

# IN-SITU TERRAIN ANALYSIS FOR PLANETARY ROVERS

Gabrielle Hedrick<sup>a</sup>, Dylan Covell<sup>a</sup>, Yu Gu<sup>a</sup>

<sup>a</sup>*Department of Mechanical and Aerospace Engineering, 1306 Evansdale Drive, PO Box 6106 Morgantown, West Virginia  
26506-6106  
glc0006@mix.wvu.edu, dmcovell@mix.wvu.edu, yu.gu@mail.wvu.edu*

## Abstract

NASA is proposing a Mars Sample Return mission to bring back material from Mars, proposed to be completed in 100 to 200 Martian days (or sols). Due to landing uncertainty, the rover would potentially need to cover distances up to 10 km to collect the samples. Recent surface missions have been limited in their driving capabilities, such as drive time and distances covered, due to a variety of factors including lack of autonomy and communication with Earth, and uncertainty about the environment. To increase drive time, it is crucial to integrate methods to gain information about the terrain into the path planning process and facilitate traversability assessments. This paper presents a rover-mounted instrument capable of obtaining good knowledge of the terrain by measuring the in-situ parameters cohesion  $c$  and angle of internal friction  $\phi$ . The instrument selected is a pocket shear vane, that gives shear stresses for given normal stresses. They are then used to compute the Mohr-Coulomb failure criterion from which the terrain parameters are derived. It is adapted to be mounted on a rover and controlled remotely, with a capability of generating stresses up to almost 400 kPa. Tests are conducted on reference soils in the laboratory as well as in-situ, with the instrument mounted on linear guide rails, and results obtained are close to the expected parameters of each soil tested. The main contributions of this work are: 1) to provide a planetary surface rover an easy option to gather intrinsic soil parameters that can be used to identify a terrain when needed; 2) to provide a method to identify a terrain with little to no human intervention; 3) to obtain results in a timely manner (a few minutes per measurement) to allow a fast traverse rover to quickly make a decision regarding its path given the terrain characteristics.

*Keywords:* Mars Rover, Shear Vane, Cohesion, Angle of Internal Friction

## 1. Introduction

Space missions are becoming increasingly complex, as shown by the proposed Mars Sample Return (MSR) mission (Witze, 2014) to be completed in a shorter amount of time than previous surface mission. Such a mission requires enhanced autonomy compared to current surface robots, to enable further and faster driving than existing Mars rovers, with limited communication.

The next journey to Mars is a proposed fetch rover to gather and bring back samples collected by Perseverance. The sample return proposal brings a new level of complexity to engineers: it is mostly designed to bring back samples (vs. performing science experiments), unlike any other missions to the red planet, and is planned to be completed in less than one Martian year (687 Martian days or sols) (MEPAG, 2008). The landing ellipse for the Perseverance rover is roughly 11 km by 8 km (Golombek et al., 2017), and assuming a similar landing area for MSR, the fetch rover might have to traverse potentially great distances in a short amount of time. A similar sized rover such as Opportunity drove an average of 3 km in 1 Earth year (Schroeder, 2019) with the designed capability of driving four hours in one sol (Biesiadecki et al., 2007). Therefore, MSR calls for more autonomy beyond the pre-planning done by humans.

To understand better how autonomy is becoming a necessary part of NASA's future missions, it is important to know how the past and current Mars missions are organized in terms of daily planning. A typical sol (about 40 minutes longer than a day on Earth) includes one downlink and one uplink through

the Deep Space Network (DSN), which means that Earth communicates with the rover only a couple of times a day (Bajracharya et al., 2008). The downlink provides the team with images and other results from the previous sols that they can use to navigate, and rover movements are therefore limited to line-of-sight driving. There is an autonomous navigation feature, “AutoNav”, that allows the rover planners to drive beyond what is seen on images, but it is very slow - for MER, speeds could drop down to 10 m/hr (Biesiadecki et al., 2007). Moreover, driving is not scheduled every sol (Gaines et al., 2016) resulting in overall speed and distances covered not adequate for MSR.

Hence, there is a need for increased autonomous driving distance due to the potential size of the landing ellipse and the plan for a duration of only about 100 to 200 Martian sols (i.e., Martian days) for the MSR mission (Klein et al., 2014). To increase driving distances and go beyond line-of-sight from downlinked images, it is necessary to know the terrain and assess its traversability, i.e., its ability to sustain a driving robot without reaching failure (Papadakis, 2013). Some research has already been performed regarding a terrain-aware path planning that efficiently takes into consideration terrain information as the rover drives (Hedrick et al., 2020). This work focuses on a method to gather in-situ information about the terrain that can be utilized in the aforementioned path planning, and significantly lower the uncertainty related to traversability. The main idea is to support the prediction of performance on the terrain ahead and help the rover make adequate decisions regarding its path using reliable information about the soil obtained from in-situ measurements.

This paper is organized as follows: after presenting relevant research, a problem statement and contributions will be detailed. This section will be followed by the technical approach, results of this work, and finally the conclusion.

## 2. Related Work

Planetary surface exploration is challenging for many reasons, including the lack of knowledge about the soil. A lot of work has been dedicated to modeling terrains in order to improve rover performance during such missions, ranging from remote sensing (such as thermal inertia, radar, optical methods, etc...) to in-situ analysis of terrain, directly or indirectly (such as rover sub-systems, instrumentation, etc...) (Chhaniyara et al., 2012).

One important approach to traversability analysis involves terramechanics, i.e., the study of the interaction between the wheels and the soil. Vibrations induced by the terrain interacting with the wheels during driving has been suggested as a mean to classify terrains (Brooks & Iagnemma, 2005). Wheel slip, torque, sinkage and drawbar pull are among the parameters suggested to sense the terrain (Iagnemma et al., 2003). More research was done towards planetary exploration in subsequent years to improve on autonomy and include terrain analysis into path planning, such as computing a path, then evaluating the terrain based on wheel/ground interaction, and finally recomputing the path if necessary (Ishigami et al., 2007). It was later suggested to integrate a full dynamic model of the robot into the path planning, comprising of a sub-model of the vehicle to get a mobility profile, and a terramechanics sub-model to obtain interaction forces on deformable soils. The research proposes to compute several feasible paths, run a dynamic simulation for each candidate and calculate a dynamic mobility index, comprised of roll, pitch, slip, elapsed time and energy required to reach a position (Ishigami et al., 2011).

Another approach involves the use of sensors and cameras to analyze the terrain: for example, using on-board sensors such as gyros, accelerometers, encoders, motor current, voltage, ultra-sonic or infrared sensors, the data are fed into neural networks that characterize and classify the terrain between five different options (gravel, sand, asphalt, grass and dirt) (Ojeda et al., 2006). Similarly, another model suggests using on-board cameras to remotely classify terrains and predict slip. The terrains are then divided into three main categories related to mobility performance: traversable, not traversable, and uncertain. This system has been specifically intended for planetary rovers after seeing the difficulties encountered on Mars (Helmick et al., 2008). Research has also been done towards making the rover a more autonomous explorer, gathering valuable scientific information autonomously and utilizing it to navigate (Girdhar & Dudek, 2016). The idea to use instruments to gain information about the terrain had already been suggested (Chhaniyara et al.,

2012), with the study of several hand-held instruments (e.g., cone penetrometers, shear vanes) as potential candidates for astronauts or rovers (Rahmatian & Metzger, 2010). It was found that a modified shear vane could acquire good insight into the soil strength. A similar instrument, called a cone-vane penetrometer was actually used on the Lunakhod rover to estimate the bearing capacity of the soil (Zacny et al., 2010) but did not make any diagnosis regarding other properties such as soil strength. A recent article proposed to integrate a more complex instrument, a spectrometer, to collect information to support rover path planning. In this scenario, the initial map is a belief map of geological units assessed by scientists (potential hypothesis about the soil). The rover updates its route as it is gaining information about the terrain (obtained from the spectrometer) through Bayesian inference (Candela et al., 2017). This paper was inspired by previous work on incorporating Bayesian processes to make planetary rovers more independent in their scientific exploration (Arora et al., 2017). However, the drawback of this method is the cost of using such an instrument (about three hours each time (Gellert et al., 2009), not counting the energy expenditure), and if it is well adapted for scientific mission, it might not be adequate for missions such as the sample return rover.

Finally, a third approach to traversability analysis worth mentioning is the use of orbital imagery. Recently, several authors have looked into predicting traverse performance by taking into account orbital data such as HiRISE (High Resolution Imaging Science Experiment) or slope in a software called Mars Terrain Traversability Tool (Ono et al., 2016). MTTT uses a terrain classifier, Soil Property and Object Classification (SPOC) that analyses HiRISE images to classify different terrains into categories (Rothrock et al., 2016). These terrain types are coupled with rock abundance (Cumulative Fractional Area (Golombek & Rapp, 1996) or CFA), hazards and slope to predict rover speed (Ono et al., 2016). However, the SPOC classifier used in MTTT has a small chance to misclassify the terrains (Rothrock et al., 2016) and the rover would need to account for such scenarios, e.g., by using terramechanics and/or instrumentation to analyze the terrain directly from the surface.

### 3. Problem statement and contribution

The objective of this work is to equip a rover with the capability of gathering in-situ terrain information at an accuracy equal to or greater than a human operator to support autonomous path assessment and planning. It would allow the vehicle to make decisions regarding its traverse before mobility difficulties could be encountered, by choosing strategic locations to gather meaningful information about the environment. This new knowledge would be used to update the map of the landing site via extrapolation techniques as proposed in Hedrick et al. (2020) and would be a useful input for local planning and replanning along the remaining route. The main constraint is to have human operated experiments widely utilized on Earth adapted to function autonomously on a rover. That is, any hand-held tool needs to be mounted on the rover, and any experiments must be performed remotely. An important assumption is that there is no significant difference between Earth and Mars, meaning, an instrument on Earth and its manipulation on a given terrain will result in similar conclusions when operated on Mars on a similar soil. For example, a cone penetration test on sand is interpreted in the same manner on both planets and results are thus assumed to be valid when carried out on Mars.

The main contributions of this work are the following: 1) to provide an easy option to gather intrinsic soil parameters that can be used to identify a terrain when needed; 2) to give a way of identifying a terrain with little to no human intervention; 3) to obtain results in a timely manner (i.e., a few minutes) to allow a fast traverse rover to quickly make a decision regarding its path given the terrain characteristics.

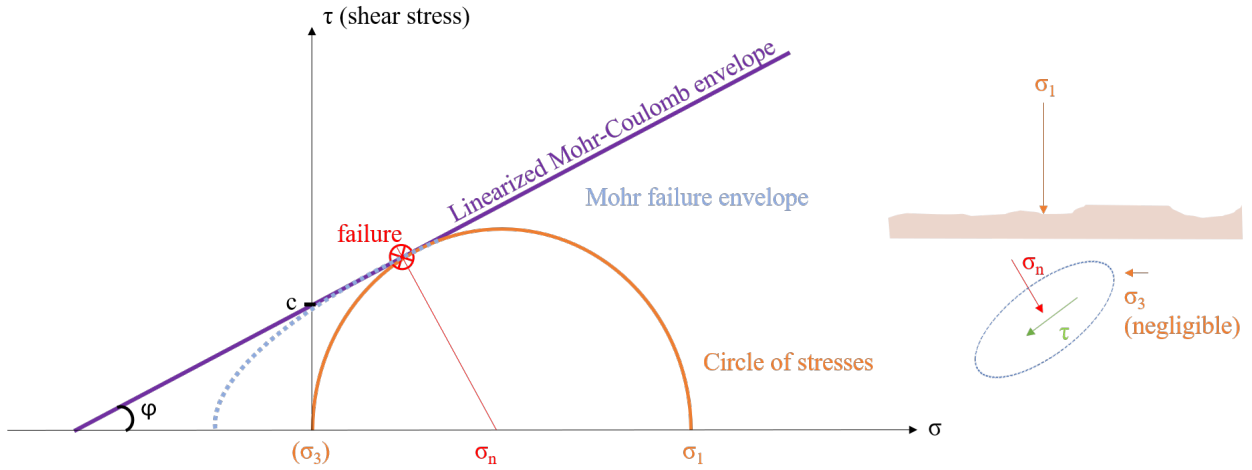
### 4. Soil parameters

Cohesion  $c$  and angle of internal friction,  $\phi$  characterize soil strength, and are both intrinsic properties of the terrain. Cohesion is the ability for a material to hold itself together and corresponds to the cohesive strength of a terrain. The angle of internal friction gives an estimate of the friction due to the material itself. For example, a terrain with zero cohesion, such as loose rock debris, will still resist deformation due to friction, unless a stress is applied (Melosh, 2011). These two parameters are extremely important in geology,

114 and can be used to classify planetary materials (Sullivan et al., 2011). They also matter in engineering as  
 115 they can be utilized to retrieve bearing capacity and soil failure point, two important variables that might  
 116 need to be estimated in the case of a planetary rover to avoid unexpected collapsing of the soil under the  
 117 wheels. Cohesion and friction are related to the normal and shear stresses via the following equation:

$$\tau = c + \sigma_n \tan \phi \quad (1)$$

118 where  $\tau$  is the maximum shear stress the soil can handle, and  $\sigma_n$  is the normal stress. Equation 1 represents  
 119 the linear approximation of the Mohr-Coulomb failure envelope (Zoback, 2010). The intersection of the  
 120 Mohr-Coulomb failure with the circle of stresses given by  $\sigma_1$  (vertical stress) and  $\sigma_3$  (confining pressure)  
 121 determines the point of failure. Since this study focuses on only one point along the envelope (the point  
 122 of failure), the linearization is an acceptable approximation (Labuz & Zang, 2012). For a driving rover,  
 assumptions can be made that  $\sigma_3$  is negligible and  $\sigma_1$  is the load of the vehicle (Sullivan et al., 2011). The



**Fig. 1:** Mohr-Coulomb envelope and failure point shown for a vertical load ( $\sigma_1$ ) on a terrain. The confining pressure is negligible.

123 line perpendicular to the point of failure and intersecting the x-axis gives  $\sigma_n$ . Assuming a known cohesion  
 124  $c$  and angle of internal friction  $\phi$ , the Mohr-Coulomb failure envelope will give the maximum load (weight,  
 125 or  $\sigma_1$ ) and shear stress (wheel rotation at constant velocity (Sullivan et al., 2011), or  $\tau$ ) a certain planetary  
 126 terrain can sustain without failing. The concept is illustrated in Fig.1.  
 127

## 128 5. Technical Approach

### 129 5.1. Hand-held pocket shear vane

130 By knowing selected terrain parameters (i.e., cohesion and friction) and the load of the vehicle, the  
 131 Mohr-Coulomb curve and circle of stresses can be obtained to find the failure point. There exists direct and  
 132 indirect methods for measuring these parameters. Direct methods include laboratory testing where samples  
 133 of soil are subjected to normal stresses and confining pressure (Bishop & Henkel, 1962) in what is called  
 134 a triaxial test. Indirect methods include shear vane testing, with three main instruments available on the  
 135 market: the pocket shear vane, the field shear vane and the geovane. Shear testing consists of applying  
 136 torque to a bladed probe, or vane, until the soil yields (i.e, no resistance is met, when the soil has reached  
 137 failure). The shear strength value can be read on the instrument dial, with the pointer staying in place when  
 138 failure occurs (Fig.2).

139 The amount of required torque varies depending on the blade geometry and depth of testing. The geovane  
 140 and field vane are lengthy instruments (minimum 30 cm with extension rods available), whereas the pocket



141 shear vane is compact and more applicable to this research, as it will be mounted on a rover. It was therefore  
 142 chosen as a field instrument for the rover (Fig.2).



**Fig. 2:** Pocket Shear Vane Instrument with different blades (optional) (Credit: Gilson inc.)

143 However, this instrument gives only the value for shear stress (i.e.,  $\tau$ ), while the normal stress is applied  
 144 by a human operator. As suggested in Rahmatian & Metzger (2010), it can be modified to bear a controlled  
 145 amount of weight (and therefore, a controlled normal stress can be applied). The modification is shown in  
 146 Fig.3 and consists of a 3D printed part with slots tailored to calibration weights.



**Fig. 3:** Modified pocket shear vane tester including: a 3D printed cast with weight slots; calibration weights.

147 The maximum total mass available is 0.55 kg and the minimum is 0.17 kg. By applying different weights  
 148 to obtain several measurements at the same location, a small range of normal stresses and their associated  
 149 shear stresses can be plotted. The Mohr-Coulomb envelope can be obtained from these measurements, the  
 150 resulting slope and y-axis intersection would give  $\phi$  and  $c$ , respectively.

## 151 5.2. Mounted pocket shear vane

152 To make the pocket shear vane remotely operable, the human actions must be simplified as electronic  
 153 motions and the stresses obtained digitally. As seen in Fig.4, the instrument is integrated into an actuated  
 154 payload equipped with a potentiometer and a pair of identical shear load cells oriented perpendicular to each  
 155 other.

156 This prototype was designed with a focus on minimizing undesired skew and payload footprint while  
 157 remaining simple to manufacture. The T-slot guide rails provide an adaptable means of connecting this  
 158 sensor to the Fast Traversing Autonomous Rover built at West Virginia University's Interactive Robotics  
 159 Laboratory. Additionally, the T-slot linear bearings contain no moving parts, an ideal trait for the dusty  
 160 Martian environment. The linear servos are used over a stepper motor and lead screw to protect from dust  
 161 and debris.

162 Once the payload frame is resting on the ground, a linear servo presses down on the shear vane and the  
 163 resulting normal force is read via the load cell oriented normal to the ground. The output voltage of the  
 164 load cell is amplified before reaching the arduino. The HX711 library takes this read voltage and outputs

165 an equivalent weight. The rated repeatability of the load cells is +/- 10 g or 0.05% of the maximum range  
 166 of 20 kg. This value is then converted to a normal stress using Eq.2:

$$\sigma_n = \frac{Wg}{S} \quad (2)$$

167 Where  $\sigma_n$  is the normal stress,  $W$  is the weight of the instrument,  $g$  is the gravity ( $9.81m/s^2$  for Earth)  
 168 and  $S$  is the surface area being forced into the soil and covered by the blades (see Fig. 4a).  $S$  is specified  
 169 in the instrument's instruction manual and is equal to  $0.000491m^2$ . The measurement from the shear vane  
 170 is given in  $kg/cm^2$  and converted to  $kg/m^2$ . The shear stress is obtained from the measurement simply by  
 171 multiplying by  $g$ .

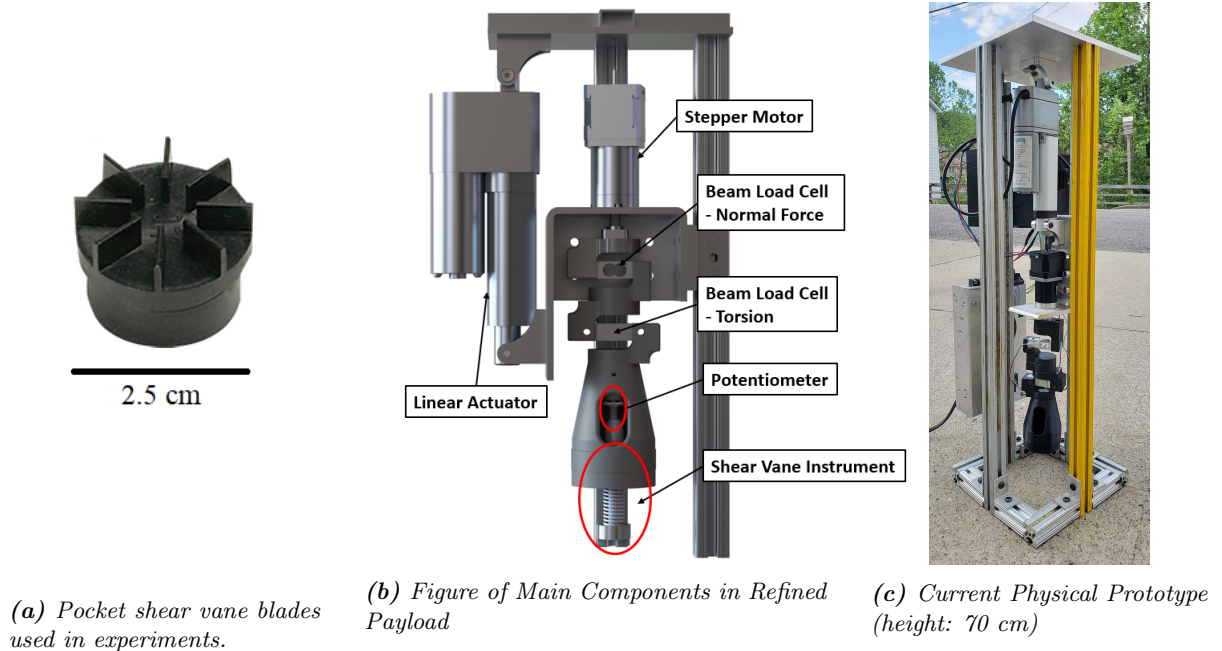
172 The resulting shear stress can then be obtained via two methods:

- 173 • The potentiometer output, that models the position of the shear vane dial.
- 174 • The output of the load cell aligned with the torsional neutral axis is amplified before the arduino reads  
 175 the voltage. The HX711 library takes this voltage and outputs an equivalent weight, which is then  
 176 converted to a shear stress using Eq.3:

$$\tau = \frac{w_m * d * r}{J} \quad (3)$$

177 Where  $\tau$  is the shear stress ( $Pa$ ),  $w_m$  is the measured weight ( $N$ ),  $d = 0.04m$  is the distance between  
 178 the effective and reactive  $w_m$  (the working length of the load cell),  $r = 0.00635m$  is the distance  
 179 between the neutral axis of the load cell and the stressed surface, and  $J = 1.10954 \times 10^{-9}m^4$  is the  
 180 second polar moment of inertia. Therefore, Eq.3 can simply be rewritten as:

$$\tau = 2.289 \times 10^5 w_m \quad (4)$$

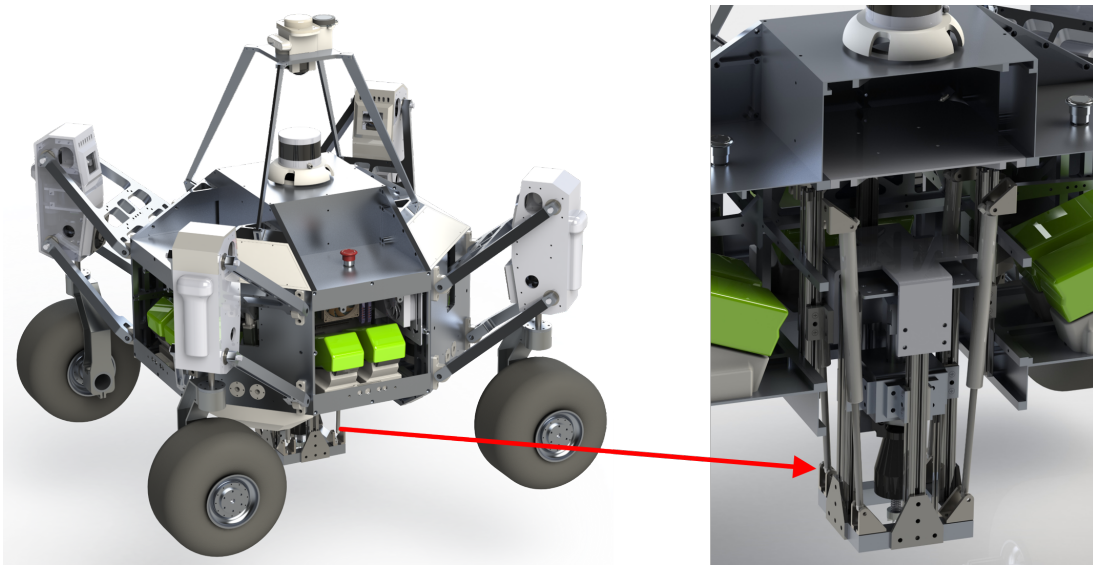


**Fig. 4:** Automated Shear Vane Test Prototype

181 This prototype is composed of the electronic hardware found in Table 1 and controlled via MATLAB<sup>®</sup>  
 182 through an Arduino Uno. The Arduino library for the load cell amplifier (Giacoboni, 2020) greatly enables  
 183 a seamless integration of the load cells into MATLAB<sup>®</sup>.

**Table 1:** Prototype Hardware and Estimated Maximum Energy Consumption

Component	Maximum Power Consumption (Watts)	Company
Arduino Uno R3	2.5	Arduino
CTS 282T33L502A26C2 Potentiometer	0.005	Digi-Key
FA-PO-150-12-2 Linear Actuator	60	Firgelli Automation
High Current DC Motor Driver	0.08	Firgelli Automation
667oz-in NEMA-17 Stepper Motor	20.4	Phidgets
CZL635 20 kg Load Cell	0.025	Phidgets
DRV8825 Stepper Motor Driver	0.005	Pololu
HX711 Load Cell Amplifier	0.008	Sparkfun

**Fig. 5:** mounted pocket shear vane Device inside Fast Traverse Rover

184 The first prototype of the refined payload shown inside the center rover compartment (Fig.5) is currently  
 185 estimated to weigh 7.3 kg and take up a 39.4 cm tall x 16.8 cm wide x 22.9 cm deep box.

## 186 6. Results

187 The modified pocket shear vane was tested in known soils (per United Soil Classification System, or  
 188 USCS), referred to as controlled samples, to verify that the modifications would lead to adequate results  
 189 from which cohesion and angle of internal friction could be retrieved (i.e, Mohr-Coulomb envelopes). Several  
 190 tests were run for each load for each sample (five per controlled sample), and a Mohr-Coulomb type curve fit  
 191 to the results before being linearized to compute the intrinsic parameters. The loads are obtain for masses  
 192 ranging from 0.176 g to 0.546 g. The initial curve to be linearized is a power equation of the form:

$$\tau = a\sigma_n^b + c \quad (5)$$

193 Where  $\tau$  is the shear stress,  $\sigma_n$  is the normal stress, and  $a, b, c$  are constants defining the curve. The  
 194 linearization is performed by fitting a tangent to the curve after it has flattens.

195 The tested soils are the following (the official United Soil Classification System (USCS) label is specified  
 196 for each one):

- 197 • Fine sand (USCS SW, SP)
- 198 • Clay of low plasticity (USCS CL)
- 199 • Silt loam, compacted (USCS ML, OL, MH, OH)

200 The instrument was then tested in different soils the field with only one to three sets of data per load, to  
 201 stay as close as possible to the conditions it would operate in when deployed on Mars. The following terrains  
 202 (estimated by human operators) were tested:

- 203 • Silt loam, saturated (USCS ML, OL, MH, OH)
- 204 • Sand (USCS SW, SP)

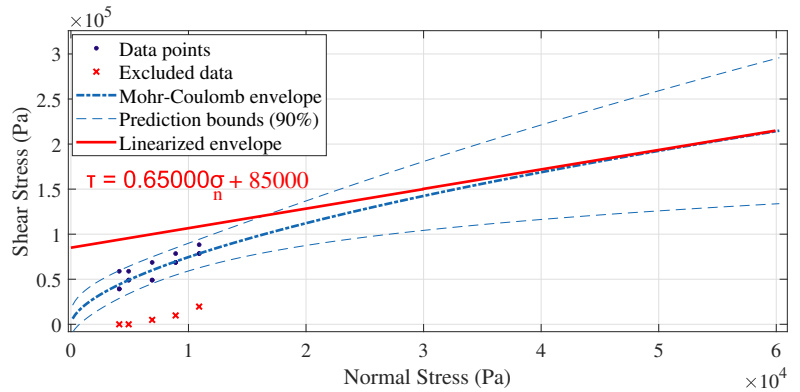
205 The normal stress is obtained from the range of weights available, using Eq.2. Certain data points were  
 206 eliminated due to the inadequacy of the testing conditions.

207 *6.1. Hand-held pocket shear vane: controlled samples*

208 Dry clay is of low plasticity, with an expected cohesion of  $86kPa$  when compacted, and an angle of  
 internal friction between  $27^\circ$  and  $35^\circ$ . The tests results for compacted dry clay are presented in Fig.6.



(a) Dry clay of low plasticity. Expected  $c = 86kPa$  and  $27^\circ \leq \phi \leq 35^\circ$ ,

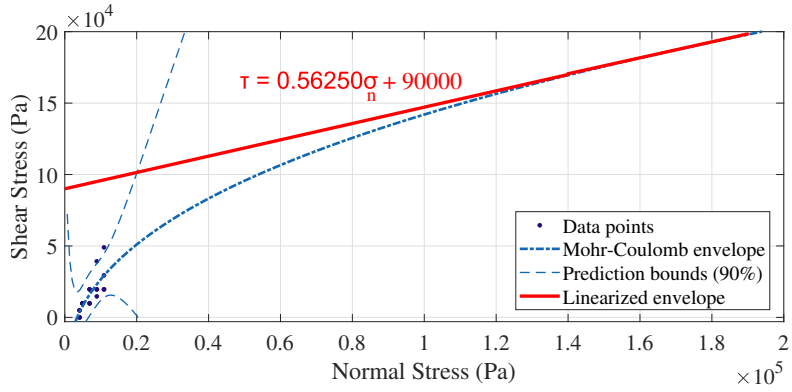


(b) Mohr-Coulomb envelope and linearization giving  $c = 85kPa$  and  $\phi = 33^\circ$ .

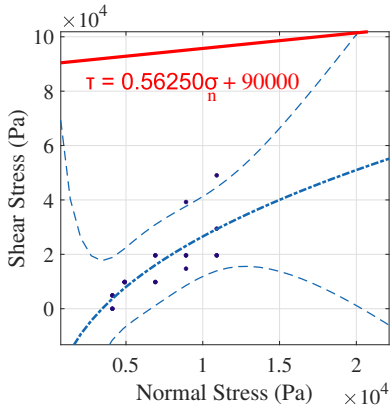
**Fig. 6:** Clay of low plasticity, compacted. Results give  $c = 85kPa$  and  $\phi = 33^\circ$ . Some data points overlap with each other.

209 The tests gave a linearized envelope showing a cohesion  $c = 85kPa$  and an angle of internal friction  
 210  $\phi = atan(0.6500) = 33.02^\circ$ . The initial Mohr-Coulomb curve is given by Eq.5 with the following coefficients  
 211 (with 95% confidence bound)  $a = 327.8(-312.6, 968.2)$ ,  $b = 0.5892(0.3713, 0.8072)$  and  $c = 0$ . The goodness  
 212 of fit for the non-linearized curve for dry clay is characterized by  $R^2 = 0.7469$ .

214 Next, compacted silt loam was tested. The sample was placed in a container and manually compacted  
 215 by applying a load to the sample before the tests were conducted. Compacted silt loam has a cohesion of 60  
 216 to 90  $kPa$  and an angle of internal friction between  $25^\circ$  and  $32^\circ$ . The results are presented in Fig.7.



(a) Mohr-Coulomb envelope and linear approximation giving  $c = 90kPa$  and  $\phi = 29^\circ$



(b) Zoom on the part of the curve with data points.



(c) Silt Loam. Expected  $c = 60 - 90kPa$  and  $25^\circ \leq \phi \leq 32^\circ$ .

**Fig. 7:** Hand-held instrument tested in compacted silt loam. Results give  $c = 90kPa$  and  $\phi = 29^\circ$ . Some data points overlap with each other.

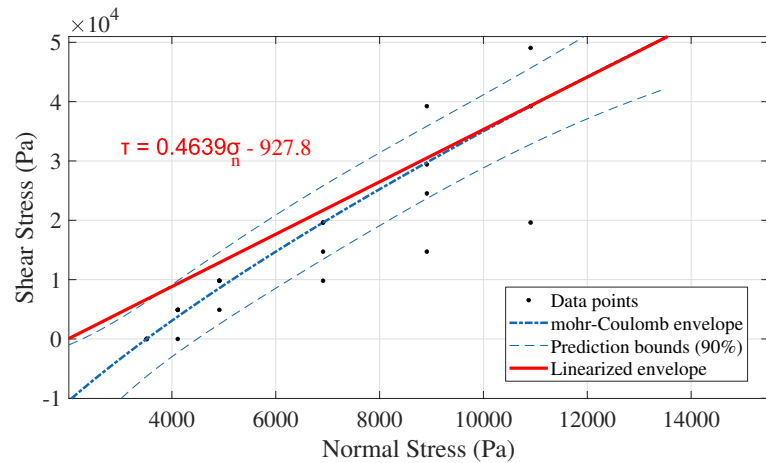
217 The compacted silt loam led to the following results after linearization of the Mohr-Coulomb curve:  
 218  $c = 90kPa$  and  $\phi = atan(0.5625) = 29.36^\circ$ . The initial Mohr-Coulomb envelope is given by Eq.5 with the  
 219 following coefficients (with 95% confidence bound)  $a = 1960(-6.63e04, 7.022e04)$ ,  $b = 0.3983(-2.671, 3.468)$   
 220 and  $c = -5.027e04(-5.559e05, 4.554e05)$ . The goodness of fit for the non-linearized curve for compacted silt  
 221 loam is characterized by  $R^2 = 0.6234$ .

222 The third sample material tested was well graded, fine grain, cohesionless sand with an expected angle  
 223 of internal friction ranging between  $36^\circ$  and  $41^\circ$ .





(a) Fine grain, well graded sand. Expected  $c = 0kPa$  and  $36^\circ \leq \phi \leq 41^\circ$



(b) Mohr-Coulomb envelope and linear approximation giving  $c = -0.92kPa$  and  $\phi = 25^\circ$

**Fig. 8:** Tests of the hand-held shear vane in fine grain sand material of known parameters.

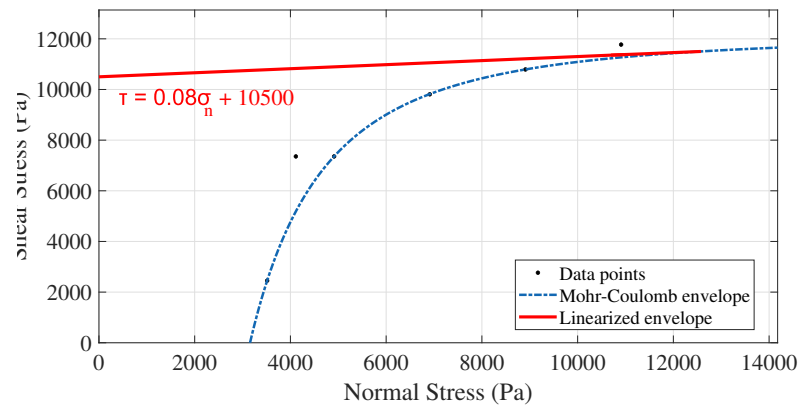
224 The results show a tangent with a slope (i.e., cohesion)  $c = -0.9278kPa$  and an angle of internal  
 225 friction  $\phi = \text{atan}(0.4639) = 24.88^\circ$  for fine, well graded sand. The Mohr-Coulomb curve before linearization  
 226 is given by Eq.5 with the following coefficients (with 95% confidence bound)  $a = 102.7(-639.8, 845.2)$ ,  
 227  $b = 0.7039(-0.003083, 1.411)$  and  $c = -3.217e04(-8.106e04, 1.671e04)$ . The goodness of fit for the non-  
 228 linearized envelope for fine, well graded sand is characterized by  $R^2 = 0.9463$ .

## 229 6.2. Hand-held pocket shear vane: field testing

230 Two tests were conducted in the field. A sand pit under dry conditions was chosen, where the cohesion  
 231 is expected to be  $0kPa$  (cohesionless) and the angle of internal friction is usually between  $37^\circ$  and  $38^\circ$  for such  
 232 material. The results are presented in Fig.9b, where the resulting Mohr-Coulomb envelope is shown.



(a) Sand pit in which tests were performed. Expected values are  $c = 0kPa$  and  $\phi = 37^\circ - 38^\circ$ .



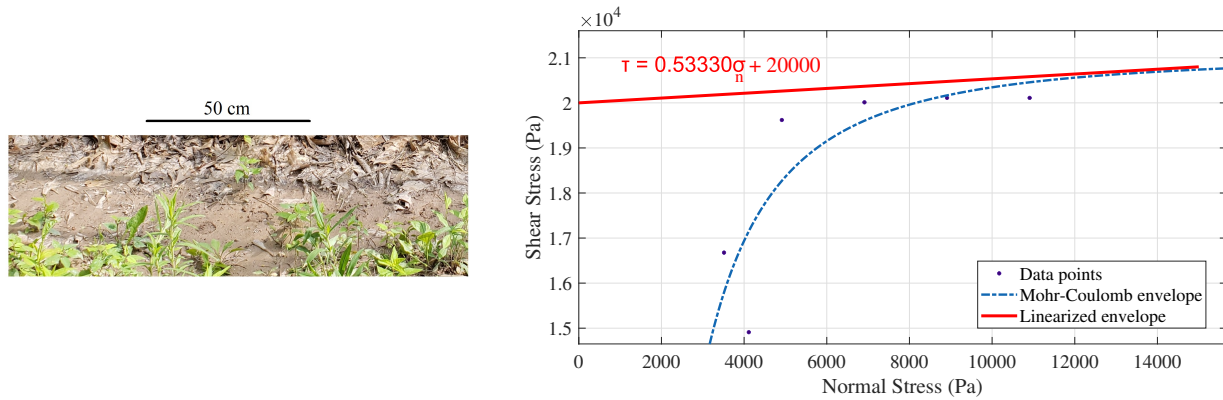
(b) Mohr-Coulomb envelope and linear approximation giving  $c = 10.5kPa$  and  $\phi = 4.57^\circ$

**Fig. 9:** Test of the hand-held pocket shear vane in cohesionless sand under dry conditions.

233 From the plotted data in Fig.9, the retrieved cohesion from the linearized Mohr-Coulomb envelope is  $c =$   
 234  $10.5kPa$  and the angle of internal friction is  $\phi = \text{atan}(0.0800) = 4.57^\circ$ . The initial Mohr-Coulomb curve is

235 given by Eq.5 with the following coefficients (with 95% confidence bound)  $a = -2.615e11$ ,  $(-5.793e12, 5.27e12)$ ,  
 236  $b = -2.095(-4.723, 0.5321)$  and  $c = 1.217e04(7598, 1.675e04)$  The goodness of fit of the Mohr-Coulomb en-  
 237 velope is characterized by  $R^2 = 0.9652$

238 The second experiment was conducted in saturated silt loam. The results are presented in Fig.10.



(a) Saturated silt loam, expected  $c = 10\text{kPa}$  –  
 $20\text{kPa}$  and  $\phi = 25^\circ - 32^\circ$

(b) Mohr-Coulomb envelopes and linear approximation giving  $c =$   
 $20\text{kPa}$  and  $\phi = 28^\circ$ .

**Fig. 10:** Hand-held instrument tested on saturated Silt Loam, with geological unit being quaternary alluvial mix of mostly silt mixed with fine sand and clay.

239 The retrieved cohesion from Fig.10 is  $c = 20\text{kPa}$  and friction angle is  $\phi = \text{atan}(0.5333) = 28.07^\circ$ .  
 240 The Mohr-Coulomb curve is given by Eq.5 with the following coefficients (with 95% confidence bound)  
 241  $a = -2.544e10(-2.006e12, 1.955e12)$ ,  $b = -1.885(-11.6, 7.829)$  and  $c = 2.108e04(1.007e04, 3.209e + 04)$ .  
 242 The goodness of fit of the Mohr-Coulomb envelope for this test is characterized by  $R^2 = 0.6821$ .

**Table 2:** Results of cohesion  $c$  and angle of internal friction  $\phi$  from testing (hand-held instrument) compared to expected values (Obrzud, 2010; MnDOT, 2007). In green are adequate results, and in red, inadequate results.

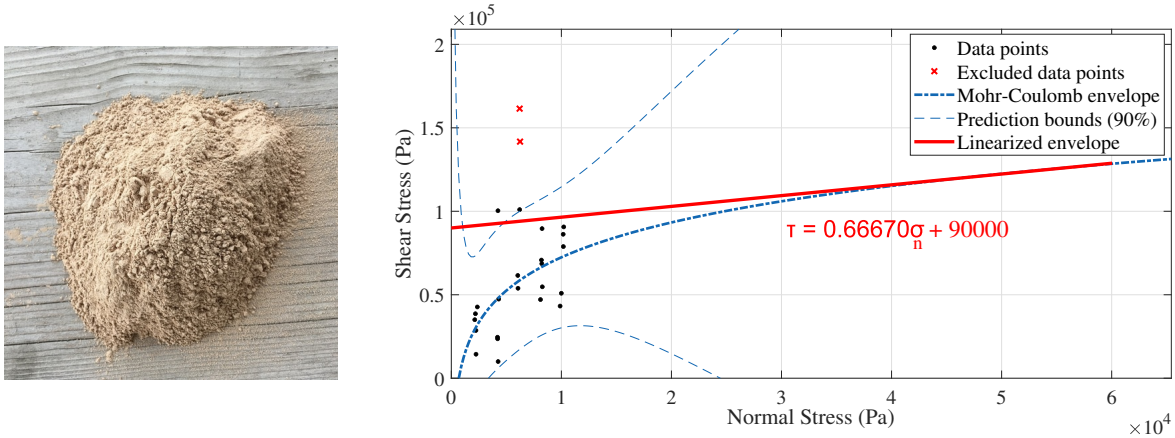
Description	USCS	Expected c (kPa)	Measured c (kPa)	Expected $\phi$ (°)	Measured $\phi$ (°)
<i>Controlled samples</i>					
Clay of low plasticity, compacted	CL	86	85	27 – 35	33
Silt loam, compacted	ML, OL, MH, OH	60 – 90	90	25 – 32	29
Sand, fine grain, well graded	SW, SP	0	-1	36 – 41	25
<i>Field testing</i>					
Sand	SP	0	10.5	37 – 38	4.9
Silt loam, saturated	ML, OL, MH, OH	10 – 20	20	25 – 32	28

### 243 6.3. Interpretation of results

244 The results are summarized in Table 2. Overall, the tests give a accurate cohesion and angle of internal  
 245 friction (within the expected range for the material). The exceptions to the rule are the sand samples, that  
 246 give a negative cohesion in one case and an unrealistic angle of internal friction in the other. This shows  
 247 that the proposed method is limited to cohesive soils (known to be present on Mars (Sullivan et al., 2011)).  
 248 For the mounted pocket shear vane, cohesionless soil testing was therefore eliminated.

#### 249 6.4. Mounted pocket shear vane: controlled samples

250 The mounted pocket shear vane has been tested in compacted clay of low plasticity. The results are  
 251 presented in Fig.11. The payload prototype currently takes between 3 to 5 minutes for each measurement (the  
 252 low linear servo precision requires a very slow pace to adjust weights accurately), and 25 measurements total  
 253 were taken to construct the Mohr-Coulomb curve (five for each mass tested, 100g to 500g in 100g increments).  
 254 Due to the unpredictability of field testing, results that are known to be unrealistic are omitted from the  
 255 data. For example, when the instrument was not properly deployed (such as tilted, not touching the ground  
 256 evenly, etc...) the resulting data were not considered.



(a) Dry clay of low plasticity. Expected  $c = 86kPa$  and  $27^\circ \leq \phi \leq 35^\circ$ ,

(b) Mohr-Coulomb envelope and linearization giving  $c = 90kPa$  and  $\phi = 34^\circ$ .

**Fig. 11:** Mounted instrument tested in Clay of low plasticity, compacted. Results give  $c = 90kPa$  and  $\phi = 34^\circ$ . Some data points overlap with each other, while others were eliminated due to being .

257 The retrieved cohesion is  $c = 90kPa$  and the angle of internal friction is  $\phi = atan(0.6667) = 33.69^\circ$ .  
 258 Unlike the tests with the hand-held instruments, the data for the mounted pocket shear vane are more  
 259 widespread, leading to a goodness of fit for the non-linearized curve characterized by  $R^2 = 0.3466$ . The  
 260 variables values for Eq.5 are the following (with a 95% confidence bound):  $a = 2.338e + 05(-1.601e +$   
 261  $07, 1.648e + 07)$ ,  $b = 0.06755(-2.917, 3.052)$  and  $c = -3.632e + 05(-1.866e + 07, 1.794e + 07)$ . The estimated  
 262 maximum power consumption is  $83.056 W$ , and the total power consumption of the payload per data  
 263 collection should not exceed  $2.5 kJ$ . These values do not include the resources expended to transition the  
 264 rover from a driving stance to a squat in order to conduct the experiment, or the minute movement needed  
 265 to position the mounted pocket shear vane over fresh soil for a new measurement.

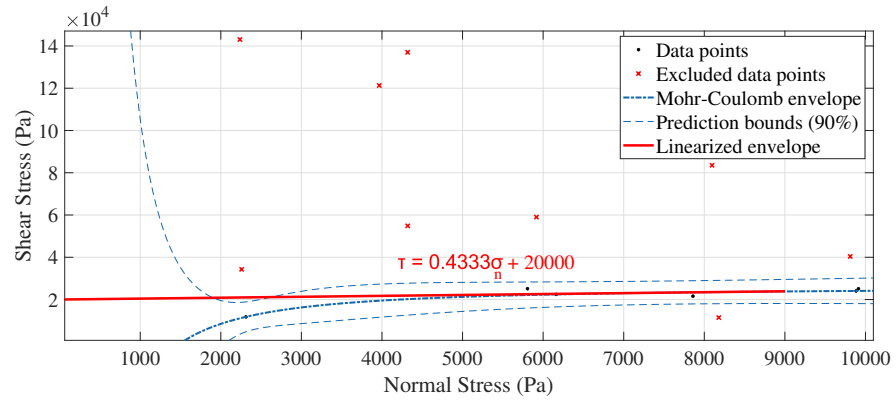
#### 266 6.5. Mounted pocket shear vane: field testing

267 Field testing was conducted in saturated silt loam, with 15 data points taken (three for each mass from  
 268 100g to 500g). The retrieved cohesion is  $c = 20kPa$  and angle of internal friction is  $\phi = atan(0.4333) =$   
 269  $23.43^\circ$ . Unlike the hand-held experiments, a lot of points had to be eliminated because the position of  
 270 the instrument led to false readings. The goodness of fit for the non-linearized curve is characterized by  
 271  $R^2 = 0.8938$ . The variables values for Eq.5 are the following (with a 95% confidence bound):  $a = 2.338e +$   
 272  $05(-1.601e + 07, 1.648e + 07)$ ,  $b = 0.06755(-2.917, 3.052)$  and  $c = -3.632e + 05(-1.866e + 07, 1.794e + 07)$ .

#### 273 6.6. Interpretation of results

274 The mounted pocket shear vane appears to be extremely sensitive to its tilt and gives false results when  
 275 operating at an angle, which affected mostly field testing. The overall results are presented in Table 3. The





(a) Silt loam, saturated. (b) Mohr-Coulomb envelope and linearization giving  $c = 20\text{kPa}$  and  $\phi = 23.5^\circ$ .  
Expected  $c = 10 - 20\text{kPa}$   
and  $25^\circ \leq \phi \leq 32^\circ$ ,

**Fig. 12:** Mounted instrument tested in Silt loam, saturated (*in-situ*). Results give  $c = 20\text{kPa}$  and  $\phi = 24^\circ$ . Some data points overlap with each other, while others were eliminated.

**Table 3:** Results of cohesion  $c$  and angle of internal friction  $\phi$  from testing (mounted pocket shear vane) compared to expected values (Obrzud, 2010; MnDOT, 2007). In green are adequate results.

Description	USCS	Expected $c$ (kPa)	Measured $c$ (kPa)	Expected $\phi$ ( $^\circ$ )	Measured $\phi$ ( $^\circ$ )
<i>Controlled samples</i>					
Clay of low plasticity, compacted	CL	86	90	27 – 35	34
<i>Field testing</i>					
Silt loam, saturated	ML, OL, MH, OH	10 – 20	20	25 – 32	23.5

276 values are not as close to the expected parameters as they were with the hand-held instrument, due to the  
277 mounted pocket shear vane being significantly affected by testing conditions (e.g., tilt). Some of the resulting  
278 outputs are a little out of range (e.g., cohesion of  $90\text{kPa}$  instead of  $86\text{kPa}$ ), but the difference being less  
279 than 5% of the actual value, it is negligible.

## 280 7. Discussion and future work

281 As seen with the hand-held experiments, the modified pocket shear vane leads to results that can be  
282 used to compute parameters of soils with a non-zero cohesion. When mounted, the instrument performs  
283 adequately on flat terrain, and is capable of giving intrinsic parameters needed to identify a terrain. The  
284 pocket shear vane's ease of use, its light weight and compact form makes it a perfect candidate for planetary  
285 surface missions.

286 The transition to remote operation shows that many improvements can be made to the mounted instru-  
287 ment. One way to improve its operation involves selecting components that are better suited for such a  
288 design. For instance, the current linear servo struggled to precisely apply axial load, especially at weights  
289 higher than  $3\text{ kg}$ . A lead screw and stepper motor would be able to apply force more precisely than the

290 potentiometer-based linear servo. Also due to lack of linear servo precision in the current prototype, it was  
 291 witnessed that the device performed better under light loads (for masses between 100 - 500g). In the future,  
 292 utilizing more sensitive load cells that have a maximum weight closer to the light loads used in the experi-  
 293 ment will greatly increase the accuracy of individual measurements. It should be noted that the prototype  
 294 shown in this paper is a proof of concept, and the device performance will be improved in the next version  
 295 by utilizing higher quality components that will increase stiffness and precision of the overall mechanism.

296 In addition to improving on the current design to obtain more accurate results, future work includes  
 297 testing under a wider range of conditions, as well as developing an algorithm to automatically interpret the  
 298 results and compute the cohesion and angle of internal friction on board. Additionally, this instrument will  
 299 be mounted on a physical robot currently under construction at the Interactive Robotics Laboratory at West  
 300 Virginia University, and further testing will be conducted with the shear vane fully integrated. This research  
 301 will also lead to studying other types of instrument, to be able to predict soil parameters in cohesionless  
 302 terrain.

### 303 Acknowledgement

304 This research is supported by NASA EPSCoR Research Cooperative Agreement WV-80NSSC17M0053  
 305 and the Benjamin M. Statler fellowship.

### 306 References

- 307 Arora, A., Fitch, R., & Sukkariéh, S. (2017). An approach to autonomous science by modeling geological  
 308 knowledge in a bayesian framework. In *2017 IEEE/RSJ International Conference on Intelligent Robots  
 309 and Systems (IROS)* (pp. 3803–3810). IEEE.
- 310 Bajracharya, M., Maimone, M. W., & Helmick, D. (2008). Autonomy for mars rovers: Past, present, and  
 311 future. *Computer*, *41*, 44–50.
- 312 Biesiadecki, J. J., Leger, P. C., & Maimone, M. W. (2007). Tradeoffs between directed and autonomous  
 313 driving on the mars exploration rovers. *The International Journal of Robotics Research*, *26*, 91–104.
- 314 Bishop, A. W., & Henkel, D. J. (1962). The measurement of soil properties in the triaxial test, .
- 315 Brooks, C. A., & Iagnemma, K. (2005). Vibration-based terrain classification for planetary exploration  
 316 rovers. *IEEE Transactions on Robotics*, *21*, 1185–1191.
- 317 Candela, A., Thompson, D., Dobrea, E. N., & Wettergreen, D. (2017). Planetary robotic exploration driven  
 318 by science hypotheses for geologic mapping. In *2017 IEEE/RSJ International Conference on Intelligent  
 319 Robots and Systems (IROS)* (pp. 3811–3818). IEEE.
- 320 Chhaniyara, S., Brunskill, C., Yeomans, B., Matthews, M., SaaJ, C., Ransom, S., & Richter, L. (2012).  
 321 Terrain trafficability analysis and soil mechanical property identification for planetary rovers: A survey.  
 322 *Journal of Terramechanics*, *49*, 115–128.
- 323 Gaines, D., Anderson, R., Doran, G., Huffman, W., Justice, H., Mackey, R., Rabideau, G., Vasavada, A.,  
 324 Verma, V., Estlin, T. et al. (2016). Productivity challenges for mars rover operations, .
- 325 Gellert, R., Campbell, J., King, P., Leshin, L., Lugmair, G., Spray, J., Squyres, S., & Yen, A. (2009). The  
 326 alpha-particle-x-ray-spectrometer (apxs) for the mars science laboratory (msl) rover mission. In *The 40th  
 327 Lunar and Planetary Science Conference, LPI Contribution* (p. 2364). volume 1468.
- 328 Giacoboni, N. (2020). Basic custom arduino library for hx711. URL: [https://www.mathworks.com/  
 329 matlabcentral/fileexchange/66641-basic-custom-arduino-library-for-hx711](https://www.mathworks.com/matlabcentral/fileexchange/66641-basic-custom-arduino-library-for-hx711).

- 330 Girdhar, Y., & Dudek, G. (2016). Modeling curiosity in a mobile robot for long-term autonomous exploration  
331 and monitoring. *Autonomous Robots*, *40*, 1267–1278.
- 332 Golombek, M., Otero, R., Heverly, M., Ono, M., Williford, K., Rothrock, B., Milkovich, S., Almeida, E.,  
333 Calef, F., Williams, N. et al. (2017). Characterization of mars rover 2020 prospective landing sites leading  
334 up to the second downselection. In *Lunar and Planetary Science Conference*. volume 48.
- 335 Golombek, M., & Rapp, D. (1996). Size-frequency distributions of rocks on mars. In *Lunar and Planetary  
336 Science Conference*. volume 27.
- 337 Hedrick, G., Ohi, N., & Gu, Y. (2020). Terrain-aware path planning and map update for mars sample return  
338 mission. *IEEE Robotics and Automation Letters*, *5*, 5181–5188.
- 339 Helmick, D., Angelova, A., Matthies, L., Brooks, C., Halatci, I., Dubowsky, S., & Iagnemma, K. (2008).  
340 Experimental results from a terrain adaptive navigation system for planetary rovers. In *the Ninth Inter-  
341 national Symposium on Artificial Intelligence, Robotics and Automation in Space, i-SAIRAS*.
- 342 Iagnemma, K., Kang, S., Brooks, C., & Dubowsky, S. (2003). Multi-sensor terrain estimation for plan-  
343 etary rovers. In *Proceedings of the 8th international symposium on artificial intelligence, robotics, and  
344 automation in space. IEEE Press, New York*.
- 345 Ishigami, G., Nagatani, K., & Yoshida, K. (2007). Path planning for planetary exploration rovers and its  
346 evaluation based on wheel slip dynamics. In *Proceedings 2007 IEEE International Conference on Robotics  
347 and Automation* (pp. 2361–2366). IEEE.
- 348 Ishigami, G., Nagatani, K., & Yoshida, K. (2011). Path planning and evaluation for planetary rovers based on  
349 dynamic mobility index. In *2011 IEEE/RSJ International Conference on Intelligent Robots and Systems*  
350 (pp. 601–606). IEEE.
- 351 Klein, E., Nilsen, E., Nicholas, A., Whetsel, C., Parrish, J., Mattingly, R., & May, L. (2014). The mobile  
352 mav concept for mars sample return. In *2014 IEEE Aerospace Conference* (pp. 1–9). IEEE.
- 353 Labuz, J. F., & Zang, A. (2012). Mohr–coulomb failure criterion. *Rock mechanics and rock engineering*, *45*,  
354 975–979.
- 355 Melosh, H. J. (2011). *Planetary surface processes* volume 13. Cambridge University Press.
- 356 MEPAG, N. D. S. A. G. (2008). Science priorities for mars sample return.
- 357 MnDOT (2007). *MnDOT Pavement Design Manual*.
- 358 Obrzud, R. (2010). *The hardening soil model: A practical guidebook*. Zace Services.
- 359 Ojeda, L., Borenstein, J., Witus, G., & Karlsen, R. (2006). Terrain characterization and classification with  
360 a mobile robot. *Journal of Field Robotics*, *23*, 103–122.
- 361 Ono, M., Rothrock, B., Almeida, E., Ansar, A., Otero, R., Huertas, A., & Heverly, M. (2016). Data-driven  
362 traversability analysis for mars 2020 landing site selection, .
- 363 Papadakis, P. (2013). Terrain traversability analysis methods for unmanned ground vehicles: A survey.  
364 *Engineering Applications of Artificial Intelligence*, *26*, 1373–1385.
- 365 Rahmatian, L. A., & Metzger, P. T. (2010). Soil test apparatus for lunar surfaces. In *Earth and Space 2010:  
366 Engineering, Science, Construction, and Operations in Challenging Environments* (pp. 239–253).
- 367 Rothrock, B., Kennedy, R., Cunningham, C., Papon, J., Heverly, M., & Ono, M. (2016). Spoc: Deep  
368 learning-based terrain classification for mars rover missions. In *AIAA SPACE 2016* (p. 5539).

- 369 Schroeder, C. (2019). Life on mars: my 15 amazing years with oppy, nasa's record-breaking rover, .
- 370 Sullivan, R., Anderson, R., Biesiadecki, J., Bond, T., & Stewart, H. (2011). Cohesions, friction angles, and  
371 other physical properties of martian regolith from mars exploration rover wheel trenches and wheel scuffs.  
372 *Journal of Geophysical Research: Planets*, 116.
- 373 Witze, A. (2014). Nasa plans mars sample-return rover. *Nature News*, 509, 272.
- 374 Zacny, K., Wilson, J., Craft, J., Asnani, V., Oravec, H., Creager, C., Johnson, J., & Fong, T. (2010). Robotic  
375 lunar geotechnical tool. In *Earth and Space 2010: Engineering, Science, Construction, and Operations in*  
376 *Challenging Environments* (pp. 166–181).
- 377 Zoback, M. D. (2010). *Reservoir geomechanics*. Cambridge University Press.