

Jet Propulsion Laboratory California Institute of Technology "Look and you will find it – what is unsought will go undetected." - Sophocles



# Content-based On-board Summarization to Monitor Infrequent Change

• SMIC

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Program: Strategic Initiative (6x)

Assigned Presentation #18

# Executive Summary – Global Change Monitoring



- Spacecraft like the Mars Reconnaissance Orbiter (MRO) scan the entire surface of Mars every few days at high resolution
- Yet only a handful of pre-selected images per day can be sent back to Earth due to limited communications bandwidth
- Transient science (active surface change) is rarely captured... can't know where/when it will be to request image
  - Fresh Impacts, rockfall, dust devils, araneiforms, active gulleys, recurring slope linea, active transverse aeolian ridges, dust uplift events, etc.
- Conventional change detection requires a before and after picture for differencing
- Global case would need the whole planet stored onboard as pixels for comparison... would take XX TB of onboard storage!
- Innovation: Use (2) onboard science autonomy capabilities to globally capture & monitor high-impact transient science targets
- 1) Store highly summarized representations of surface (rather than raw pixels) onboard and identify change in summary
  - Agnostic to expectation, can find new kinds of change at many spatial scales
- 2) Directly recognize known kinds of active/recent transient events in still images
  - Identify, map, and return cutouts of hi-res data only around targets of interest
- Outcome: Demonstrated utility of system via "Mission into the data" and discovered ~100 new, previously unknown fresh
  impacts in the existing MRO CTX image collection





## Few big images / day



Here's what changed on Mars today. Would you like to downlink?

# Challenge: Changing Planet, minimal downlink









Araneiforms

(all Mars images courtesy NASA/JPL

## **Active Dust Devils**

Why transient science matters? Seismic activity

- Regolith geomechanical properties
- Active aeolian processes / wind characterization
- Replenishment / transport of atmospheric dust
- Subsurface composition

courtesy Vaneltin Bickel

- Modern planetary bombardment rate
- CO<sub>2</sub> & water ice transport via sublimination and melting
- Dry vs. wet erosive processes
- Broader implications for planetary chronology

### Direct change detection on Earth? Simple!

- Align two images pixel-for-pixel
- Difference them
- Look for change... voila!





For a single 17-day MRO repeat cycle

Single global mosaic of Mars ~400 TB (~2 Gbps) Downlink possible ~0.3 TB (~1.5 Mbps)

### **State of Art: Exhausting Manual Search**

Missed findings, unscalable human effort

### **Planetary Change Detection? Hard.**

- Can't send back whole planet's images
- Especially repeatedly via overpasses
- Can't store whole planet onboard
- Hence, direct change detection won't work
- How to recognize, capture, and characterize global Martian change?

# Innovation: Onboard Summarization for Science



Many kinds of change can be captured

(courtesy NASA

Camera + compute + COSMIC is new smart instrument!

Miss a comm pass? COSMIC keeps working.

# **Change Detection Capabilities & Progress**

## Known Science Targets (surface classification)

- Know what, but not where/when
- Can look for any surface feature
- Needs example images for training
- Alerts provide time/space map
- Works on entire globe

## Agnostic Change Detection (image-summary-based change)

- Don't know what, where, or when
- Discover new transients
- Global reports on change characterization
- Agnostic to expectation or known targets

### Araneiform (Spiders) Tracks Fresh Impacts Swiss Cheese Terrain



## **Region Monitoring** (Pixel change for small areas)

- Know where, but not what/when
- Classical change detection
- Only a ~1k small regions at a time
- Alerts when something starts to happen



### TRL ~5

- Tested on entire CTX database
- Actual science discoveries attained!
- Related task infused into PDS archive
- Ready for flight-oriented maturation

## TRL ~3

- Feasibility study complete on real data
- Computational goals met
- New change-detection req: Stability
- Refinement needed for specific science use-cases and rigorous V&V

## TRL ~3.5

- Excellent performance on small tests using real & synthetic data
- Tolerant to expected operational environment
- Ready for rigorous V&V





# Protecting the Autonomy: Initial Image Filtration 🕩

**Goal**: Filter inputs that are either uninformative or problematic for downstream algorithms with a sequence of classifiers built for each problem class

# **Requirements:** Low false positive rate in order to minimize missed science



**Dark Images / Artifact** Dark images contain no information. Artifacts will trigger false positives from

downstream algorithms

Atmospheric dust obscures Martian surface

Dust

Examples of Problematic Input from HiRISE



### Miscalibration

Vertical striping caused by miscalibration will trigger false positives from the downstream algorithms



### Featureless

Uniformly featureless landscapes are uninformative and may trigger false positives from downstream algorithms

**Evaluation:** 260 image HiRISE test set with an equal number of classes. Inputs ran sequentially through the system. Positive classifications rejected and negatives allowed through.



### Results

ccuracy:	0.91
alse Negative Rate:	0.02
alse Positive Rate:	0.10

# Localization - Where Are You Looking?

- Images must be <u>localized</u> accurately to be compared across multiple observations.
- Spacecraft pointing/position awareness isn't good enough. Prevents any change detection match-up.
- Use content-based, multi-tiered approach to reach accuracy required by each science use case (1-100 m):



### Novel method for content-based localization

Initial Localization Use Spacecraft Ephemeris Error: O(1 km)

Coarse Alignment Register to Global CTX Mosaic Tie-points Error: O(10-100 m)

Fine Alignment Direct Pixelwise Registration/Comparison Error: O(1-10 m)







Narration: Dr. Gary Doran

# Target Monitor: Pixel-based change in select regions

Target Monitoring: when you know where to look but not when

- Store raw pixels of select regions onboard, perform pixel-level registration to align
- Use a classifier (TextureCam) that uses before/after image pair to detect change
- Pixel-wise classifier uses a context window to tolerate some pixel misalignment
- Send alerts to ground when, after months of calm, something starts to happen



Above: seasonal frost

**Right**: avalanche in polar region

(all images courtesy COSMIC project)



"Before" image registered into the frame of "After" image

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Results of change detection: robust to mis-alignment

Narration: Dr. Gary Doran

# Transient-suggestive Surface Features: Landmarks

"Landmarks" are surface features of interest, including craters, slope streaks, dust devil tracks, etc.

- Several new labeled data sets created with landmarks in CTX (low resolution) and HiRISE (high resolution)
- Enabled assessment of landmark <u>detection</u> methods previously developed (Wagstaff et al., 2012) as well as creation of new landmark <u>classifiers</u> using state-of-the-art technology
- Finding: Random forest achieves same or better performance as convolutional neural network (CNN) on high-resolution HiRISE image tiles, at 50% of the CNN runtime cost. Good news for potential onboard deployment.

<u>Case study</u>: Deployed "fresh impact" CNN classifier across 112,000 CTX images of Mars (Munje et al., 2020). Currently reviewing candidates in order of probability of fresh impact. Of top 1000, 69% are fresh impacts and 9% were previously unknown (new discoveries!)

- CNN out-performed random forest on CTX image classification, and ground-based analysis can accommodate more costly models
- New discoveries are being submitted for HiRISE close-up imaging (via HiWish)
- Significance: Machine learning helps accelerate scientific discoveries (at current rate: increasing fresh impact catalog by 50%)



Fresh impact found in HiRISE image by Random Forest classifier



(all images courtesy NASA or COSMIC project)

Narration: Dr. Kiri Wagstaff

# **Global Agnostic Change Detection**

Global agnostic change detection scours the entire Martian surface for any change, without targeting.

- Salience-based approach: Save "interesting" region summaries for later comparison
- Tiling-based approach: Save summaries for whole planetary grid (HEALPix) for later comparison



Salience-based change detection



Tiling-based change detection



Stability analysis to assess how imaging parameters affect change detection performance; genetic algorithm (GA) to optimize stability.



all images courtesy NASA or COSMIC project)



Conclusion: Both salience-based and tiling-based change detection methods are feasible for the onboard settings with improvements to feature representations and stability of salient landmarks detection.

	Storage	Compute	Stability
Salience-based change detection	Only features of salient landmarks will be stored (smaller database)	More compute resources to find salient landmarks	ID of salient regions over- sensitive to unimportant changes
Tiling-based change detection	Features of all sub- tiles will be stored (larger database)	Tiling algorithm is computational efficient	Robust to small errors in localization



## **Custom Science Payload Example: FOLD**

COSMIC framework can accommodate custom science algorithms.

Example: FOLD (Fast Oriented Layer Detector)

Originally designed for analyzing rover images to find layered rock, the same algorithm can identify layers within orbital imagery.



all images courtesy NASA or COSMIC project

Above: layers detected in a Mastcam Image Right: layers detected from orbital CTX imagery



## **On-Demand Global Mars Database**



### **Onboard Databases of Current Martian Information**



## Final Status & Future Plans

**(**)

Ambitious original goal: all functionality + flight-like system implementation + extensive validation (aim high!) Successes:

- Identified / retired significant risks for onboard change detection
  - Stability (underlying systems need more than accuracy for change; they must also say the same answer, right or wrong, every overflight)
  - Targeted surface feature detection (Demonstrated real science yield, accuracy levels, and stability appropriate to deployment)
  - Tile-based change detection (Feasibility confirmed, requires additional research to summarize tile contents efficiently)
  - Salient-region change detection (Identified as not yet sufficiently stable for change detection purposes)
  - Localization (identified as major dependency, addressed by new content-based, achieved required performance)
  - Onboard storage (identified a few TB onboard storage sufficient to enable global change detection)
- Demonstrated ACTUAL science yield during SRTD
- Contributed heavily to <u>KISS Nebula</u> workshops (superset of COSMIC capabilities)
- Ongoing FY21 formulation plans
  - (formulation) Combine COSMIC + sister SRTD tasks + KISS Nebula workshop into JNEXT submission: a global change monitoring smart camera
  - (pending) Submitted NASA Game Changing Development proposal
  - (pending) NASA AMMOS tile-based change detection system
  - (not selected) Submitted Topical RTD for Onboard Cartographer localization

## **Publications & Press Releases**

### Published Papers & Presentations

Gary Doran, Steven Lu, Maria Liukis, Lukas Mandrake, Umaa Rebbapragada, Kiri L. Wagstaff, Jimmie Young, Erik Langert, Anneliese Braunegg, Paul Horton, Daniel Jeong, and Asher Trockman. <u>COSMIC: Content-based Onboard Summarization to Monitor Infrequent Change</u>. Proceedings of the IEEE Aerospace Conference, 2020.

Michael J. Munje, Ingrid J Daubar, Gary Doran, Kiri Wagstaff, and Lukas Mandrake. <u>Large-Scale Automated Detection of Fresh Impacts on Mars</u> <u>Using Machine Learning with CTX Observations</u>. 11th Planetary Crater Consortium, Abstract #2065, August 2020.

(several datasets and additional manuscripts in process)

Significant NASA/JPL press release resulting in dozens of featured articles:

- "Machine Learning Software is Now Doing the Exhausting Task of Counting Craters On Mars," Universe Today
- https://www.universetoday.com/148165/machine-learning-software-is-now-doing-the-exhausting-task-of-counting-craters-on-mars/
- "NASA Unveils a Cluster of Mars Craters Discovered by Al" on Analytics Insight
- <u>https://www.analyticsinsight.net/nasa-unveils-cluster-mars-craters-discovered-ai/</u>
- Interview by Austin Cross from KPCC (NPR)
- "Scientists are Finding a Ton of New and Mysterious Craters on Mars" on vice.com
- <u>https://www.vice.com/en/article/4ayjkq/scientists-are-finding-a-ton-of-new-and-mysterious-craters-on-mars</u>
- "NASA uses AI to find new craters on Mars" on Clarksville Online
- https://www.clarksvilleonline.com/tag/hirise/
- Follow-up image on HiRISE database: <u>https://www.uahirise.org/ESP\_066018\_1765</u>





## **Virtual Research Presentation Conference**

### Content-based Onboard Summarization to Monitor Infrequent Change (COSMIC)

Principal Investigator: Lukas Mandrake (398) Co-Is: Gary Doran (398), Masha Liukis (398), Steven Lu (398), Umaa Rebbapragada (398), Jimmie Young (398), Kiri Wagstaff (398) Program: Strategic Initiative (6x) Assigned Presentation #



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