

Development of a Measure While Drilling System

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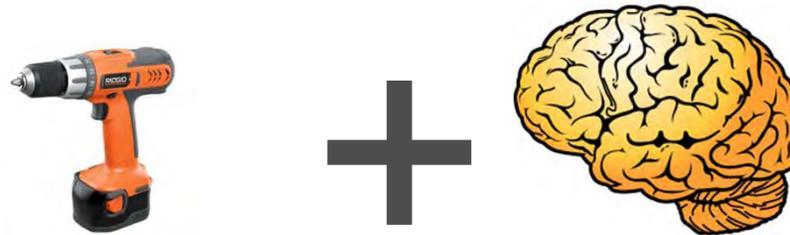
Intelligent MWD System

MWD Planetary Substrate Identification System Development

- Proposed technology is part of a larger NORCAT project involving a TRL 5 prototype of a rover-based drill
 - Can also be mounted on a fixed base (e.g. lander)
- Technology will demonstrate Canadian expertise in planetary drilling for future exploration missions
- Supports the concept of NASA's next generation RESOLVE mission.

Measure While Drilling: Overview

- End objective:
 - Develop additional functionality to planetary drills. From a tool for gathering core samples
 - Decision support
 - Science instrument
 - a robust system architecture that would permit a MWD system to be readily integrated into a NORCAT planetary drill system
 - develop an understanding of which parameters are most useful for MWD



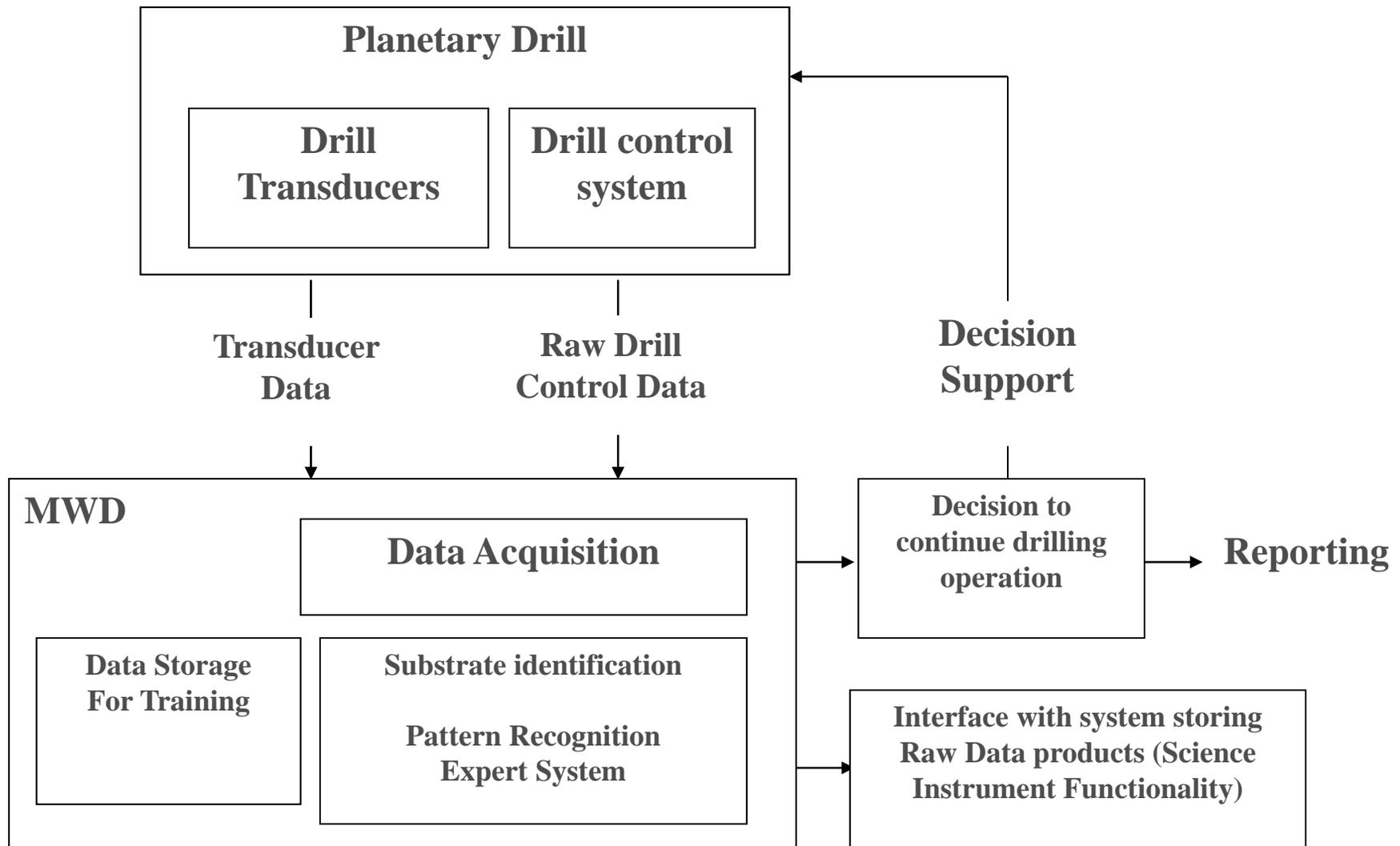
Decision Support Scenario

- In situ substrate identification
 - Rock or ice bearing regolith
 - Decision support: continue drilling or to abort the hole
- Avoidance of Hazardous Drilling Conditions
 - Ex: drill encounters a partially exposed rock protruding into the bore hole
 - This condition can result in accelerated bit wear or damage to diamond cutters.
 - Extend equipment life

Science Instrument Scenario

- Use drill to gather key geotechnical and mechanical characteristics of the material (Strength, hardness etc.)
- Data can also be stored and corroborated with data collected from sample analysis
- Data from multiple holes could be used to predict geological makeup of larger areas

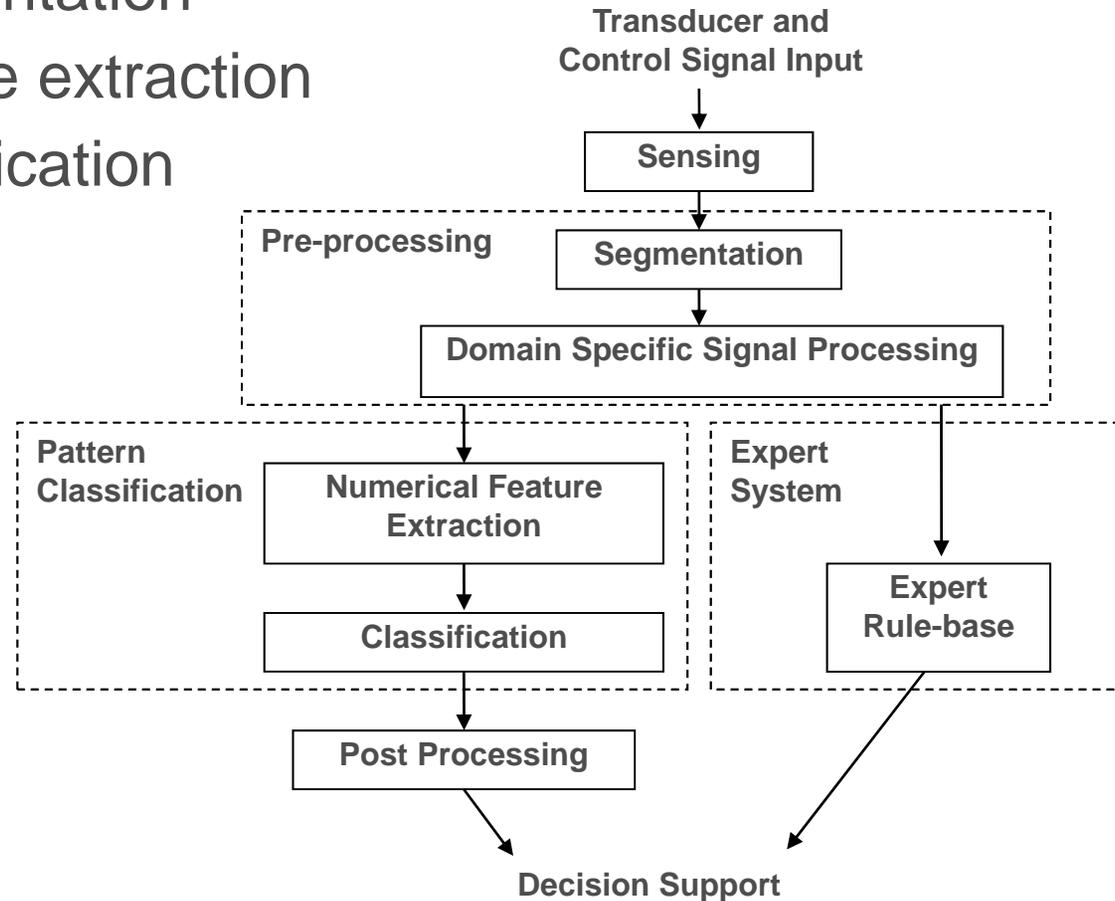
Conceptual Implementation



Data Flow

- Map data from the sensor signals to the end goal of substrate identification through:

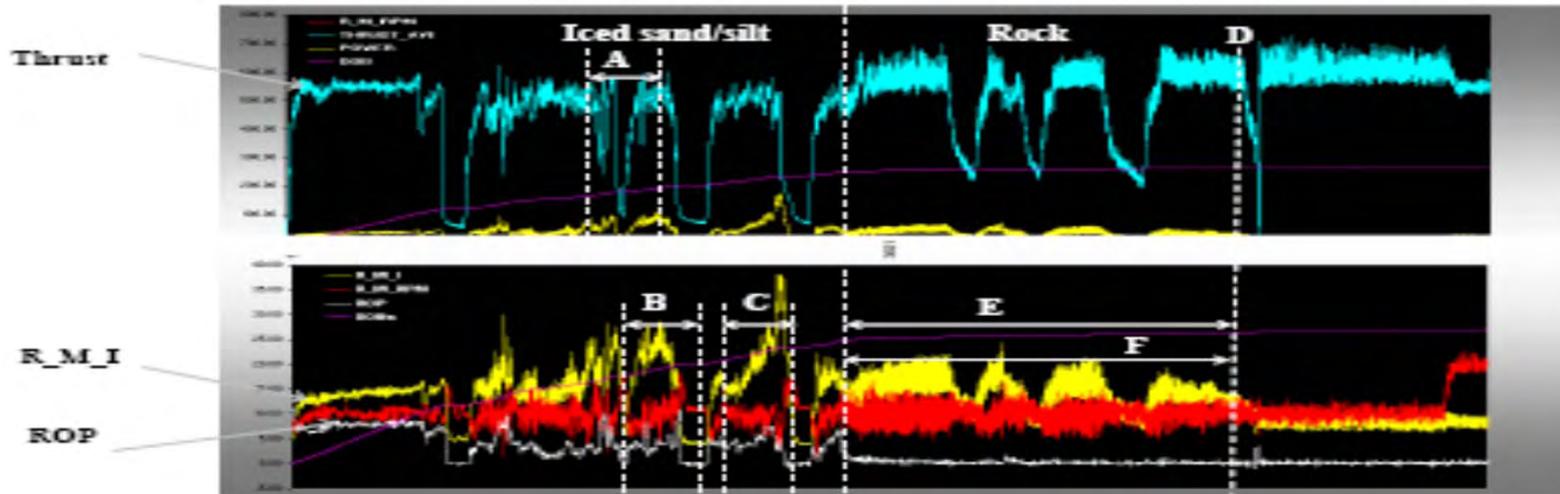
- Segmentation
- Feature extraction
- Classification



Is this feasible?

Is this feasible?

- Technically feasible based on:
 - Previous findings at NORCAT
 - Literature review
 - Relatively straightforward to incorporate into NORCAT's existing drill systems
- Various researchers have shown that the task of identifying substrates is possible using an instrumented drill
 - Prediction of strata in coal mine using thrust, ROP (95% confidence)
 - Characterise the structural integrity of tunnels before material is excavated
 - Statistical pattern recognition to ID rock types to optimize blasting

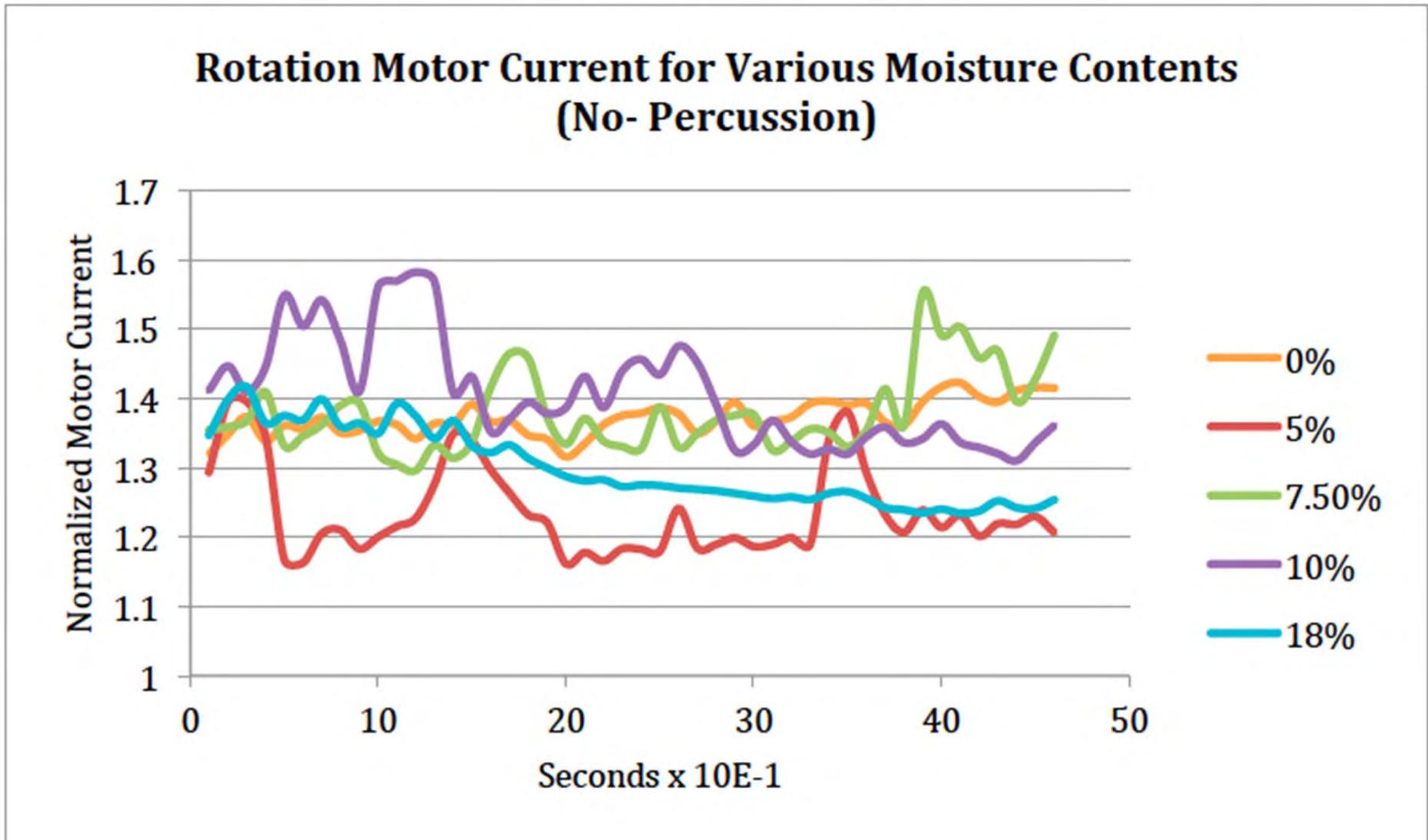


<u>Parameter</u>	<u>Iced Sand/Silt</u>	<u>Rock</u>
Thrust	Significant variations	Relatively constant
R_M_I (current)	Continually ramped up, paused to clear	Lower, comparatively flat
ROP (rate of penetration)	High	Minimal, constant

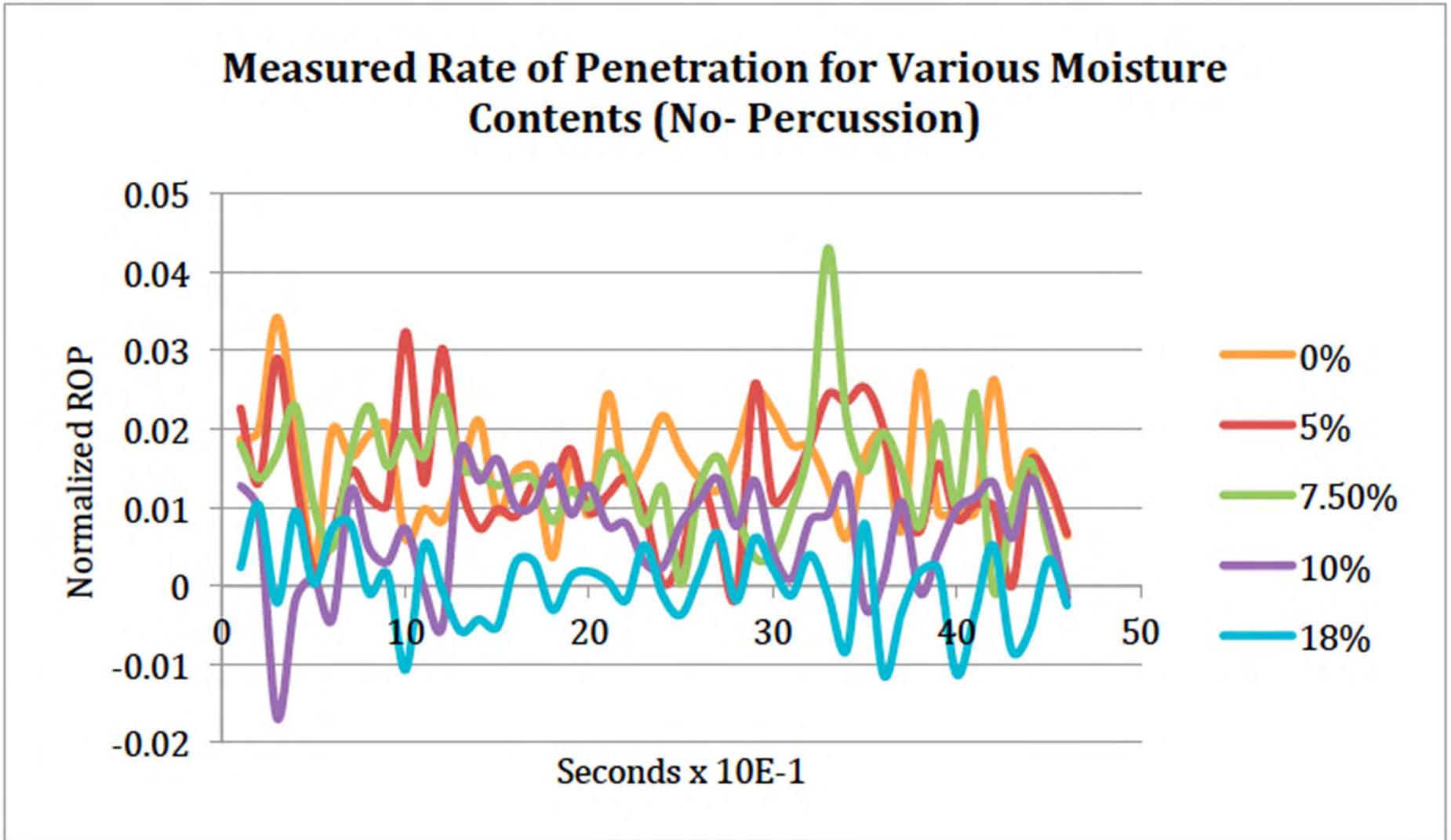
Initial Investigation: Is this a trivial task?

- Before resorting to:
 - Advanced and more computationally intensive methods of signal processing or high dimensional feature extraction and classification for MWD.....
- The work first investigated the basic response and relationships of the control parameters to moisture content
 - rotation motor current, measured rate of penetration and measured thrust to the varied levels of moisture content and substrate types in the unconsolidated material.

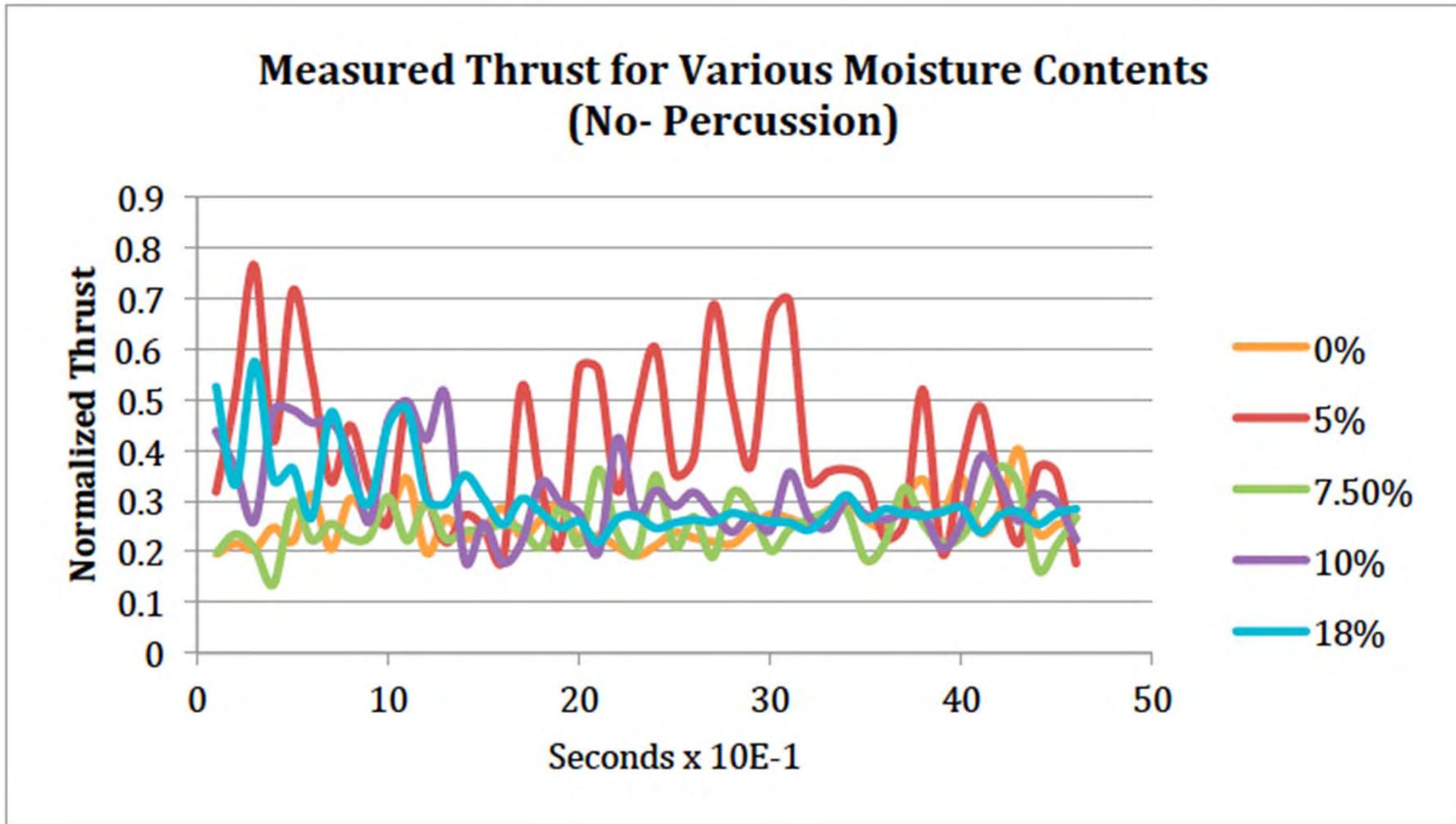
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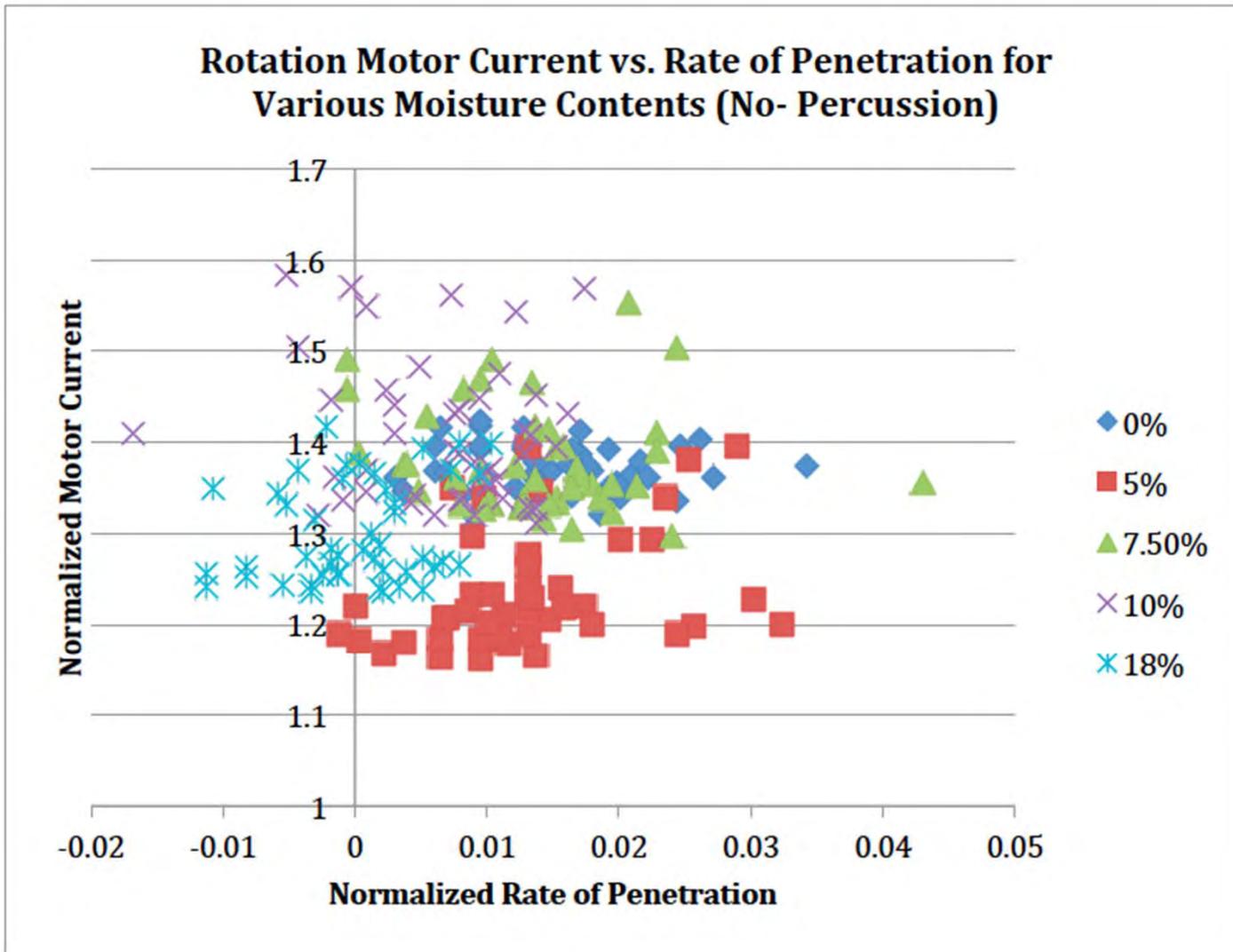


Initial Investigation: Is this a trivial task?

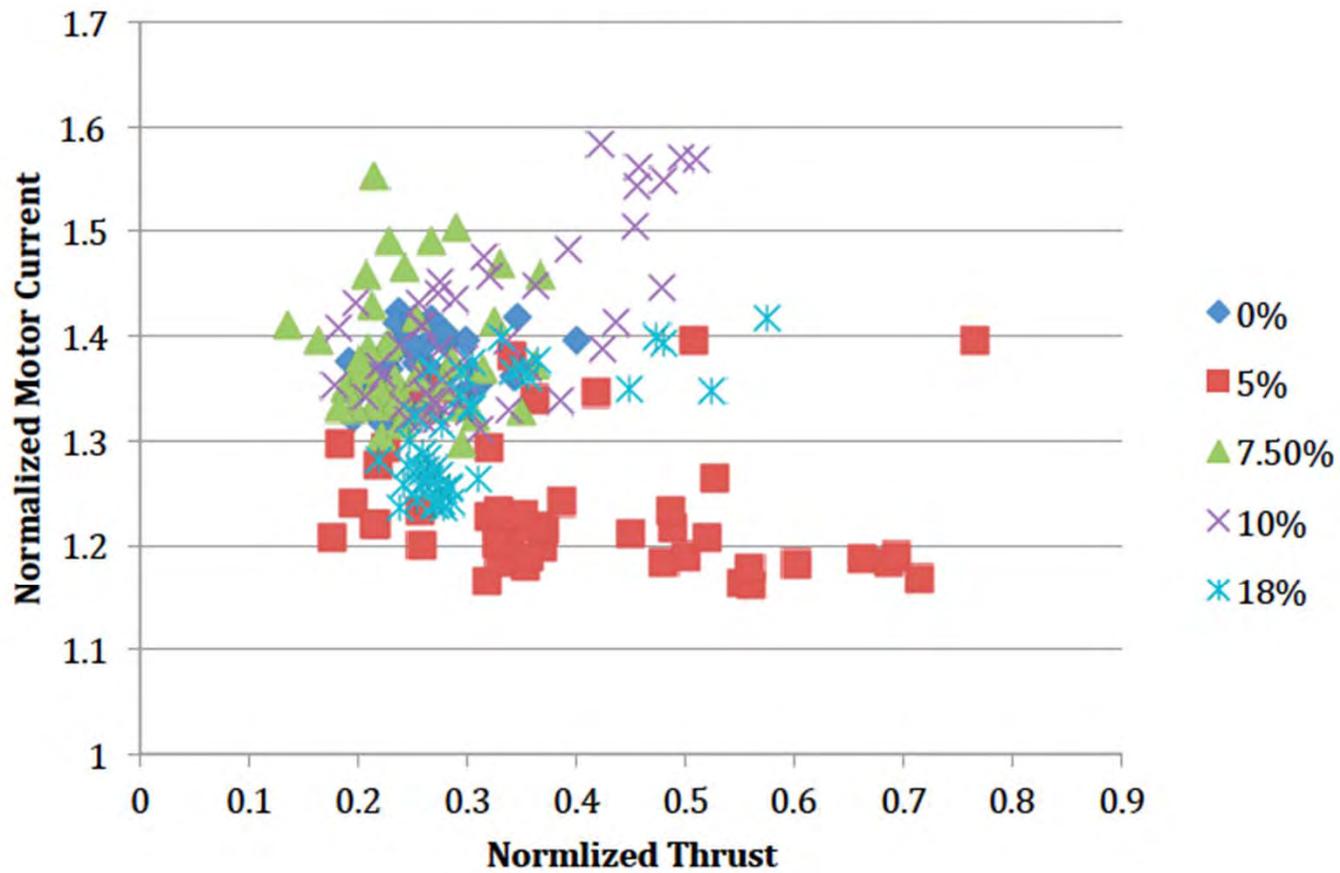


Initial Investigation: Is this a trivial task?





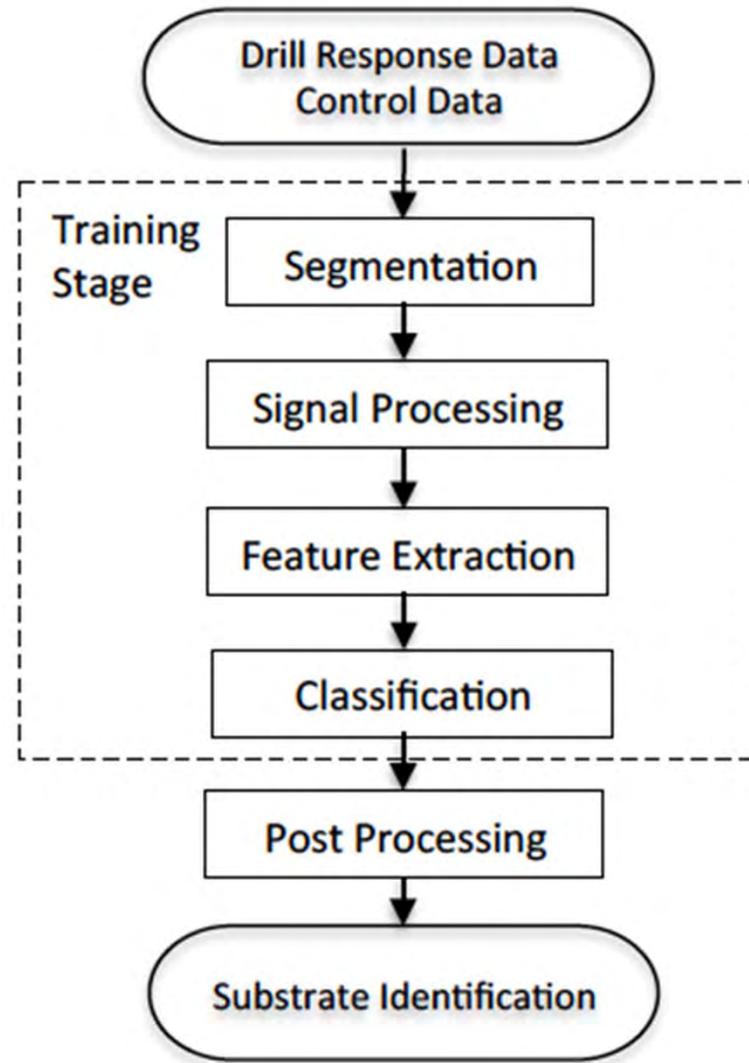
Rotation Motor Current vs. Thrust for Various Moisture Contents (No- Percussion)



Initial Investigation: Is this a trivial task?

Conclusion:

- MWD is a non-trivial classification problem.
- MWD will require the use of additional:
 - parameters,
 - Features
 - Multi-dimensional classification tools.



Data Collection

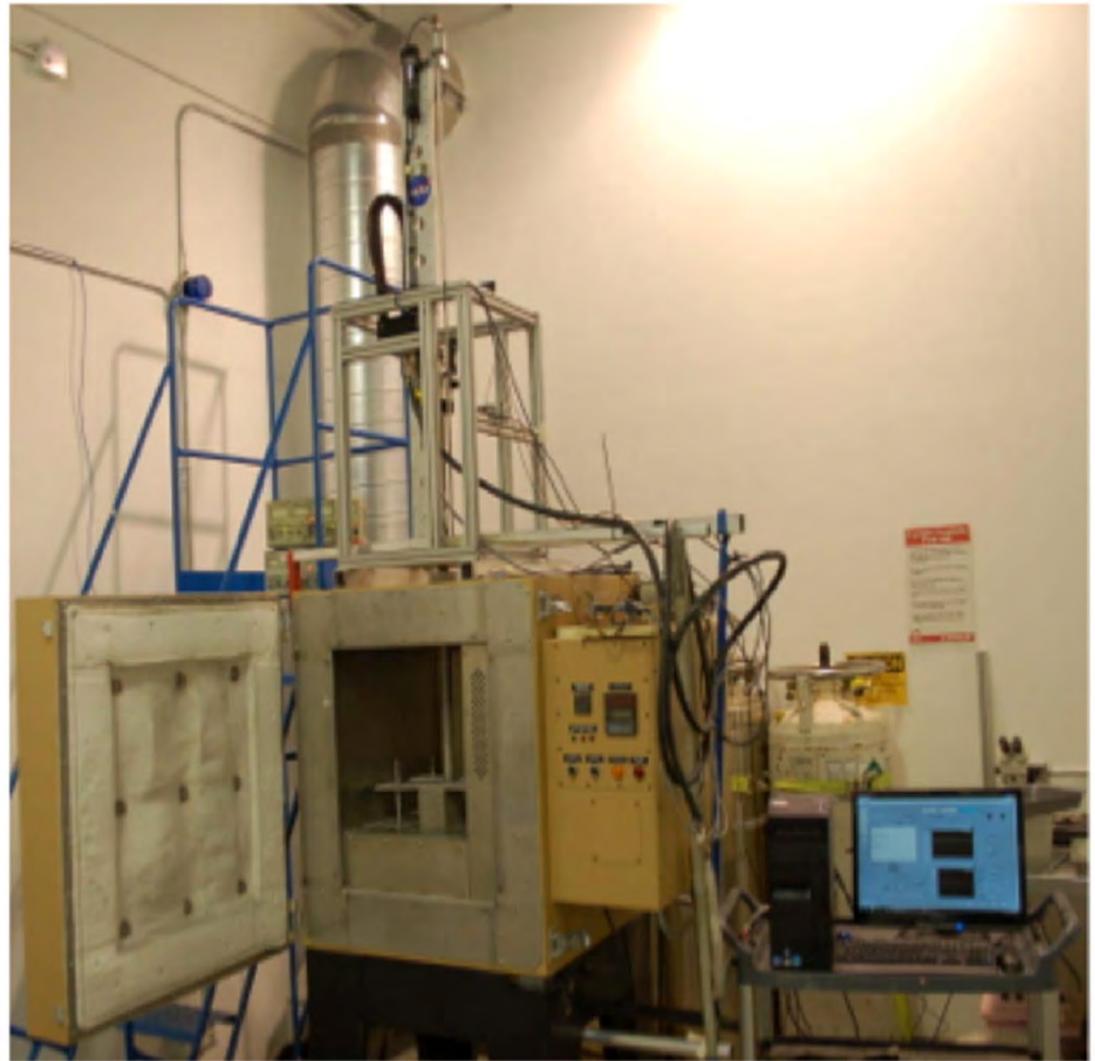
Decision Support System Development: Data Collection

Requires Data That:

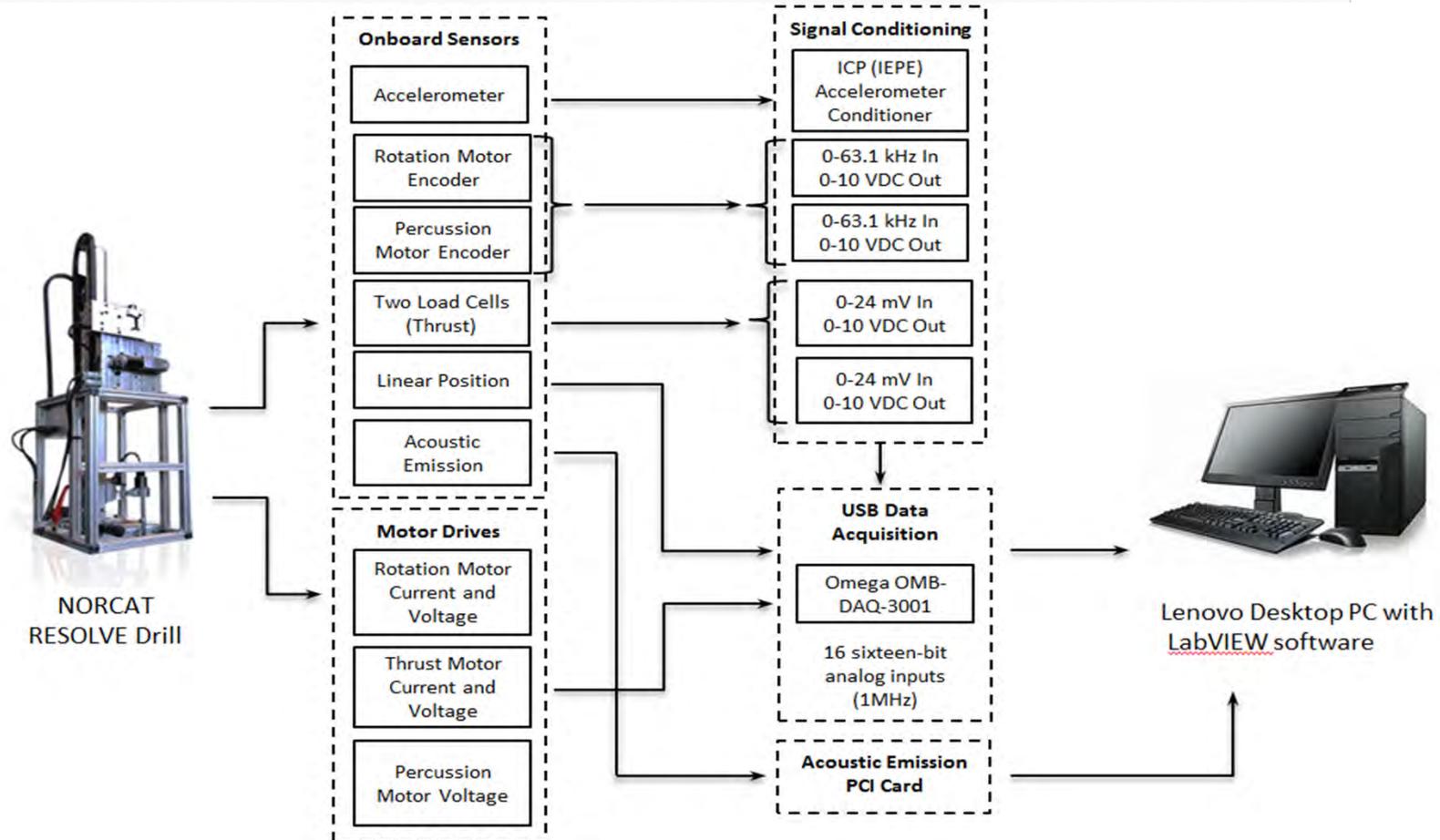
- Characterises the response of the modified NORCAT RESOLVE EBU2 drill penetrating a range of substrates under controlled conditions
- Can be used for training, optimization and testing of classification system
- Can also be used to demonstrate science Instrument capability

Data Collection Setup

- The prepared vessel containing the sample placed in the cryotemper (-175° C).
- The instrumented drill will be placed on top of the cryotemper
- the drill bit accessed the frozen CHENOBI within the chamber by means of the top port in the cryotemper.
- Drilling depth range between 10 to 90 mm.

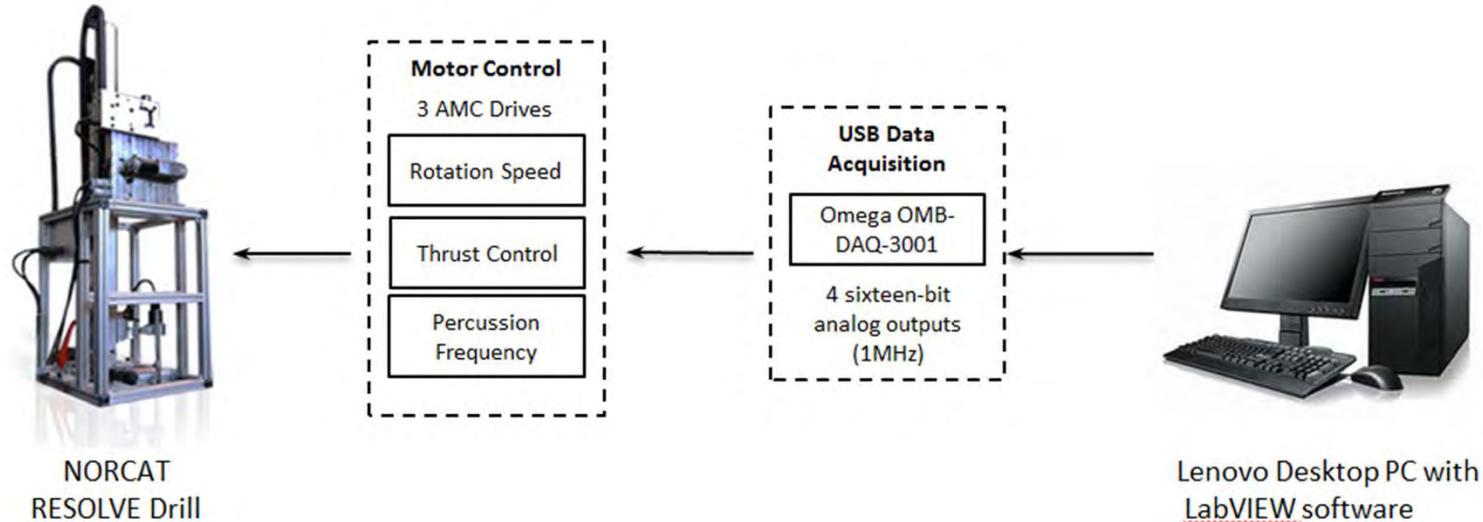


Drill Instrumentation



- Direct instrumentation of the control system of drill and sensors mounted on the drill
- Training data collection and analysis
 - Collection of large, time dependent multivariate data set
 - Sensory and control data collected during supervised drilling of known substrates

Drill Control System



Control of drilling process:

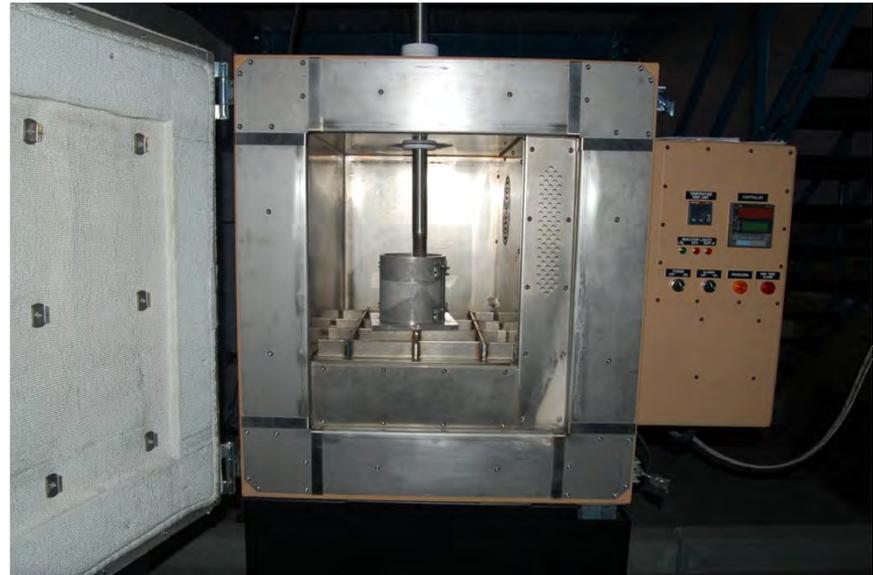
- The instrumented drill will penetrate the sample under specified operating conditions regulated by the MWD control system.
- With and without percussion
- Closed loop (PID) control will be used to control
 - Drill speed (encoder feedback signal)
 - Thrust (load cell feedback)

Samples Tested

- Data collected from drilling of both consolidated and unconsolidated material

Substrates:

- Basalt
- Anorthosite
- Dolomite
- Regolith Simulant
 - compacted Chemically Enhanced OB-1 (CHENOBI) lunar highlands regolith physical simulant
 - 0%, 5%, 7.5%, 10% and 18% moisture content by weight
 - at cryogenic temperatures and standard temperature pressure.



Data analysis

Pre Training Scenario

- The most direct approach to the training and deployment of a MWD system is to train the system to recognize the range of expected substrates under simulated laboratory conditions prior to the mission.
- Once deployed, the system would then compare extracted features from new (unclassified) measurements to features from the database assembled in the laboratory prior to the mission.
- The clear limitation of this approach is the dependency of MWD system on the similarity of data from the (simulated) laboratory conditions to the conditions in-situ.
- This type of system would also not be capable of identifying any data that it was not trained on.

Analysis: Feature Extraction

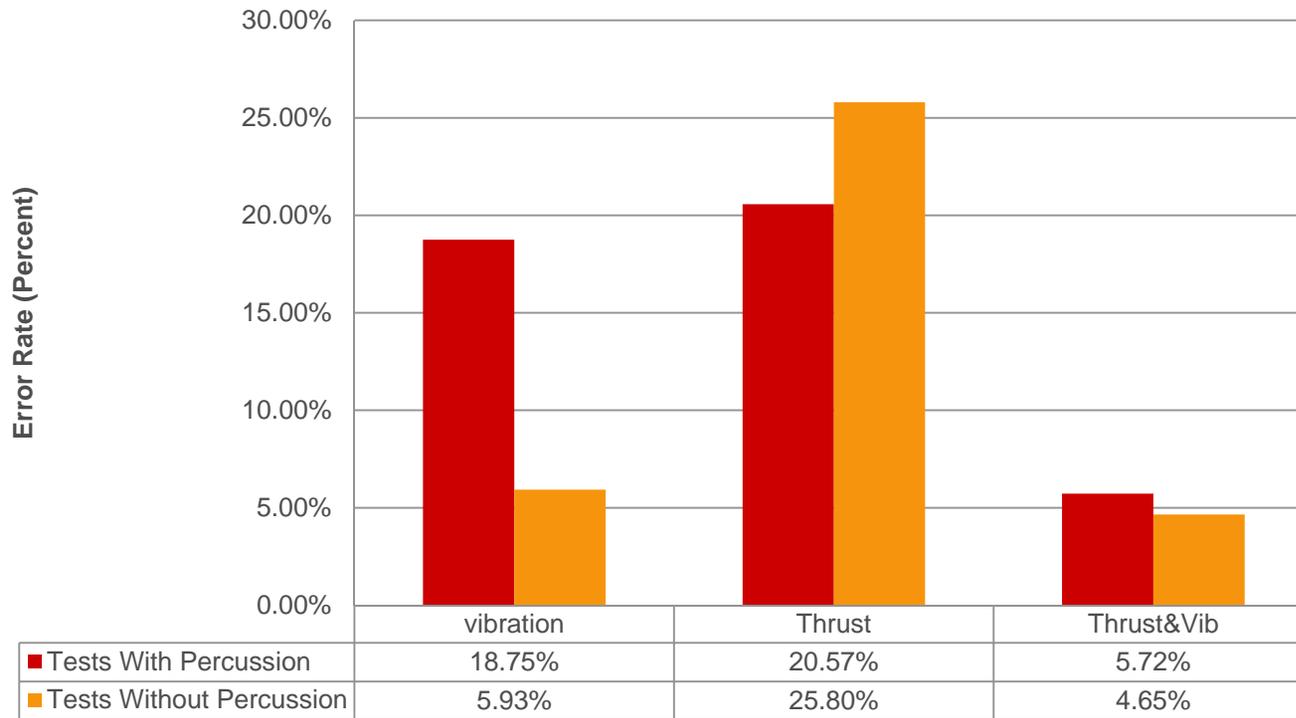
- Features included drill control parameters (speed, thrust, rate of penetration) and vibration
- The vibration feature vectors were built from six statistical features and an autoregressive model (AR) of the vibration time waveform.
- The features included: RMS, kurtosis, impulse factor, crest factor, standard deviation, absolute mean and the coefficients of an autoregressive model of the tenth order.
- The thrust measurement time signal was processed to form a feature vector (referred to here as the thrust feature vectors) built from 7 statistical features: RMS, kurtosis impulse factor, crest factor, standard deviation, average and absolute mean.

Analysis: Classification

- A range of classifiers were trained in order to determine the separation of the data representing different substrates and drilling modes.
- Classification was achieved via feed forward neural networks, with three hidden layers, trained by backpropagation network training function (Levenberg-Marquardt back-propagation algorithm).
- Future work will involve more advanced features and classification methods.

Preliminary Results

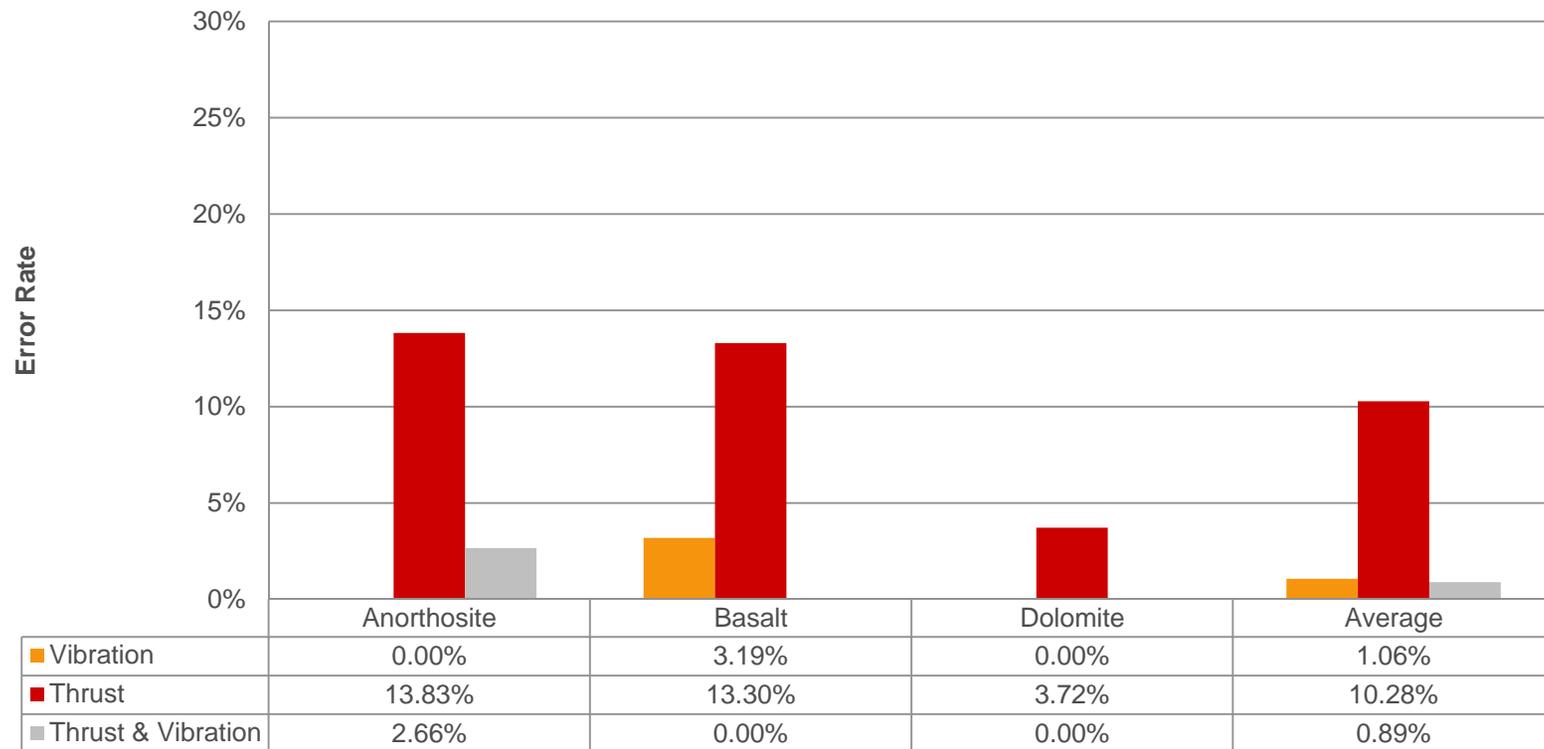
Differentiating between Rock and Unconsolidated



- Initial classification results for drilling response in all materials

Preliminary Results

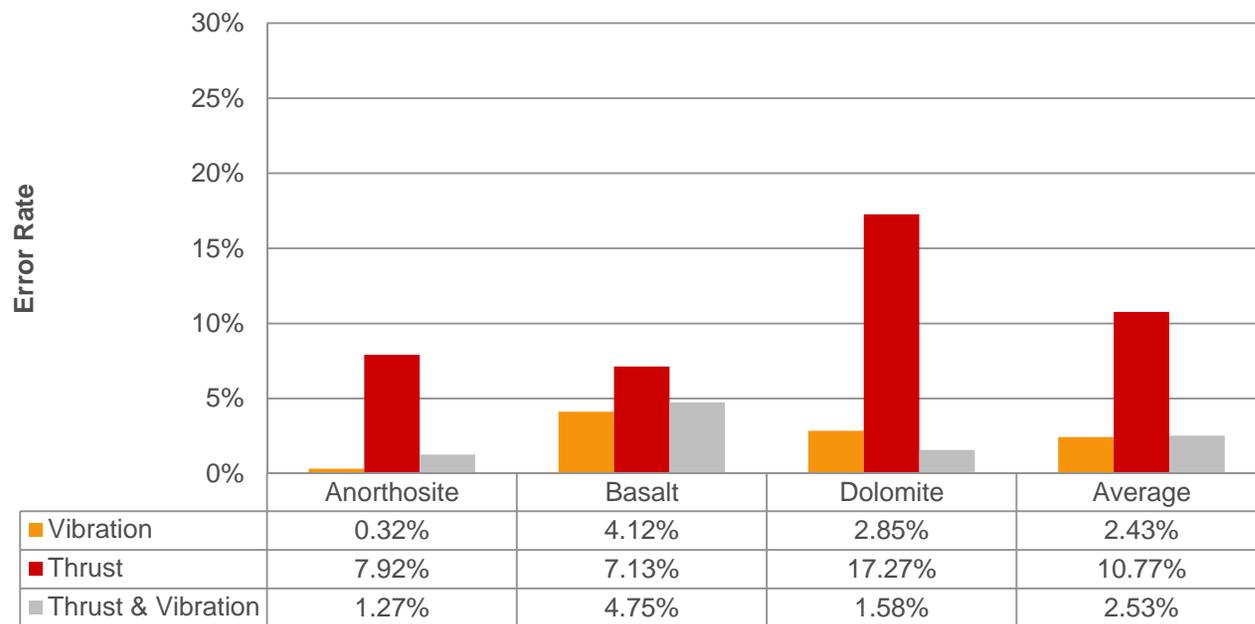
Distinguishing Rock Type From Tests Without Percussion



- Initial classification results for drilling response in rock without percussion

Preliminary Results

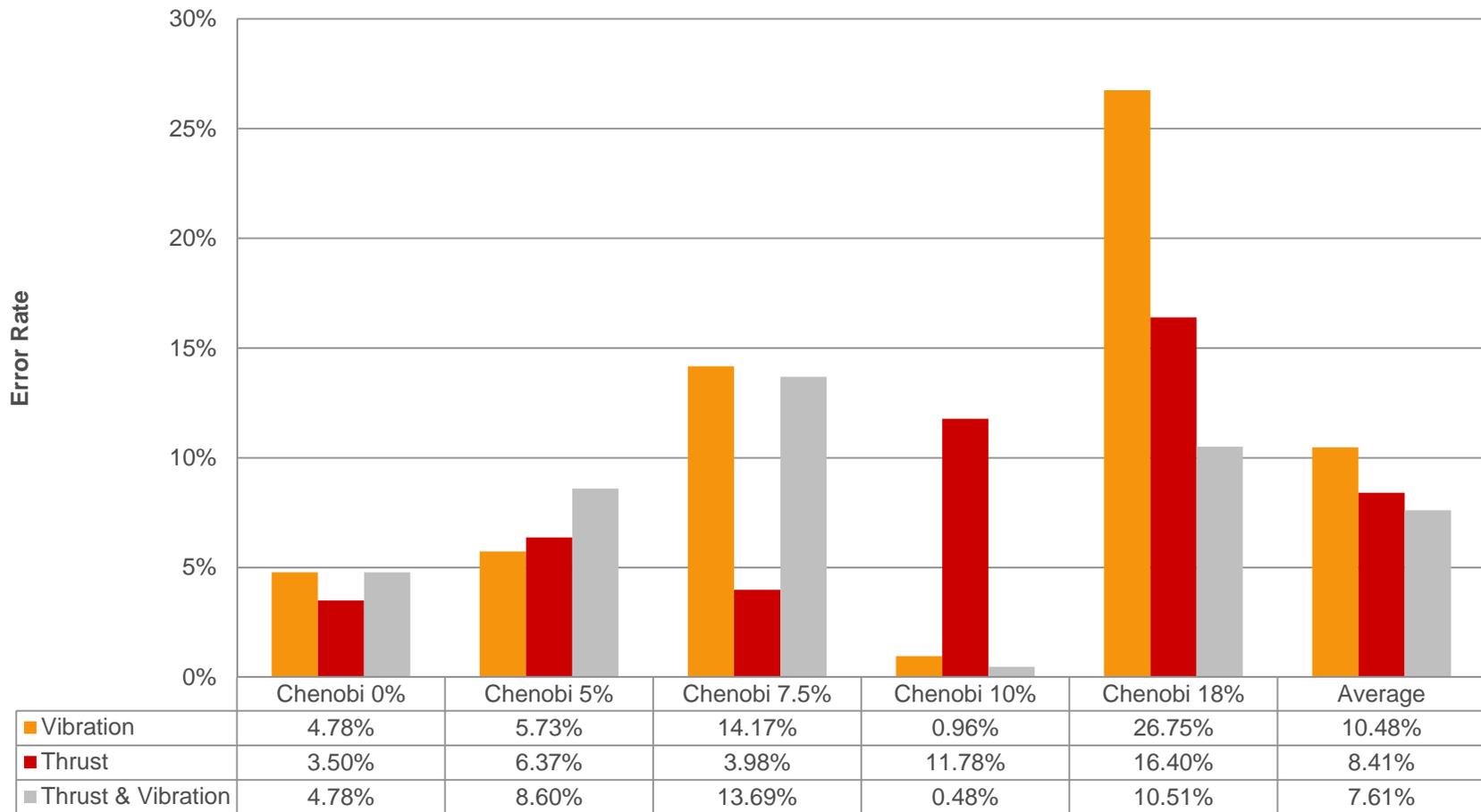
Distinguishing Rock Type From Tests With Percussion



- Initial classification results for drilling response in rock with percussion

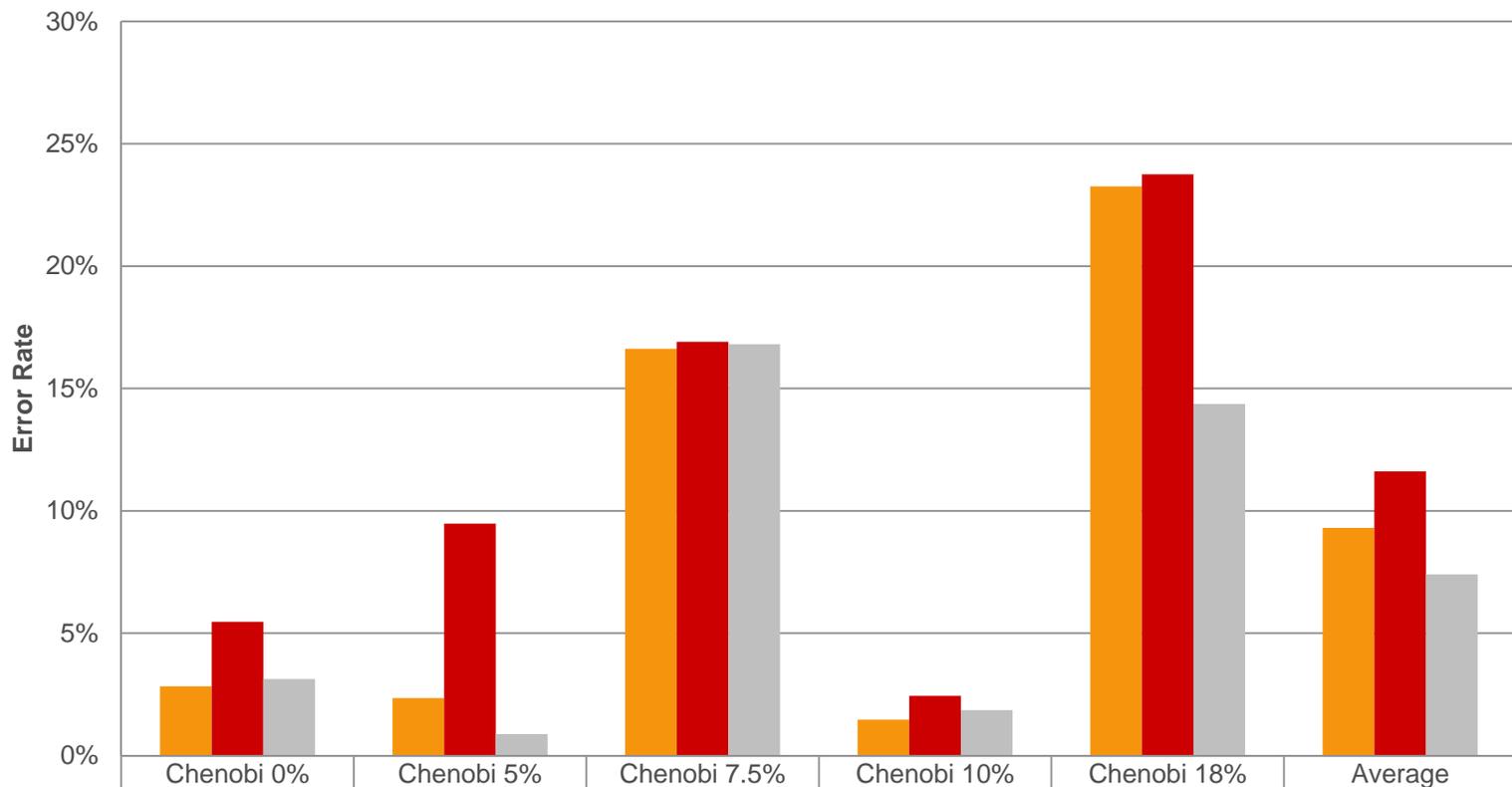
Preliminary Results: Moisture Content

Distinguishing Moisture Content From Tests With Percussion



Preliminary Results: Moisture Content

Distinguishing Moisture Content From Tests Without Percussion



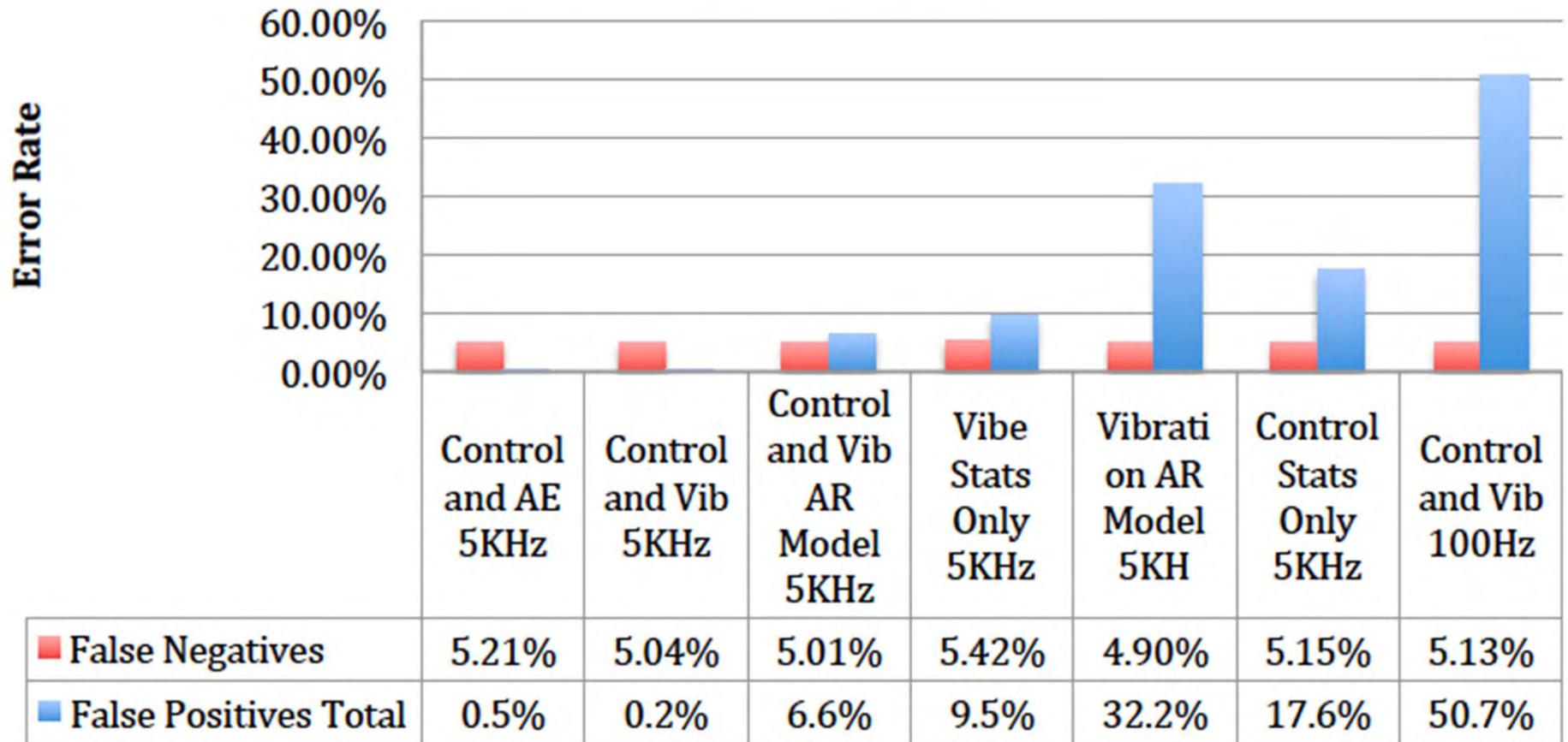
■ Vibration	2.83%	2.35%	16.62%	1.47%	23.26%	9.31%
■ Thrust	5.47%	9.48%	16.91%	2.44%	23.75%	11.61%
■ Thrust & Vibration	3.13%	0.88%	16.81%	1.86%	14.37%	7.41%

In-Situ Training Scenario

- In-Situ Training Scenario: the MWD system depends on the confirmation of material type **during the mission** resulting from the analysis of the captured drill core samples that is processed through the LAVA Gas Chromatograph/Mass Spectrometer and OVEN systems.
- The classification approach of novelty detection offers several practical advantages when applied to the task of substrate identification. These advantages relate to the data available:
 - The ability to identify substrates without relying on laboratory simulated training data fault data.
 - The ability to classify previously unseen substrates.

Novelty Detection Error Rates for Classifying

All Moisture Contents



Overall Conclusions

From the first round of testing (terrestrial pre-training):

- MWD system is capable of differentiating between various ranges of moisture contents of unconsolidated material and identify types of rock.
- This approach would require prior training of the drill in simulated conditions.
- System would be only as good as the training data.

Overall Conclusions

The second round of testing (in-situ training):

- employed the novelty detection approach and assumed that there was no data available in advance of the mission.
- This round resulted in excellent classification results.
- However, the capability of MWD would be limited to the data generated by processing, analyzing samples and confirming substrates during the mission.
- The system would get smarter as it processed more samples.

Conclusion

- The Concept
 - Concept is technically feasible:
 - It is possible to determine rock type and moisture content
 - Could assist with mission decision support
 - Would make a valuable science instrument
 - Could be implemented with minimal additional payload
- Next Steps
 - Further data collection of more samples
 - Pattern recognition system development
 - Optimization of features, signal processing, data compression
 - Pilot System Development
 - System Validation

Acknowledgements

