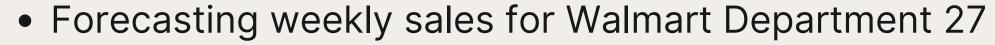
Department 27









• Dataset: 143 weekly records (Feb 2010 – Oct 2012)

6 Features: Date - IsHoliday - Temperature - Fuel\_Price CPI - Unemployment

Target variable: Weekly\_Sales





### Data Exploration

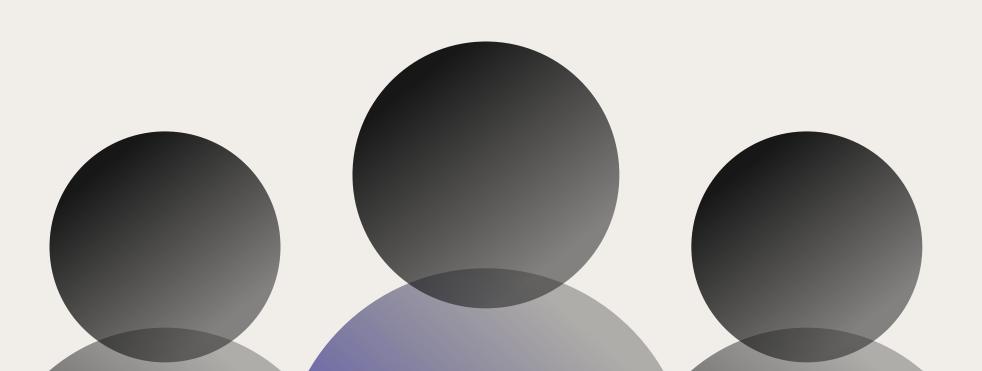
Clear yearly seasonality

ACF/PACF analysis to inform ARIMA/SARIMA

- Explored classical time series models: Holt's Linear Trend, Holt-Winters, ARIMA and SARIMA.
- Applied machine learning regressors:
  Random Forest and XGBoost
- Tested deep learning:
  MLP (Multi-Layer Perceptron) and LSTM
  - Investigated an innovative model:

**N-BEATS** 

• Final model chosen for deployment: **Prophet** 



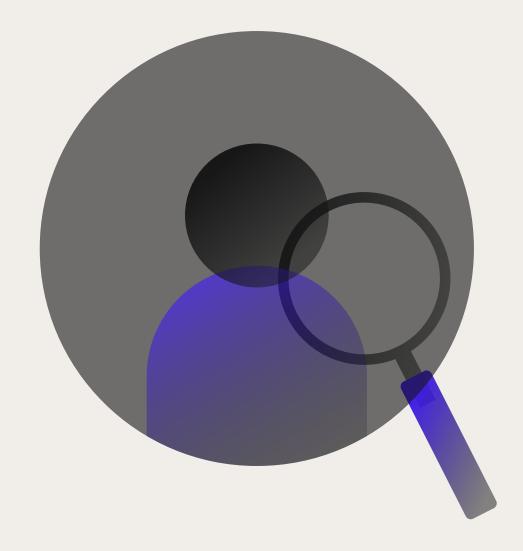
### Modeling Approach



# Model metrics sorted by RMSE

## Modeling Approach

Model	RMSE	MAE
Prophet	156.24	131.36
MLP	187.13	144.20
SARIMA(2,1,2)(1,0,1,52)	213.22	165.94
Random Forest	229.62	186.40
XGBoost	243.16	183.70
LSTM	243.72	202.48
Holt-Winters	265.32	215.23
Holt Linear	354.11	314.54
ARIMA(3,1,2)	428.87	384.86
N-BEATS	1403.73	1087.43

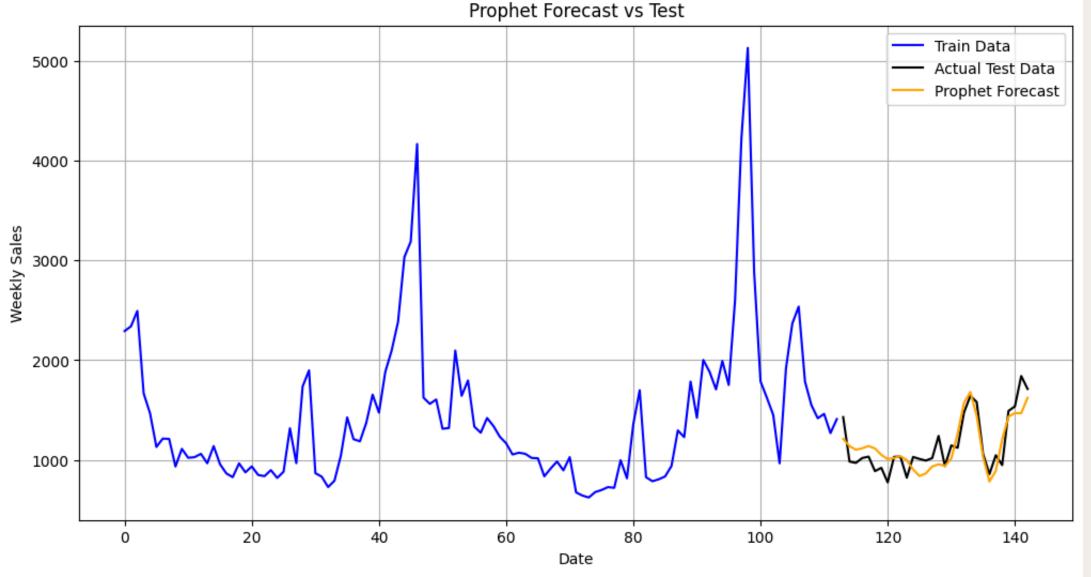


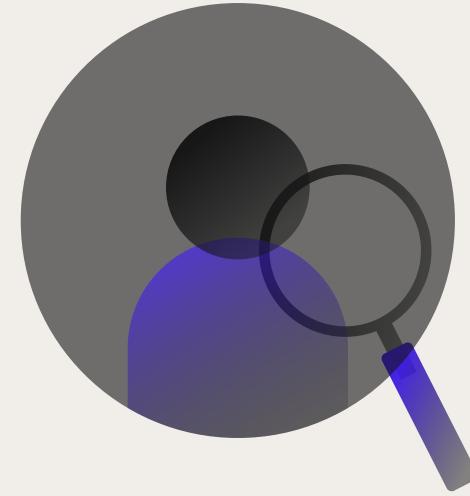
#### **Prophet Metrics:**

MAE: 131.36

RMSE: 156.24

# Modeling Approach







### Deployment

1.

2.

3.

4.

5

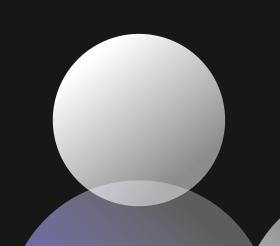
Prophet model pickled for reuse Built interactive Streamlit app:

User inputs forecast horizon

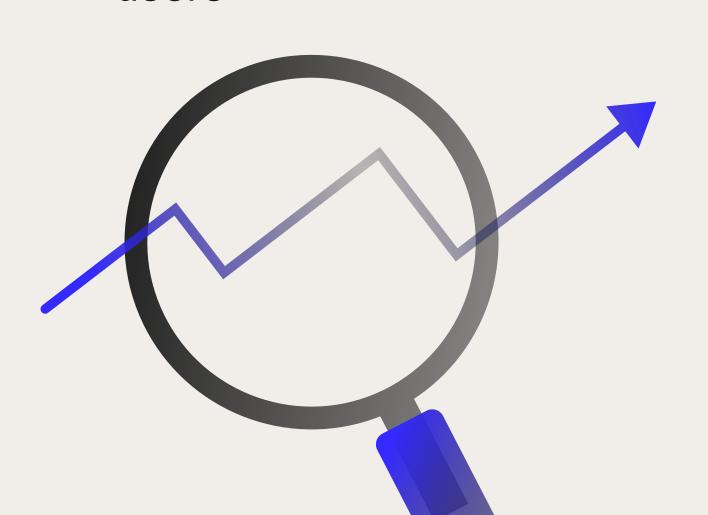
Generates table of forecast

Ready for integration / hosting





- Aligning forecasts with correct future dates
- Tuning models without overfitting (especially for DL/ML models)
- Formatting outputs for non-technical users





### Challenges Faced

