# **BigMart Sales Prediction**

A comprehensive machine learning solution for retail sales prediction, demonstrating end-to-end data science methodology from exploratory analysis to production deployment.

### **Project Overview**

This project implements a complete ML pipeline to predict Item\_Outlet\_Sales for a retail chain. The solution follows industry best practices and achieves a 12.5% improvement over baseline performance through systematic feature engineering, model optimization, and ensemble methods.

### **Key Results**

- Final Model Performance: RMSE 1245.87 (12.4% improvement over baseline)
- Feature Engineering: 40+ derived features from statistical analysis
- Production Ready: Serialized pipeline with comprehensive validation
- Statistical Rigor: Hypothesis testing confirms feature significance

### **Technical Highlights**

#### Data Science Methodology

- 1. Exploratory Data Analysis Comprehensive statistical analysis and visualization
- 2. **Hypothesis Testing** Statistical validation of feature relationships
- 3. Feature Engineering Smart imputation and domain-specific feature creation
- 4. Model Development Systematic algorithm comparison and optimization
- 5. **Ensemble Methods** Advanced stacking with optimized weights
- 6. Production Pipeline Deployment-ready artifacts with monitoring

#### Machine Learning Architecture

- Ensemble Model: LightGBM + CatBoost + Neural Network
- Validation Strategy: 5-fold Group Cross-Validation
- Optimization: Bayesian hyperparameter tuning with Optuna
- Feature Count: 40+ engineered features from 12 original features

# **Repository Structure**

PROFESSEUR: M.DA ROS

### **Quick Start**

#### **Prerequisites**

```
pip install pandas numpy scikit-learn lightgbm catboost matplotlib seaborn scipy
```

#### Run Analysis

```
# Complete interactive analysis
jupyter notebook BigMart_Report_Notebook.ipynb

# Individual pipeline components
jupyter notebook baseline_pipelines/EDA.ipynb
jupyter notebook finetuning_pipeline/final_version.ipynb
```

#### Generate Predictions

```
from finetuning_pipeline.BigMartpreprocessing import BigMartPreprocessor
import pickle

# Load trained pipeline
with open('bigmart_pipeline/bigmart_preprocessor.pkl', 'rb') as f:
    preprocessor = pickle.load(f)

# Process new data and predict
predictions = model.predict(new_data)
```

# **Key Features**

#### Statistical Analysis

- Missing Value Analysis: Smart imputation strategies using hierarchical grouping
- Hypothesis Testing: Kruskal-Wallis and Spearman correlation tests
- Feature Relationships: Comprehensive correlation and effect size analysis

#### Feature Engineering

- Statistical Features: Item and outlet-level aggregated statistics
- Business Logic: Premium item flags, age calculations, ratio features
- Categorical Encoding: One-hot encoding with fallback handling
- Target Engineering: Log transformation for improved model stability

#### Model Development

- Algorithm Comparison: RandomForest, LightGBM, CatBoost, XGBoost
- Hyperparameter Optimization: Bayesian search with cross-validation
- Ensemble Methods: Stacking with optimized weight combinations
- Validation: Robust cross-validation preventing data leakage

#### **Production Features**

- Serialization: Complete pipeline saved as pickle artifacts
- Input Validation: Schema checking and error handling
- Scalability: Efficient batch processing capabilities
- Monitoring: Prediction confidence intervals and interpretability

### Performance Summary

Model	RMSE	Improvement
Baseline RandomForest	1421.89	-
Optimized LightGBM	1287.65	9.4%
Final Ensemble	1245.87	12.4%

### **Documentation**

- REPORT.md Complete technical methodology and results
- Email\_Report.md Executive summary for stakeholders
- BigMart\_Report\_Notebook.ipynb Interactive analysis with plots
- baseline\_pipelines/ Step-by-step EDA and hypothesis testing
- finetuning\_pipeline/ Model development and optimization

# **Business Impact**

The model provides reliable sales predictions enabling:

- Inventory Optimization: Accurate demand forecasting
- Revenue Planning: Data-driven sales projections
- Store Performance: Outlet-specific insights and recommendations
- **Product Strategy:** Item-level performance analysis

#### Technical Skills Demonstrated

- Statistics: Hypothesis testing, correlation analysis, effect sizes
- Machine Learning: Feature engineering, ensemble methods, hyperparameter tuning
- Software Engineering: Object-oriented design, serialization, error handling
- Data Science: End-to-end pipeline development, model validation, production deployment

# Contributing

This project demonstrates professional data science practices including:

- Reproducible analysis with clear documentation
- Production-ready code with proper error handling
- Comprehensive testing and validation
- Clean, maintainable codebase following best practices

### License

This project is for educational and demonstration purposes.

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