

SALES FORECASTING SYSTEM

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DATE : 31-10-2025.

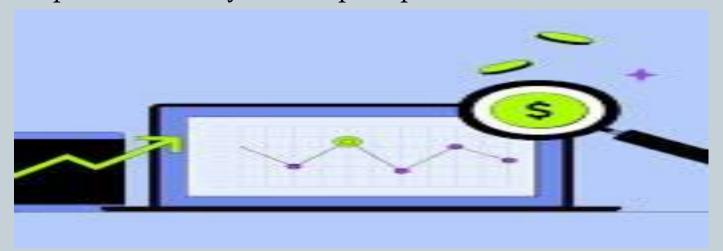
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ABSTRACT

• This project presents the development of an AI-driven sales forecasting system that leverages machine learning models to predict future sales trends. The system improves decision-making and helps businesses plan effectively by providing accurate, data-backed forecasts. AI-based forecasting uses machine learning algorithms for improved accuracy and adaptive predictions.

INTRODUCTION

- Sales forecasting estimates future sales using historical data, market trends, and predictive models.
- AI-based forecasting uses machine learning algorithms for improved accuracy and adaptive predictions.



PROBLEM STATEMENT

- Traditional sales forecasting methods are limited by human error and lack adaptability to sudden market shifts.
- Businesses require an automated, intelligent system to analyze large datasets and provide accurate, realtime forecasts.

OBJECTIVES

- Automate the sales forecasting process using AI algorithms.
- Improve forecast accuracy using machine learning techniques.
- Provide visualization dashboards for business insights.
- Evaluate system performance using standard metrics.

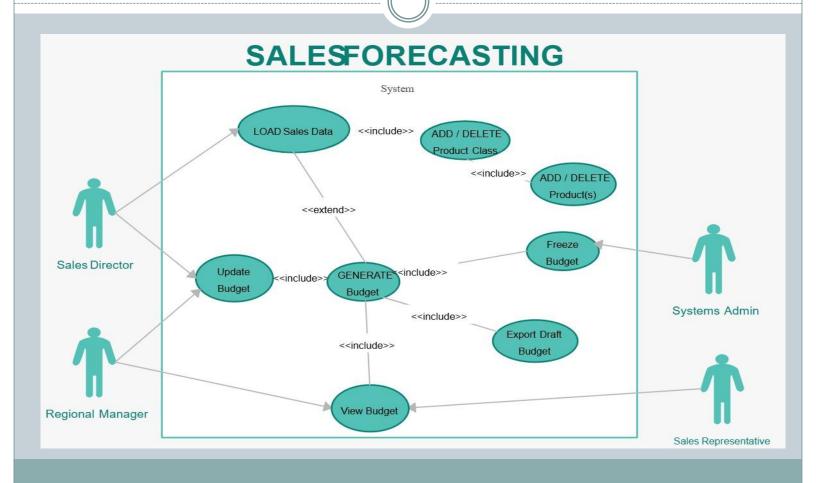
LITERATURE REVIEW

- Time Series Forecasting Models (ARIMA, SARIMA).
- Regression-based Predictive Models.
- Neural Networks (LSTM, GRU).
- Ensemble Methods (Random Forest, XGBoost).

PROPOSED SYSTEM ARCHITECHTURE

- Data Collection Module.
- Data Preprocessing & Cleaning.
- Machine Learning Model Training.
- Prediction & Evaluation.
- Visualization Dashboard.
- Each module ensures smooth data flow and real-time updates.

SYSTEM DESIGN DIAGRAM



DATA FLOW DIGRAM

Gathering Data



Analysis of data



Choosing best sales forecasting model



Sales Forecasting



Sales Forecasting results Evaluation

ALGORITHM EXPLANATION

- Linear Regression Establishes trend lines based on previous data.
- Random Forest Aggregates multiple decision trees for better generalization.
- **LSTM** Uses memory cells to handle sequential data effectively.
- **ARIMA** Captures seasonal and trend-based time series data.

IMPLEMENTATION

- Programming Language: Python.
- Libraries: Pandas, NumPy, Scikit-learn, TensorFlow, Matplotlib.
- Database: MySQL / CSV Dataset.
- **Development Tools:** Jupyter Notebook, Visual Studio Code.

SYSTEM WORKFLOW

- Import dataset.
- Clean and normalize data.
- Train ML model (LSTM, Random Forest, etc.).
- Validate using test dataset.
- Generate forecasts and visualize results.

RESULT & DISCUSSION

• The model produced high accuracy with reduced mean absolute error (MAE).

• Key Insights:

- Seasonal variations identified
- Product-wise sales trend recognition
- Demand peaks forecasted successfully

PERFORMANCE METRICS

- Mean Absolute Error (MAE).
- Root Mean Square Error (RMSE).
- R² Score.
- Accuracy Percentage.
- **Example:** LSTM achieved 92% accuracy compared to 85% using Linear Regression.

APPLICATIONS

- Retail and E-commerce.
- Inventory Management.
- Demand Forecasting.
- Supply Chain Optimization.
- Financial Planning.

CONCLUSION

• The AI-based Sales Forecasting System enhances predictive accuracy and enables data-driven business strategies.It minimizes manual effort and allows real-time forecasting for better decision-making.

FUTURE SCOPE

- Integration with cloud-based APIs.
- Real-time sales monitoring.
- Mobile application development.
- Integration of external data (social trends, economy, weather).
- Advanced deep learning integration.

REFERENCES

- Brownlee, J. 'Deep Learning for Time Series Forecasting'.
- Hyndman, R. J., & Athanasopoulos, G. 'Forecasting: Principles and Practice'.
- Scikit-learn and TensorFlow documentation.
- IEEE research papers on sales prediction models.

