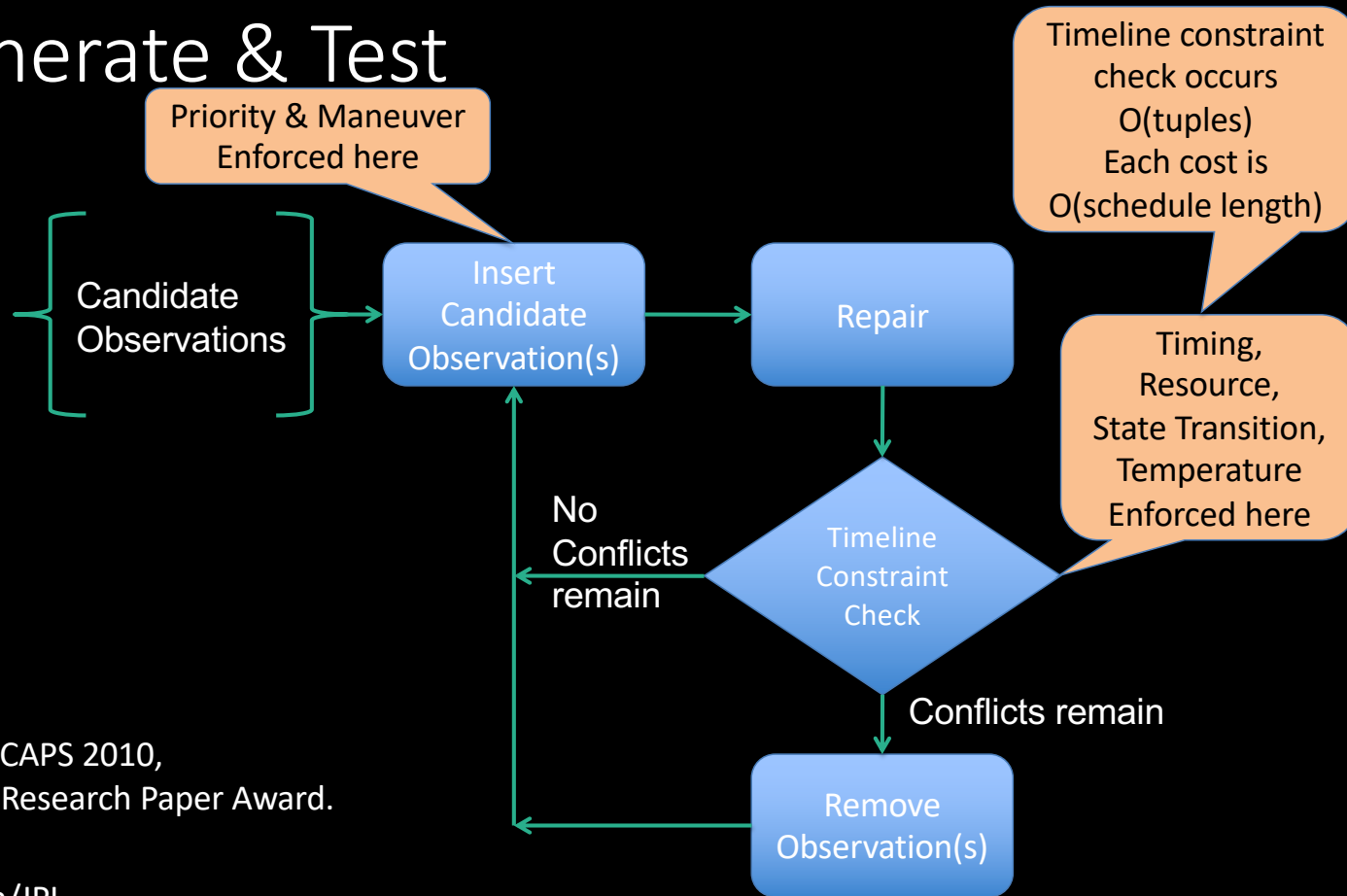
d : Initial shadow length
 d' : Shadow length after projecting up to DEM & down along sun vector
 h : Plume point height

Timeline-based Scheduling



Many timeline based schedulers in use for space missions.
See
[Chien et al. 2012, SpaceOps]

Generate & Test




See:
Chien et al., ICAPS 2010,
Best Applied Research Paper Award.

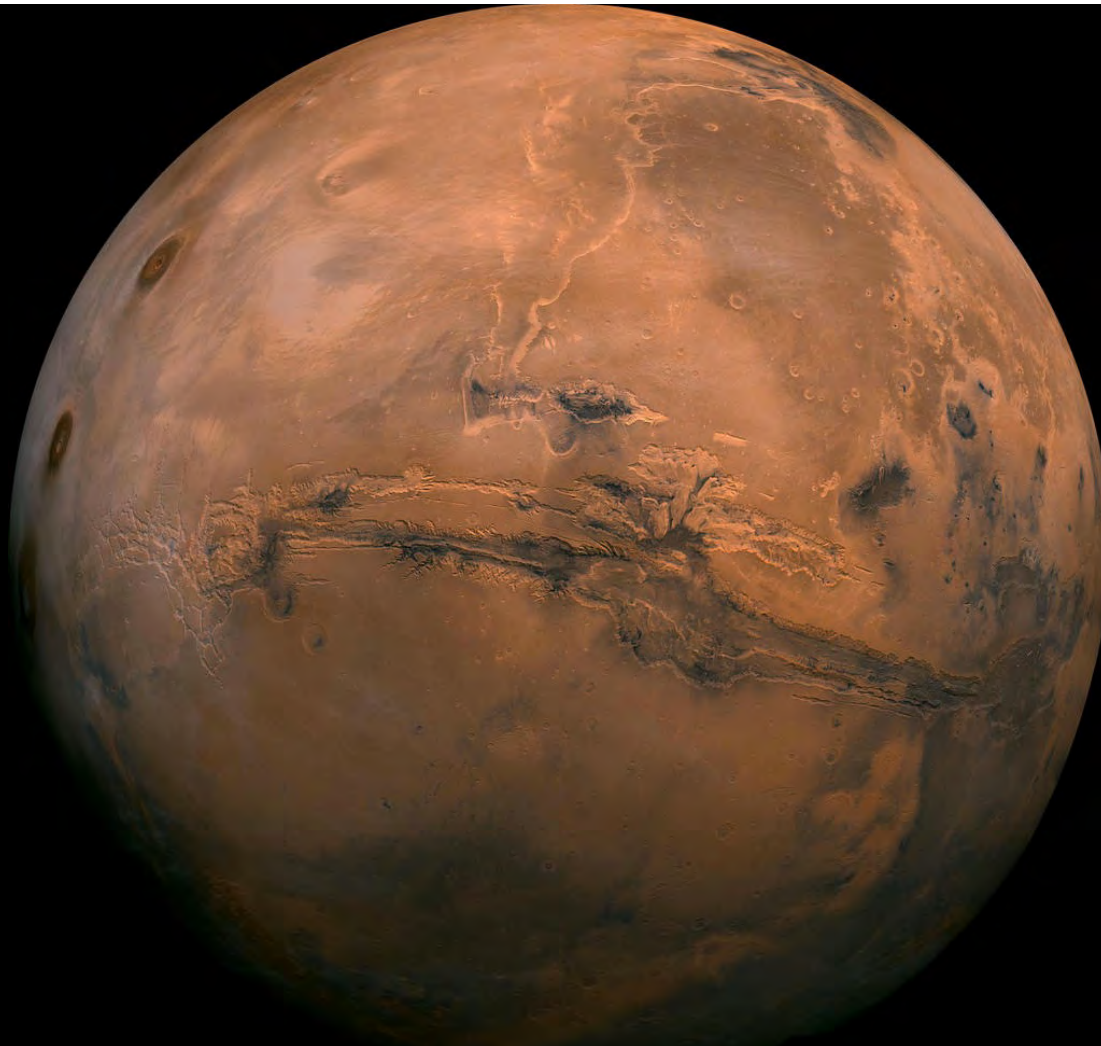
POC: S. Chien/JPL

SHARE



Manam Volcano viewed from EO-1 on June 28th 2010  ALAMY

<https://www.wired.com/2017/03/say-farewell-eo-1-nasas-smartest-satellite/>



ars TECHNICA

ARS UNITE

NASA's next Mars rover will use AI to be a better science partner

Experience gleaned from EO-1 satellite will help JPL build science smarts into next rover.

ALEXON BISH | 12/02/16, 12:00 PM



Image 7 The Mars 2020 rover will likely carry artificial intelligence software to help manage the science workload.

32

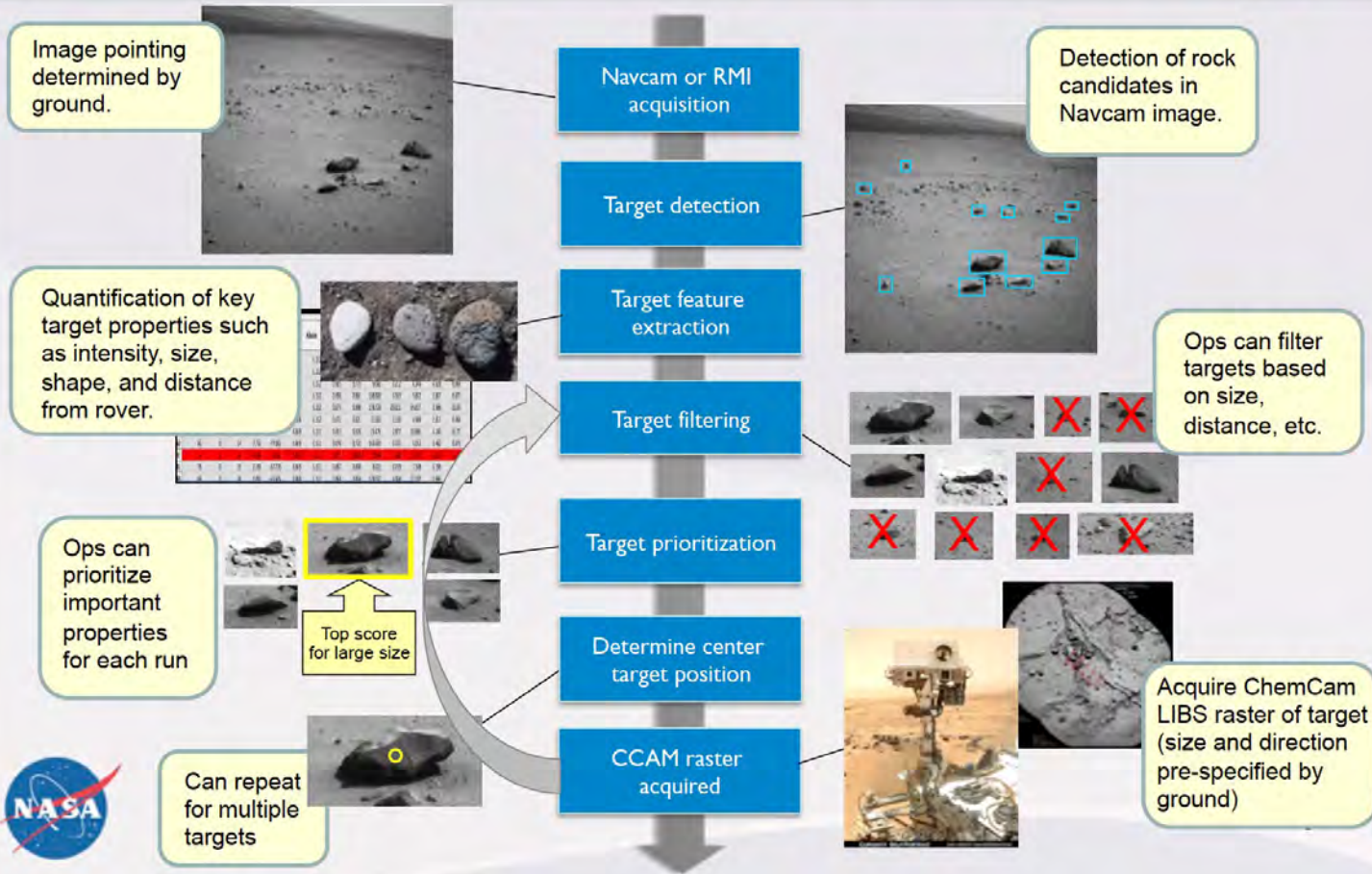
Welcome to Ars UNITE, our week-long virtual conference on the ways that innovation brings unusual pairings together. Each day this week from Wednesday through Friday, we're bringing you a pair of stories about facing the future. Today's focus is on AI in manufacturing and space—stand by to blast off!

NASA can't yet put a scientist on Mars. But in its next rover mission to the Red Planet, NASA's Jet Propulsion Laboratory is hoping to use artificial intelligence to at least put the equivalent of a talented research assistant there. Steve Chen, head of the AI Group at NASA JPL, envisions working with the Mars 2020 Rover "much more like [how] you would interact with a graduate student instead of a rover that you typically have to micromanage."

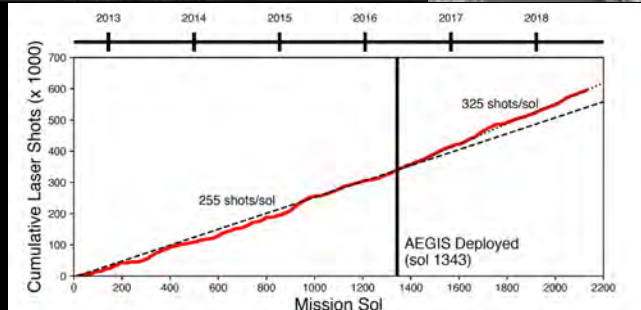
Ars Unite
Why driving is hard—even for AIs
All hail the AI overlord: Smart cities and the AI Internet of Things

Predecisional, for planning and discussion only.

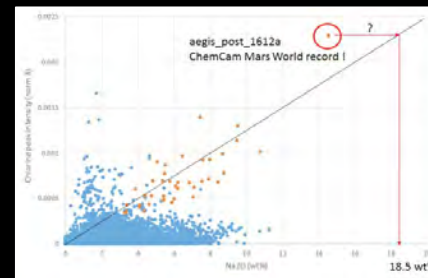




AEGIS Results on MSL Curiosity



Significant increase in rate of data return from ChemCam



Sol 1612: AEGIS found highest concentration of chlorine ever measured by ChemCam on Mars

Targets Selected in First M2020 Run

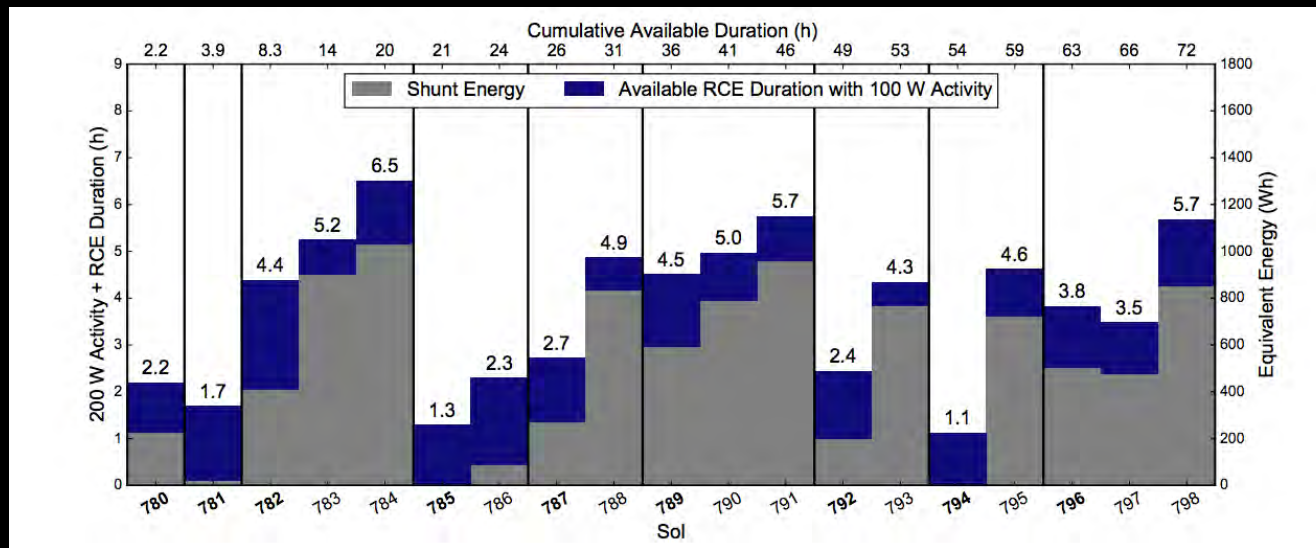


SCAM and high-res Navcam data acquired on above two identified rock targets (Sol 383)

Sol 383 =
March 19, 2022

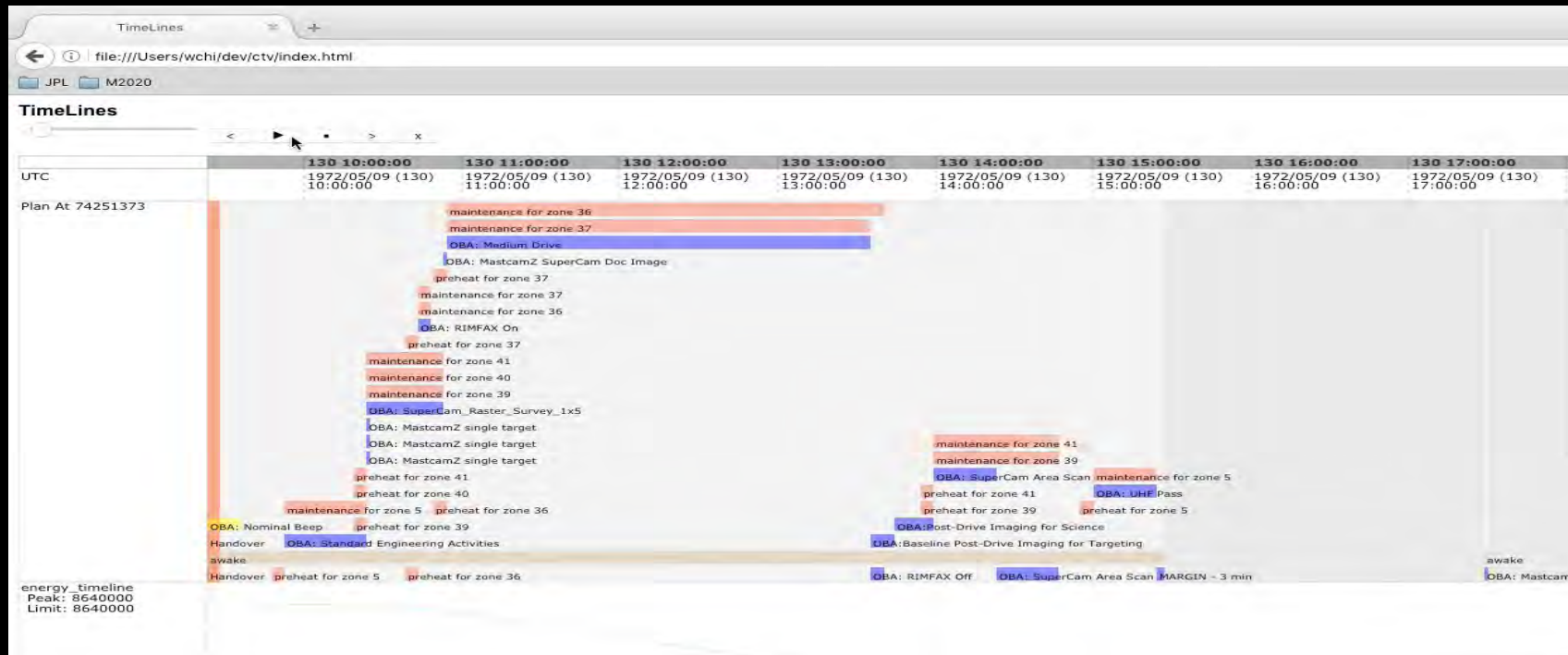
AEGIS POCs: Tara Estlin, Dan Gaines, Raymond Francis, Jet Propulsion Laboratory, California Institute of Technology

MSL unused Time/Energy



MSL submasters on average execution time 28% less than planned time (+ cleanups). [Gaines et al. 2016]

Example M2020 Sol Type:
Medium Drive with post drive imaging.



Rabideau et al. 2017 IWPSS; Chi et al 2018 ICAPS; Chi et al. 2019 ICAPS; Chi et al. 2020 ICAPS

Predecisional, for planning and discussion only.

POC: D. Gaines, S. Kuhn, S. Chien



ESA's Rosetta Orbiter Operations (2014-2016)

Overview

- Ground Automated Science Planning used to develop over 30 Medium Term Plans (MTP) (~ 4 weeks each).
- Ground use of automated buffer scheduling from MTP through execution.

Details

- Critical to handling multiple contingencies around Philae Lander deployment, re-planning 4 weeks of operations within 24h due to Philae updates.
- MTP plans of up to 2000 activities, 2000 slews and pointings, 60 active science campaigns.



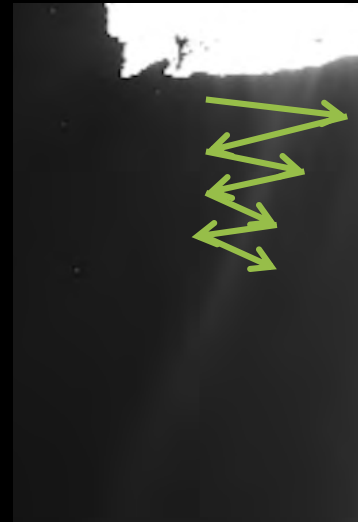
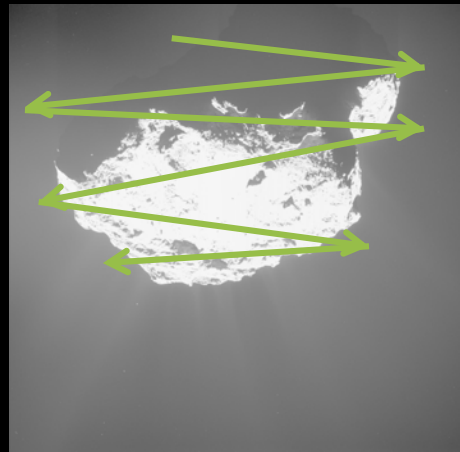
MTP 006 01 Aug – 01 Sep 2014: 32 days, 2027 observations, 2160 pointings and slews, 63 science campaigns, 10,000's constraints checked and over 1400 downlink dumps



MTP 006 01 Aug – 01 Sep 2014: Zoom in on downlink plan for portion of MTP006

See [Chien et al. 2021 JAIS] and [Rabideau et al. 2017 JAIS].

Broad Sweeps vs Targeted Sweeps

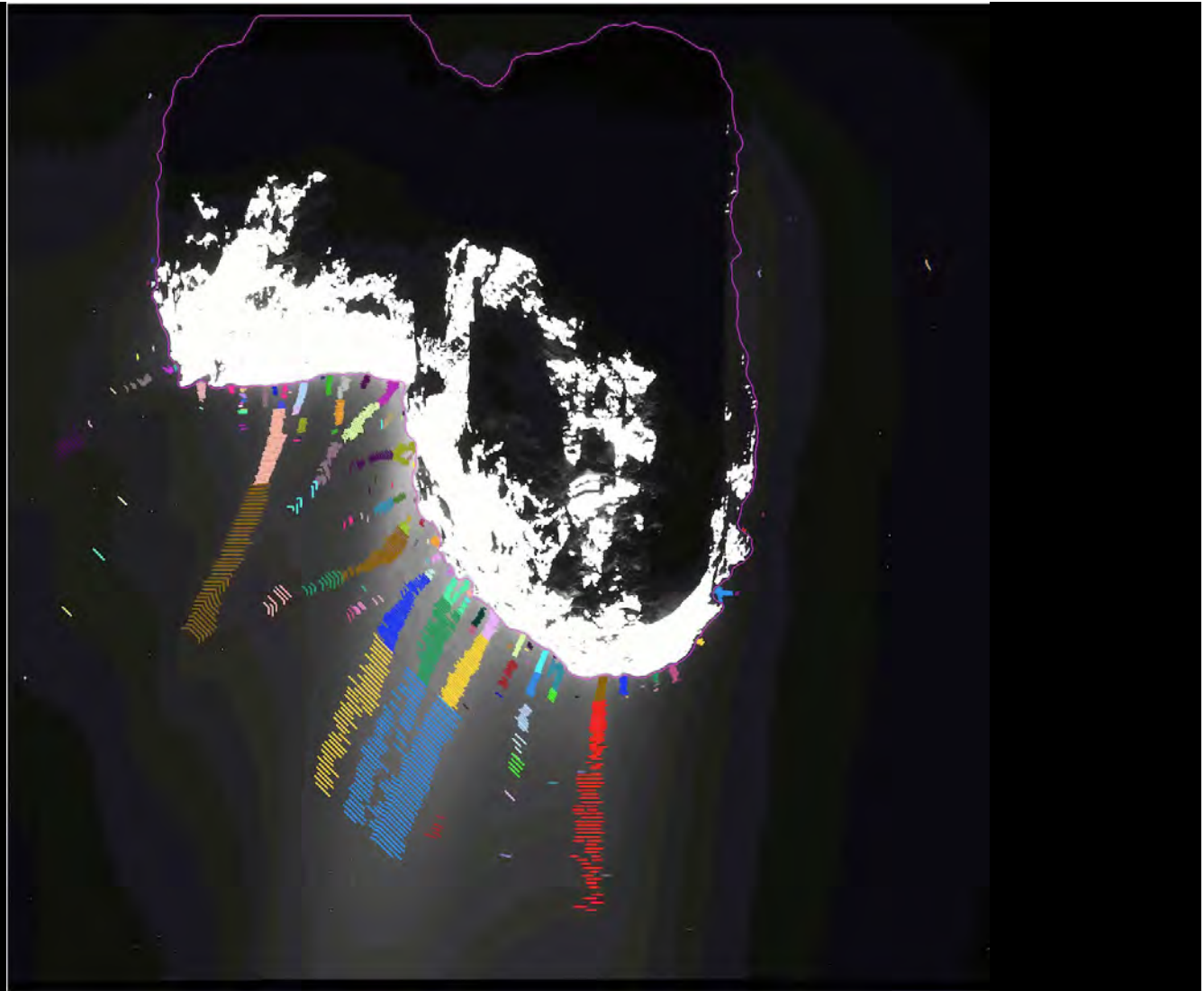


Plume Detection Rosetta OSIRIS

Brown et al. 2019 *Astronomical J.*

Collaboration w. H. Sierks/MPI

Original Image sequence credit:
OSIRIS/MPI, Rosetta/ESA



Coverage Scheduling

- Numerous missions involve variations of coverage scheduling, this technology is mature and in use for several NASA missions
 - ECOSTRESS: coverage, illumination, priority and background mapping, radiation keepouts, data management
 - OCO-3: visibility, illumination, complex geometry, area map prioritization, PMA calibration, complex rapid pointing and flip constraints
 - NISAR: data volume, power/energy, complex coverage campaigns

POC: Chien, Wells, Doubleday

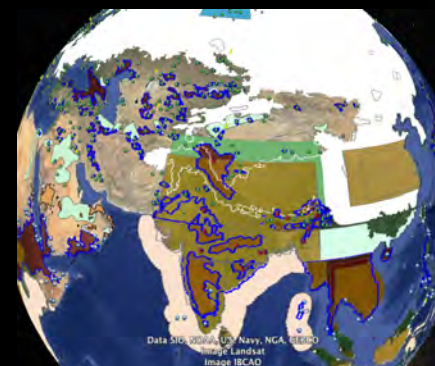
ECOSTRESS

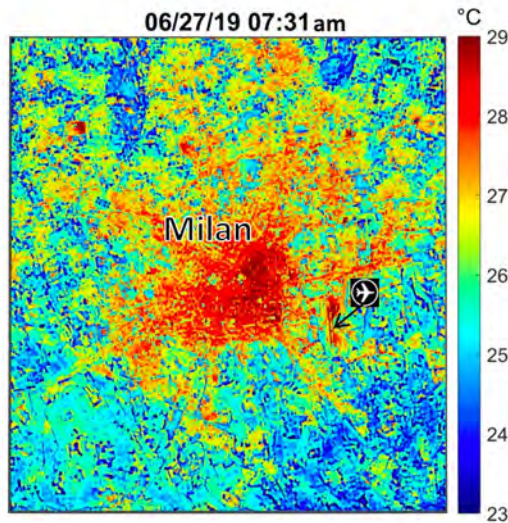
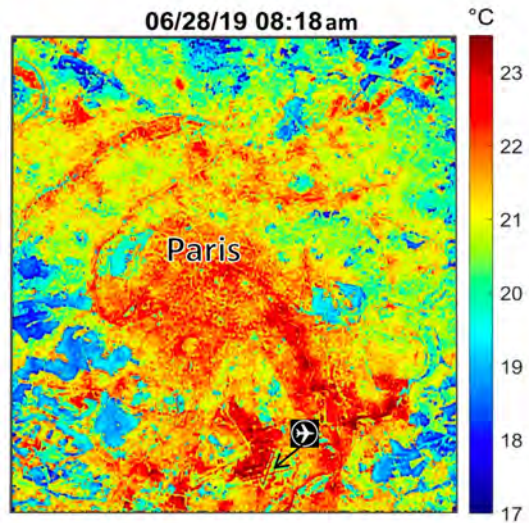
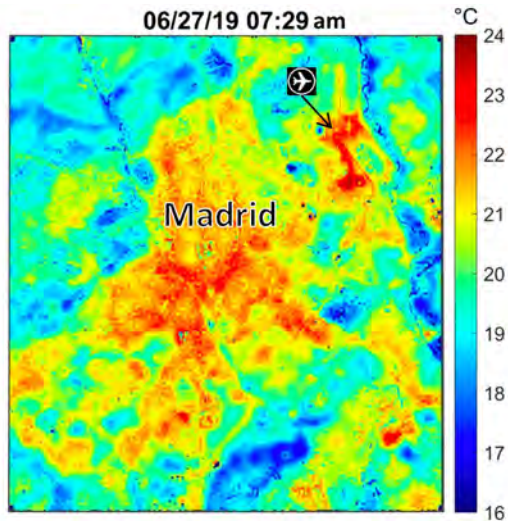
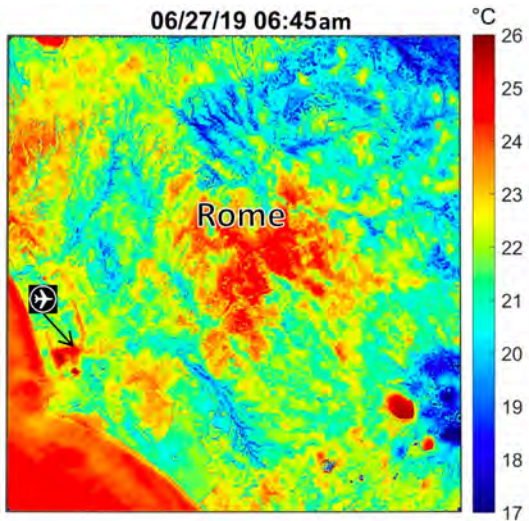


OCO-3



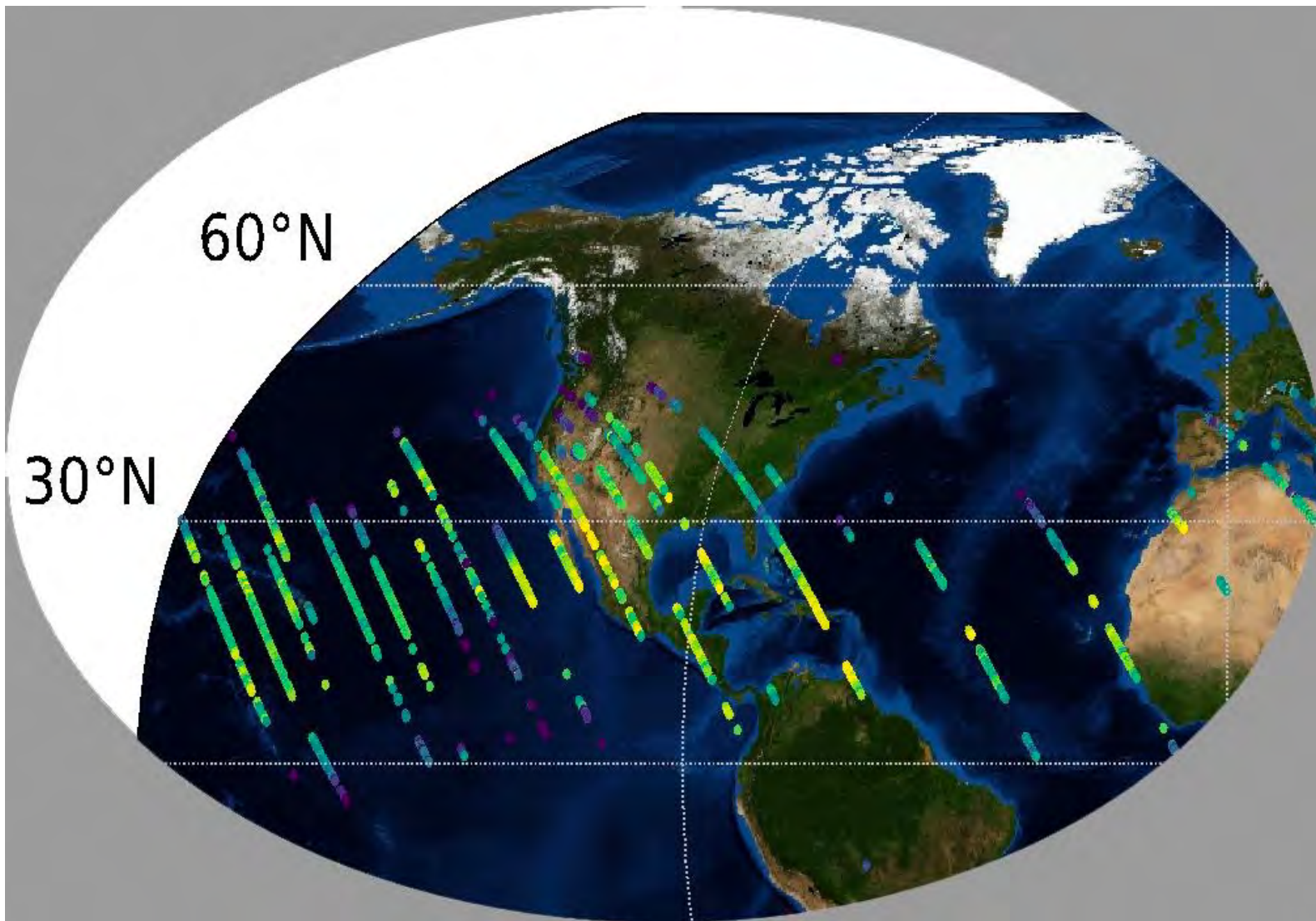
NISAR





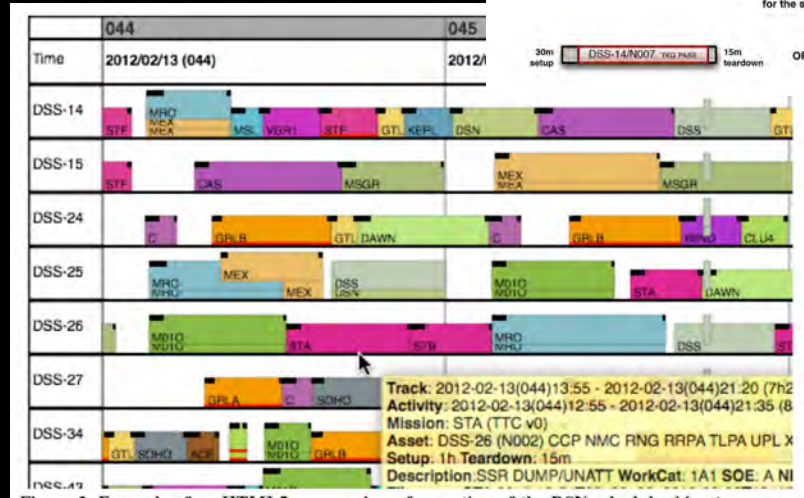
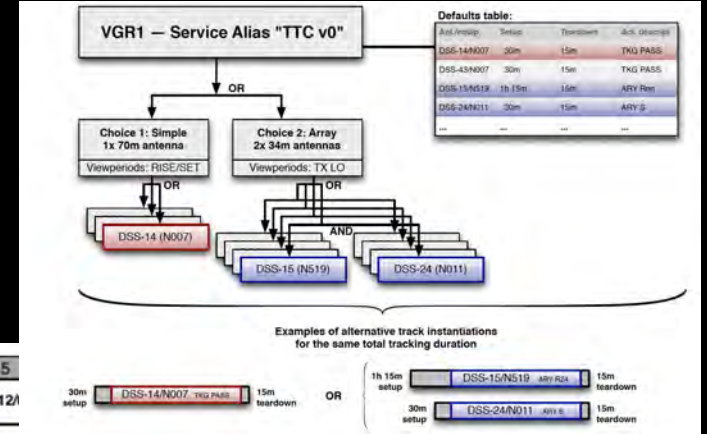
NEWS | JULY 2, 2019
NASA's ECOSTRESS Maps
European Heat Wave From
Space

<https://www.jpl.nasa.gov/news/news.php?feature=7445>



NEWS | JULY 12,
2019
**NASA's Orbiting
Carbon
Observatory-3
Gets First Data**

[https://www.jpl.n
asa.gov/news/ne
ws.php?feature=
7452](https://www.jpl.nasa.gov/news/news.php?feature=7452)

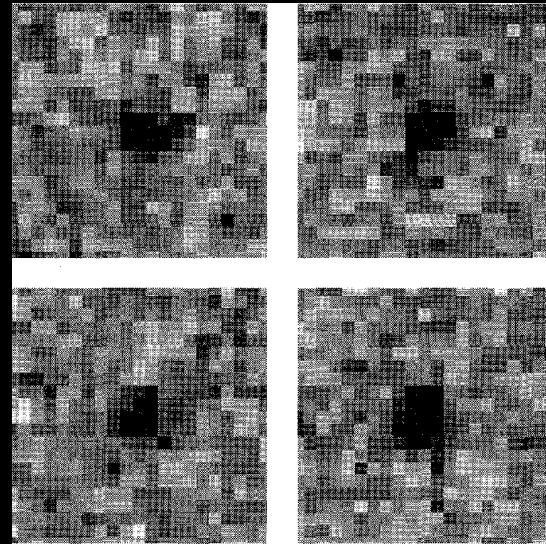


Constraint based scheduling for NASA's Deep Space Network
 Demand forecasting [SpaceOps 2018], Midrange scheduling [AIMAG 2014], near real-time scheduling
 Link complexity based scheduling [SpaceOps 2018]
 POC: M. Johnston

Machine Learning

As far back as 1993 – Sky Image Classification Tool (SkICAT)

- Machine Learning (Decision Trees) used to classify 2nd Palomar Observatory Sky Survey
- Bootstrap from Digital to non digital data allows classification of more faint objects
- > 90% accuracy
- See Fayyad et al. 1993 ICML and Weir et al. 1995 Astronomical Society of the Pacific.
- But more recently...

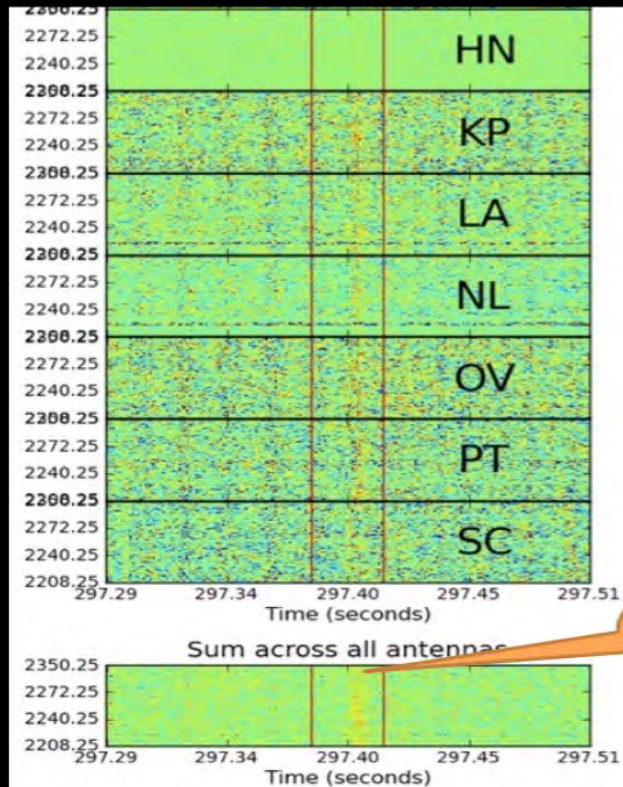




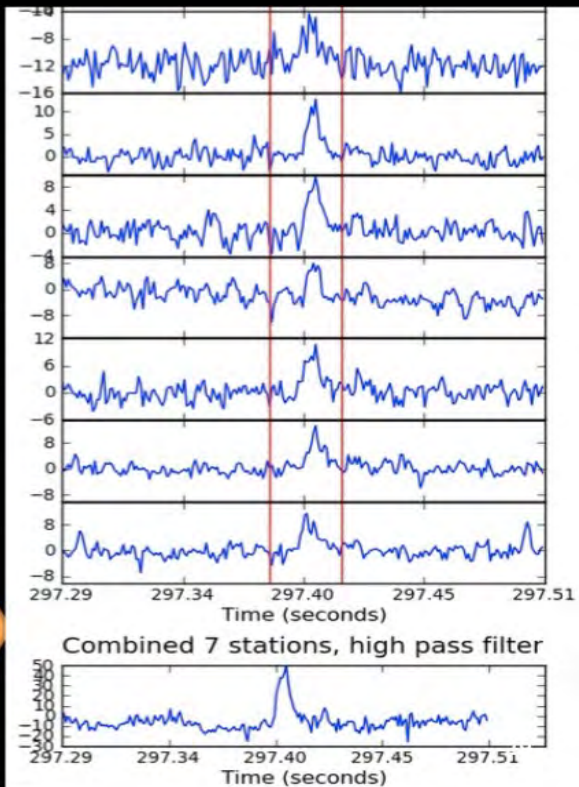
Machine learning is used for optical astronomy to **triage millions of optical events nightly** for the Intermediate Palomar Transient Factory (i-PTF) [Waszczak et al. 2017, Masci et al. 2017] and the Zwicky Transient Facility (ZTF) [Mahabal et al. 2019, Masci et al. 2019].

Science (left) and reference (center) images used with Machine Learning to find true astronomical transient and variable objects (right).

POC: U. Rebbapragada/JPL



Pulsar!



Machine Learning for Automated Triage/classification of Radio Transient Events
 Very Long Baseline Array (VLBA) Fast Transients Experiment (V-FASTR)
 $10^5 \rightarrow 50$ per 24h
 Random Decision Forests, continuous quality control and retraining.

Deep Mars CNN Classification of Mars Imagery for the PDS Imaging Atlas

- MSL Rover data set¹
 - 6,691 labeled images (Mastcam L/R eye, MAHLI)
 - 24 classes
- MRO HiRISE data set²
 - 10,433 labeled images
 - 8 classes
 - Augmentation: rotation, flipping, brightness adjustment

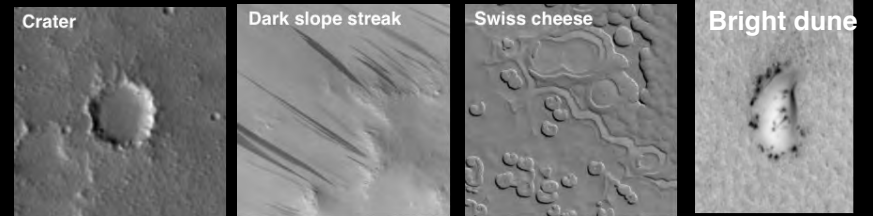
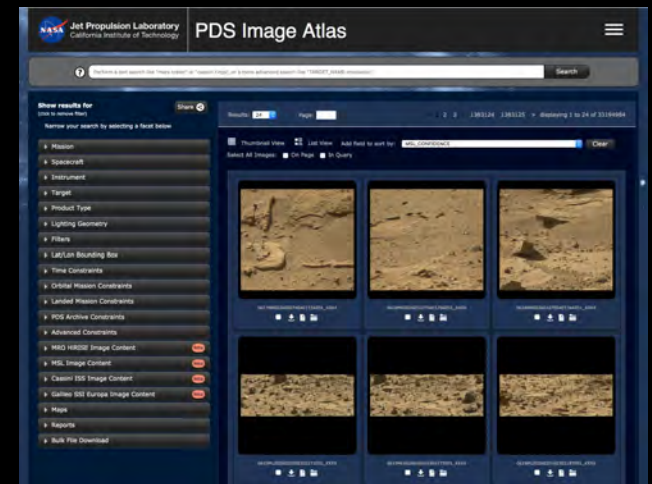


Figure 2. Example images for HiRISENet



Figure 1. Example images for MSLNet



Result: Content indexing for PDS Atlas

Mars Target Encyclopedia

- Lunar and Planetary Science Conference

- Three years
- 5,920 documents
- 2-page abstracts
- 7.2M words



Entities

Find
Elements,
Minerals,
Targets

Relations

Classify pairs
of Target +
(Element or Mineral)

MTE
Database

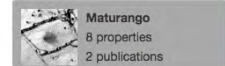
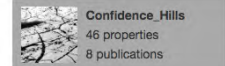
Wagstaff et al. 2018 IAAI

Mars Target Encyclopedia

Compositional information from publications about MSL ChemCam surface targets
Publications currently indexed: abstracts from LPSC 2015 and 2016

hematite

9 targets found

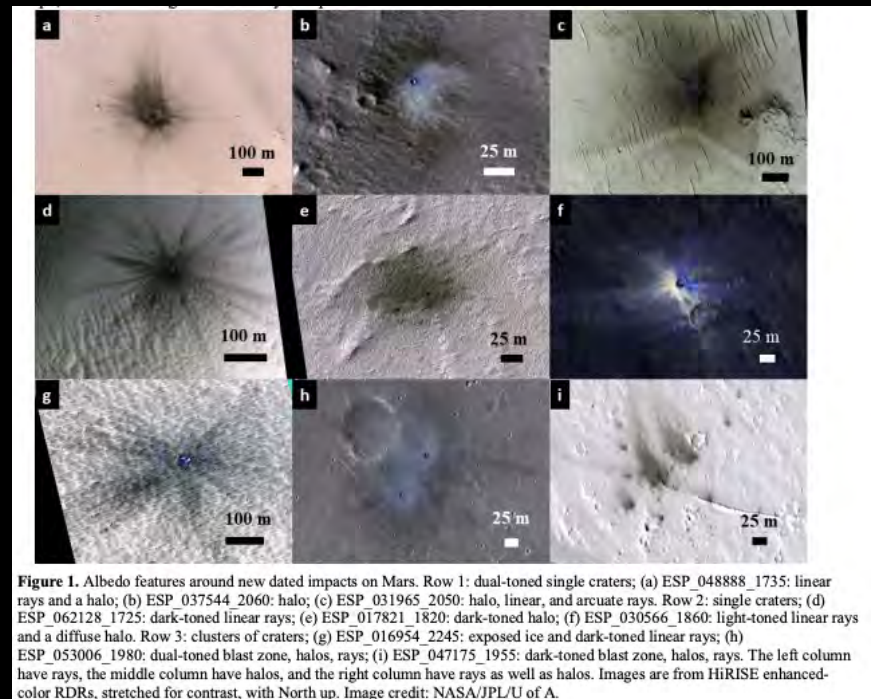


MMGIS

login



Deep Learning for detection of fresh impact craters at Mars



See Daubar et al. JGR Planets 2022
Wagstaff et al. Icarus 2022.

AI in Space: The Future!

CADRE (2024 Launch)

- CADRE is a flight technology demonstration manifested as a payload on CP11 (CLPS)/ Intuitive Machines (IM-3) mission, targeting launch Apr 15th, 2024 on Falcon-9.
- 4 lunar rovers autonomously execute coordinated measurements
- Flight system uses Mexec onboard planner as “Strategic Planner” on leader with Mexec exec on each of 3 rovers

<https://ai.jpl.nasa.gov/public/projects/cadre/>

POC: JP de la Croix, PI, JPL
CL#22-6066



Europa Lander - Autonomy



On the Matanuska Glacier, AK July 2022

- Mission Concept Challenges:
 - Limited energy + radiation
→ limited lifetime
 - Large communication blackouts with Earth (> 42 out of every 85 hours).
 - Unprecedented level of model uncertainty.
- Autonomy required for Mission Success
 - Decision theoretic Approach
Utility and Probability

CL#22-5344

POC: S. Chien

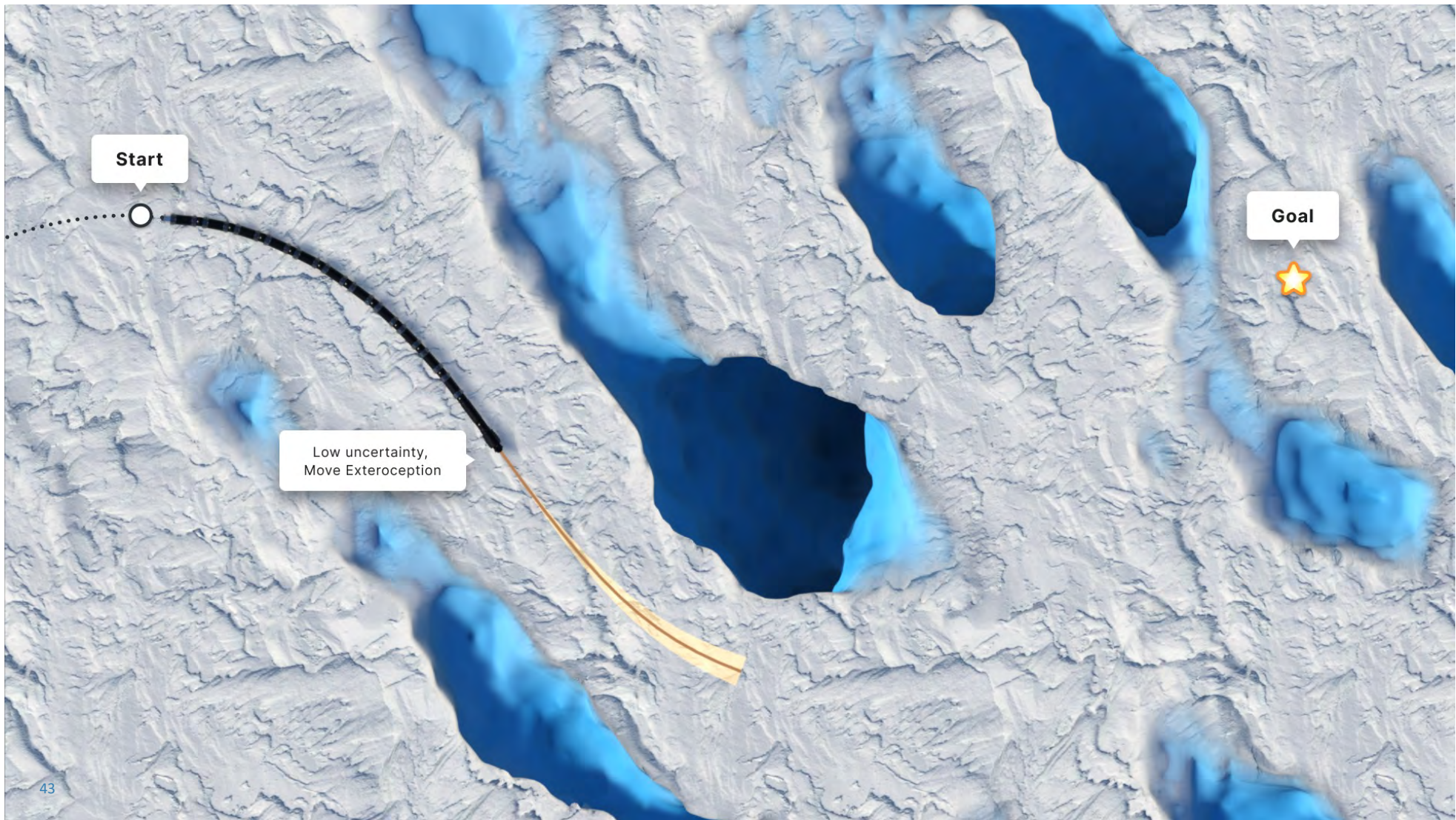
40

<https://ai.jpl.nasa.gov/public/projects/europa-lander/>



Exobiology Extant Life Surveyor (EELS) for
Enceladus Exploration Mission Concept

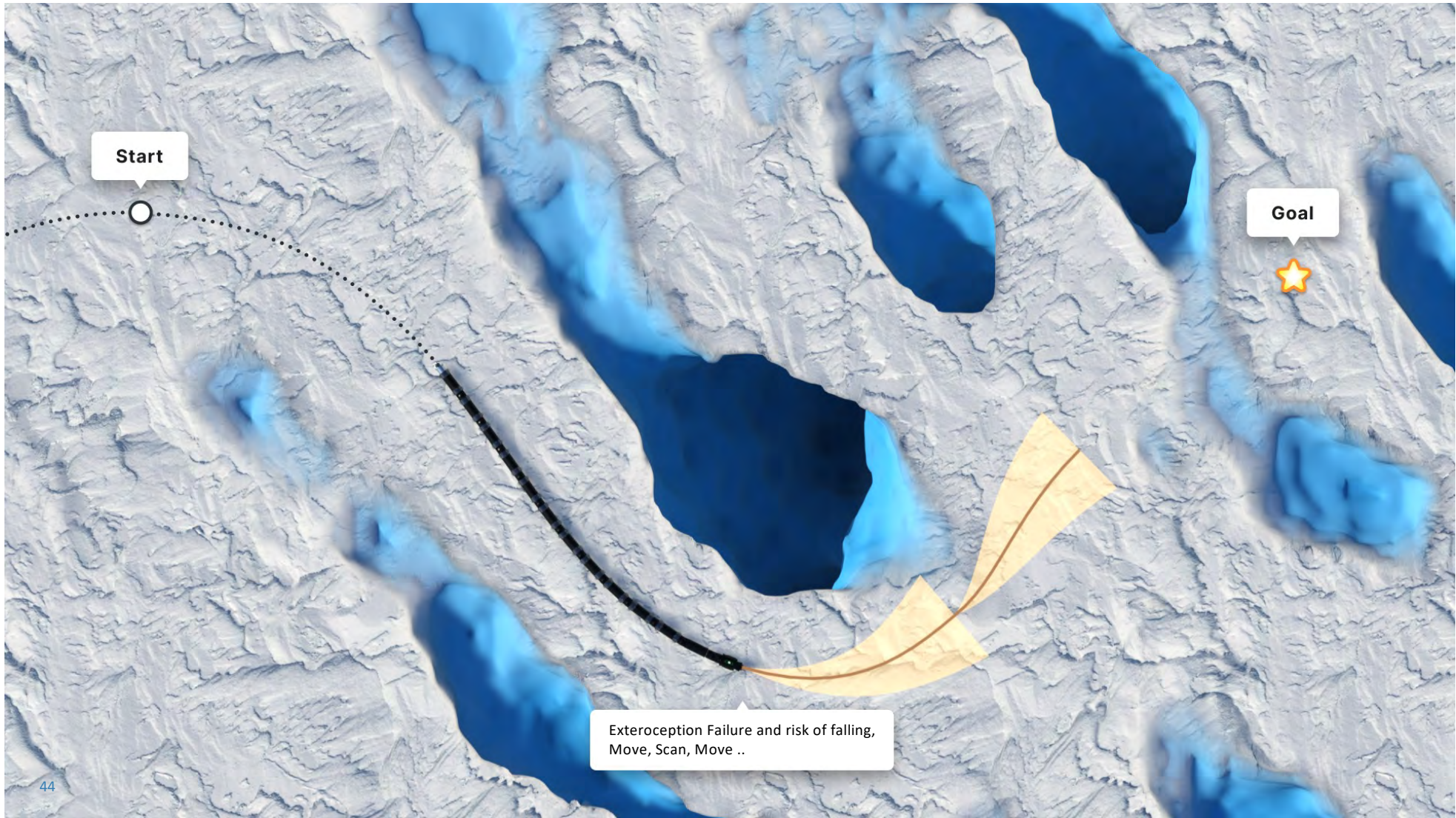
H. Ono PI



Start

Goal

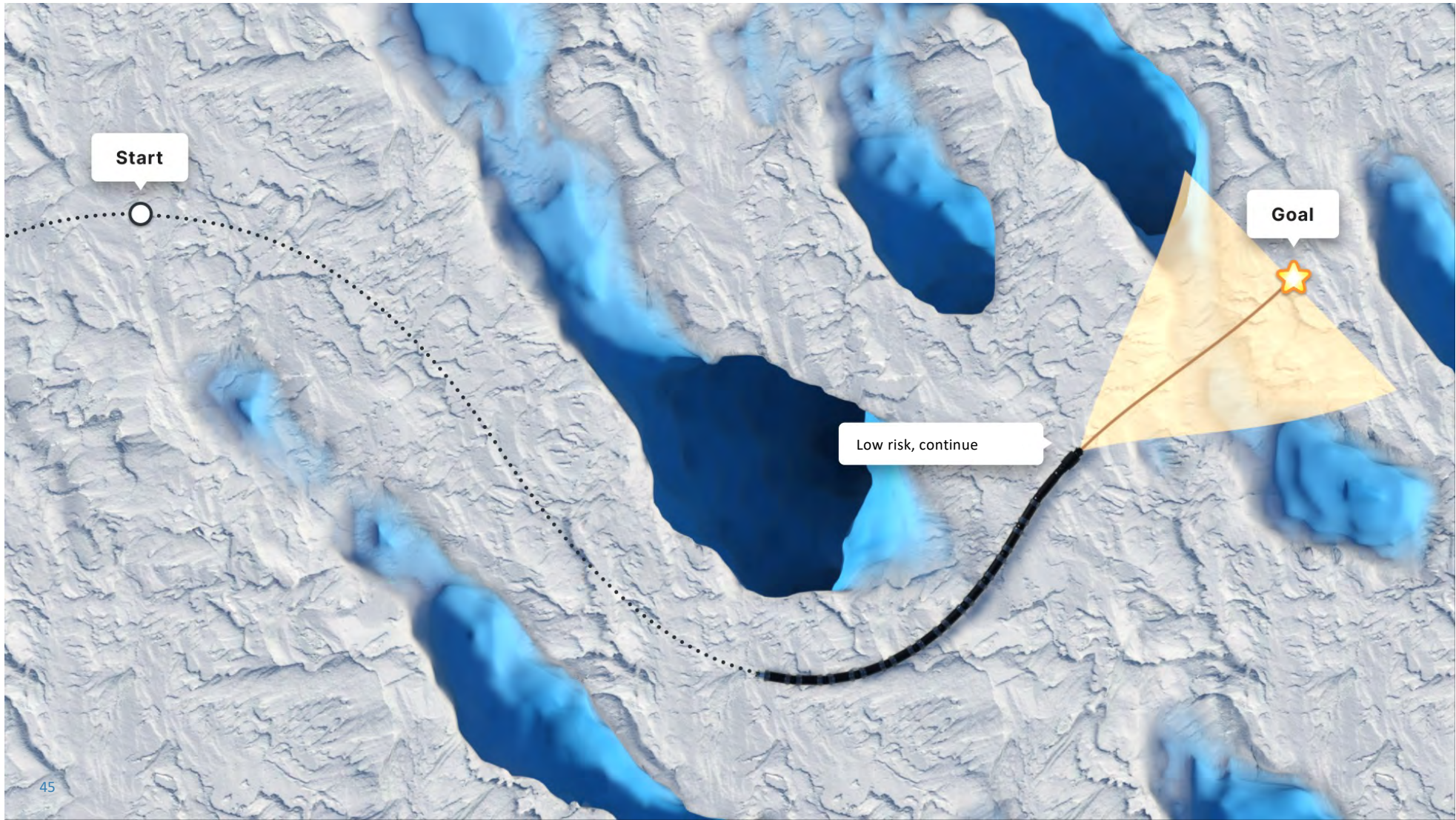
Low uncertainty,
Move Exteroception



Start

Goal

Exteroception Failure and risk of falling,
Move, Scan, Move ..



Start

Goal

Low risk, continue

EELS Surface Mobility

- Screw-based gaits (e.g. leader-follower)
- Shape-based gaits
- Hybrid gaits



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Steve Chien interview: Why NASA is inventing curious AI for deep space

Space probes will be the first to explore the furthest reaches of our solar system and beyond. To make discoveries like finding alien life, they will need to think more like humans, says NASA's Steve Chien



SPACE 20 October 2021

By [Neil Briscoe](#)



Why we need curiosity?

Curiosity is a quality related to inquisitive thinking such as exploration, investigation, and learning -- Wikipedia

Filling in the blanks versus investigation of the *true unknown*.

Lewis and Clark 1804-1806

Pacific Ocean



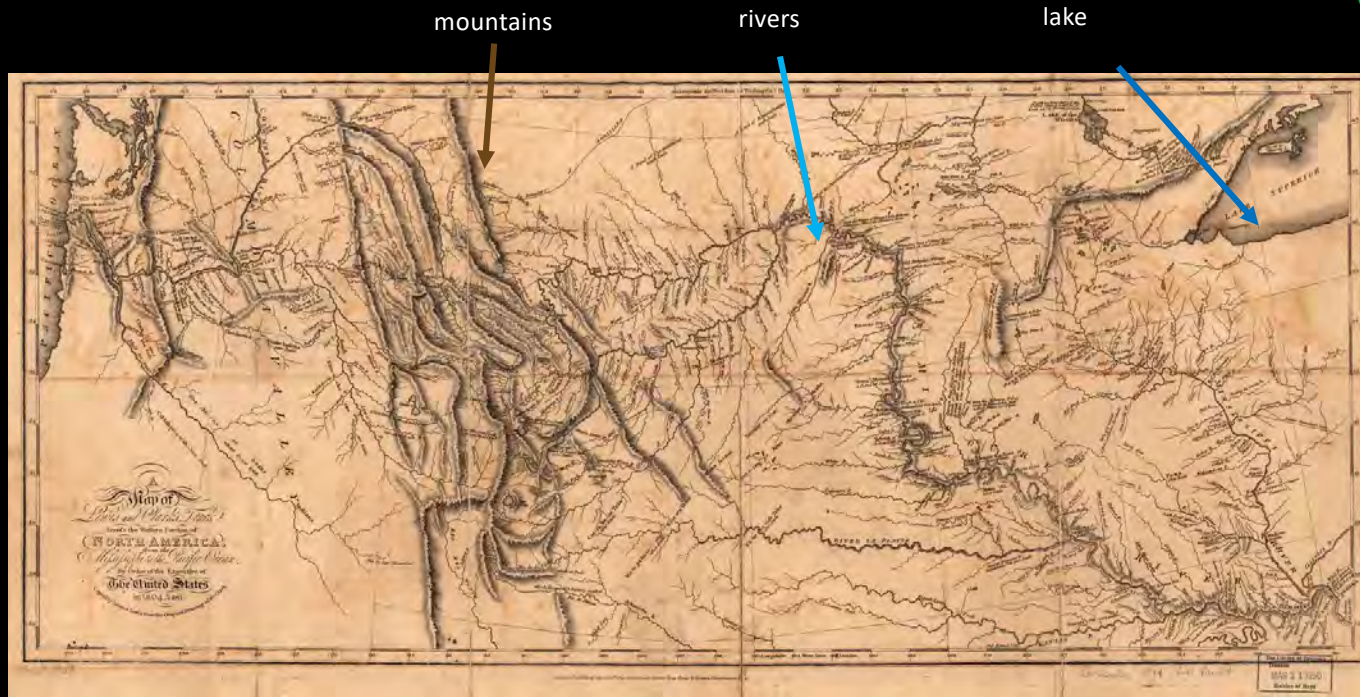
8000 miles Over 2 years

Modern Day St. Louis, MO

Products

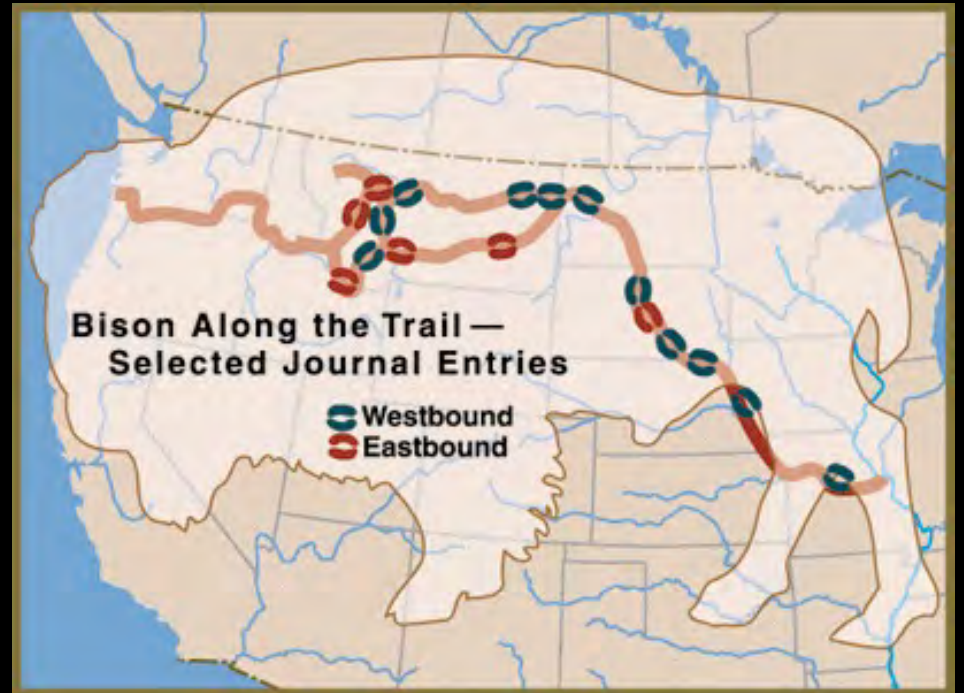
- 2 years
- Maps, 120 animal specimens, catalogued 200 plants
- → 18 4x6" notebooks
- ~ 750 pages
- In contrast modern space mission brings down gigabytes per day
1 GB = 678,000 pages of text *per day*

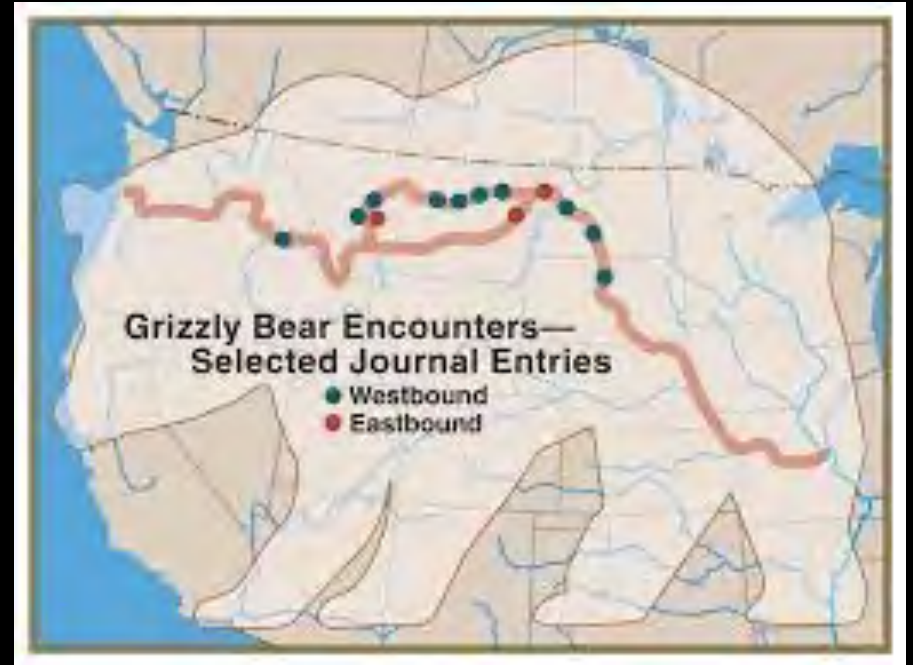
Lewis and Clark 1804-1806

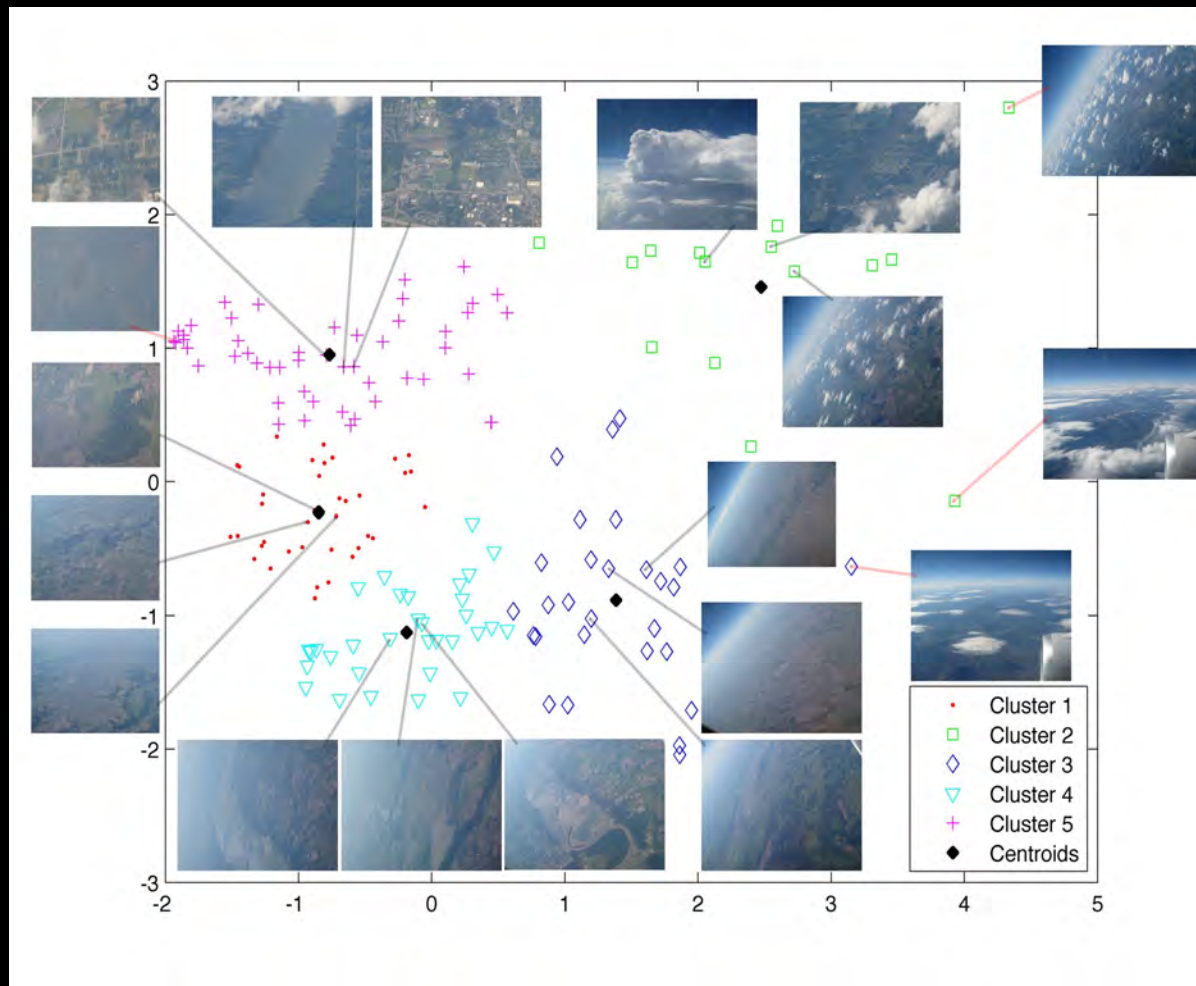


wildlife





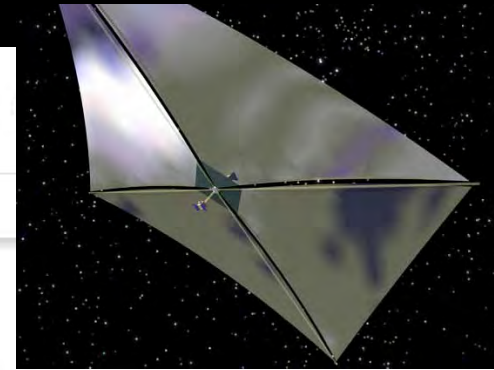




Clustering and Metric Learning of Aerial Imagery [Hayden et al. 2012 ACM TIST]

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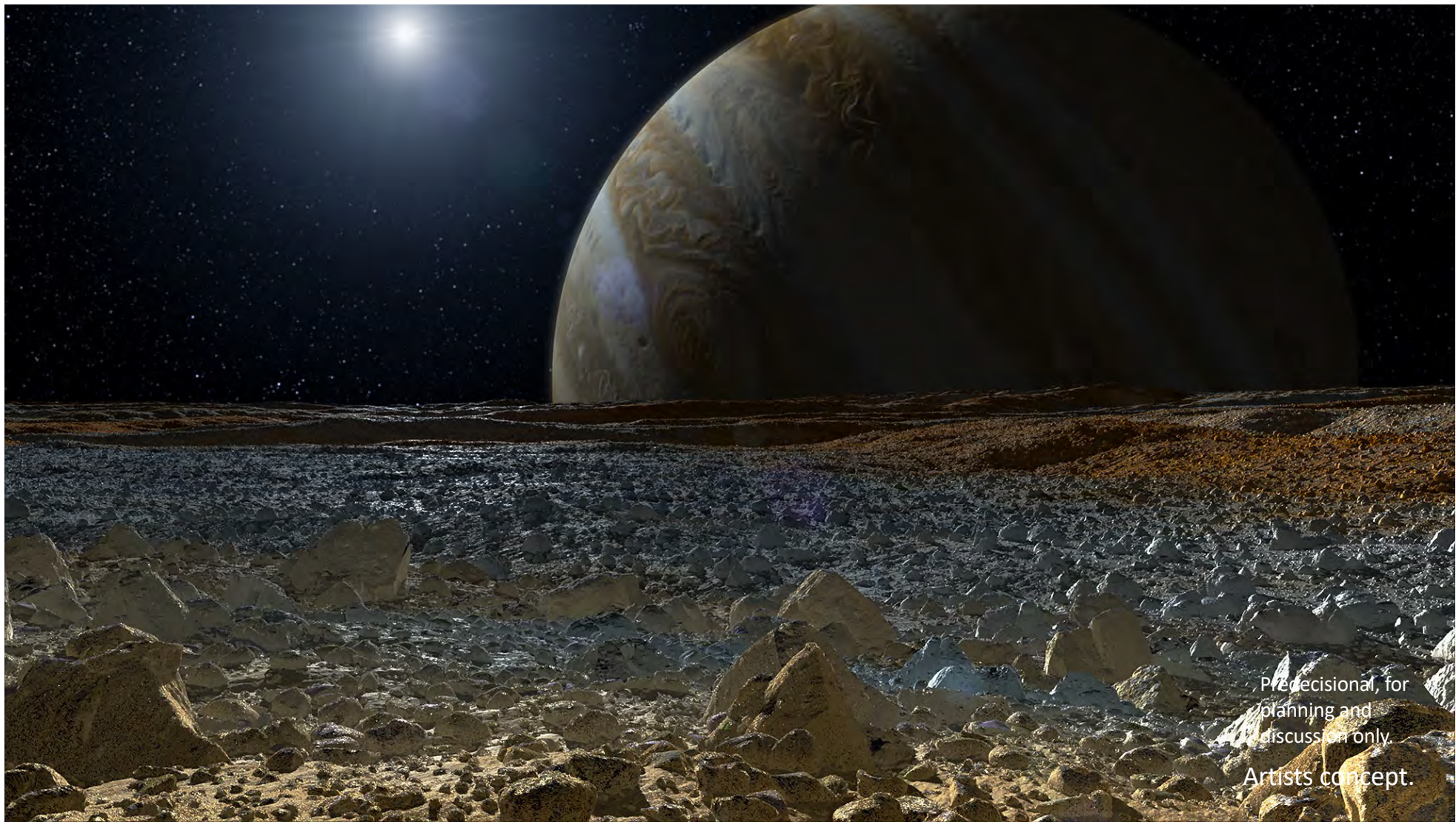
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How NASA's Search for ET Relies on Advanced AI

Jet Propulsion Laboratory's artificial intelligence chief describes the "ultimate" test for AI in space exploration

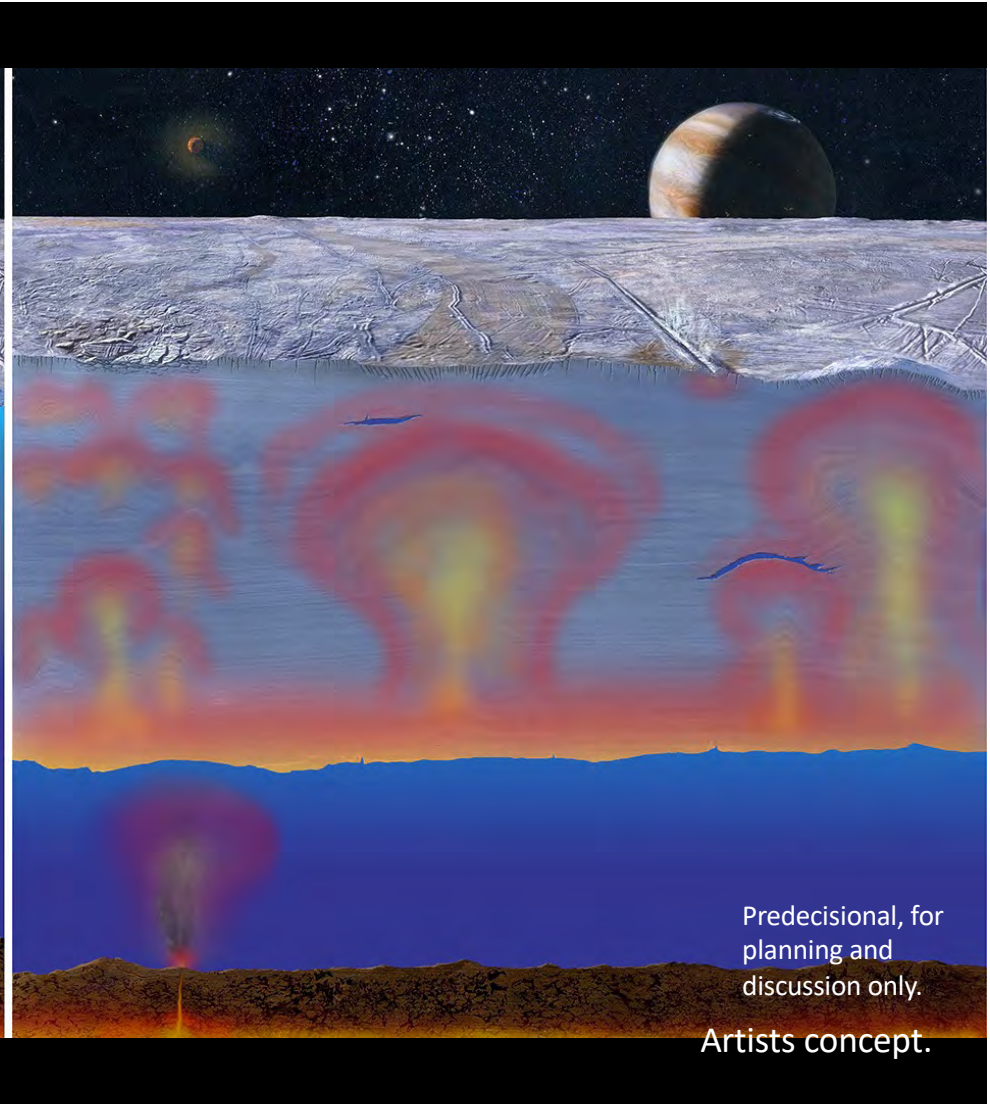
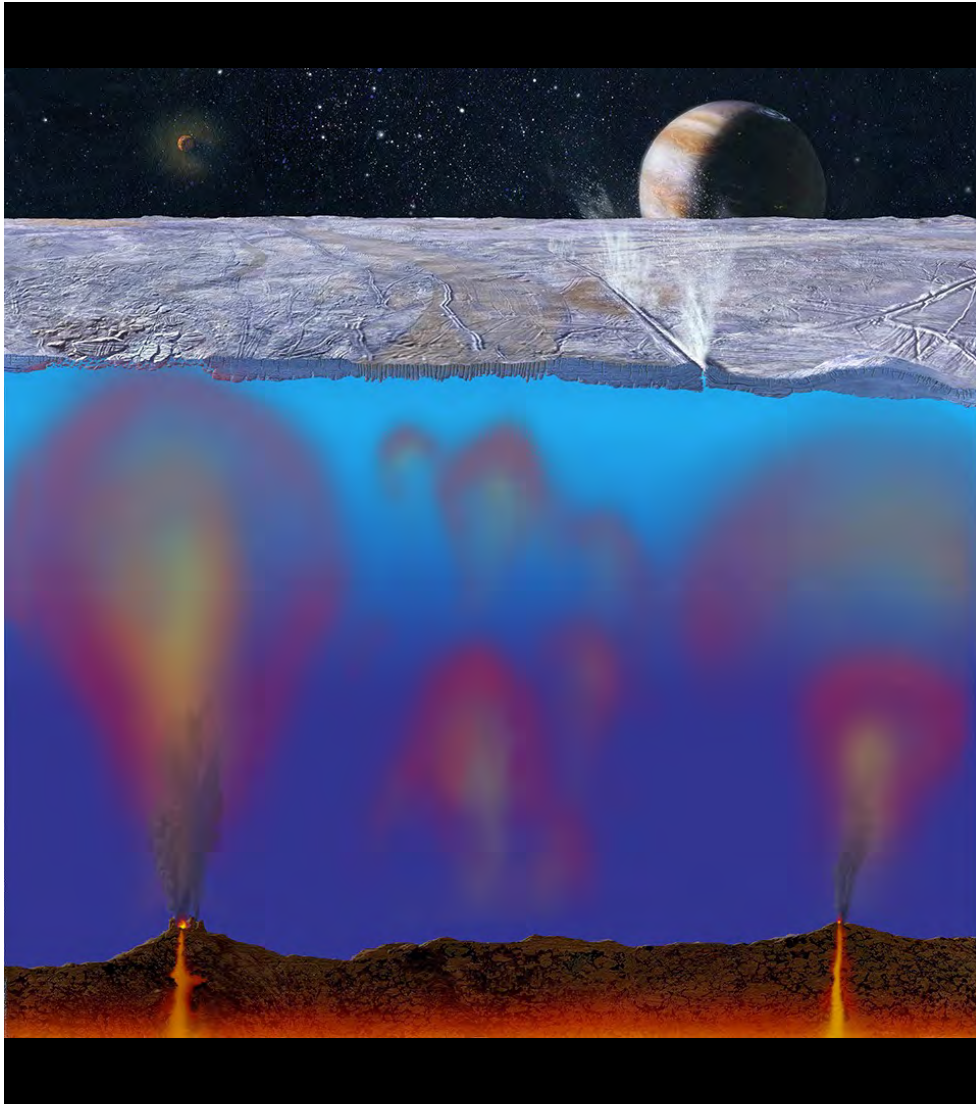
By Larry Greenemeier on December 28, 2017



Predecisional, for
planning and
discussion only.
Artists concept.



Predecisional, for
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Artists concept.



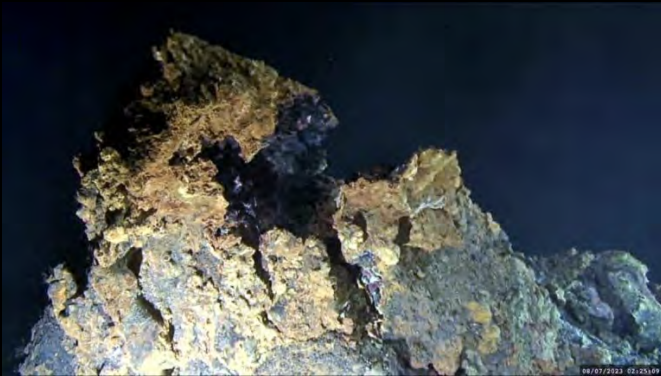
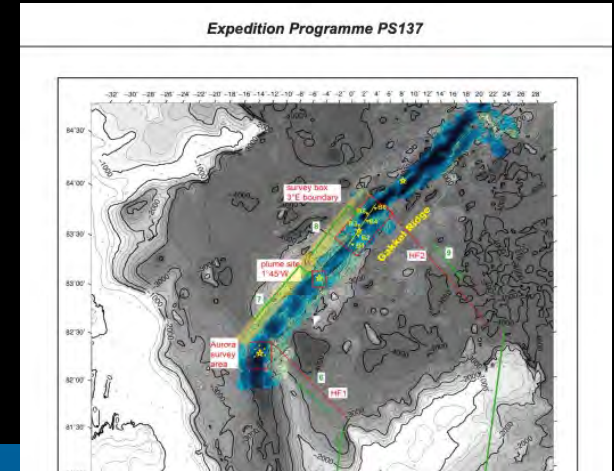
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Ocean
Worlds
submersible
concept.

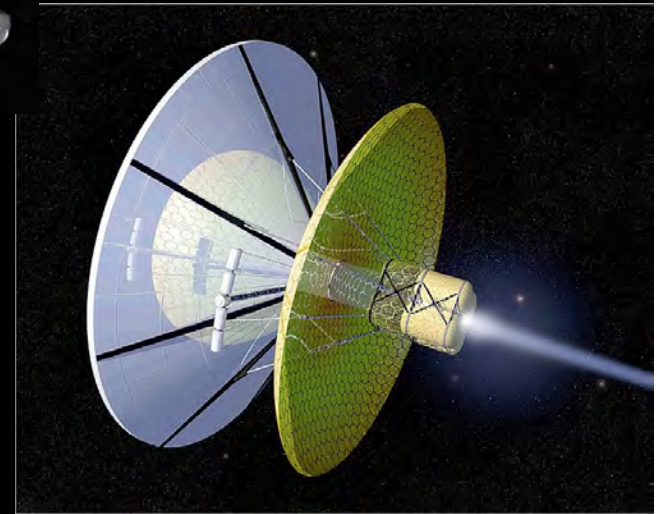
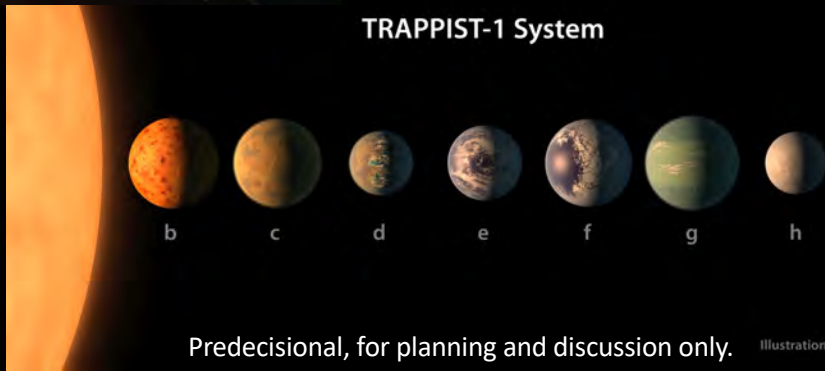
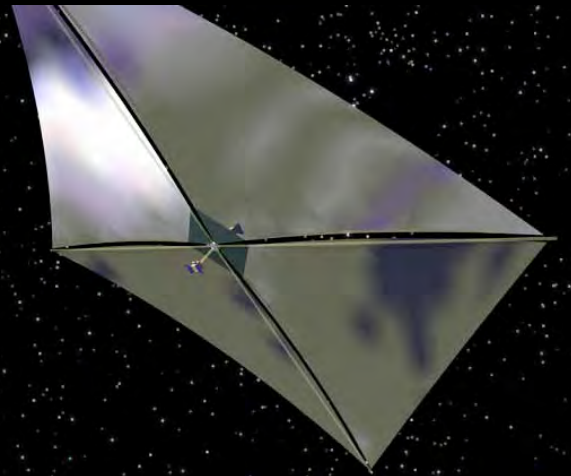
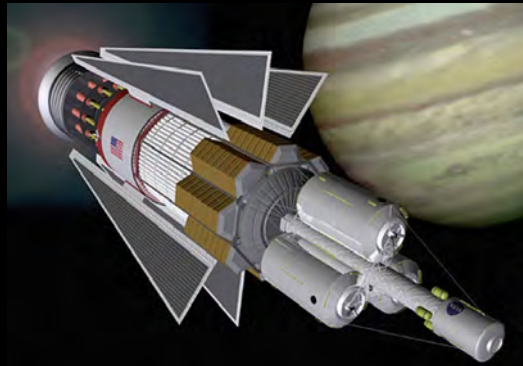
From the Polarstern Cruise PS137 (Alois) Gakkel Ridge



Courtesy PS137 NUI team. © Copyright WHOI.

26 June – 1 August 2023

Interstellar Mission Concepts



Artists concepts.

Predecisional, for planning and discussion only. Illustration

DARE MIGHTY THINGS

The logo consists of the letters 'JPL' in a bold, red, sans-serif font. The 'J' and 'L' have a distinctive shape with a horizontal base that tapers to a point. The 'P' is also bold and red, with a vertical stem and a curved top.

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