



Aberdeen Section

# ESP shutdown prediction using hybrid support vector machine and improved parallel particle swarm optimization

Moradeyo Adesanwo, Oladele Bello, Lajith Murali, Ajish Kb, William Milne, BHGE

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#### June 27, 2018

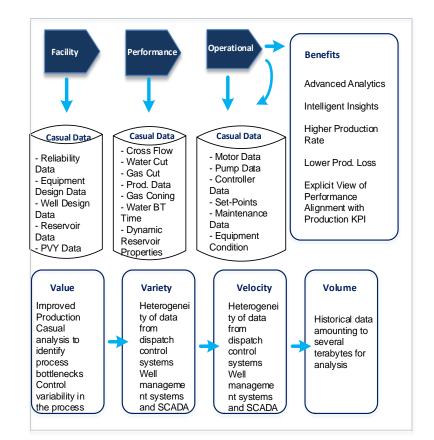
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# Outline

- Introduction and current challenges
- Current challenges in why predict ESP shutdown?
- Objectives
- Methodology
- Field examples results and discussion
- Conclusions

# Background and motivation for ESP-based well operations management

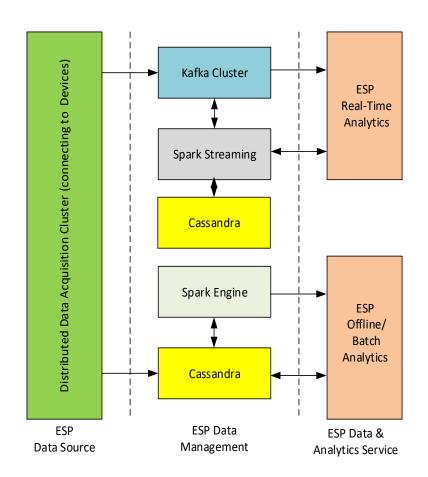
- Big data technologies are now being used to address many industrial problems & challenges related to high volume, high velocity, high variety
- Digitization using advanced sensors in intelligent oilfield solutions has resulted in the generation of huge amount of data that can be used to enhance the performance, operational health surveillance & management of ESP
- Need to use powerful analytics to estimate inexpensively and accurately equipment performance degradation, equipment failures, or detrimental interactions among subsystems
- Enhanced monitorability to reduce downtime and improve hydrocarbon recovery



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# **Current challenges in smart ESP-based well operations management**

- Overwhelming downhole & surface sensor data has become too large and complex to be effectively processed by traditional approaches
- Acquiring, cleansing, processing & analyzing data has to be done in shorter time frames
- Downhole big data analytics requires a range of technologies (big data platforms – Spark; data warehouses and databases; data ingestion, pipeline, integration; virtualization and containerization of workloads; servers, storage and networking infrastructure; analytics model tuning, visualization and reporting applications;
- Current upscaling of online ESP surveillance practices, are largely determined by the size of the training data and the limited computational infrastructure for storing, managing, analyzing and visualizing big datasets



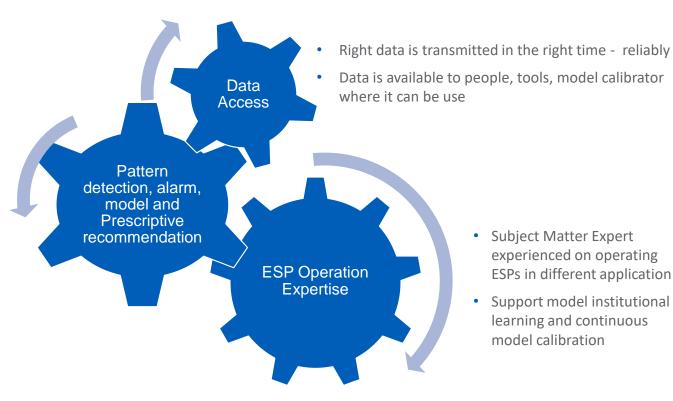
# **Objectives**

- Develop a general purpose system architecture for predicting ESP shut-down focusing on low-cost cloud –based computing platform
- Explore the feasibility of using real-time data collected by downhole sensors as the only source of information about the normal and abnormal operating conditions of a ESP system
- Investigate the potential of using data-driven artificial intelligence and statistical techniques for efficiently extracting and capturing information about the expected patterns of the signals coming from the ESP sensors, controller data, power data, surface data during the normal operating conditions of an ESP (normal operating patterns)

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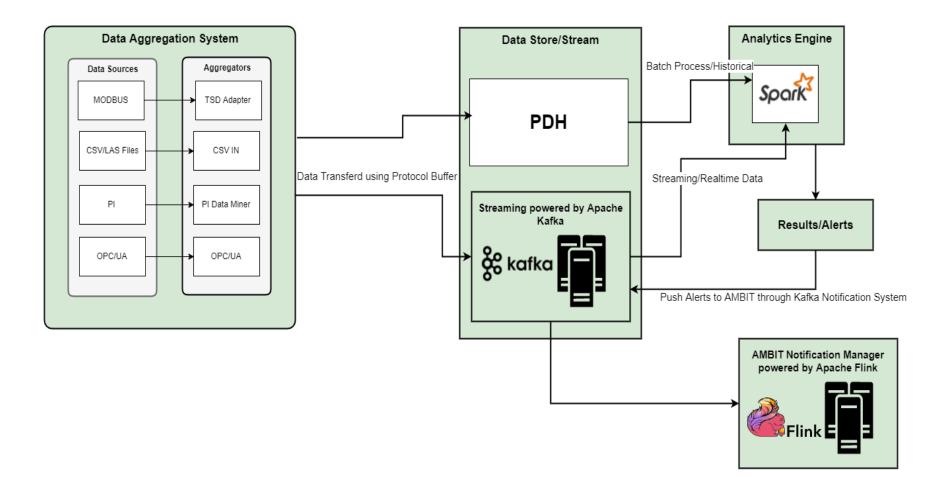
# Methodology: proposed solution framework

- No guess work on data -Fuzzy logic based analysis for finding whether data is increasing, decreasing or constant
- Easy to recognize ESPs with issues – exception based surveillance (Exceptions are cross verified with the Prescriptive rules)
- Once the analyzed data and rules/pattern matches, alarms are triggered based on the prescription

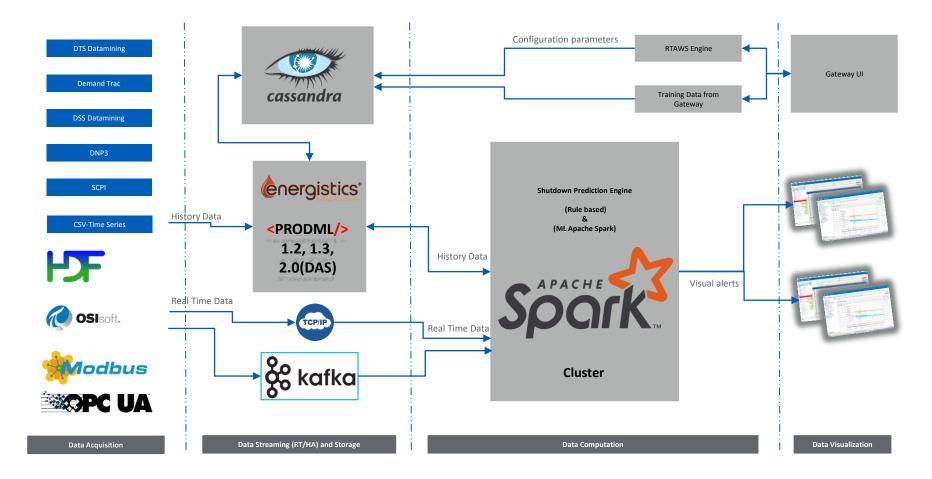


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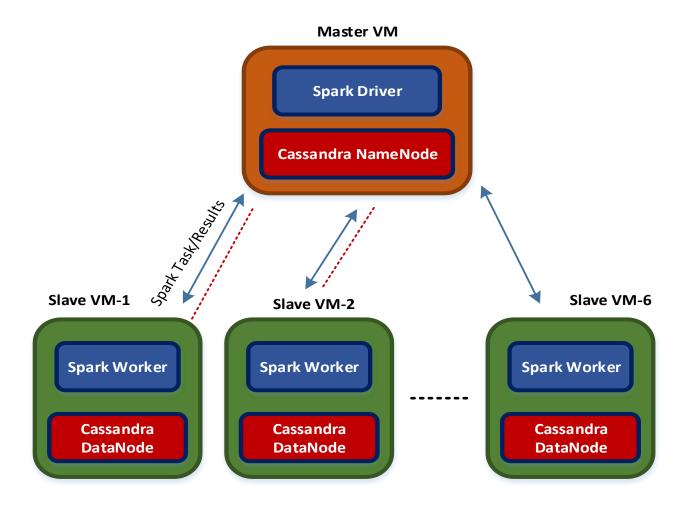
## Methodology: proposed deployment service platform



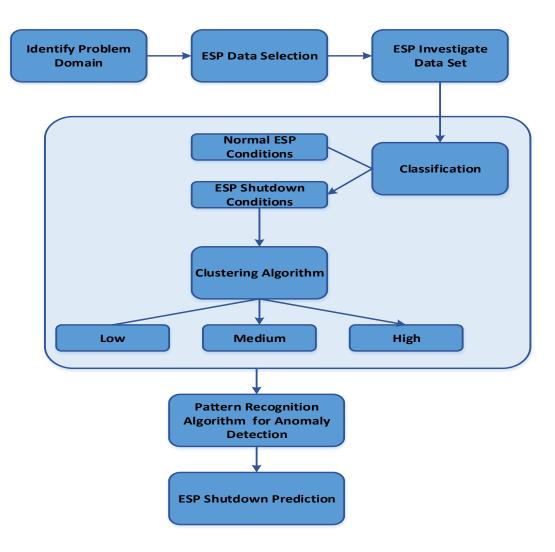
## Methodology: proposed system architecture



### Methodology: proposed spark computational cluster architecture



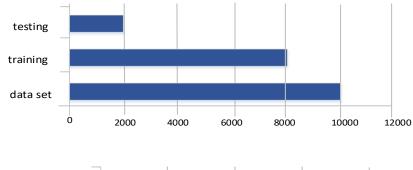
### Methodology: ESP shutdown prediction model

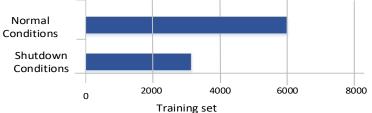


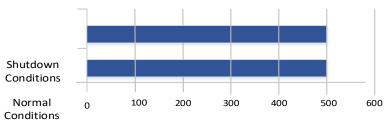
#### Dataset for the case study

TOTAL NUMBER OF WELLS	~350
Total number of shutdown	3,984
Maximum shutdowns per month	597
Minimum shutdowns per month	111
Average shutdowns per month	332
Total SCADA shutdowns	1,751
Total input power quality	1,516
Total operations/downhole	635
Total external signal shutdown	46
Total internal drive component	36

#### **Distribution of dataset**







Testing set

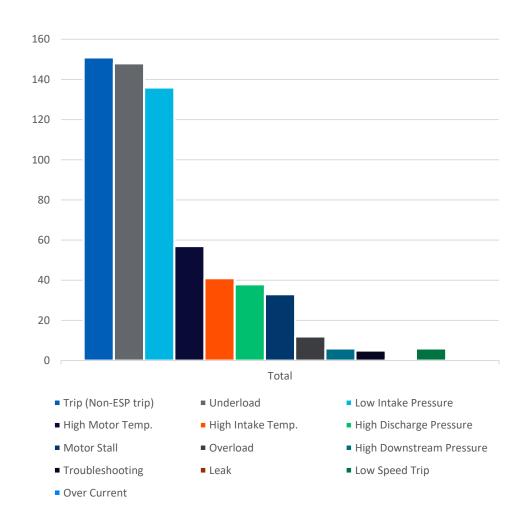
### Important operating parameters that is typically monitored include ...

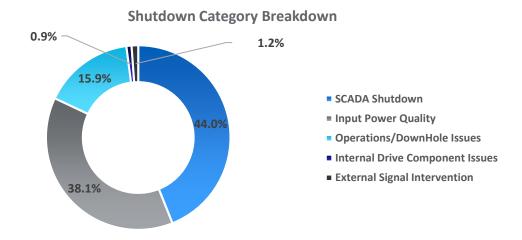
- Frequency
- Pump intake pressure and temperature
- Pump discharge pressure and temperature
- Internal motor operating temperature
- Flow rate
- Surface current and voltage
- Voltage imbalance
- Tubing pressure and casing pressure
- Current leakage
- Unit vibration

### Important operating parameters that may be Inferred include ...

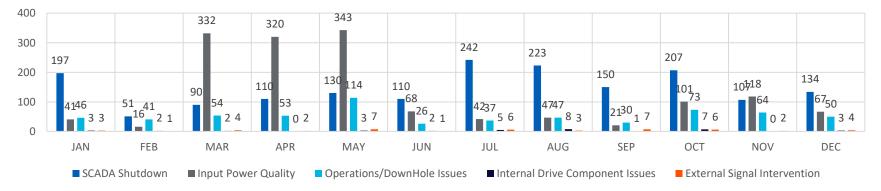
- Seal section oil expansion
- Driveline torque
- Thrust bearing loading
- Motor terminal voltage
- Motor terminal current
- Motor load factor
- Pump operating range
- Stator temperature
- Seal oil temperature

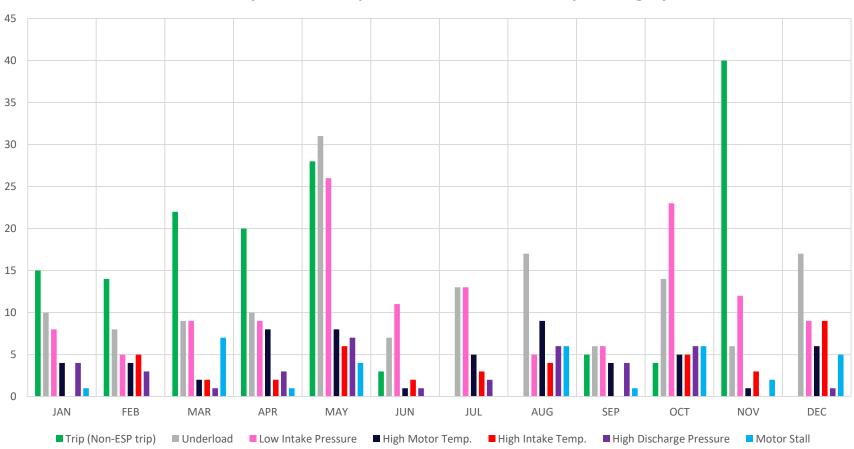
OPERATIONAL/DOWNHOLE	TOTAL
Trip (non-ESP trip)	151
Underload	148
Low intake pressure	136
High motor temp.	57
High intake temp.	41
High discharge pressure	38
Motor stall	33
Overload	12
High downstream pressure	6
Troubleshooting	5
Leak	1
Low speed trip	6
Over current	1





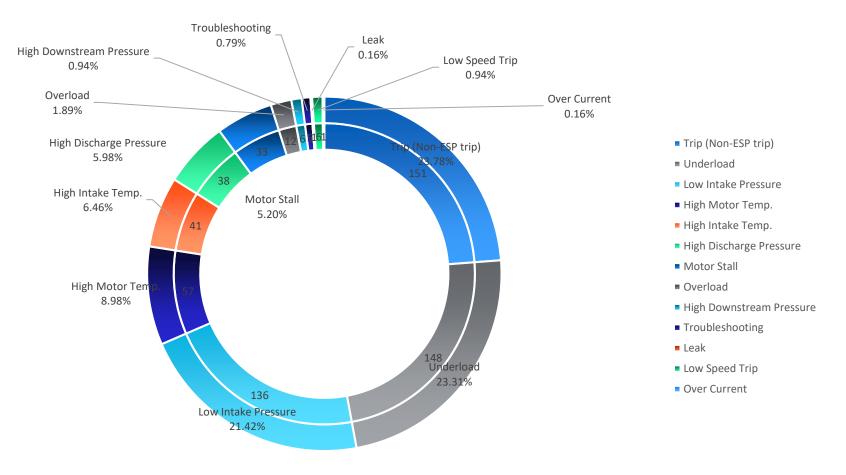
#### Monthly Num. of Shutdowns per Category





Monthly number of operations/downhole issues per category

**Operations/downhole issues breakdown** 



### Conclusions

- An automated detection and prediction of shutdown has been designed and implemented for real-time Electrical Submersible Pumps (ESP) performance monitoring and diagnostics
- The main novelties are that the proposed approach extracts features from operating conditions based on clustering analysis and predict via data classification
- Test, verify and demonstrate the effectiveness and efficiency of the novel methodology for ESP shutdown prediction on a number of real-life case studies from the field
- Robust performance is guaranteed and versatile to various ESP applications onshore, offshore, SAGD, etc.
- Future directions more sophisticated anomaly detection algorithms for online selflearning on cloud-based distributed clusters

