Summary of the whole <u>Lead Scoring</u> Project in short (max 500 words).

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X Education Lead Scoring Model – Summary Report

This case study details a systematic approach to refining lead conversion for X Education, an online education provider, by developing a logistic regression model that scores leads from 0 to 100. The business objective was to efficiently prioritize prospects with a high likelihood of conversion, thereby optimizing resource allocation and driving revenue growth.

Process Overview

1. Data Acquisition and Preprocessing:

The project began with importing the dataset (Leads.csv) and performing rigorous data cleaning. This step involved handling missing values, correcting inconsistencies, and addressing outliers to ensure the dataset's reliability. These preprocessing measures were critical in laying a solid foundation for the subsequent analysis.

2. Exploratory Data Analysis (EDA):

An in-depth exploratory analysis was conducted to understand the distribution and relationships between various features. By identifying potential predictors and gaining insights into the underlying data patterns, the team could effectively plan the model development phase. This analysis not only highlighted key trends but also guided the selection of appropriate variables for modelling.

3. Feature Engineering:

To enhance the model's performance, categorical features were transformed using dummy variables while continuous attributes were normalized. This step ensured that the data fed into the model was optimized for accuracy, thereby increasing the reliability of the predictive outcomes.

4. Model Development:

The prepared dataset was split into training and testing subsets. A logistic regression model was then developed to predict the probability of lead conversion. This stage was iterative, with the team fine-tuning parameters to maximize the model's predictive power while ensuring robustness.

Model Evaluation:

Evaluation metrics such as ROC curves, accuracy, recall, and specificity were used to assess model performance. Continuous monitoring and refinement during this phase ensured that the model met predefined business objectives and performance benchmarks.

6. Lead Scoring and Business Application:

The logistic regression output was translated into a lead score on a scale of 0 to 100. A higher score signifies a "hot lead," enabling the sales team to prioritize their engagement efforts effectively. This scoring system directly supports strategic decision-making and resource allocation, making it an invaluable tool for improving conversion rates.

Key Learnings

• Data Quality is Paramount:

Rigorous data cleaning and preprocessing are essential to build a reliable predictive model. Inconsistent or missing data can significantly skew results.

• Iterative Analysis Enhances Accuracy:

The EDA phase provided critical insights that informed both feature engineering and model selection, underscoring the value of iterative analysis in predictive modelling.

• Agile Model Tuning:

The iterative development and evaluation process allowed for flexible adjustments, ensuring that the model remained aligned with evolving business requirements.

• Scalable and Adaptable Framework:

The modular design of the process facilitates easy updates and integration of new features, supporting future enhancements as additional data becomes available.
