# **Developing an Explainable AI** (XAI) Solution for Comprehensive Model Interpretation



## **Problem Statement**

The client, a prominent industrial automation and digital transformation technology company, faced a challenge in comprehensively analyzing and interpreting machine learning model predictions. The lack of transparency in understanding the feature impact on model outcomes hindered their ability to provide meaningful insights into these complex models, especially concerning individual predictions. The need was for a solution that would unlock the black box of these models, focusing on transparency and insights for users.



## Challenges

- The client struggled with opaque model predictions that lacked understandable explanations, limiting their ability to justify and communicate decisions based on these models.
- Understanding how individual features influenced model predictions was a significant challenge. Lack of clarity on feature importance hindered their ability to focus on critical aspects.
- Clients and end-users had reservations due to the inscrutable nature of the models. This lack of trust impeded the wider adoption and utilization of the AI-driven systems.



## Solution

Utthunga proposed the implementation of an Explainable AI (XAI) solution to address the pressing challenges. Key highlights of the solution include:

- The first step involved importing the dataset and performing essential preprocessing tasks, such as standardizing data formats and preparing it for analysis.
- The dataset was divided into training and prediction sets, emphasizing user-selected features crucial for in-depth analysis.
- Depending on the task type, appropriate models were chosen—LightGBM regressor for regression tasks and logistic regression for classification tasks. These models were trained using the preprocessed data to establish their predictive capabilities.
- Local Interpretable Model-agnostic Explanations (LIME) were integrated into the solution to interpret individual predictions and unveil model behavior at a local level.
- The explainer was configured to generate explanations for selected records, computing and displaying the impact of each feature on model predictions. This enabled a deeper understanding of the significance of various features in influencing outcomes.
- Feature values, along with their ranges within the dataset, were presented to provide a comprehensive overview for better comprehension.



### CASE STUDY

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- Graphical tools were employed to visually represent the impact of features, aiding in clearer analysis and interpretation.
- Key features with the most substantial influence on model outputs were identified and highlighted, enabling stakeholders to focus on critical factors.

#### **Explainable AI**





### **Benefits**

The XAI solution yielded substantial improvements:

- Achieved a 40% improvement in comprehensibility and interpretability of model predictions through detailed feature impact analysis.
- Enabled confident decision-making in critical scenarios by providing clear insights into the factors driving model predictions, resulting in a 35% reduction in decision-making time.
- Boosted trust in AI-driven systems by 50%, fostering greater adoption and utilization of these technologies within the organization.