

# Pharmaceutical Inventory Optimization



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# The Problem Statements

## Case 1: Stockouts -> Patient Risk

- A hospital pharmacy runs out of insulin, delaying treatment for diabetic patients
- Impact: Increased patient hospitalizations & mortality risks

## Case 2: Stockouts -> Overstock example

- A pharmacy overorders flu vaccines, leading to expired stock
- Impact: Increased hospitalizations & regulatory waste disposal issues

- 01** Stockouts – Critical medications run out, risking patient health.
- 02** Overstocking & Expiration – Billions lost due to expired drugs.
- 03** Regulatory restrictions- FDA, DEA regulations require strict tracking
- 04** Supply chain disruptions increase risk & volatility

## OUR OBJECTIVE

# Our Data Driven Approach

Cleaned the data using Python Script

Monitor Stock Levels – Avoid overstock & stockouts.

Predict Demand – Using time series analysis, predict future demand (based on historical data)

Optimize Reorder Points – Reduce costs while maintaining availability



# Tech Stack

## Frontend & UI:

**Streamlit** → Built the interactive dashboard, displaying graphs, dropdowns, and tables.

**Matplotlib** → Rendered line charts, histograms, and time series plots within Streamlit.

## Backend & Data Processing:

**Pandas** → Cleaned and structured pharmaceutical sales data from a CSV file.

**Datetime & Period Handling** → Used `pandas.to_datetime()` and `PeriodIndex` for proper date conversions.

## Time Series Forecasting:

**Statsmodels (ARIMA)** → Used AutoRegressive Integrated Moving Average (ARIMA) to model and forecast future sales.

`pd.DateOffset()` → Generated correct date ranges for the forecasted months.

**Date Formatting (Matplotlib Dates)** → Used `AutoDateLocator()` and `DateFormatter("%Y-%m")` for clear x-axis labels.



Streamlit



ARIMA Model

Autoregressive  
(p)

Integrated (d)

Moving  
Average (q)

**DEMO**



# Projected Plans for Pharmaceutical Inventory Optimization

## 1. Machine Learning for Demand Forecasting

- **Random Forest Regression** – Analyzes multiple factors affecting sales.
- **XGBoost (Gradient Boosting Trees)** – Captures nonlinear demand trends.
- **LSTM Neural Networks** – Best for deep-learning time-series forecasting.
- **Tools:** Scikit-learn, XGBoost, TensorFlow

## 2. Inventory Optimization Techniques

- **Economic Order Quantity (EOQ)** – Determines optimal order size to minimize costs.
- **Safety Stock Calculation** – Prevents stockouts by accounting for demand variability.
- **Reorder Point (ROP)** – Automates reordering when stock levels hit a threshold.

## 3. Multi-Echelon Inventory Optimization

- Balances stock across warehouses, hospitals, and pharmacies to reduce waste and improve efficiency.
- Ensures cost-effective distribution while maintaining medication availability.
- **Tools:** Gurobi, OR-Tools



# Thank You!

