

How can AI help us explore and understand the universe?

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Special Advisor on AI
OSU Libraries



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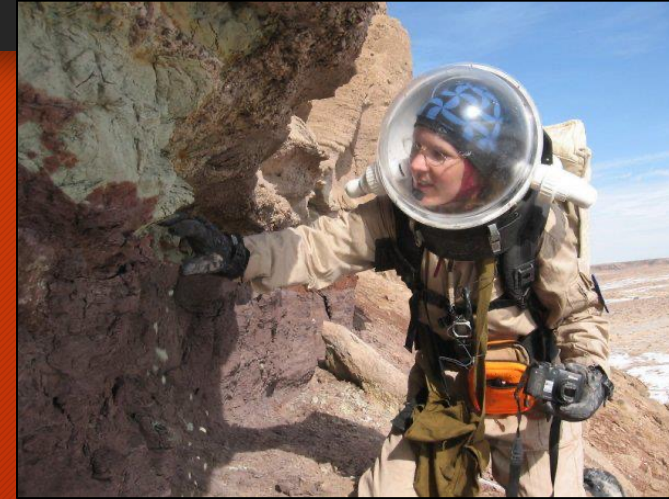
OSU AI Club



Dr. Kiri Wagstaff



- Researcher – Ph.D. Computer Science
 - NASA Jet Propulsion Laboratory (2003-2022)
 - AI and machine learning for space exploration
 - Mars rover operations (Opportunity rover)
 - Fellow of the Association for the Advancement of Artificial Intelligence (AAAI)
- Educator – Oregon State University since 2019
 - “Machine Learning Challenges in the Real World”
 - “How to Be Less Wrong: A Study in Common Misconceptions”
 - AI Advisor to U.S. Senator Mark Kelly (2023-2024)
 - Special Advisor on AI for the OSU Libraries

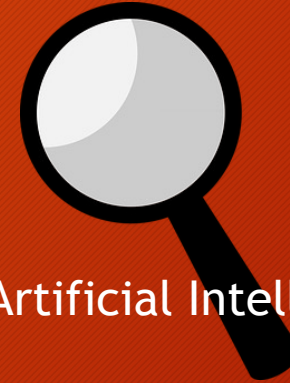


Mars Desert Research Station



U.S. Congress

Tools in the quest for scientific discoveries



Artificial Intelligence

“Equipped with our five senses – along with telescopes and microscopes and mass spectrometers and seismographs and magnetometers and particle accelerators and detectors sensitive to the entire electromagnetic spectrum – we explore the universe around us and call the adventure science.” – Edwin Hubble

What is AI?

- There is no widely agreed-on definition
 - Even among AI researchers
- “Artificial intelligence is the intelligence exhibited by machines and software.”
 - Wikipedia
- “AI refers to computer systems capable of **performing tasks** that typically require **human intelligence**, such as reasoning, learning, perception, and language understanding.”
 - ISO (International Organization for Standardization)

Personalize

Automate

Create

Predict

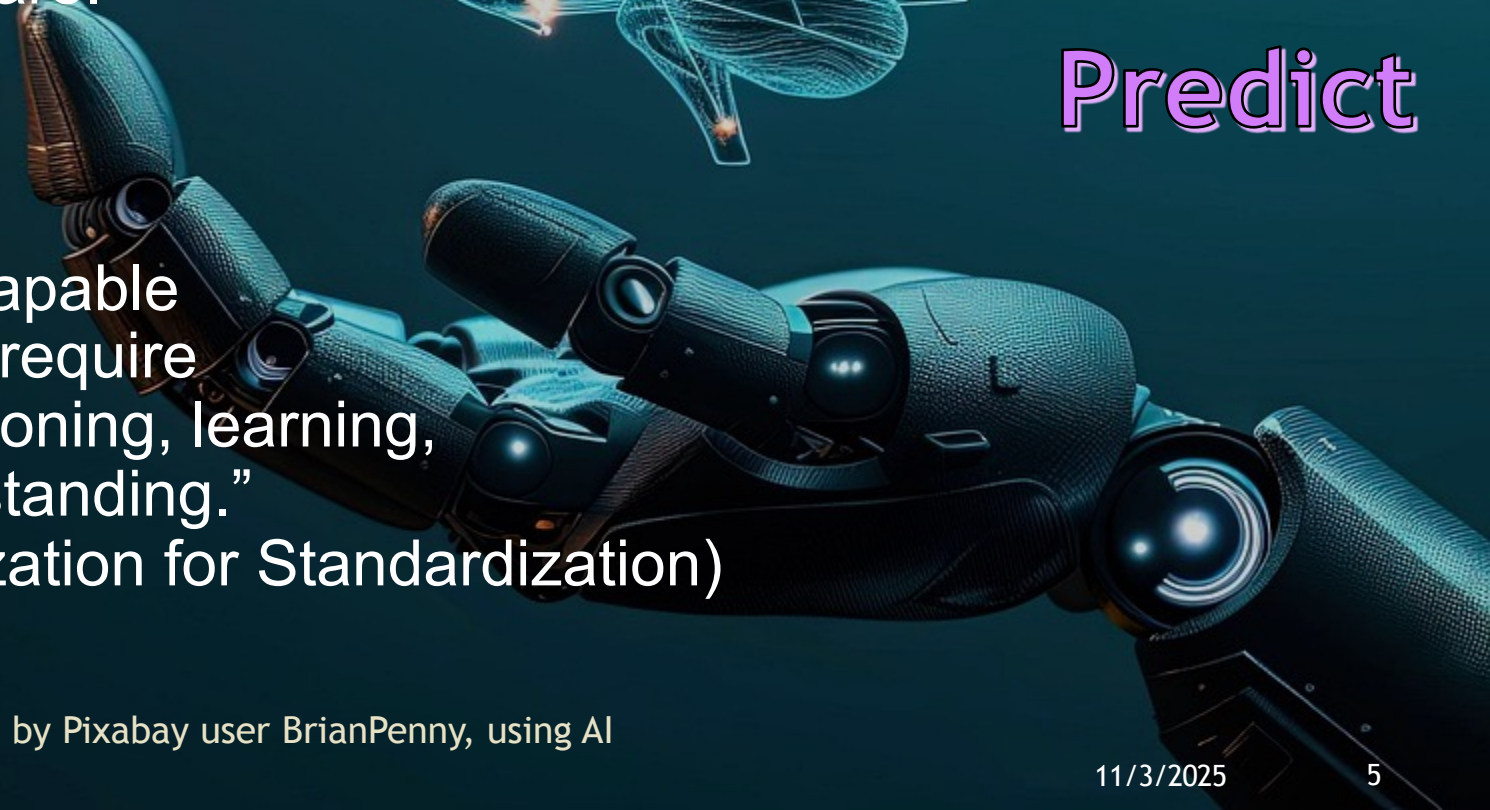
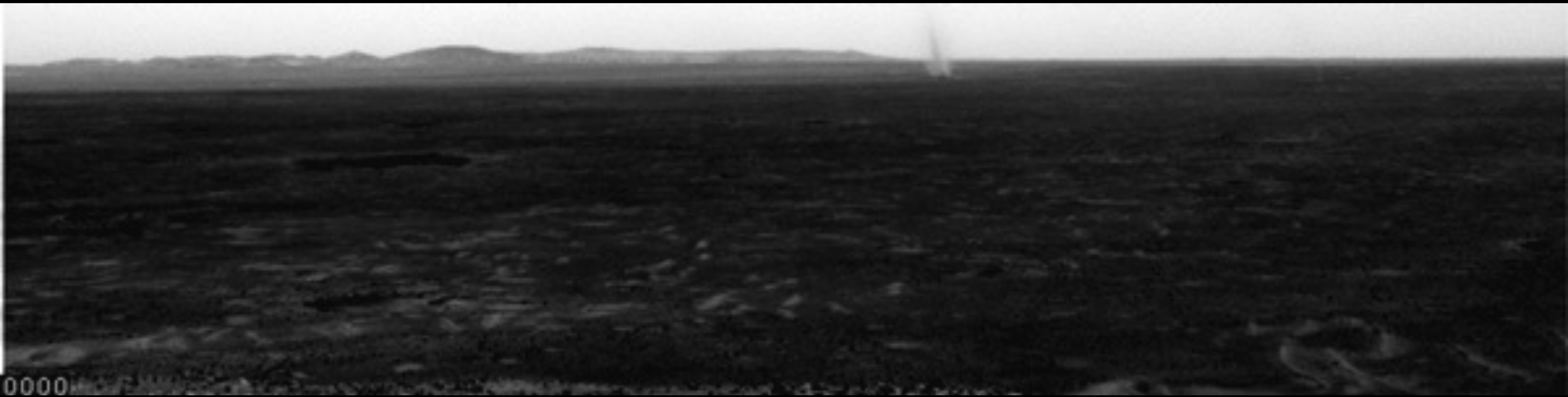
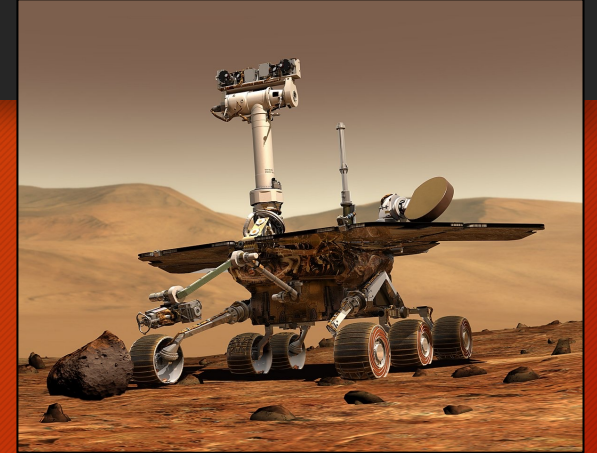


Image created by Pixabay user BrianPenny, using AI

1. Use AI in places humans can't go

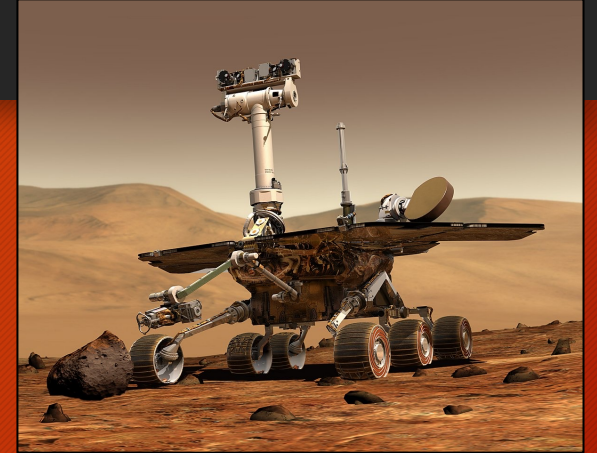
- A signal takes 20-40 minutes to go to Mars and back
- Meanwhile, exciting things are happening on Mars



Spirit Mars rover, Sol 568 (August 5, 2005) – Image credit NASA/JPL/Texas A&M

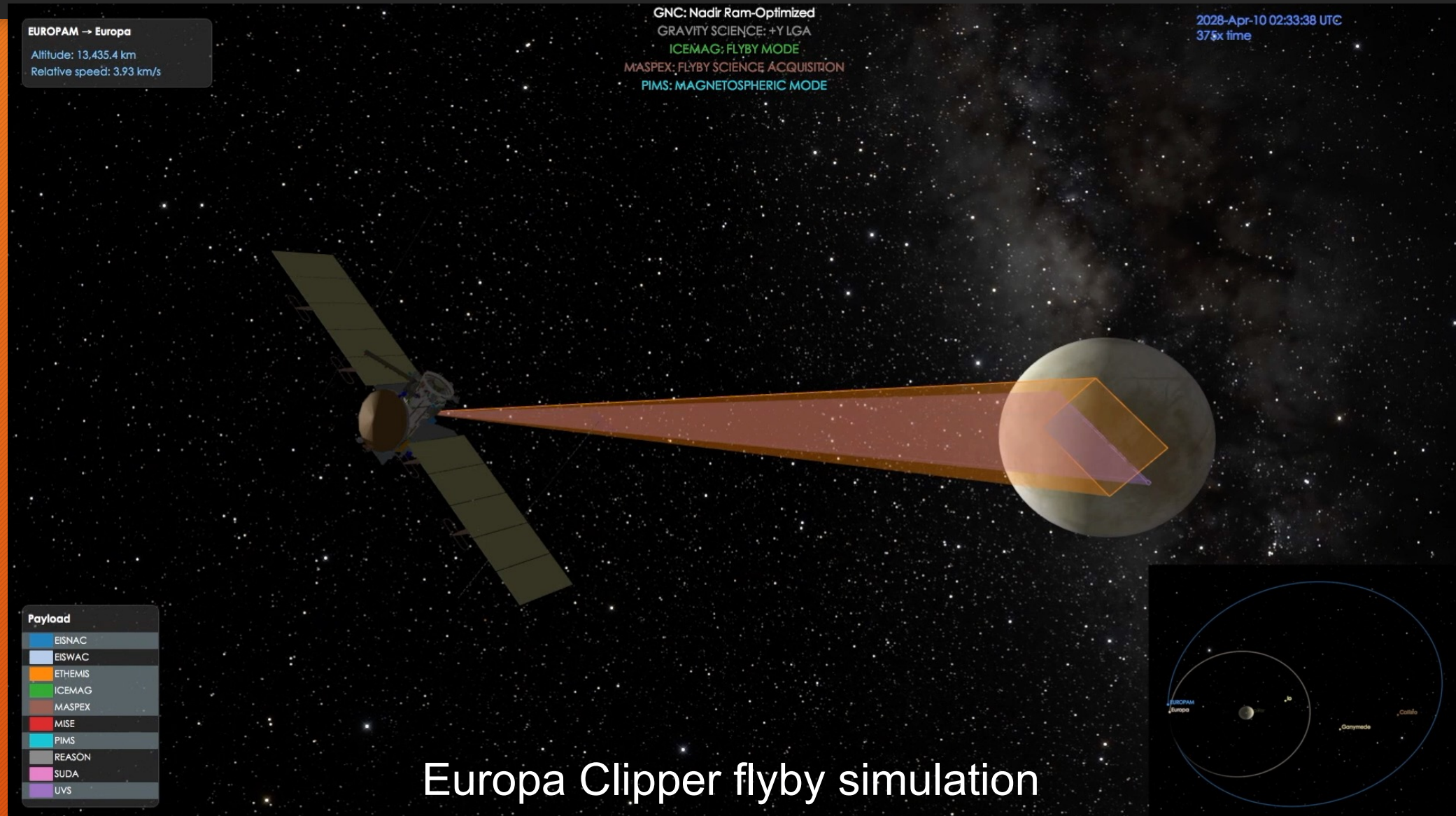
1. Use AI in places humans can't go

- A signal takes 20-40 minutes to go to Mars and back
- Meanwhile, exciting things are happening on Mars
 - Use AI to detect interesting events and capture data to send back to Earth – no other way to respond fast enough
 - 70% reduction in data volume sent back to Earth; but carefully weigh risks and benefits



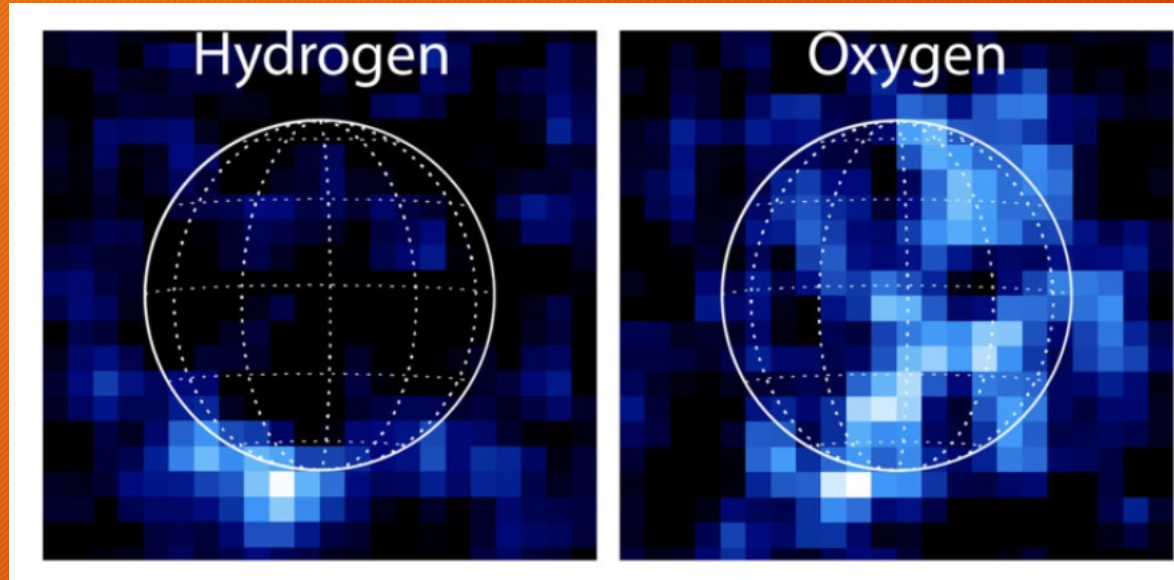
A. Castaño et al., 2007. *Machine Vision Applications* 19:467-482.

Jupiter's moon Europa: Up to 1 light-hour from Earth!



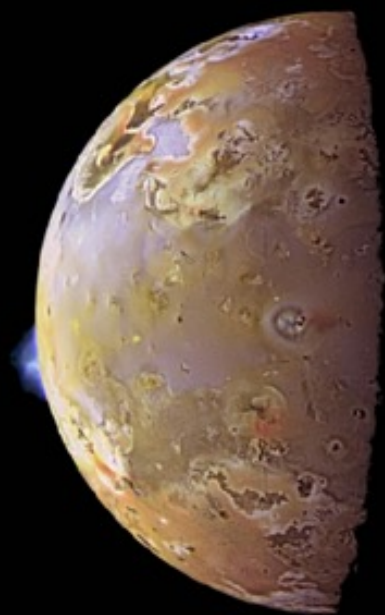
Could Europa have icy plumes?

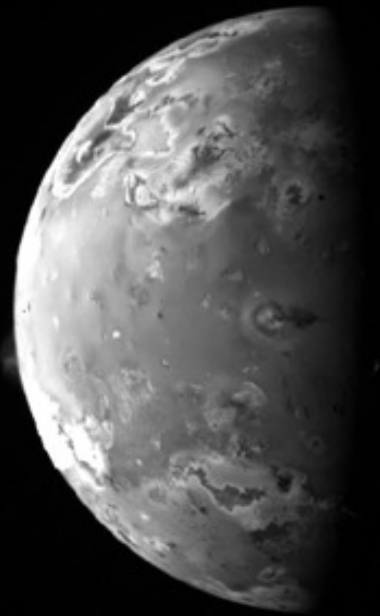
- Tantalizing observation from 2014 using Hubble Space Telescope



[Roth *et al.*, 2025]

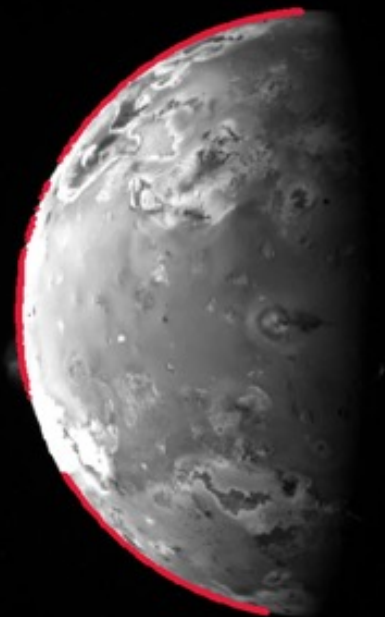
- What will we see up close?





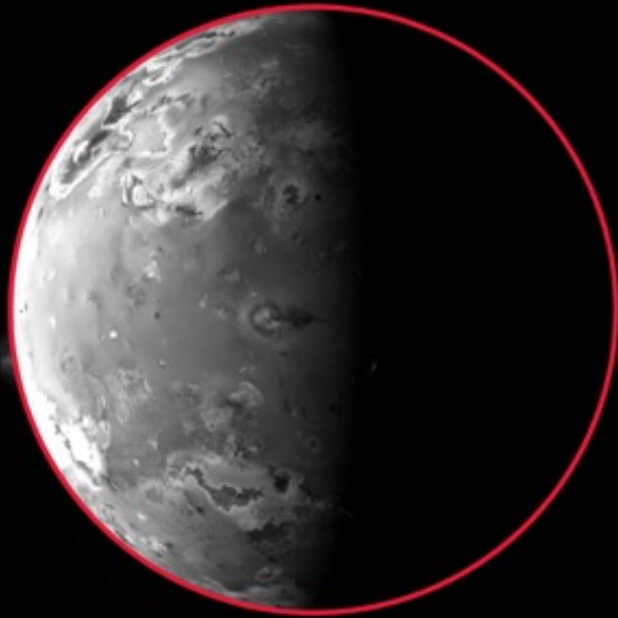
Galileo image of Io (analogue with known active plumes)

Step 1: Detect Edges



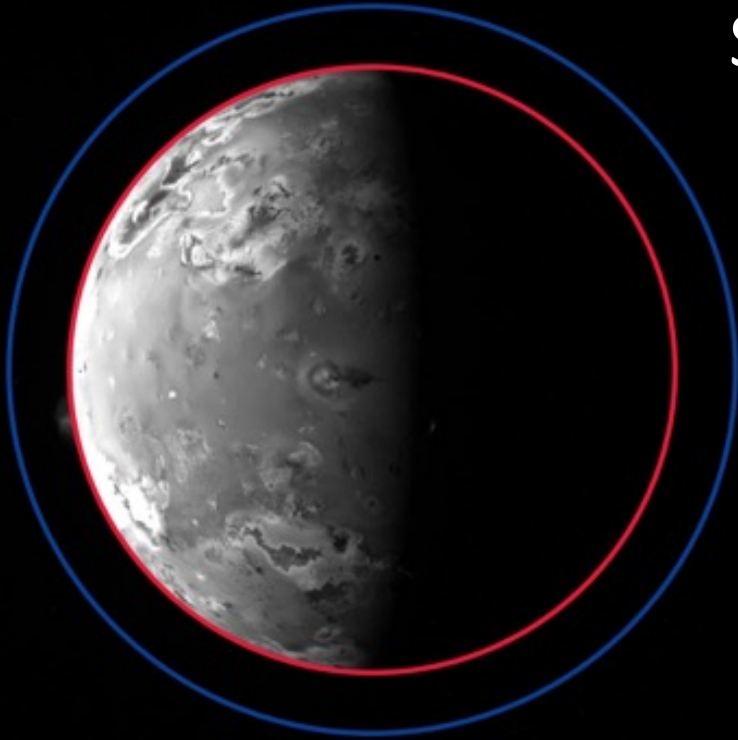
Galileo image of Io (analogue with known active plumes)

Step 2: Fit Limb Model



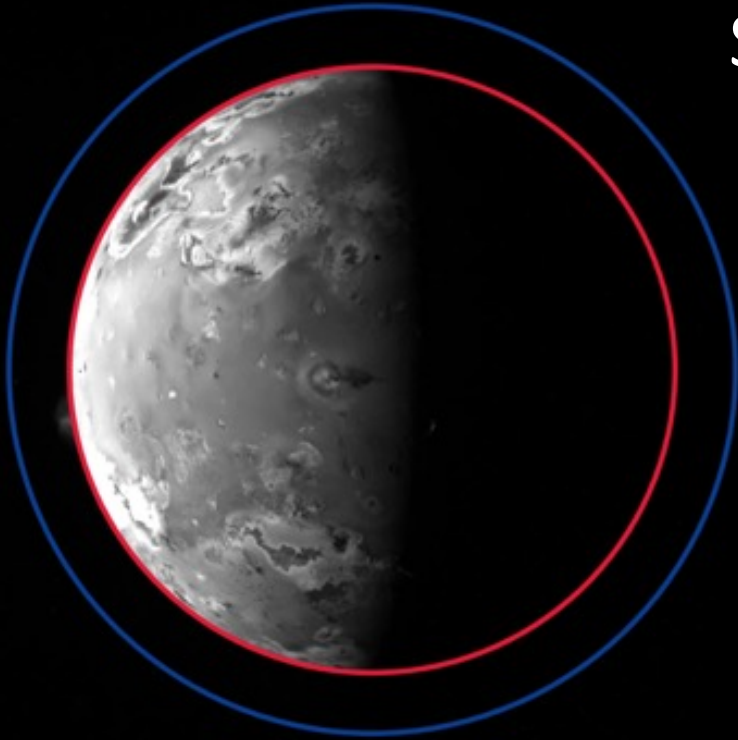
Galileo image of Io (analogue with known active plumes)

Step 3: Search within annulus



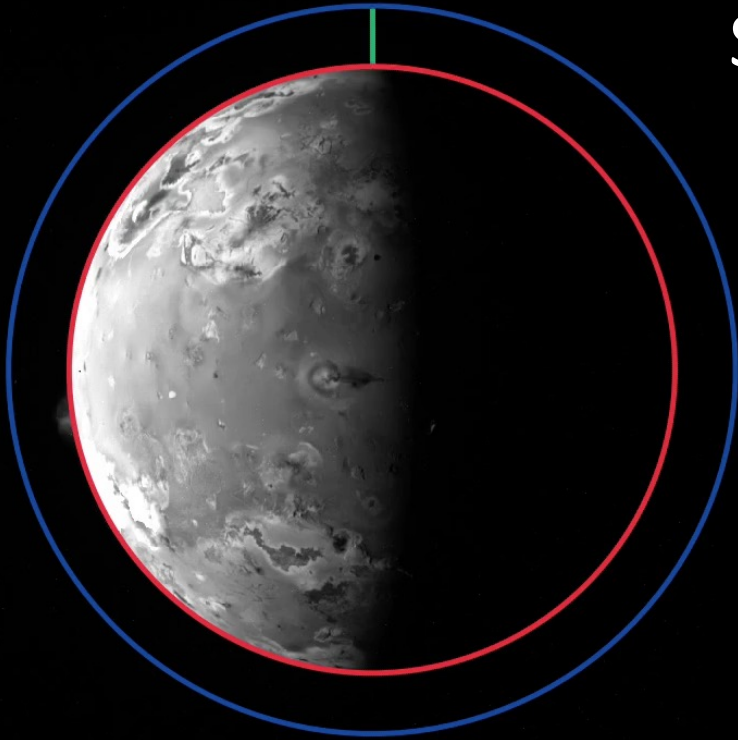
Galileo image of Io (analogue with known active plumes)

Step 3: Search within annulus



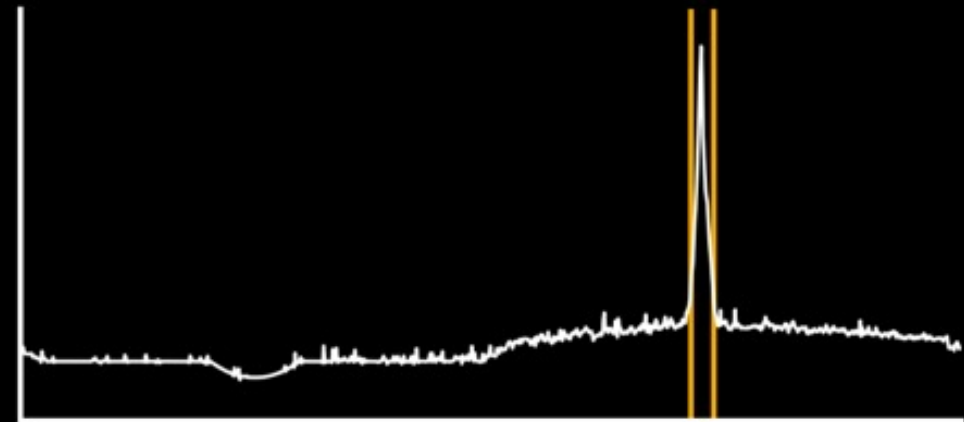
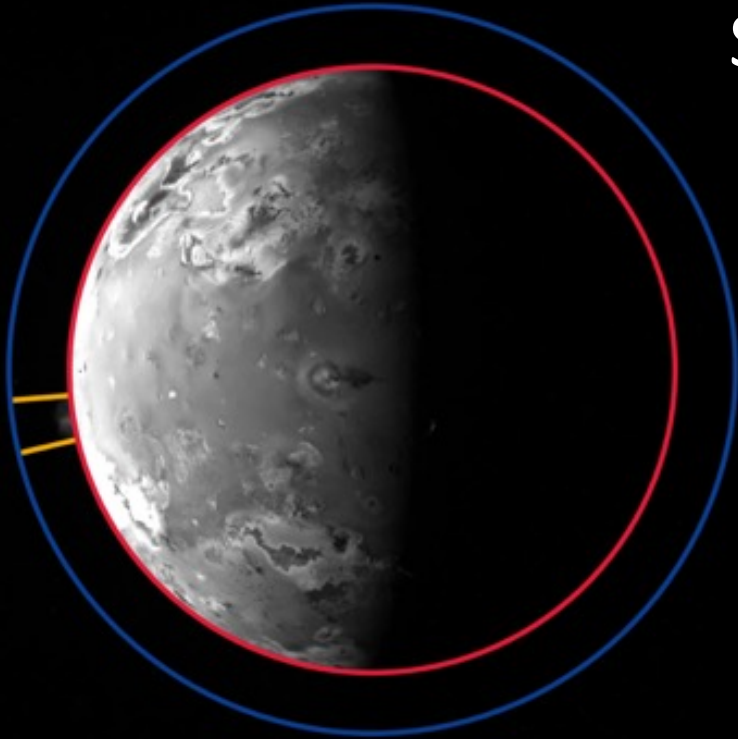
Galileo image of Io (analogue with known active plumes)

Step 3: Search within annulus



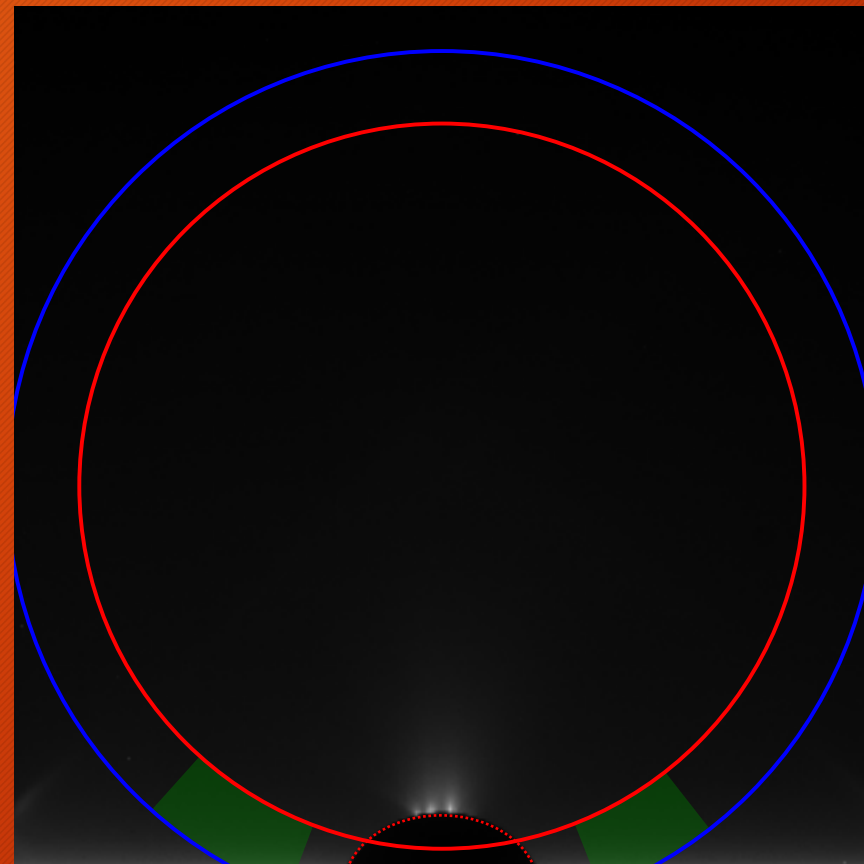
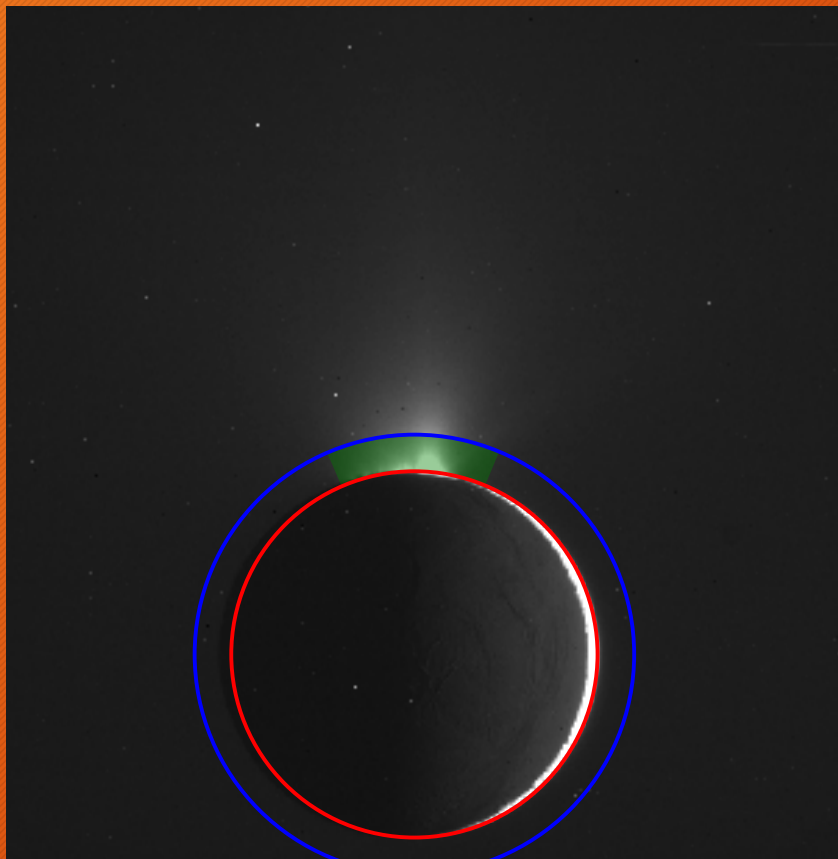
Galileo image of Io (analogue with known active plumes)

Step 4: Find peaks in intensity



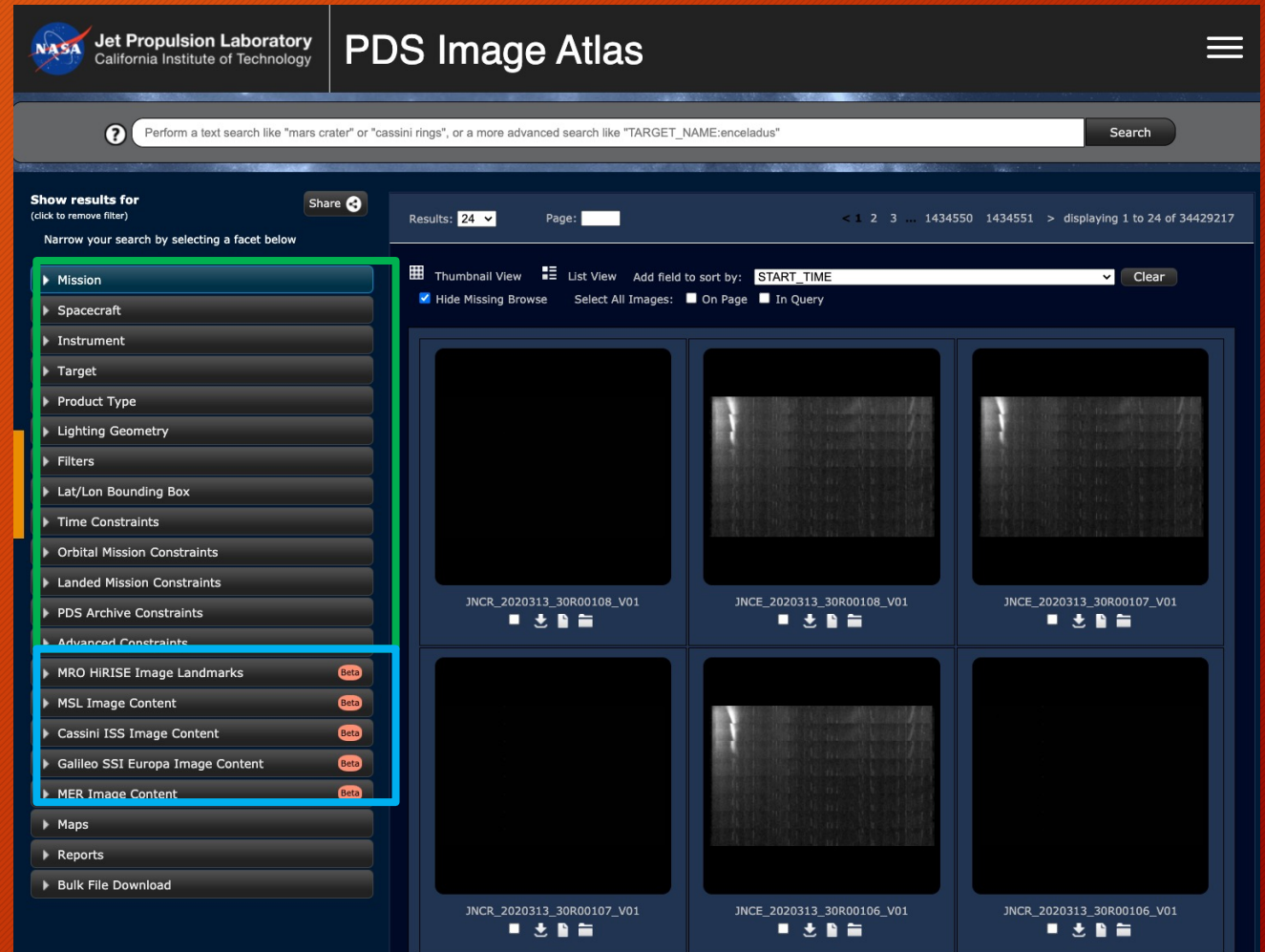
Galileo image of Io (analogue with known active plumes)

Plumes from Enceladus (moon of Saturn)



2. Use AI to understand what's in our data archives

- You can search NASA images by specifying *prior* info
(prior = known before image is collected)
 - Mission
 - Instrument
 - Target
 - Location
 - Time / orbit number
- You want to search by *content*
(only known after image is collected)
 - Crater, dune, impact, streak, ...
 - Layers, sand, drill hole, sky, ...

A screenshot of the NASA Jet Propulsion Laboratory's PDS Image Atlas search interface. The header shows the NASA logo, JPL logo, and the title 'PDS Image Atlas'. Below the header is a search bar with a placeholder text: 'Perform a text search like "mars crater" or "cassini rings", or a more advanced search like "TARGET_NAME:enceladus"'. The main content area is divided into two columns. The left column contains a list of search facets, including Mission, Spacecraft, Instrument, Target, Product Type, Lighting Geometry, Filters, Lat/Lon Bounding Box, Time Constraints, Orbital Mission Constraints, Landed Mission Constraints, PDS Archive Constraints, Advanced Constraints, MRO HIRISE Image Landmarks, MSL Image Content, Cassini ISS Image Content, Galileo SSI Europa Image Content, MER Image Content, Maps, Reports, and Bulk File Download. The right column shows search results, including a grid of image thumbnails and a list of image IDs (e.g., JNCR_2020313_30R00108_V01, JNCE_2020313_30R00108_V01, JNCE_2020313_30R00107_V01, JNCR_2020313_30R00106_V01, JNCE_2020313_30R00106_V01, JNCR_2020313_30R00106_V01). The interface also includes a 'Show results for' section, a 'Share' button, and a 'Results: 24' indicator.

<http://pds-imaging.jpl.nasa.gov/search/>

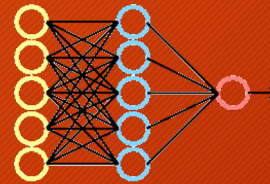
Machine learning for image content annotation

- Transfer learning: “Fine-tune” a neural network trained on Earth images to work on Mars images
- Deployed on >1M Mars **rover** images

ImageNet data set
[Russakovsky et al., 2015]



Train



Transfer



PDS image search for “wheel”

<http://pds-imaging.jpl.nasa.gov/search/>

Landed Mission Constraints

PDS Archive Constraints

Advanced Constraints

MRO HiRISE Image LandmarksBeta

MSL Image ContentBeta

Cassini ISS Image ContentBeta

Galileo SSI Europa Image ContentBeta

MER Image ContentBeta

Maps

Reports

Bulk File Download

Image Class

wheel (2320)

Confidence Level

0.9 to 1



0549ML0022140030204115E01_DRCX



0549ML0022140030204115E01_DRLX



0603ML0025430020301345E01_DRCX



0940ML0041300020403055E01_DRXX



0940ML0041300020403055E01_DRCL



0940ML0041300020403055E01_DRCL



0603ML0025430020301345E01_DRXX



0603ML0025430020301345E01_DRLX



0603ML0025430020301345E01_DRCL



Kiri Wagstaff



Steven Lu



Kevin Grimes



Gary Doran

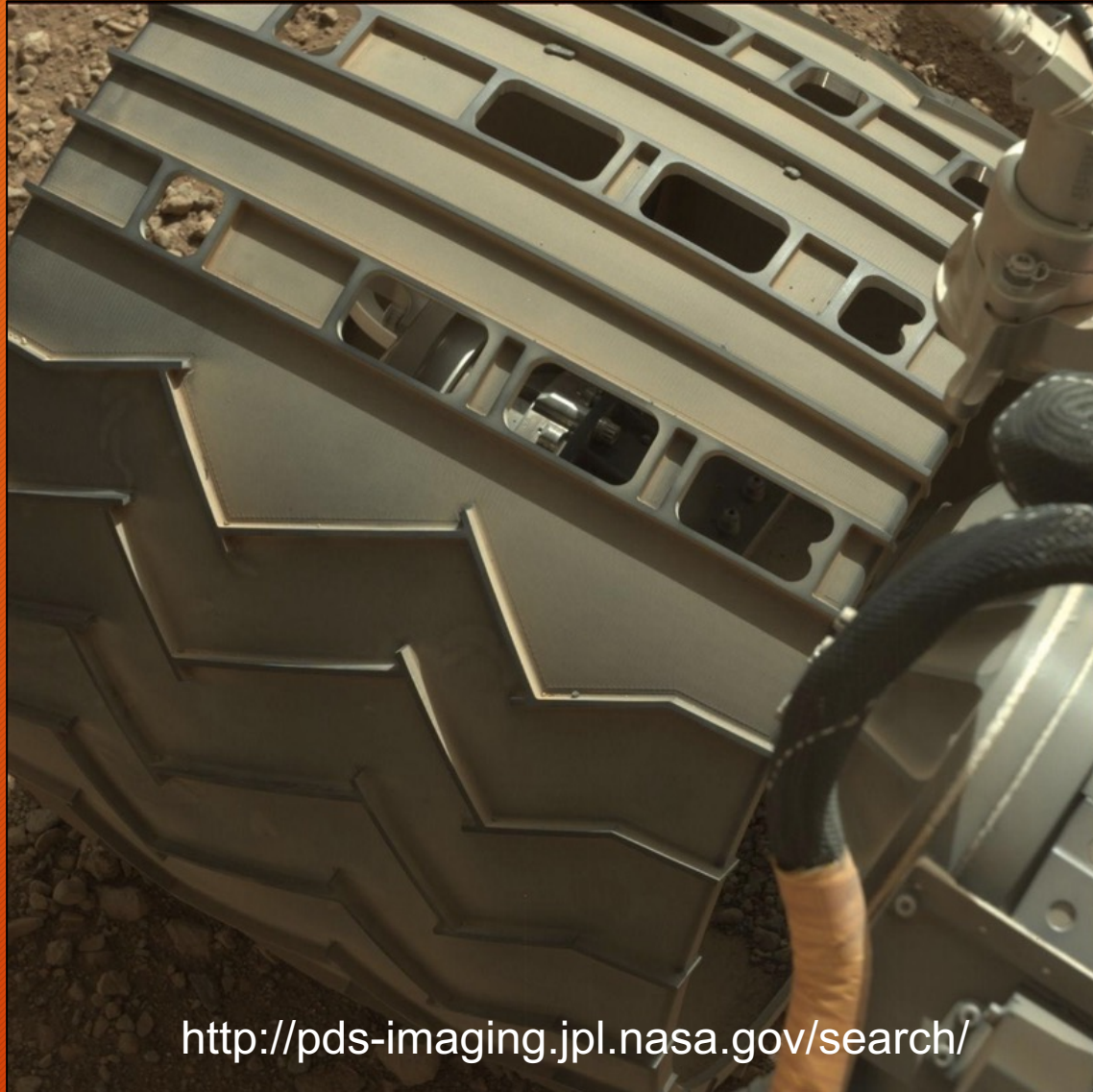


Thamme Gowda



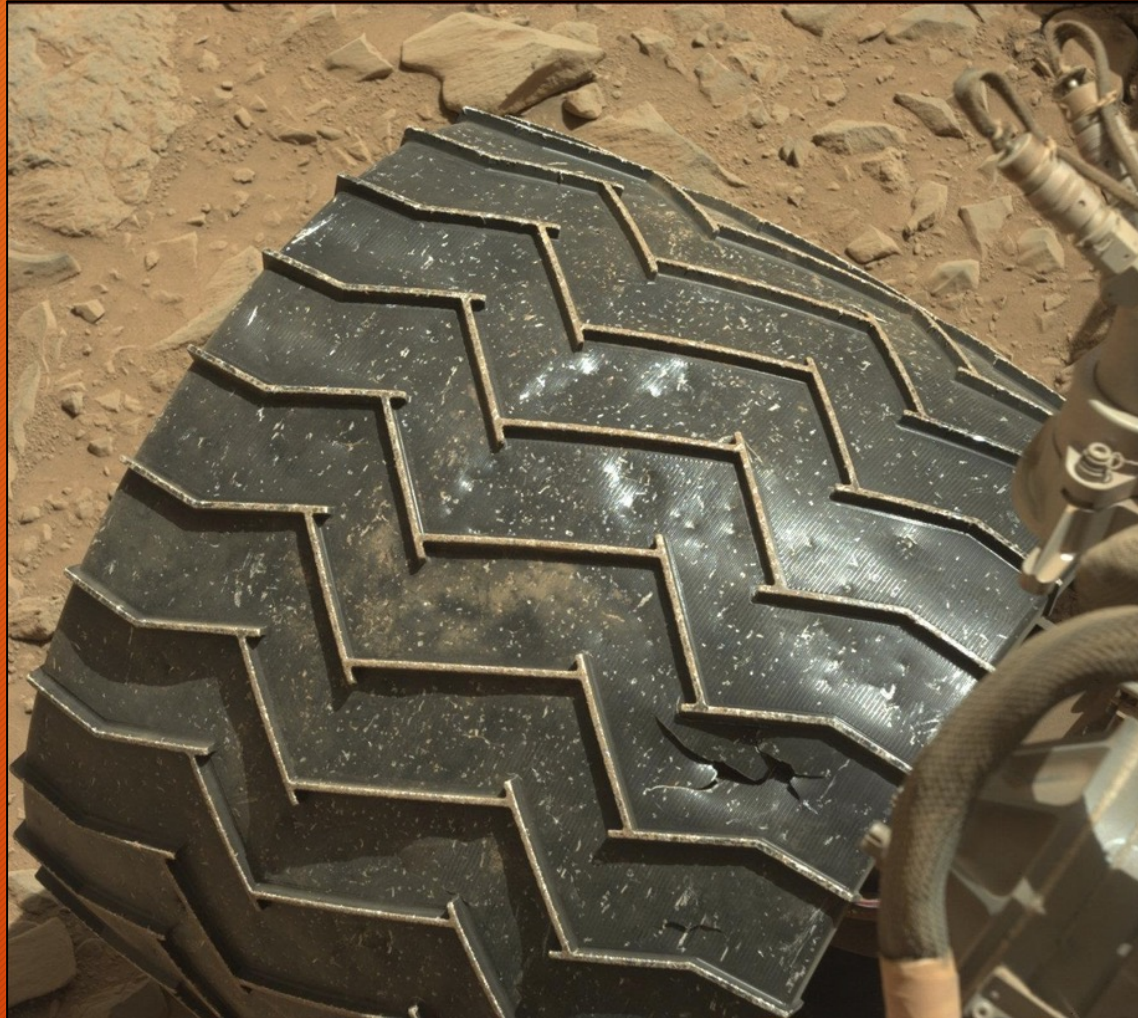
Emily Dunkel

August 2012



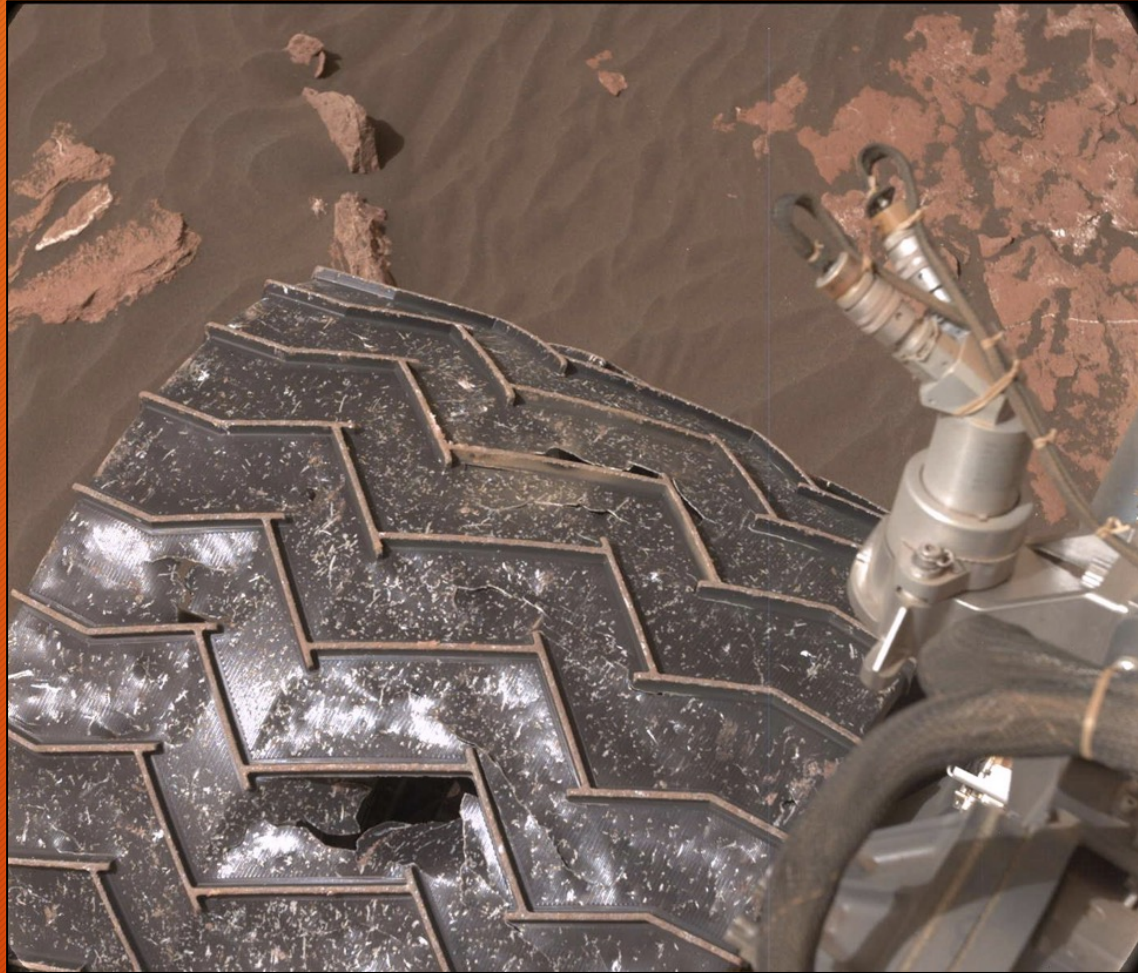
<http://pds-imaging.jpl.nasa.gov/search/>

December 2013



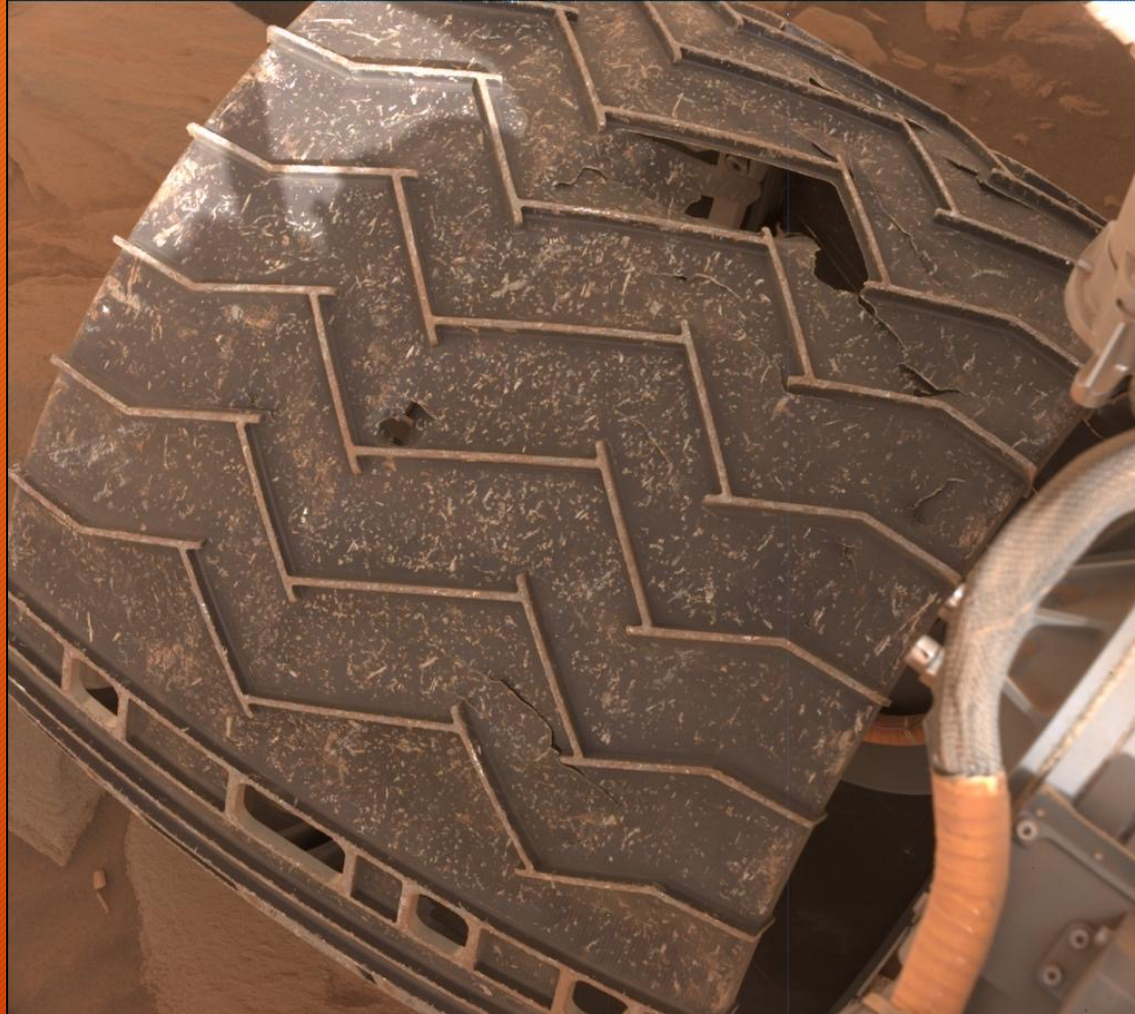
<http://pds-imaging.jpl.nasa.gov/search/>

November 2016



<http://pds-imaging.jpl.nasa.gov/search/>

March 2020



<http://pds-imaging.jpl.nasa.gov/search/>

It works for orbital images, too

- E.g., search for images containing craters
- Shows landmarks with ≥ 0.9 probability
- Landmarks are localized

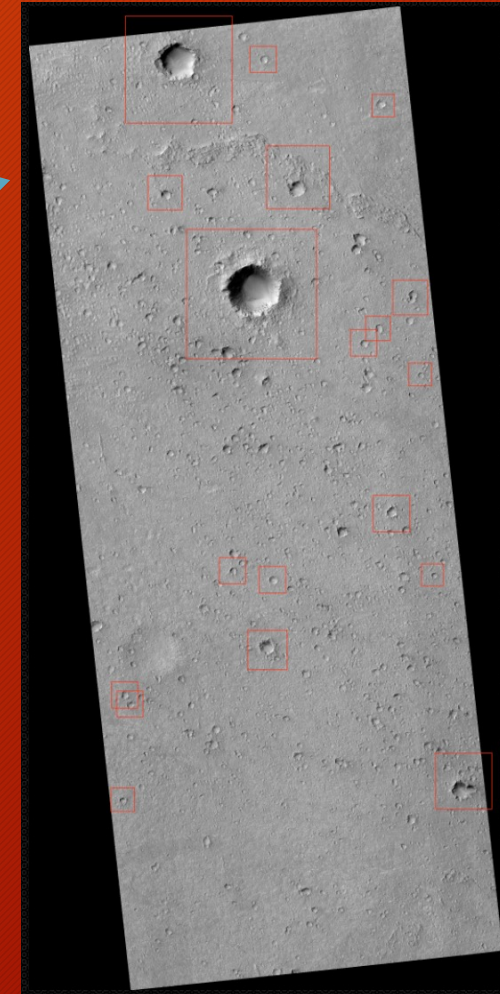
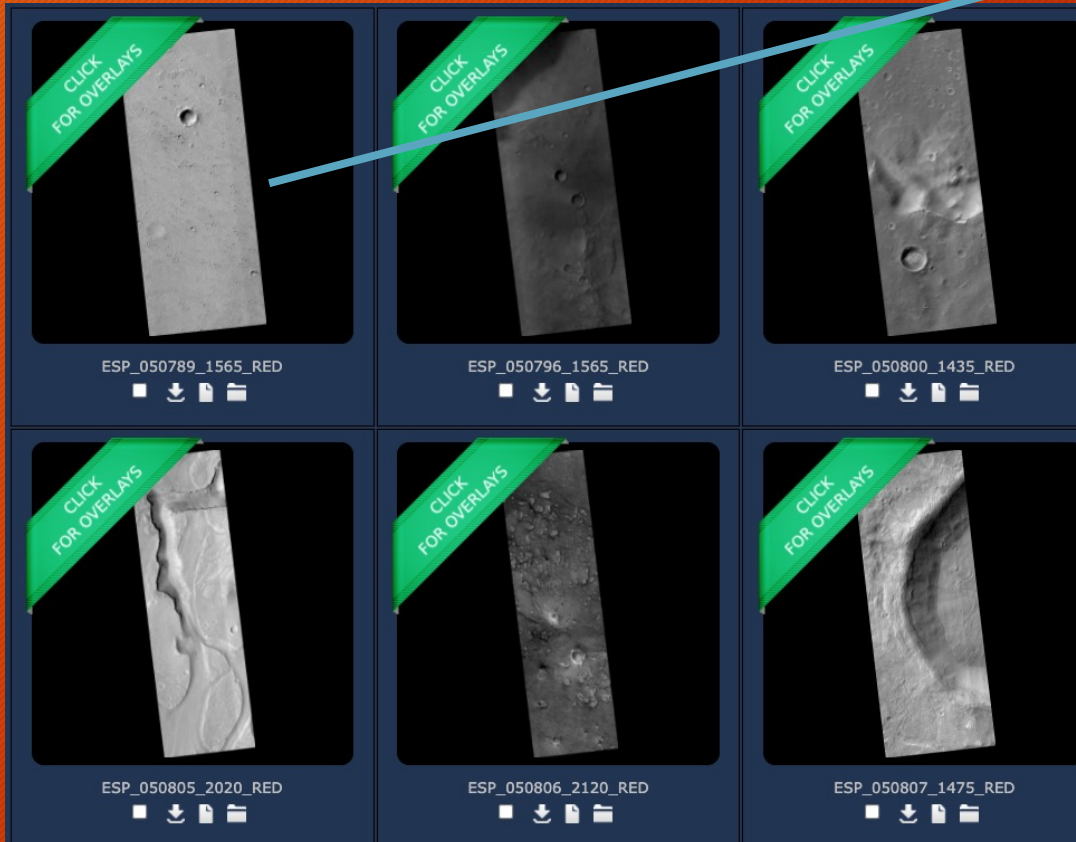
First 6 results from “crater” search

PDS Image Atlas

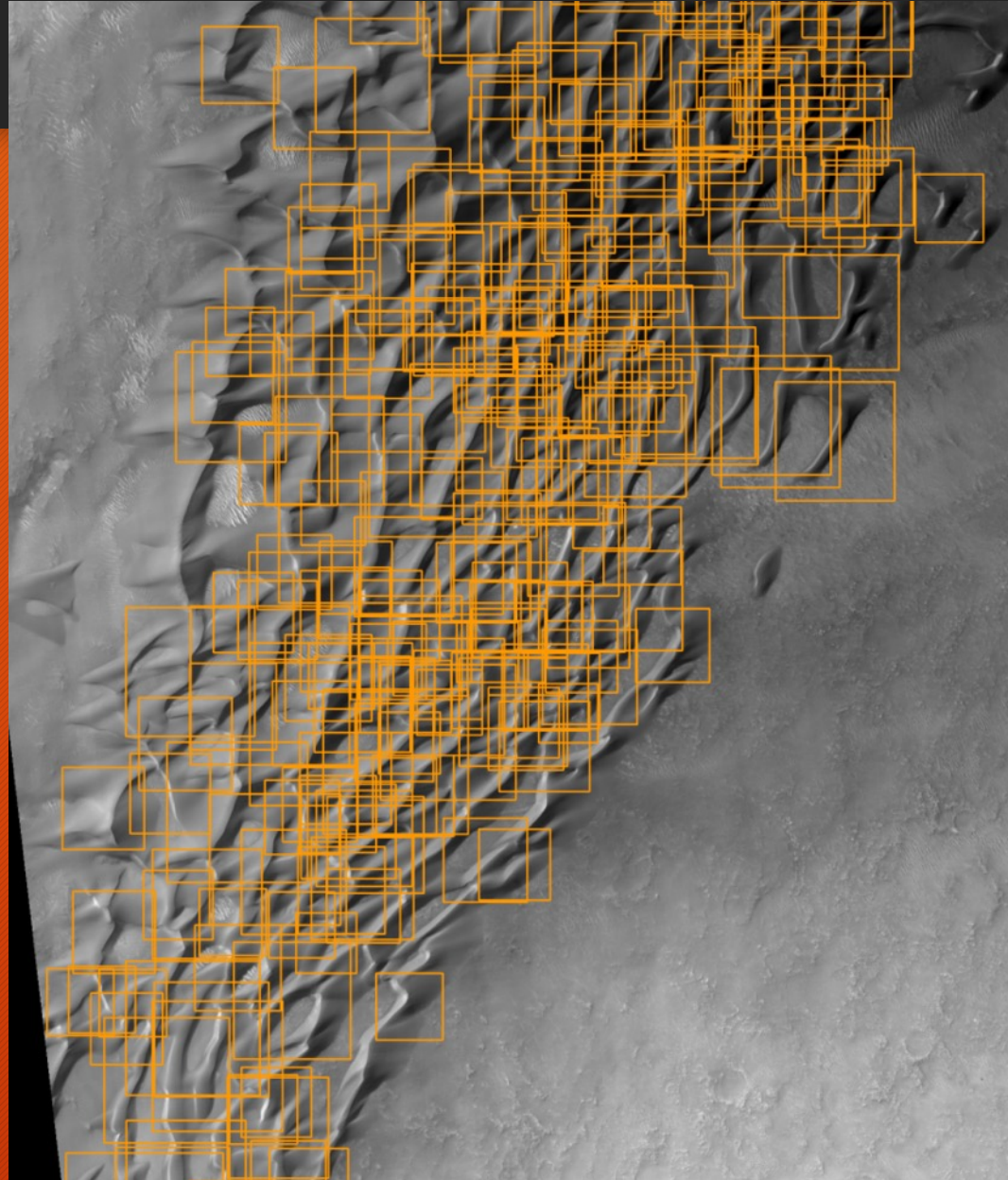
▼ MRO HiRISE Image Landmarks Beta

Image Class
(MRO HiRISE RDRs Only)

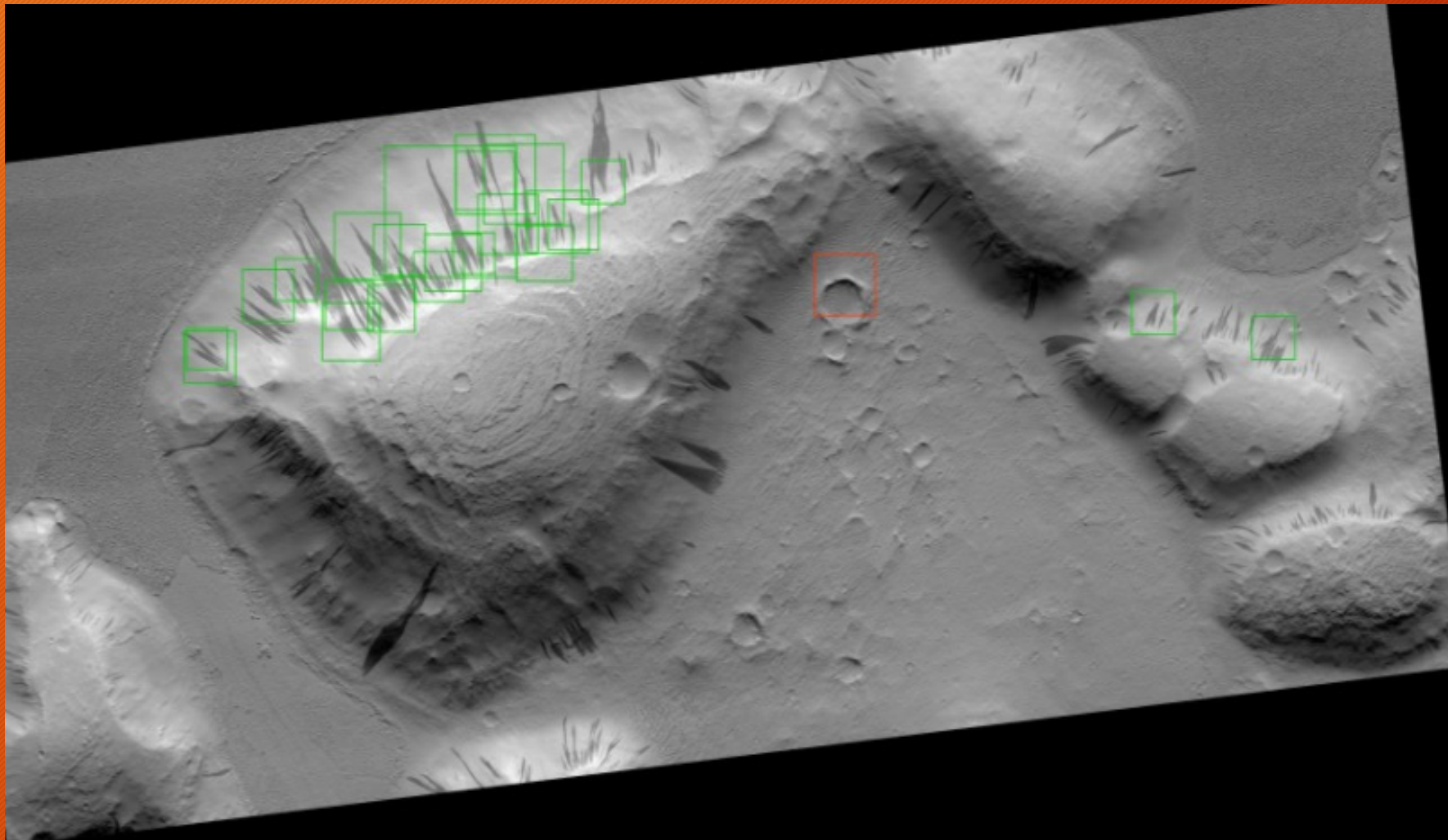
- bright dune (527)
- crater (25071)
- dark dune (2502)
- impact ejecta (2480)
- slope streak (2153)
- spider (228)
- swiss cheese (1058)



Dark dunes

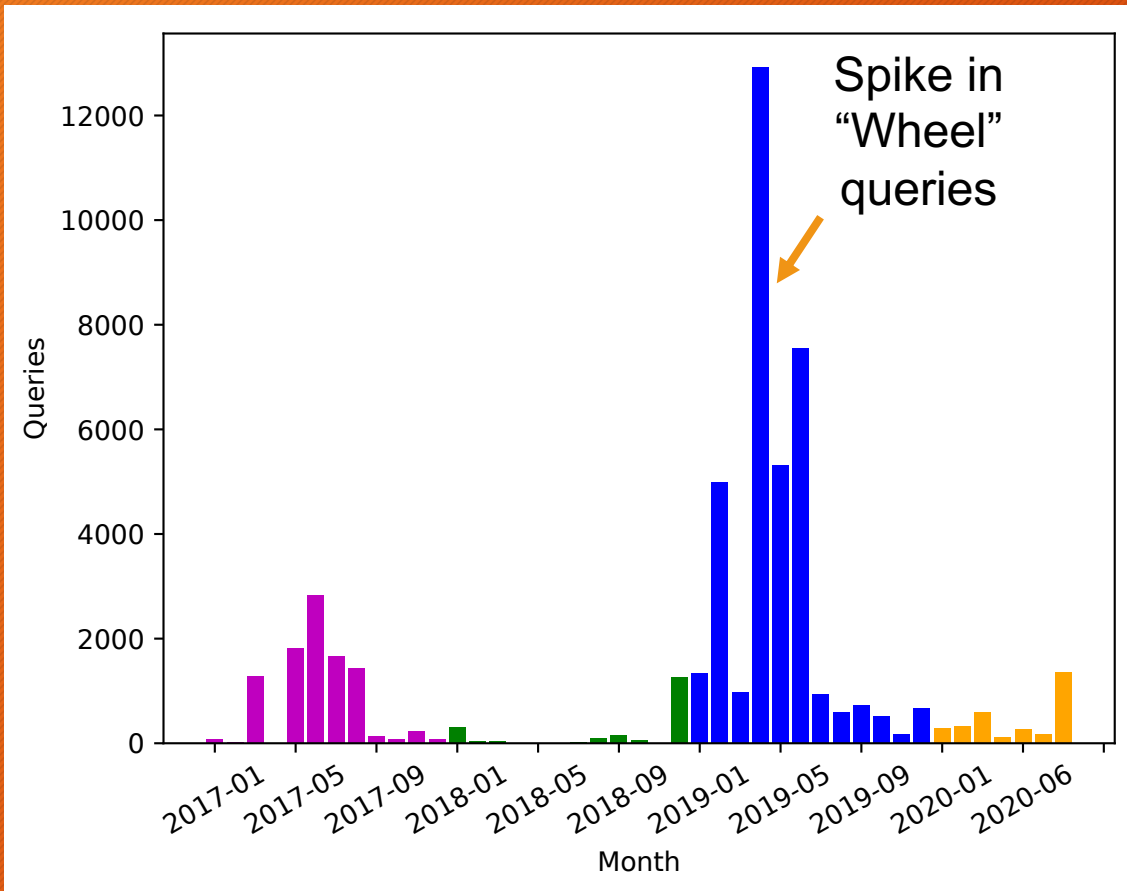


Dark slope streaks

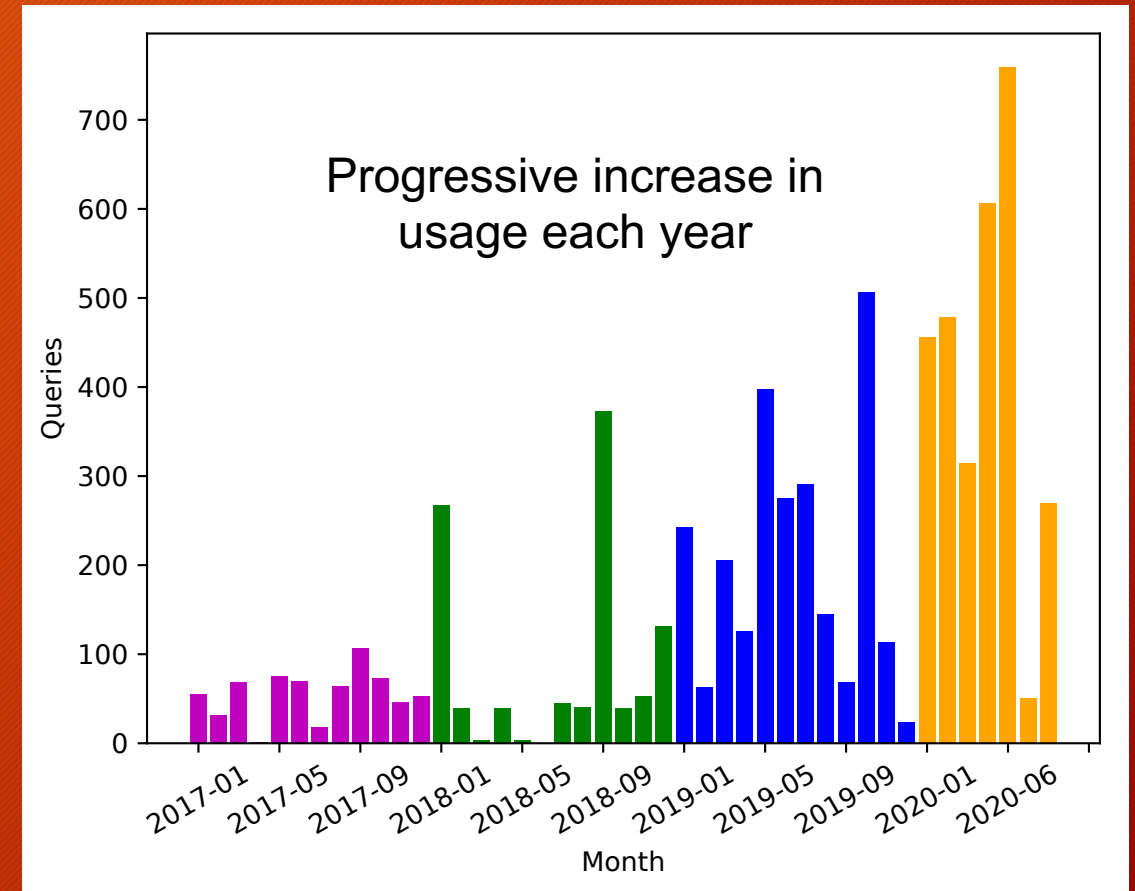


Content-Based Searches Over 3.5 Years

MSL (rover)



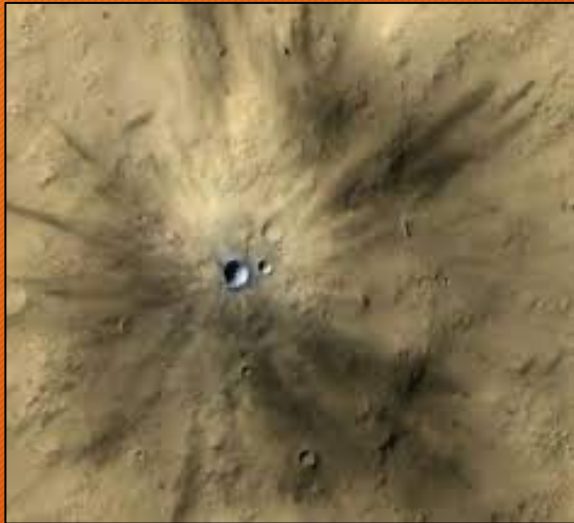
HiRISE (orbiter)



(Color-coded by year)

3. Use AI to search an entire planet

- Mars orbital images show evidence of recent meteorite impacts on Mars



Credit: NASA/JPL-Caltech/MSSS/UA



Credit: NASA/JPL-Caltech/UA

- ~1000 have been manually catalogued
- Previous orbital image classifier uses HiRISE data; this covers only ~3% of the planet
- To find more, we need an image classifier for CTX (global) data



Kiri Wagstaff



Ingrid Daubar



Michael Munje



Gary Doran



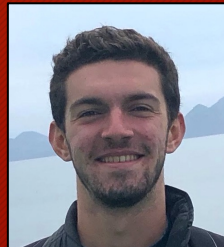
Valentin Bickel



Joe Pate



Annabelle Gao



Daniel Wexler

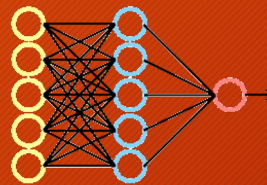
Fresh impact image classifier

- Transfer learning: “Fine-tune” a neural network trained on Earth images to work on Mars images
- Deployed on 12 TB of Mars **orbiter** images

ImageNet data set
[Russakovsky et al., 2015]

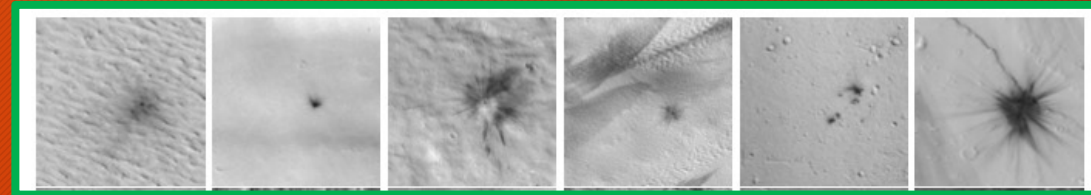


Train

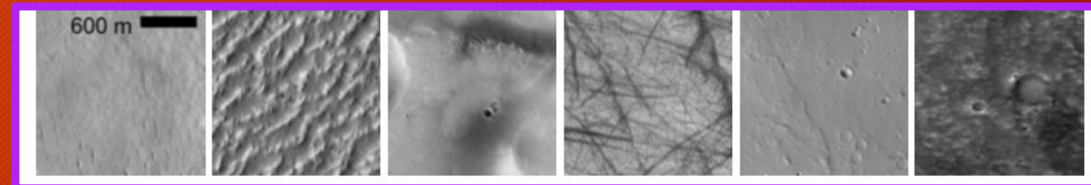


Transfer

Fresh impacts



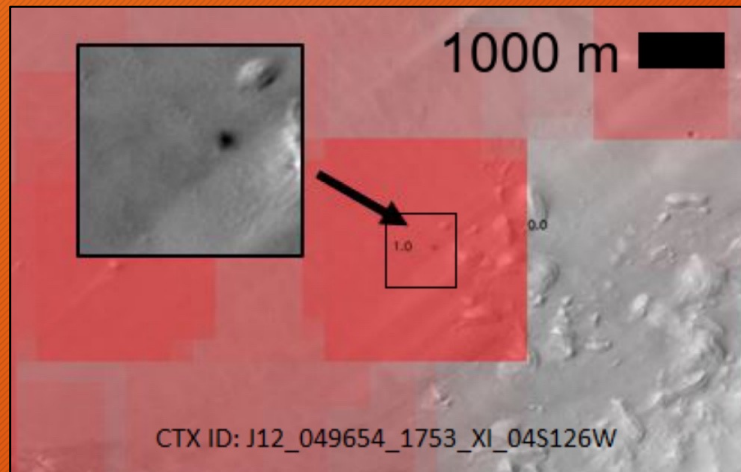
Not impacts



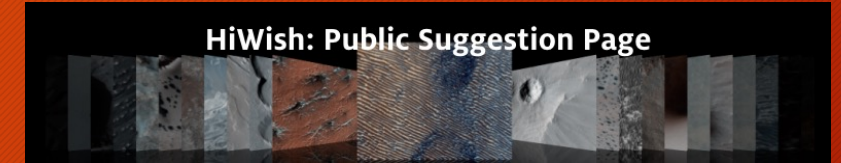
Hunting for new Mars impacts



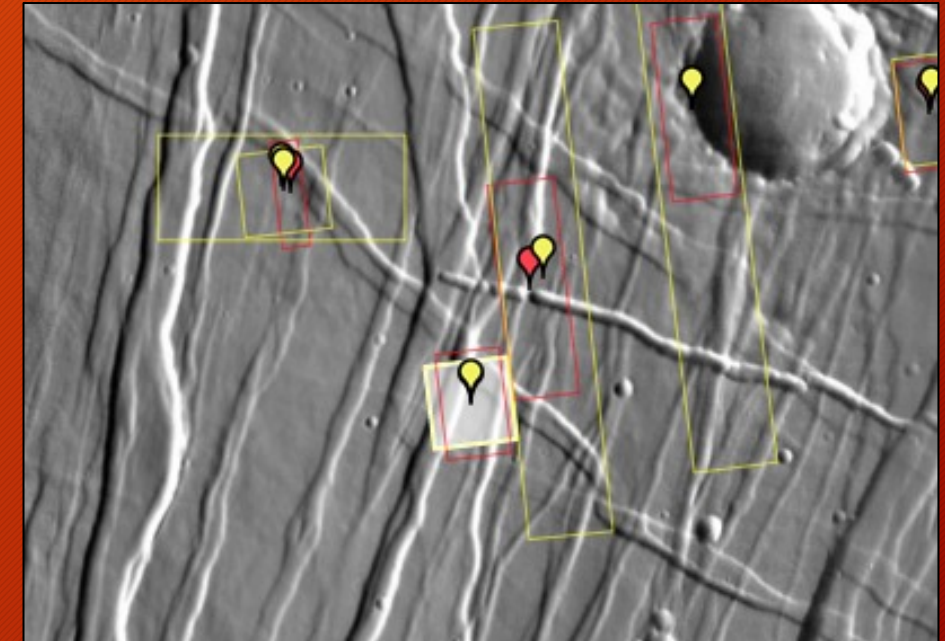
- Slide classifier window over global Mars images
 - 2 billion classifications
 - A cluster of ~500 CPUs took a week (~5 sec/image)
 - Manual hunt (40 min/image) would require ~8.5 person-years (no breaks!)



[Munje et al., 2020]



<https://www.uahirise.org/hiwish/>



- Possible new impact? Request high-resolution image with HiWish!

New Mars impact discoveries

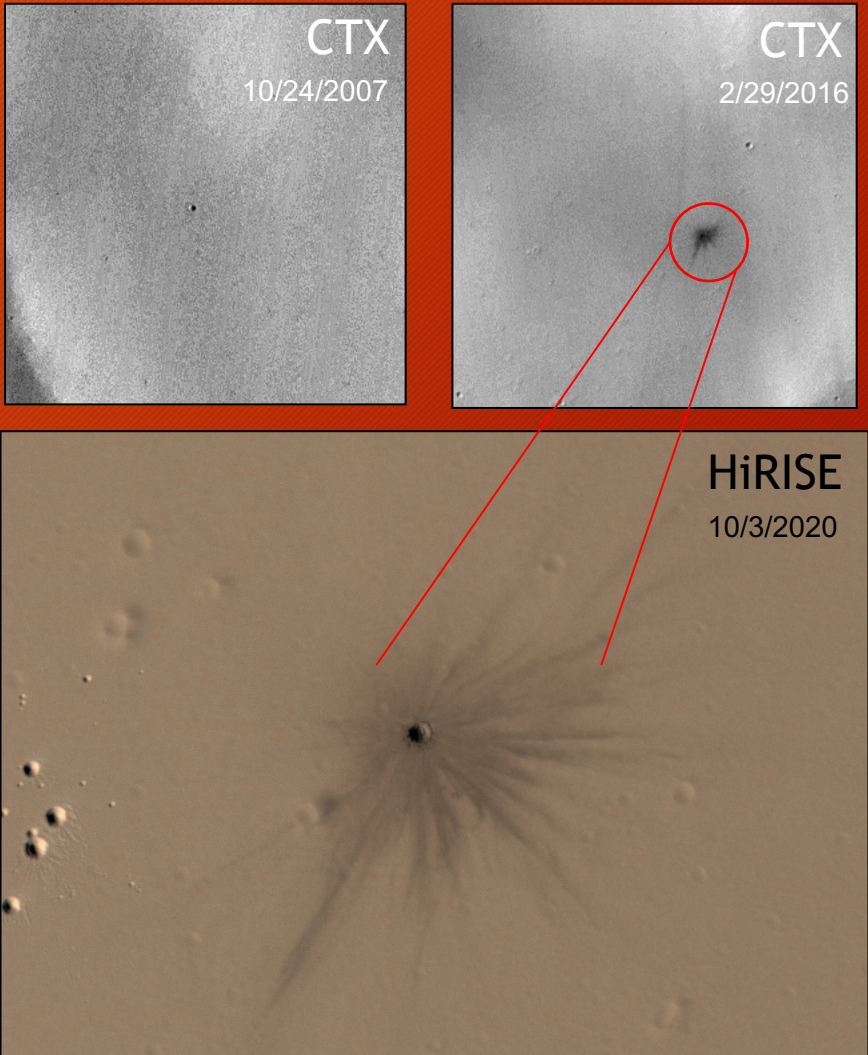
Noctis Fossae



CTX: 6 m/pixel

HiRISE: 0.3 m/pixel

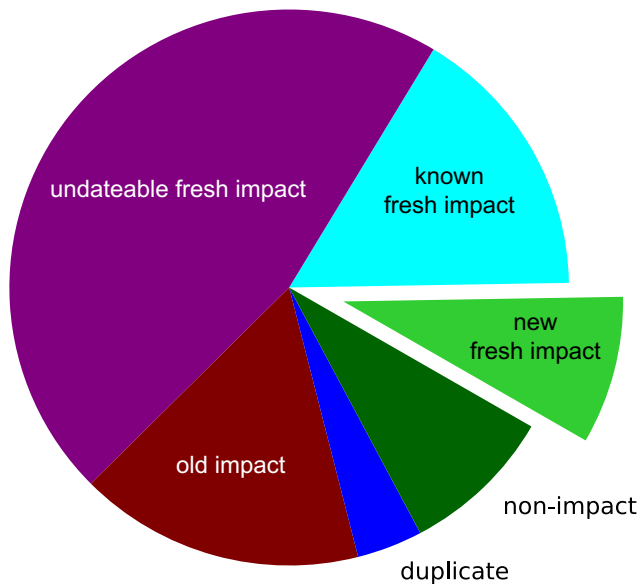
Noctis Labyrinthus



70+ new Mars impact discoveries

- >1,000,000 candidates with $P(\text{impact}) \geq 0.99$
- Group by location and time, then manually review the top 1000 (by probability)
 - New impact? Request high-resolution image with HiWish

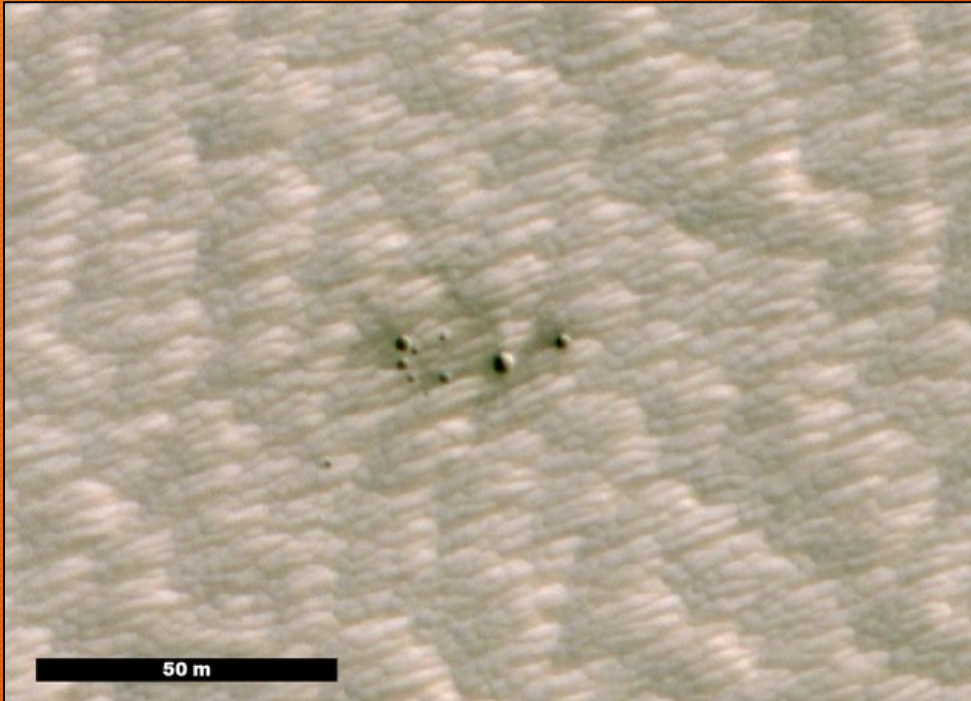
Summary of top 1000 candidates



- 99 already known (confirmation)
- 69 new discoveries
- 465 are impacts without a “before” image (undateable)
- 166 old impacts
- Only 90 (9%) non-impacts

Where's the generative AI?

1. Can it annotate image content? (E.g. Image caption generation)



Please generate a caption for this image of the Martian surface.

ChatGPT:

Caption:

High-resolution image of the Martian surface captured by NASA's HiRISE camera aboard the Mars Reconnaissance Orbiter, showing the landing site of the Curiosity rover in Gale Crater. The dark spots and streaks mark the descent stage impact and the rover's touchdown area on the patterned surface of Mars. Scale bar: 50 meters.

We want to know what's out there, not what *could* be out there (great for sci-fi stories, but not for science)

- Generative AI creates a common / likely caption, not an accurate one

2. Can it find the fresh impact craters?

ChatGPT:

1. ~48.5 m diameter crater: Located at ~3.34° N, 219.38° E, formed between March 27–28, 2012.



Actual caption: These labels show *named geographic features* on Mars. Someday, people from Earth will visit these places! Image via [NASA/ JPL-Caltech/ MSSS](#).

So: How can AI help us explore and understand the universe?

- Classification and discovery methods help accelerate science



Detect events of interest
(e.g., dust devils!) onboard
a rover or orbiter

Classify and annotate large
data archives to enable
content-based search



Quickly scan global data
sets for features of interest
(fresh impacts)

- ML leverages, but does not replace, human expertise
 - Human review is a key part of this process (and it takes time and effort!)
 - ML/AI can focus attention where it is most likely to pay off



Artificial Intelligence

Contact: wkiri@wkiri.com