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# Artificial intelligence for predictive maintenance in oil and gas operations

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# Abstract

Oil and Gas companies are maximizing capabilities using AI in the area of predictive maintenance to create data-driven insights leading towards higher operational efficiencies and safety. AI-driven predictive maintenance means analyzing data collected from various sensors and equipment to predict when a machinery fault might occur. And this approach helps cut back on unanticipated downtimes and drops maintenance costs to increase the life cycle of key assets.

AI driven systems with predictive analytics capabilities analyze equipment behavior for patterns, anomalies using AI techniques like machine learning algorithms, deep learning etc. Those methods enable detecting issues in early stages, so intervention can be done before the failure. The accuracy of learning models is enhanced by AI collaborating with IoT devices to provide real-time data and continuous monitoring.

Keywords: Predictive maintenance; Artificial Intelligence; machine learning; Oil and gas; IoT; Equipment reliability

# 1. Introduction

Predictive maintenance is among a critical process in modern industrial operations which now can leverage the techniques of Artificial Intelligence (AI) to foresee equipment failures before they occur. Research shows that predictive maintenance use in the oil and gas sector can result in substantial financial costs as well as safety risks when machines are down. AI based machine learning technology offers a way to help successfully implement an optimal solution. Scheduled maintenance and reactive repair are two commonly used traditional strategies that can be expensive, inefficient or simply non-suitable. Adopting AI can help optimize more process and data-driven approaches for the industry, hence improving operation efficiency as well reducing untimely failure.

A predictive maintenance system powered by AI uses machine learning algorithms to crunch mountains of operational data from sensors on equipment. These ML algorithms discover patterns and deviations that lead to equipment failures before it can compel you into action, pre-empting downtime. This is especially beneficial in the Oil and Gas industry where components are prone to high stress working conditions, leading to wear and tear. Research has indicated that the incorporation of AI in maintenance approaches can signal a reduction in unplanned downtime and an expansion on capital asset life, presenting opportunities for cost savings over time as well as operational reliability.

The deployment of AI powered predictive maintenance solutions also contribute to a greater degree of security and environmental protection. This minimizes risks of safety incidents and environmental spills by forecasting future failures, allowing businesses to correct these before they become more severe. While the oil and gas industry is grinding, with regulators breathing down its neck to enhance safety & decrease environmental impact, AI-driven predictive maintenance rises as a critical technology to its aid. This is a key milestone in the industry's overarching objective of operating safer and greener operations, leveraging operational excellence to meet regulatory requirements.

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Enhancing the management and monitoring of oil and gas processes demands the development of precise predictive analytic techniques. Over the past two years, oil and its prediction have advanced significantly using conventional and modern machine learning techniques,[1]. As stated in the International Energy Agency's 2020 report, the oil and gas (0&G) sector plays an important role in the global economy and substantially contributes to fulfilling the world's energy needs. The efficient management and optimization of operations within this sector are important for ensuring a dependable energy supply, mitigating environmental impacts, and maximizing economic returns [2,3]

# 2. Literature Review

**Table 1** Summary of Key Studies on Artificial intelligence for predictive maintenance in oil and gas operations, IncludingAuthor, Year, Methodology, and Main Findings

| Study                            | Objective  | Methodologies/Technologies                                | Key Findings   | Contributions   |
|----------------------------------|--|---|--|---|
| Zhang et al.<br>(2020)           | -  | Machine Learning, Neural<br>Networks                      | ML models significantly<br>improve prediction<br>accuracy for equipment<br>failures.       | potential of ML in  |
| Kumar et<br>al. (2022)           | 5  | Anomaly Detection, Time-Series<br>Forecasting             | with AI leads to reduced   |   |
| Chen et al.<br>(2021)            | To assess the role of<br>AI in enhancing<br>safety through<br>predictive<br>maintenance. | Predictive Analytics, Risk<br>Assessment Models           | AI improves safety by<br>predicting failures and<br>mitigating risks before<br>they occur. | the safety benefits of  |
| Wang et al.<br>(2022)            | To evaluate AI<br>applications in<br>environmental<br>monitoring and<br>maintenance.     | Real-Time Monitoring, Emission<br>Detection               | real-time detection of   |   |
| Al-Majed &<br>Ghabraie<br>(2021) | 0  | Predictive Maintenance Models,<br>Optimization Algorithms | operations and   | Demonstrates AI's role<br>in improving efficiency<br>and reducing costs in<br>drilling. |

### 3. What are predictive analytics

By definition, predictive analytics is a broad category of data analysis that includes statistical techniques, machine learning algorithms and other methodologies used to find the likelihood of future results based on historical or current data. In the oil and gas industry, this approach means applying these techniques to predict equipment failure in order to support business operations, decisions. AI powered applications swift through large volumes of data which are collected from various sources, such as sensors and also include historical records, predictive analytics helps companies become proactive and take data driven decisions.

AI has the potential to revolutionize maintenance techniques, and its ongoing development will indeed influence how the O&G sector develops in the future [4]. This is because there are still issues with AI methods and tools, such as overfitting, coincidence effects, and overtraining [5]. In the recent past, predictive analytics has emerged as a transformative capability, and platforms are looking to integrate these techniques to forecast future trends, and probable outcomes,[6].

Here are the few of the areas where predictive analytics in Oil & Gas industry could be used effectively:

- Equipment Maintenance: Predictive analytics are an excellent method for predicting equipment failures by analyzing the patterns and anomalies in your sensor data, operational logs, or maintenance records. This empowers Companies to undertake primary maintenance actions long before costly failures could develop. For example, predictive models are used to identify early signs of trouble in such critical equipment as pumps and compressors through vibration patterns and temperature readings.
- Drilling Optimization: When talking about drilling operations, predictive analytics can be used effectively to optimize the drilling parameters and reduce non-productive time (NPT). Predictive models use data from advanced drilling operations, to answer questions about ideal parameters such as mud weights and optimal speeds which will ensure risk mitigation while increasing speed of progress. That allows more precision and less expensive drilling methods.
- Reservoir Management: Predictive analytics is used to forecast production rates and enhance resource extraction in the management of reservoirs. Predictive models, developed by data analysis of geological and seismic as well as production information will generate important features such that predictions can be made for reservoir behavior or to highlight those sections with potential difficulties in the application of enhanced recovery techniques.
- Safety and Risk Management: Safety measures can be implemented in advance to mitigate potential hazards or risks that may not manifest until it is too late. Predictive models based on historical incident data and real-time monitoring can predict safety problems to avoid accidents happening in the future from memory as well raising its landscape level of operational risk visibility.

#### **3.1. Benefits of Predictive Analytics**

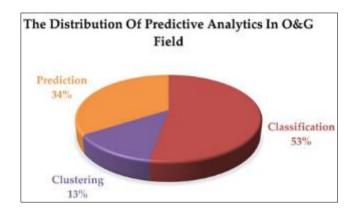
- Cost Reduction: Predicts equipment failures and optimal maintenance schedules to reduce the operational costs of unexpected downtimes & repairs.
- Increased Efficiency: With an optimized drilling parameter and production processes, businesses could potentially increase operational efficiency translating to higher output with less waste
- Enhanced Safety: By anticipating potential safety concerns early on, intervention can take place earlier reducing the chances of an accident to happen promoting workplace safety.
- Improved Decision-Making: Predictive analytics helps not only in addressing cost, efficiency & safety but also provides organizations ample actionable insights that lead to more data-driven decisions which will ultimately bring about better strategic as well as operational choices.

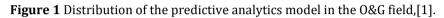
AI algorithms and data analytics have enabled more accurate predictions of equipment failures and optimized maintenance scheduling, leading to reduced downtime and operational costs,[7].

In general, the integration of predictive analytics technology is a modernization initiative for oil and gas operating companies that seeks to better coordinate corporate asset management practices with production objectives.

#### 4. Algorithms used in Oil & Gas Industry for Predictive maintenance

Variety of algorithms are employed to predict equipment malfunction and optimize schedule maintenance and sometimes combinations of algorithms are also used. The oil gas industry is no exception when it comes to using predictive analytics for preventive maintenance.





#### 4.1.1. Some major algorithms used, include:

Time-Series Forecasting Algorithms

- ARIMA (Autoregressive Integrated Moving Average): ARIMA models are used for forecasting based on historical time-series data. This model can predict future values by considering trends and seasonality in equipment performance data.
- Exponential Smoothing (ETS): applies smoothing techniques to time-series data to make predictions. They are useful for forecasting maintenance needs by smoothing out past observations.
- Long Short-Term Memory (LSTM): A type of Recurrent Neural Network (RNN). LSTM networks are effective in capturing long-term dependencies in time-series data, making them suitable for predicting equipment failures based on sequential sensor readings.

#### Classification Algorithms

- Support Vector Machines (SVM): this model is used to classify data into categories, such as "failure" or "no failure." They are useful for identifying patterns that precede equipment failures.
- Decision Trees and Random Forests: Classify data based on decision rules and are effective in identifying conditions that lead to maintenance events.
- Gradient Boosting Machines (GBM): predictive model which combines multiple weak models to create a strong predictive model. They are effective for predicting failure events by analyzing complex interactions between features.

#### Anomaly Detection Algorithms

Isolation Forest: The model isolates the outliers in the dataset, thereby making it useful for detecting unusual patterns or anomalies that may indicate impending equipment failures.

Autoencoders: A type of neural network used for anomaly detection, autoencoders learn a compressed representation of data and can identify deviations from normal behavior that suggest failures.

**Regression Algorithms** 

- Linear Regression: One of the most popular models and is based on the relationship between dependent and independent variables. This model can predict equipment wear or failure rates based on various operational parameters.
- Polynomial Regression: This extends linear regression by fitting a polynomial equation to data, allowing for more complex relationships between variables and better modelling of non-linear trends.

The model is used to predict the asset's lifetime based on readings collected from the sensors of each machine. From result, our prediction method using XGBoost for asset maintenance has presented a 6.43% increase in classification accuracy as compared to the Random Forest algorithm,[8].

Predictive analytics algorithms in preventive maintenance within the oil and gas industry are evolving and promising. By integrating advanced technologies, enhancing data utilization, and focusing on safety and sustainability, predictive analytics will continue to drive significant improvements in operational efficiency and asset management.

### 5. Future scope

Future of predictive analytics in the oil and gas realm is going to be driven by artificial intelligence (AI) & machine learning technology advancements, which will further improve both accuracy as well as maintenance strategies. As this technology develops, it will be able to support the programming of more advanced algorithms that can handle complex patterns from massive datasets. More accurate forecasts of equipment breakdowns will be possible, enabling operators to predict not only failure but also how they can improve operational parameters and hence manage their overall assets. When AI is combined with other new technology such as digital twins and edge computing, predictive capabilities will be improved even more with real-time simulations and on-site data processing which can help preemptive maintenance, being proactive in decision making.

Moreover, as broader adoption of big data and other cloud computing technologies gain momentum in the industry the way that predictive analytics is used will experience major paradigm shifts. The largest advantage is that companies will be able to develop predictive models on a more comprehensive and reliable level as they are now being exposed in handling disparate & very big amounts of data from many kinds of sources. This will allow for more predictable maintenance and operational efficiency, significantly reducing costs while improving performance.

The oil and gas industry relies heavily on aging infrastructure to extract, transport, and process hydrocarbons. As these assets age, the risk of failures and downtime increases, leading to safety hazards and costly repairs,[9]. With the recent advent of data recording sensors in exploration, drilling, and production operations, oil and gas industry has become a massive data intensive industry. Analyzing seismic and micro-seismic data, improving reservoir characterization and simulation, reducing drilling time and increasing drilling safety, optimization of the performance of production pumps, improved petrochemical asset management, improved shipping and transportation, and improved occupational safety are among some of the applications of Big Data in oil and gas industry,[10].In the future, the industry will see an increased focus on incorporating predictive analytics in combination with other asset management strategies to offer a more complete efficiency plan for both preventative maintenance and overall operational planning.

The continuous advancements of data-gathering techniques, the advent of smart wearables and the increasing popularity of the Internet of Things (IoT) devices in the field have significantly increased the amount of relevant data [11].

Predictive analytics will also assist the next generation of oil and gas services in these objectives around safety, sustainability. Using predictive analytics to enhance the precision over hazard detection while reducing environmental footprints, and hence facilitating net Neutral component on fuel retail operations from an enviro-ethno-economic perspective. An intuitive decision support system will continue to evolve, giving industry professionals real-time access for predictive insights that translate into improved safety mechanisms or operational efficiency. Predictive analytics is on the rise, and as that technology matures it will serve an increasingly important role in shaping how markets innovate toward a more sustainable future.

### 6. Conclusion

Concluding, advanced ML models combined with a wealth of data have poised predictive analytics to revolutionize preventive maintenance methods in the oil & gas sector. Over time, highly developed AI and machine learning algorithms will improve the pave for evidence-based forecasting of equipment failures as well to pinpoint maintenance requirements much more precisely. It will give businesses the ability to detect real-time sensor and other data, enabling quicker and more targeted interventions when downtime or operational disruptions occur. The transition from reactive to proactive maintenance will eventually become key in asset optimization as well as increasing the mean time between failures of critical equipment.

In addition, the combination of predictive analytics with new technologies like digital twins and edge computing promises to deliver much greater depth of insight plus more effective maintenance plans. The digital twins, which are virtual replicas of physical objects and devices allowing professionals to simulate equipment behavior in real time. Data processing and making decisions right from the point of data generation. That will allow better response times,

operational efficiency. These approaches will take maintenance to the next level and further translate into improved asset life, performance and greater operational excellence.

Predictive analytics will be mission-critical for ensuring that maintenance practices are in line with environmental and safety goals as the industry moves toward greater sustainability and safety. Predictive analytics will play a crucial role in industry achieving the energy-saving and waste-reduction benchmarks required by regulations (geared toward cleaning up hazardous sites, preventing petroleum contamination-incident fines). The use of predictive analytics in oil and gas promises a lot for the future: higher efficiency, lower cost, increased safety; setting it up neatly for further growth and innovation.

#### **Compliance with ethical standards**

#### Disclosure of conflict of interest

The author declares that there are no conflicts of interest associated with this publication

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