

Using edge-detection methods and DEMs to identify and characterize craters on Pluto

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Abstract

While sample return presents the only way to directly determine the age of other planetary surfaces, it is infeasible for outer solar system bodies like Triton, Pluto, and Charon. Thus, using crater size frequency distributions complemented with modeled production functions is the only technique to accurately assess the ages of planetary surfaces. However, until recently, images of Pluto and Charon were limited to low-resolution telescopic images. The New Horizons flyby of the Pluto-Charon system in 2015 has dramatically improved our conception of these worlds by providing, among other data, a wealth of high-resolution optical images. Craters are typically identified in optical images manually, but this is a time-intensive endeavor and subject to differences between individuals. Because crater rims produce distinct changes in brightness, I propose to use edge-detection techniques to identify the crater boundaries in Pluto imagery. I also propose to use Pluto digital elevation models to identify craters based on their distinct bowl-shaped topography. Once a crater has been detected, basic parameters such as the crater's diameter, shape, and depth can be extracted. I will test these algorithms on several images of craters, analyze the successes and failures of each, and make any improvements necessary to extend to further data. Once this has been completed, I will apply each method to an image with many craters as well as other surface expressions. Here, I will determine what features lead to false positives and negatives such that the technique could be improved. Furthermore, I will determine the size frequency distribution and the depth-to-diameter relationship for the given region. From previously modeled crater production functions for Pluto, I will determine the age of the study area. I intend for the final algorithms to be easily scalable to identify craters and compute their basic parameters across the entirety of Pluto's surface in order to create an automated database of all craters.

1. Introduction

Craters have routinely been used to estimate the relative ages of planetary surfaces across the solar system. Because any resurfacing event in a given region of a planetary body covers up any previously produced craters, the number of identifiable craters increases with older surfaces. Thus, by comparing the number of craters in different regions, one can assess their relative ages. As plate tectonics has continued to resurface the Earth, there are very few craters across the planet. Mercury, the moon, and Mars, are pockmarked with holes due to the paucity of resurfacing events over the past few billion years. During the Apollo and Luna missions, lunar samples were returned and dated precisely in laboratories, enabling absolute ages of those locations, and thus crater densities, on the moon [1]. From the size-frequency distribution of observed craters, Neukum et al. [2] showed that the size-frequency distribution of the impacting bodies could be determined. Since then, production functions (the density of craters as a function of crater diameter) have been used ubiquitously to estimate surface ages of planetary bodies.

To determine the density of craters of a given region or globally across a planet, craters of all diameters must be identified. Typically, this is done manually using software to locate and measure crater diameters and shapes [3, 4, 5]. This is time-consuming and prone to systematic error. Recently, however, several crater detection algorithms (CDAs) have been developed to recognize craters from optical remote sensing instruments. Two categories of CDAs have emerged — one which uses high resolution imagery and one which uses digital elevation models (DEMs) [6, 7]. High resolution imagery takes advantage of brightness changes across crater walls and the distinct shapes of craters. From these, edge detection can be used to identify craters and fit ellipses (see, for example, [8]). Machine learning algorithms applied to the high-resolution data take advantage of compiled databases. Both of these techniques are resolution limited and dependent on lighting angle. From DEMs, one can use slopes and curvature in elevation data to identify craters by their bowl-shape [6].

Our goal is to identify craters on Pluto's surface. Prior to the recent flyby (July 2015) of the New Horizons spacecraft [9], this would have been impossible. But the cameras onboard have produced a rich data set of high resolution imagery in which craters large and small can be detected. I intend to use the Long-Range Reconnaissance Orbiter (LORRI) for our edge-detection technique. LORRI is a high-resolution monochromatic imager designed to work under low light conditions (because Pluto receives only 10^{-3} of the sunlight received at Earth) and has a field of view of 0.29° [10]. I will also utilize Ralph - the Multi-spectral Visible Infrared Camera (MVIC). MVIC can pro-

41 duce panchromatic images and has filters for blue, red, and infrared mapping
42 [11]

43 Below, I propose to adapt CDAs to detect craters on Pluto. I will do
44 this using both high-resolution imagery and DEMs. First, a high-resolution
45 image from LORRI with only a single crater will be used to test an edge-
46 detection technique. Then, I will construct a DEM to test an identification
47 algorithm presented by Li et al. [6]. Each algorithm will also determine basic
48 crater properties. I will then assess the merits of each algorithm, improve
49 them as necessary, and repeat the process for several images with different
50 lighting conditions and craters of different morphologies. Finally, I will apply
51 the methods to an image with many craters of various sizes. I will determine
52 an age of this region based on the size-frequency distribution of craters and
53 compare the results to the manually-derived database of Robbins et al. [5].

54 2. Proposed Work

55 ***Project Goal*** - Automatically detect craters on Pluto's surface utilizing
56 edge-detection methods, high-resolution imagery, and digital elevation mod-
57 els. Then, use the resulting crater properties to determine the age of the
58 study region.

59 ***Project Objectives***

- 60 1. Identify single crater through edge detection and determine its basic
61 properties
- 62 2. Create DEM of region of interest and identify single crater through
63 elevation data
- 64 3. Test algorithms on craters with different illumination conditions and
65 morphologies
- 66 4. Automate for a large region

67 ***2.1. Objective 1: Identify single crater through edge detection and determine 68 its basic properties***

69 First, an image of Pluto's surface from the LORRI instrument aboard the
70 New Horizons spacecraft with favorable lighting conditions (visible crater rim
71 and interior) will be selected. Some of the difficulties with this are displayed
72 in Fig. 1. I will crop the selected image to include only a single clear and
73 fresh crater. The aim is to use an image of a crater from Pluto like that
74 shown in Fig. 2.

75 I will apply any necessary filters (such as a Gaussian filter) to reduce
76 noise in the image that could preclude an accurate edge detection of the
77

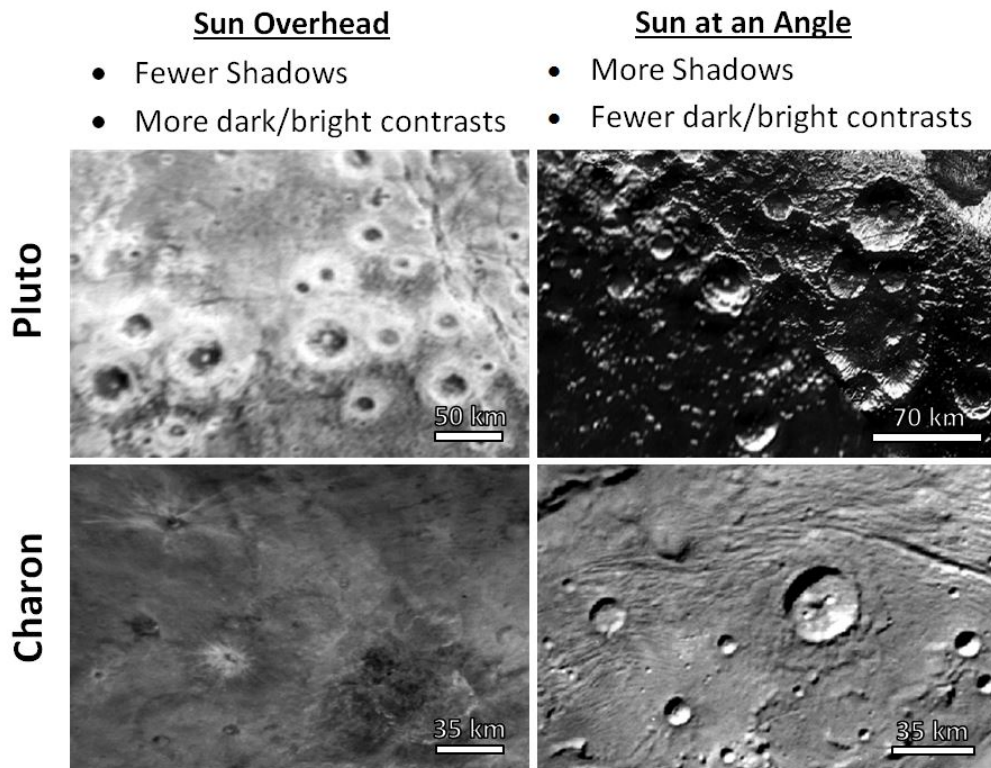


Figure 1: Potential problems resulting from different lighting conditions of LORRI images of Pluto (top row) and Charon (bottom row). (Credit: SwRI/Kelsi Singer <https://blogs.nasa.gov/pluto/2015/10/13/the-impact-of-craters/>)

78 crater. From derivatives in the image intensity the edge of the crater can be
 79 found. Thresholding the image will remove other edges in the image due to
 80 surrounding geologic or topographic boundaries. I will fit an ellipse to the
 81 crater boundaries to determine the shape and diameter of the crater.

82 2.2. Objective 2: Create a DEM and use to identify a single crater

83 I intend to use images from LORRI and MVIC to produce an original
 84 DEM of a region of interest for this project. However, I recognize that time
 85 will be limited this semester, so I present this objective as a possible descope
 86 because a Pluto DEM is readily available at [https://astrogeology.usgs.gov/search/map/Pluto/NewHorizons/Pluto_NewHorizons_Global_DEM_300m_Jul20](https://astrogeology.usgs.gov/search/map/Pluto/NewHorizons/Pluto_NewHorizons_Global_DEM_300m_Jul2017)
 87 17 [12].

88 With the DEM I will follow the procedure outlined by Li et al. [6] to
 89 identify a single crater. This is accomplished by locating depressions with
 90 large slopes. The concavity of the depression is computed from the second
 91

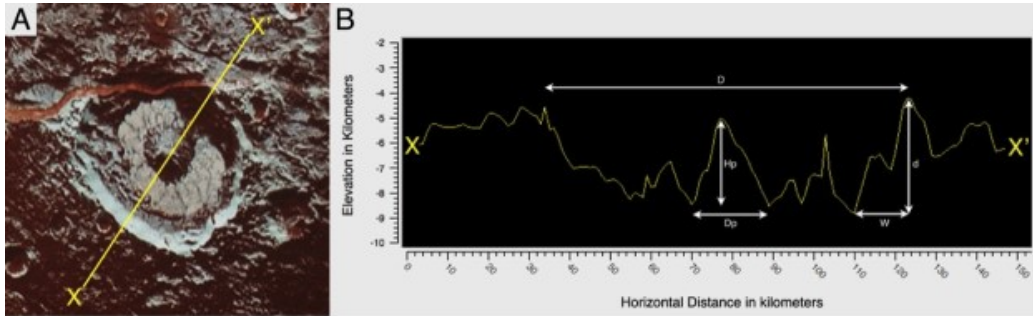


Figure 2: Image of an impact crater on the surface of Pluto from the LORRI instrument. The topography of across the yellow line in (A) is shown in (B). (Credit: Veronica Bray and Paul Schenk, NASA/JHUAPL/SwRI blogs.nasa.gov/pluto/2016/05/06/a-picture-of-pluto-is-worth-a-thousand-words/)

derivative of 4 lines across DEM elevations that have been transformed into a 2D spatial dimensions (A Gaussian filter will be applied to the DEM elevation data). The lines in the pseudo-spatial dimension are 45 degrees apart. Then, the center of a depression is found. This is followed by determining boundary of the depression, which gives the diameter and the shape in addition to the depth from the elevation data. I will compare the results to those obtained in Objective 1.

2.3. Objective 3: Repeat for several images of craters

After developing the CDAs in Objectives 1 and 2, I will apply them to images that will test the robustness of the algorithms. The images will still only cover a small area, but the sizes, shapes, and light conditions will be different. The intention of this objective is to assess problems in identifying craters with the established techniques and amend the algorithms as necessary in order to apply to a broad range of images. I expect problems to arise when an image contains overlapping craters, other circular or bowl-shaped depressions, heavily shadowed craters, etc. This step is crucial to scale the algorithms to larger regions containing a multitude of craters.

2.4. Objective 4: Automate for large region

I will run the proposed CDAs to an image containing 20-50 craters and compare the results to those crater found manually. This will further aid in determining what features lead to false positives and false negatives. What circular or bowl-shaped depressions did the algorithm wrongfully identify as a crater? What craters were missed? This information can be used to construct a decision tree that may help improve detections once scaled up a larger data set.

117 Also, with craters identified within a large region, I will construct a size-
118 frequency distribution which can be compared to those in the literature for
119 Pluto [5]. I will construct an isochron for the region and estimate its age
120 based on the crater production function.

121 Our approach outlined above could be automated for all available images
122 to create a database of all craters identifiable with the resolution of the New
123 Horizons cameras. Training data for a machine learning algorithm requires
124 manually identified craters, but to do this globally is an undertaking out of
125 the scope of this project. However, a global “consensus crater database” [5]
126 has very recently been made public ([https://astrogeology.usgs.gov/search/
127 map/Pluto/Research/Craters/Craters_PlutoCharon_System_Robbins](https://astrogeology.usgs.gov/search/map/Pluto/Research/Craters/Craters_PlutoCharon_System_Robbins)). This
128 database was constructed by the manual identification of craters and deter-
129 mination of their properties by several individuals. Thus, a global analysis
130 is readily scalable once that data set becomes available.

131 3. Relevance

132 Determining accurate ages across the surface of Pluto is necessary to
133 understand the evolution of the planet. The stark contrast between heavily
134 cratered regions and the smooth Sputnik Planum region proves that this
135 evolution contains complex processes. As geologic investigations continue
136 using data from the recent flyby of New Horizons, it will be imperative to
137 have robust methods for detecting craters of all sizes in a fast and consistent
138 manner. Additionally, with a compiled data set of craters, other geologic
139 features in the images or further characterization of the craters (simple vs.
140 complex, layered ejecta, etc.) may be automated. Finally, this endeavor
141 is timely as the release of a crater database has just been made public for
142 Pluto and Charon with which to test our proposed methods against. The
143 techniques I have proposed could be extended to Charon and to other solar
144 system bodies in the future. Specifically, they could be implemented for
145 asteroid Bennu upon the arrival of OSIRIS-Rex next year.

- 146 [1] D. Stöffler, G. Ryder, Stratigraphy and isotope ages of lunar geologic
147 units: Chronological standard for the inner solar system, in: Chronology
148 and evolution of Mars, Springer, 2001, pp. 9–54.
- 149 [2] G. Neukum, B. A. Ivanov, W. K. Hartmann, Cratering records in the
150 inner solar system in relation to the lunar reference system, in: Chronol-
151 ogy and evolution of Mars, Springer, 2001, pp. 55–86.
- 152 [3] G. Michael, G. Neukum, Planetary surface dating from crater size-
153 frequency distribution measurements: Partial resurfacing events and
154 statistical age uncertainty, Earth and Planetary Science Letters 294
155 (2010) 223–229.
- 156 [4] S. J. Robbins, B. M. Hynek, A new global database of mars impact
157 craters 1 km: 1. database creation, properties, and parameters, Journal
158 of Geophysical Research: Planets 117 (2012).
- 159 [5] S. J. Robbins, K. N. Singer, V. J. Bray, P. Schenk, T. R. Lauer, H. A.
160 Weaver, K. Runyon, W. B. McKinnon, R. A. Beyer, S. Porter, et al.,
161 Craters of the pluto-charon system, Icarus 287 (2017) 187–206.
- 162 [6] B. Li, Z. Ling, J. Zhang, Z. Wu, Automatic detection and boundary
163 extraction of lunar craters based on lola dem data, Earth, Moon, and
164 Planets 115 (2015) 59–69.
- 165 [7] A. L. Salih, M. Mühlbauer, A. Grumpe, J. Pasckert, C. Wöhler,
166 H. Hiesinger, Mapping of planetary surface age based on crater statistics
167 obtained by an automatic detection algorithm., International Archives
168 of the Photogrammetry, Remote Sensing & Spatial Information Sciences
169 41 (2016).
- 170 [8] T. Barata, E. I. Alves, J. Saraiva, P. Pina, Automatic recognition of
171 impact craters on the surface of mars, in: International Conference
172 Image Analysis and Recognition, Springer, pp. 489–496.
- 173 [9] S. A. Stern, The new horizons pluto kuiper belt mission: an overview
174 with historical context, in: New Horizons, Springer, 2009, pp. 3–21.
- 175 [10] A. Cheng, H. Weaver, S. Conard, M. Morgan, O. Barnouin-Jha, J. Boldt,
176 K. Cooper, E. Darlington, M. Grey, J. Hayes, et al., Long-range recon-
177 naissance imager on new horizons, in: New Horizons, Springer, 2009,
178 pp. 189–215.

- 179 [11] D. C. Reuter, S. A. Stern, J. Scherrer, D. E. Jennings, J. W. Baer,
180 J. Hanley, L. Hardaway, A. Lunsford, S. McMuldroch, J. Moore, et al.,
181 Ralph: A visible/infrared imager for the new horizons pluto/kuiper belt
182 mission, *Space Science Reviews* 140 (2008) 129–154.
- 183 [12] J. M. Moore, W. B. McKinnon, J. R. Spencer, A. D. Howard, P. M.
184 Schenk, R. A. Beyer, F. Nimmo, K. N. Singer, O. M. Umurhan, O. L.
185 White, et al., The geology of pluto and charon through the eyes of new
186 horizons, *Science* 351 (2016) 1284–1293.