No Forecast? No Problem

Improving customer demand forecasting



Thermo Fisher





Project sponsor: Sarah Englert

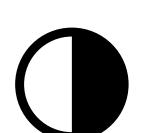
Faculty advisor: Dr. Jonas Jonasson

MIT team: Zack Horton & Tanner Street

Problem statement



The S&OP team manually generates 12-month forecasts for revenue and job steps each month.



50% of revenue is from clients that don't provide forecasts, making them harder to predict.



Goal: Automate entire forecasting procedure and improve accuracy for 'no forecast' clients.

Data sources



Salesforce: Information on upcoming contracts (features predictive of revenue)



Revenue: Historical monthly revenue

Job Steps: Historical monthly job steps

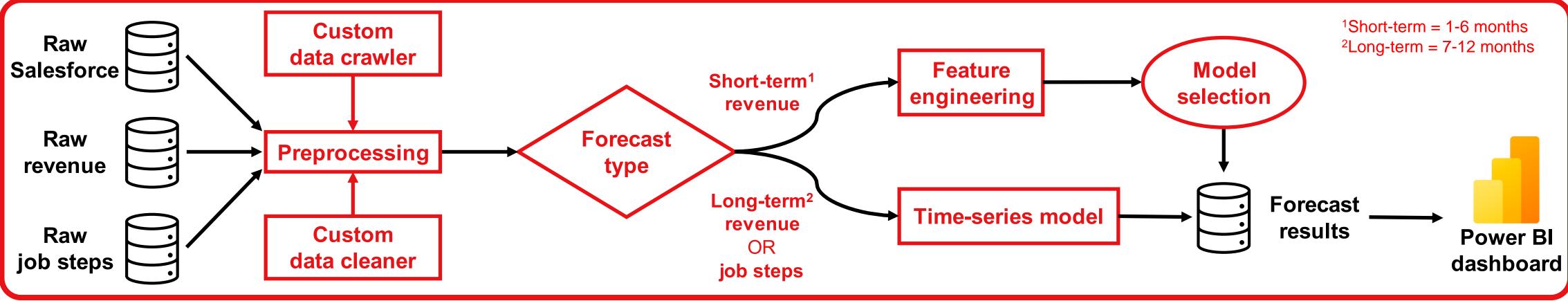
42

months

global sites

client categories

Automated system overview



Short-term revenue

Feature engineering

1.6K Features from

data crawler

Sparse PCA, aggregated by client bucket

Features to capture maturity of future opportunities

Seasonality and trend

Model selection



Walk-forward validation procedure selects model with lowest percent error



Select between Elastic Net, Random Forest, and XGBoost models

Long-term revenue & job steps

Time-series model

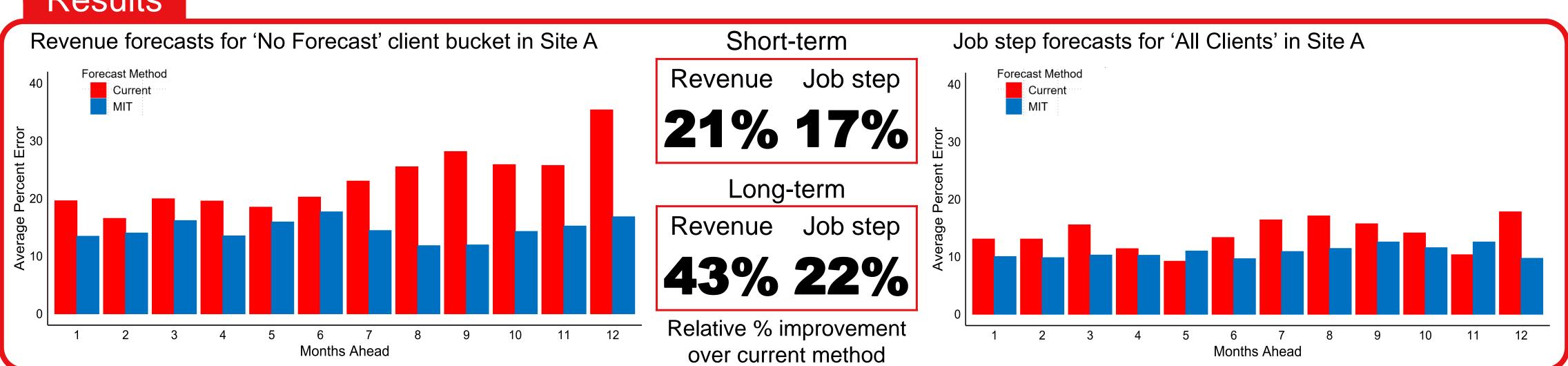


Structural time-series methods that do not require exogenous variables

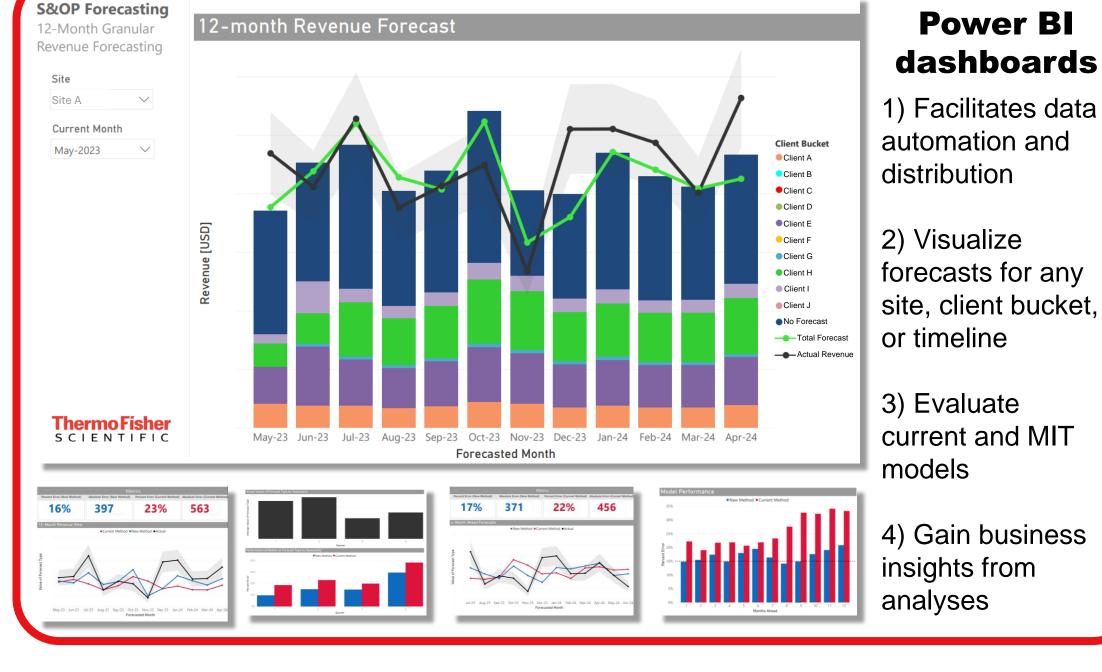


Models include Seasonal-Trend Decomposition using Loess and Holt-Winters Exponential Smoothing

Results



Reporting



Impact Relative % decrease in operating costs from forecast improvements **78** 46 **Asset** Labor Revenue scheduling planning expectations 3-months ahead 12-months ahead 6-months ahead

10%

Reduction in monthly working hours for senior demand planners due to automated system



Nicole Barker, senior demand planner

"This tool saves us hours of time in manual work each month and it allows us to more accurately forecast our customers' demand to ensure we are well equipped to meet their needs."