

Predicting Advertisement Clicks Using Logistic Regression

A Data-Driven Approach to Optimize Digital Marketing

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Executive Summary

Brief overview of the project

Key outcomes:

- Data preprocessing and feature engineering
- Model performance with a log loss of 0.426
- Business impact: Improved resource allocation and user engagement

Introduction

- Importance of predicting advertisement click-through rates
- Dataset:
 - 13 million entries
 - 24 categorical variables
 - Target variable: `click`
- Objective: Create a logistic regression model to minimize log loss and optimize predictions.

Data Processing

- **Steps Taken:**

- Transformation of the `hour` column into `day_of_week` and `hour_of_day`
- Encoding techniques:
 - One-hot encoding for low-cardinality variables
 - Frequency encoding for high-cardinality variables
- Removal of ID-related and highly correlated variables

Model Building

- **Model Used:** Logistic Regression
- **Why Logistic Regression?**
 - Handles binary classification effectively
 - Scalable for large datasets
 - Works well with categorical variable encoding
- **Feature Selection:**
 - Backward elimination
 - P-value significance testing

Model Evaluation

- **Metric:** Log loss
 - Penalizes overconfident wrong predictions
 - Encourages calibrated probability estimates
- **Results:**
 - Log loss: 0.426 on validation data
 - 90%-10% split for training and validation

Business Impact

- Enables targeted advertising by predicting user interaction
- Helps allocate resources effectively
- Improves ad placement strategies and revenue generation

Future Improvements

- Explore advanced machine learning techniques:
 - Random Forest
 - Gradient Boosting
- Integration with real-time advertising platforms for dynamic updates

Conclusion

- Logistic regression model optimized for CTR prediction
- Data preprocessing and feature engineering significantly impact outcomes
- Model aligns with business goals to enhance digital marketing efforts

Thank you!!

Q&A