



INNOVATE

DATA AND AI/ML EDITION

22 February 2023

From accuracy to business case: Scoping a successful demand forecasting PoC

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Solutions Architect

Amazon Web Services

Agenda

1. Building the business case for forecasting
2. Creating machine learning based forecasts on AWS
3. Scoping a successful PoC
4. Demo
5. Next steps

Building the business case for forecasting



Cost



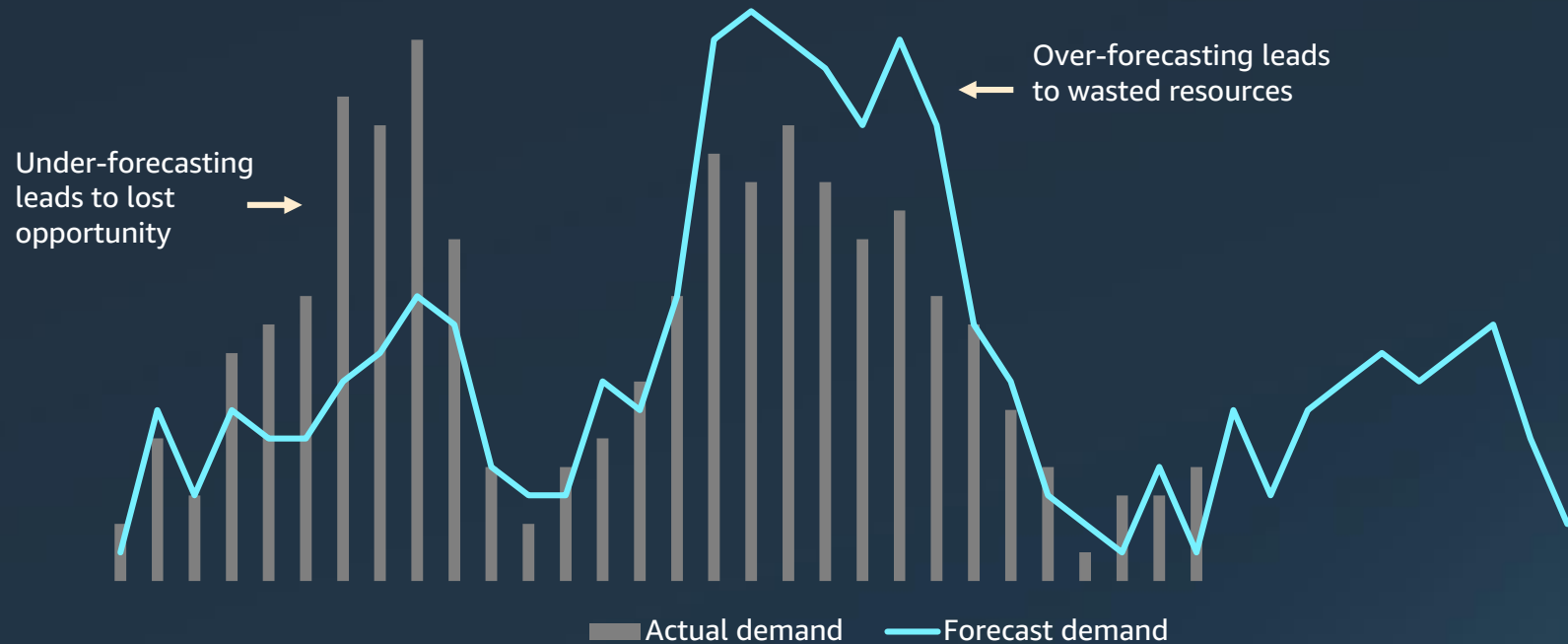
**Revenue /
Margin**

The case for forecasting

Forecasting is the science of predicting the future

Product demand

Actual demand vs. forecasted demand (\$ Millions)



The case for forecasting

Impact of under and over forecasting



How customers built a case for ML-based forecasting



Heroleads

99%

increase in
forecast
accuracy to
track
marketing
campaign KPIs

More Retail

30%

reduced food
waste by 30%
while
improving
in-stock
availability

Foxconn

\$553K

projected
annual savings
with an
increase in
forecasting
accuracy of
8%

Forenamics

35%

reduction in
overproduction

25%

reduction in
underproduction

Fabulyst

10%

boost in
incremental
revenue for
retailers by
forecasting
monthly
trends

Creating machine learning based forecasts on AWS

AWS low/no code options to incorporate ML in forecasts



Amazon SageMaker Canvas

- No ML knowledge required
- No code, visual point-and-click interface
- Automatically identifies the problem type, using ML techniques such as linear regression, logistic regression, deep learning, time-series forecasting, and gradient boosting
- Share your Amazon SageMaker Canvas models with data scientists who use Amazon SageMaker Studio



Amazon Forecast

- No ML knowledge required
- API integration
- Provides the option to include related time-series data and item metadata; can incorporate holiday and weather data
- Deploy recurring workflows with no-code using AWS CloudFormation and AWS Step Functions

Amazon SageMaker Canvas

Build ML models and generate accurate predictions — no code required



Quickly access and prepare data for machine learning



Built-in AutoML to build models and generate accurate predictions



Share ML models and collaborate with data science teams



Usage-based pricing to avoid licensing fees and reduce total cost of ownership

Amazon SageMaker Canvas

Ama
Canv

store_sales_forecast_model

Prereq

Select

Build

Analyze

Predict

Lab 1
(Mark

Lab 2
Estate

Lab 3

Lab 4
(Finan

Lab 5
Readn

Lab 6
(Healt

Lab 7
Types

Lab 8
time

Lab 9
Logist


Select a column to predict

Choose the target column. The model that you build predicts values for the column that you select.

Target column

sales

Value distribution



Model type

SageMaker Canvas automatically recommends the appropriate model type for your analysis.

Time series forecasting

Your model will forecast sales by using past data values to predict future data values.

Change type

Configure

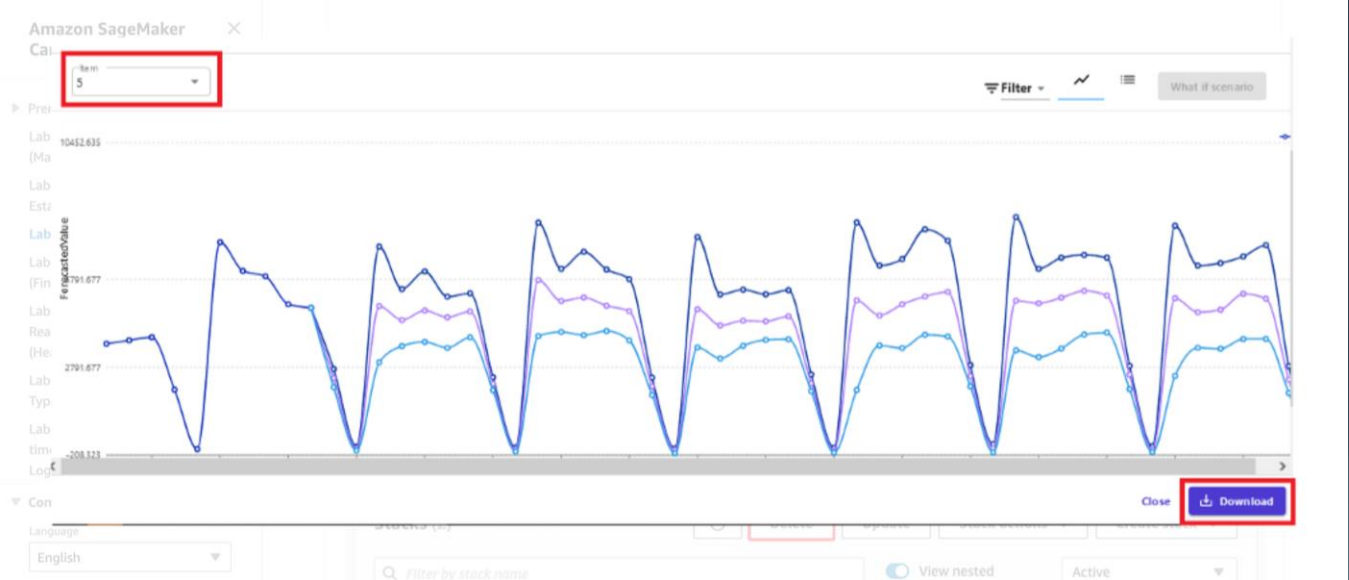
Conte

store_sales_data

Sample

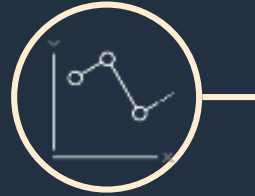
Language

Engli



Amazon Forecast

**Advanced machine learning time-series
forecasting**



Create highly accurate models without deep learning expertise



Get started easily with data in Amazon S3 using the console or API



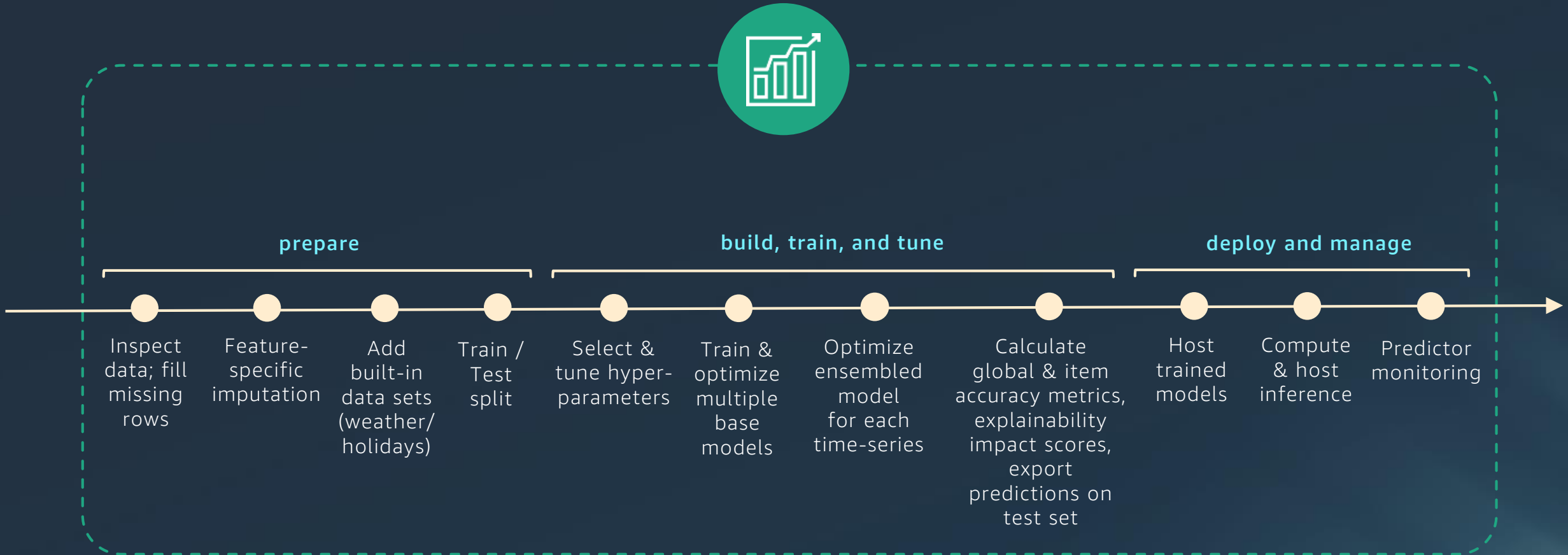
Automatically train and select the best model using AutoML



Incorporate item metadata, related time-series, holiday and weather information

Amazon Forecast is a fully managed ML service

PREPARE, BUILD, TRAIN, TUNE, DEPLOY, AND MANAGE



Examples of related time-series and item metadata

Use Case	Related time-series	Item metadata
Quick service restaurants	Store hours	Menu categorization (breakfast, lunch, dinner)
Grocery retail	Price, store hours, holidays, promotions	Department, store size, shelf life, type of stores, affluent cluster, competitive store, stores throughput
Manufacturing	Economic indicators from Bloomberg, backlog data	Marketing family, department
Hyper local on-demand services	Holidays, weather	City characteristic

Note:

- Most of the forecasting power comes from the historical demand data
- Adding additional related data does not always increase forecasting accuracy
- First create a model with only the target time-series and assess the accuracy gain, before adding additional related data
- Gaining benefit from adding related data is dependent on feature engineering

Ensemble model algorithms to improve accuracy

Ensemble modeling selects the best combination of algorithms for each product

Neural Networks <ul style="list-style-type: none">• Designed for larger data sets• External data		Statistical Algorithms <ul style="list-style-type: none">• Higher level forecasting• Fast & accurate			
CNN-QR	DeepAR+	Prophet	NPTS	ARIMA	ETS
Uses causal convolutional neural networks (CNNs)	Global neural model	Additive model with Gaussian likelihood	Non-parametric time-series	Auto-regressive integrated moving average	Statistical algorithm that uses exponential smoothing
Works best with large datasets containing hundreds of time-series; accepts related time-series data without future values	Uses related time-series and attributes to train a model	Can find trend, seasonality, cyclical, and holiday effects	Performs well for intermittent spikes	Works well with a small number of time-series; classical approach to model autocorrelations	Works well with a small number of time-series Finds trends, seasonality, and residual

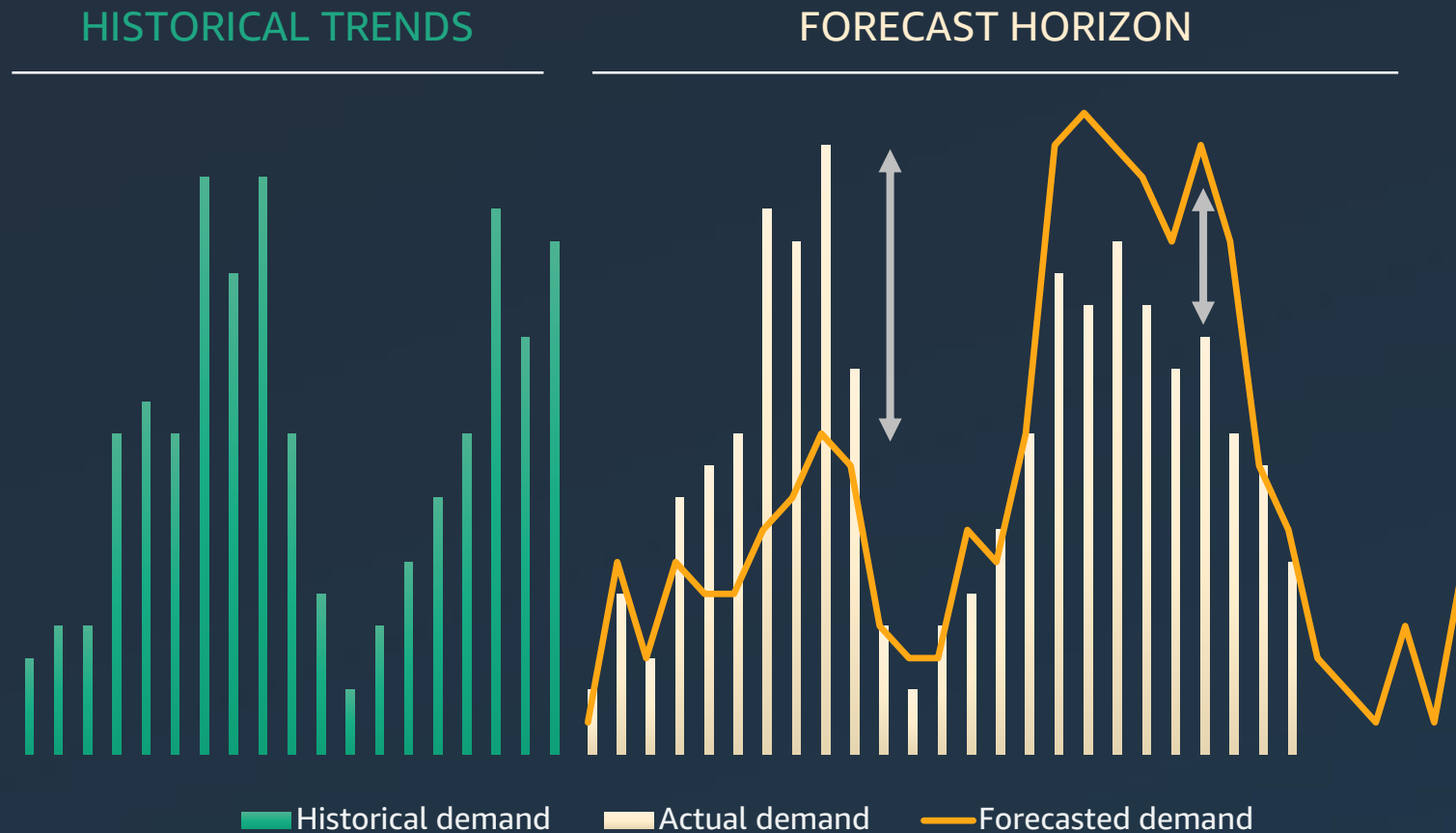
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AutoPredictor					

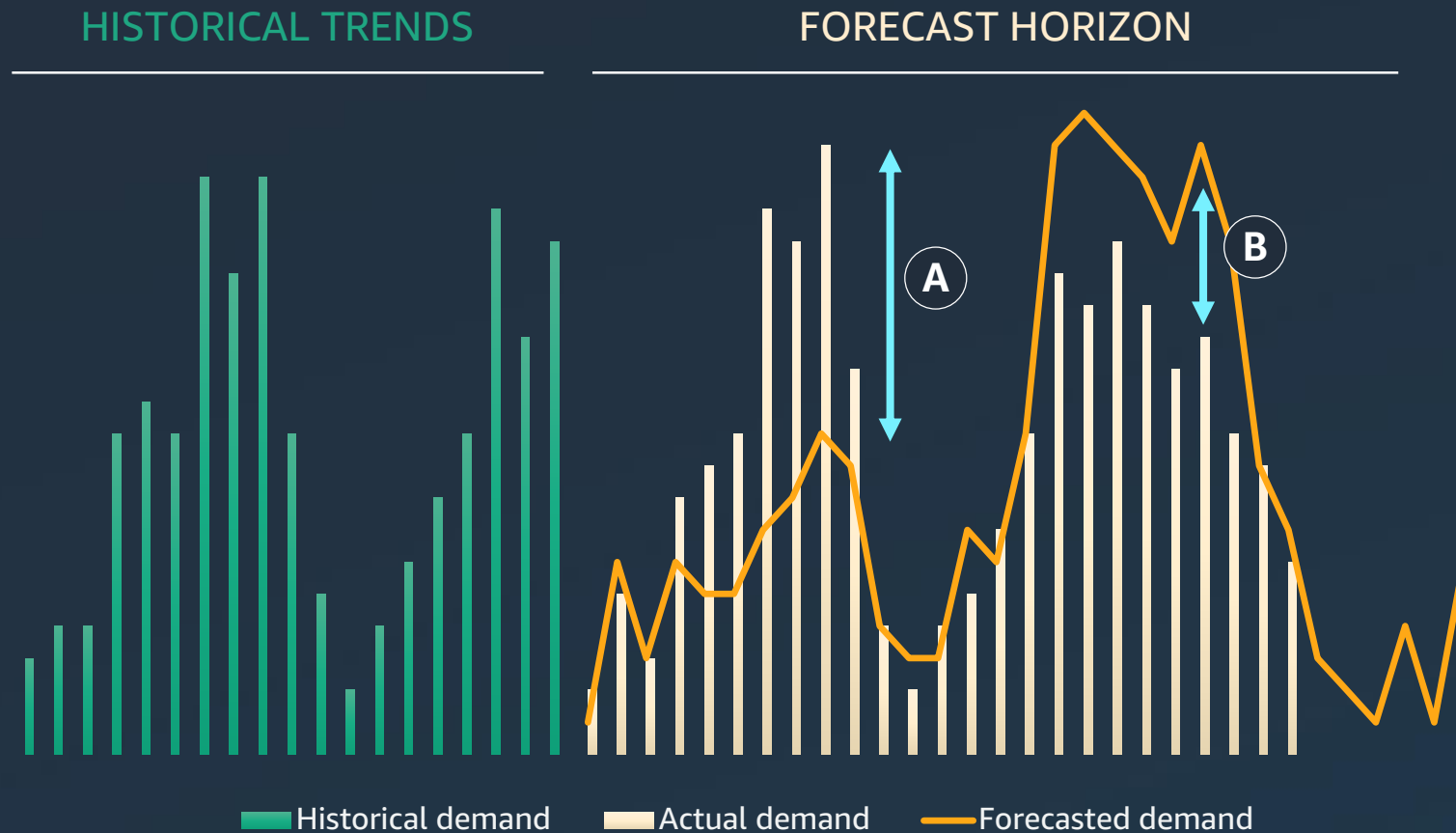
Scoping a successful PoC

Traditional metrics for evaluating your forecast



- Weighted Quantile Loss (wQL)
- Weighted Absolute Percentage Error (WAPE)
- Root Mean Square Error (RMSE)
- Mean Absolute Percentage Error (MAPE)
- Mean Absolute Scaled Error (MASE)

Traditional metrics for evaluating your forecast



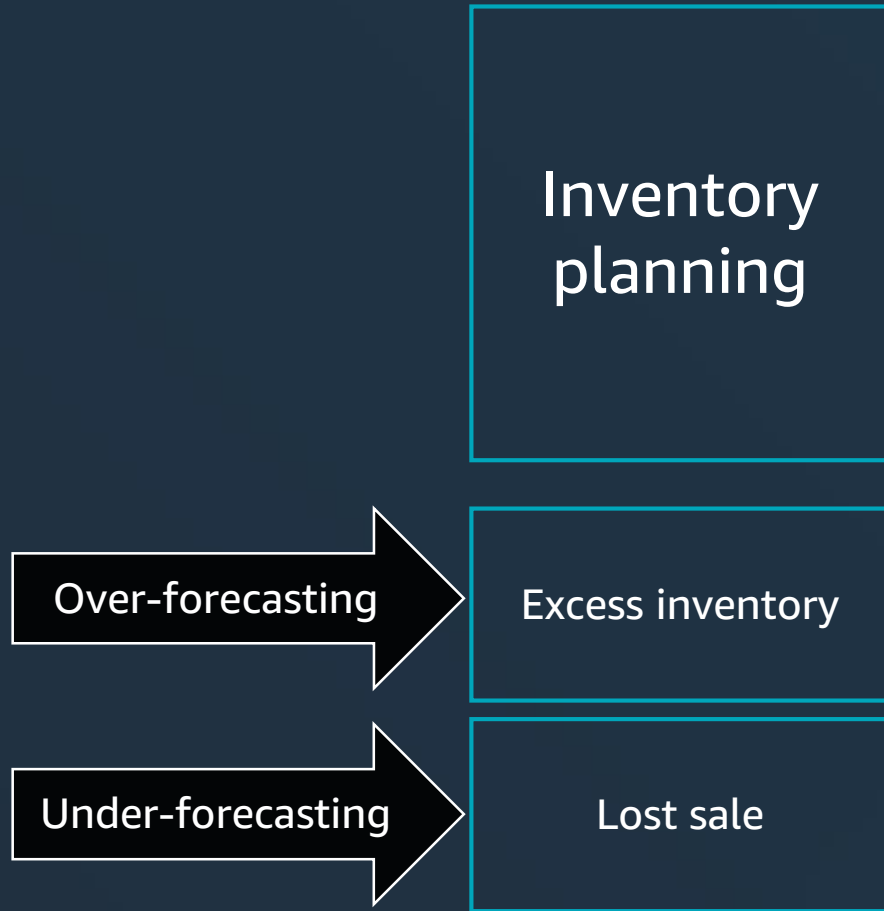
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Making sense of these errors:

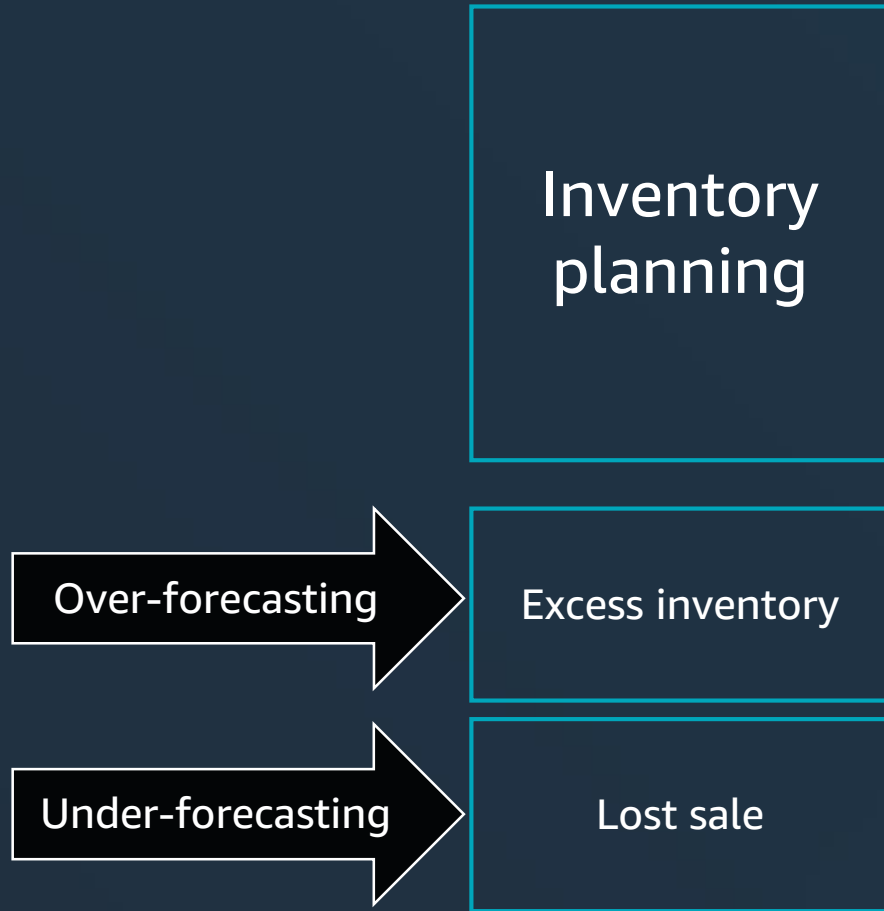
A - Stockouts

B - Excess inventory

Tying back to business outcomes



Tying back to business outcomes



The cost of excess inventory is affected by:



Storage space

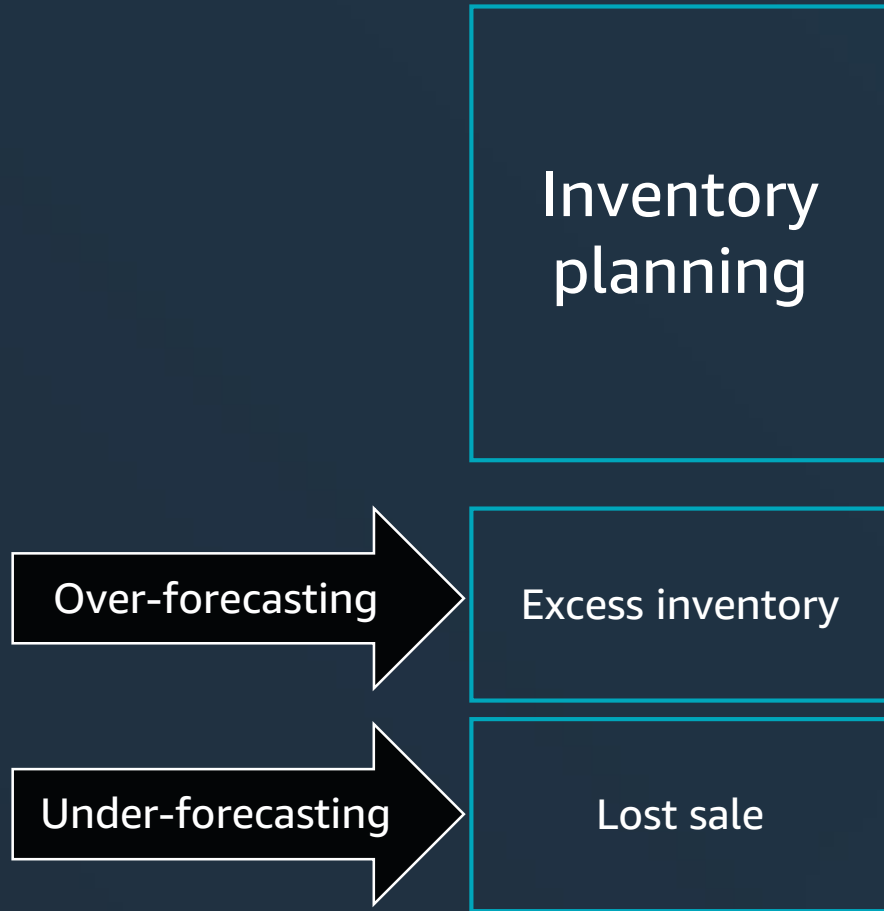


Shelf life



Exceeding capacity

Tying back to business outcomes



High level approximation for each item per month:

$$\frac{\textit{Total storage costs}}{\textit{Capacity for incremental storage costs}} \times \textit{factor for the cost of an unsold item}$$

Decide on your evaluation strategy

Backtesting

- Create a baseline using a selected timeframe of historical data
- Generate forecasts with the new algorithm using historical time-series data to compare against the baseline
- Calculate statistical and/or business metrics and compare between the algorithms
- Amazon Forecast uses backtesting to tune parameters and produce accuracy metrics

Decide on your evaluation strategy

Backtesting

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Forward-looking A/B testing

- Choose items to include in the test group
- Generate forecasts with the new algorithm using historical time-series data for this group, while generating forecasts as per usual using any existing algorithm
- Implement a purchasing policy with input from the generated forecasts
- Calculate statistical and/or business metrics and compare between the algorithms

A/B testing considerations



Metric(s) being measured should be assignable to dollar values



Results from offline model metrics and/or past tests should justify the risk of testing



Factor in the difference in the price or margin of items among groups

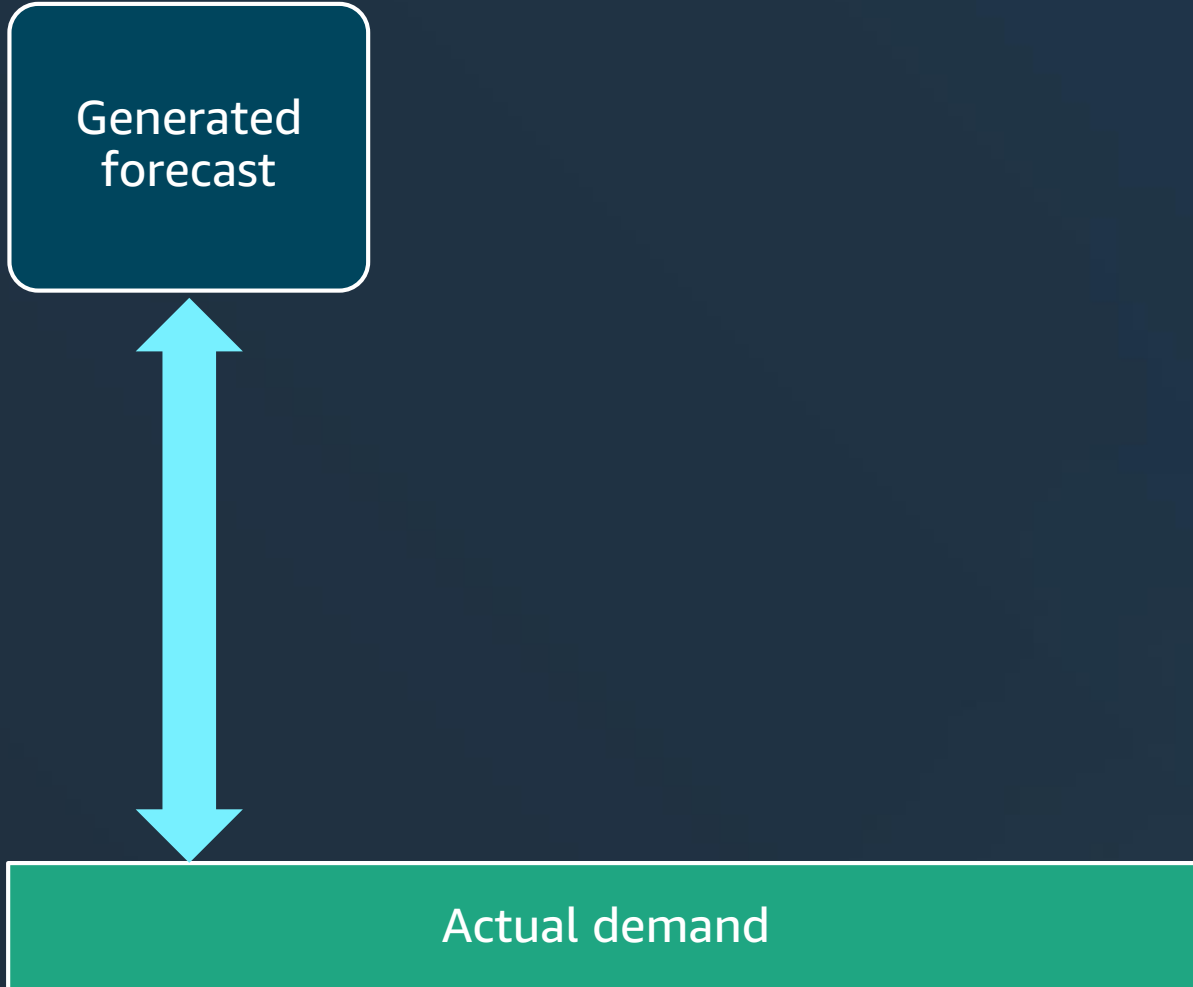


Take note of difference in the sales volumes and variability of items among groups

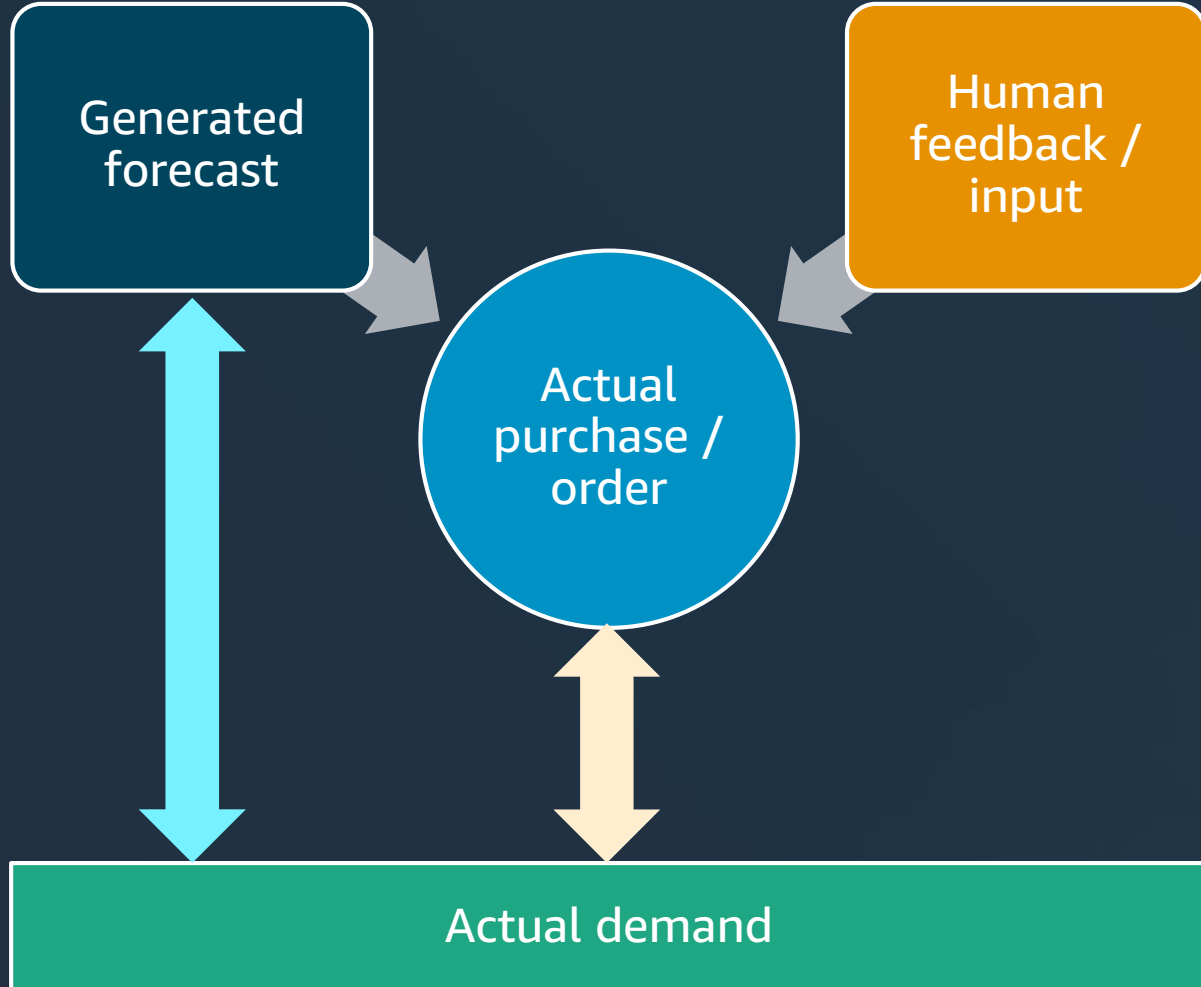


Consider the spread of items in A and B among different category managers

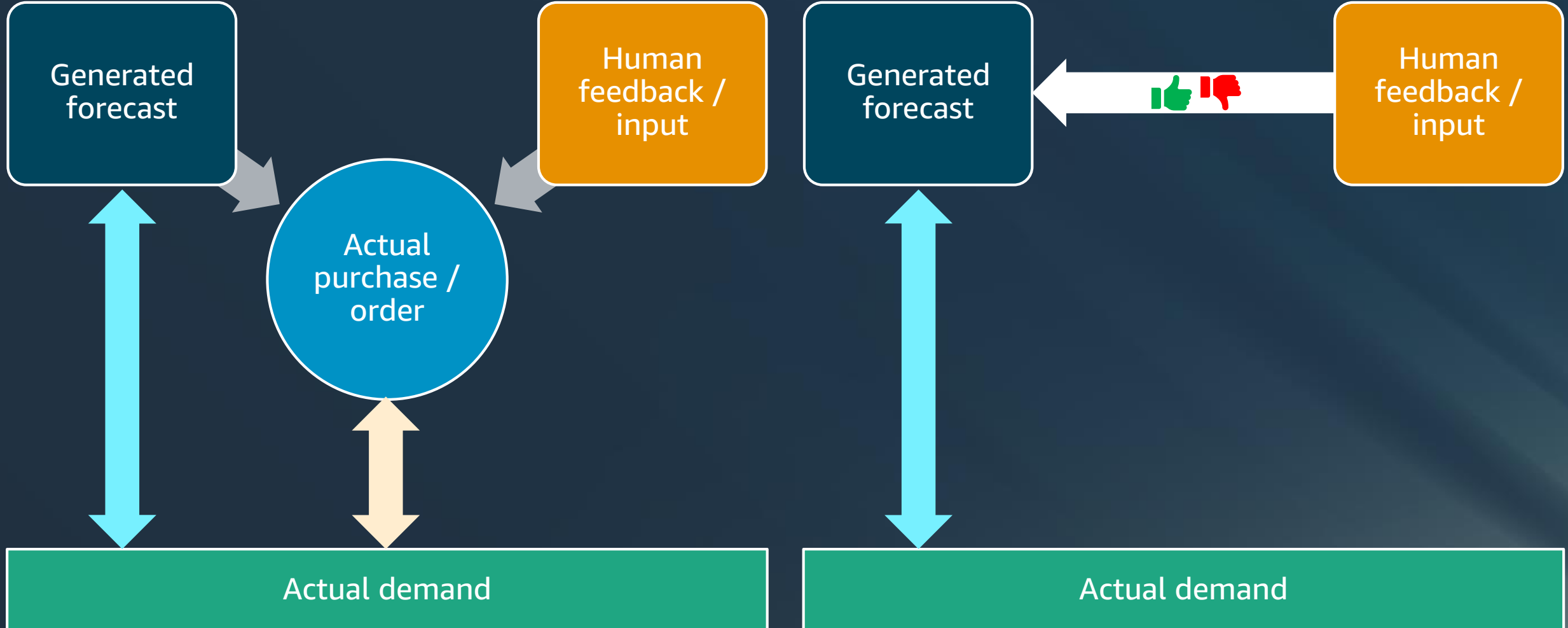
Factoring in human feedback



Factoring in human feedback



Factoring in human feedback



Demo



Architecture components

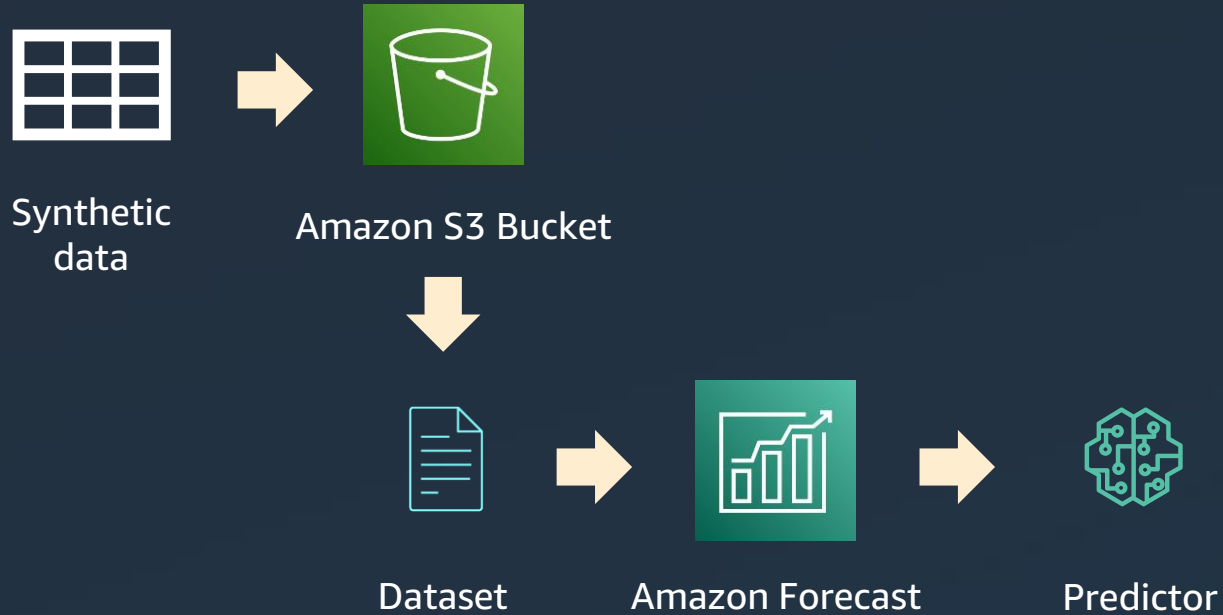


Synthetic
data

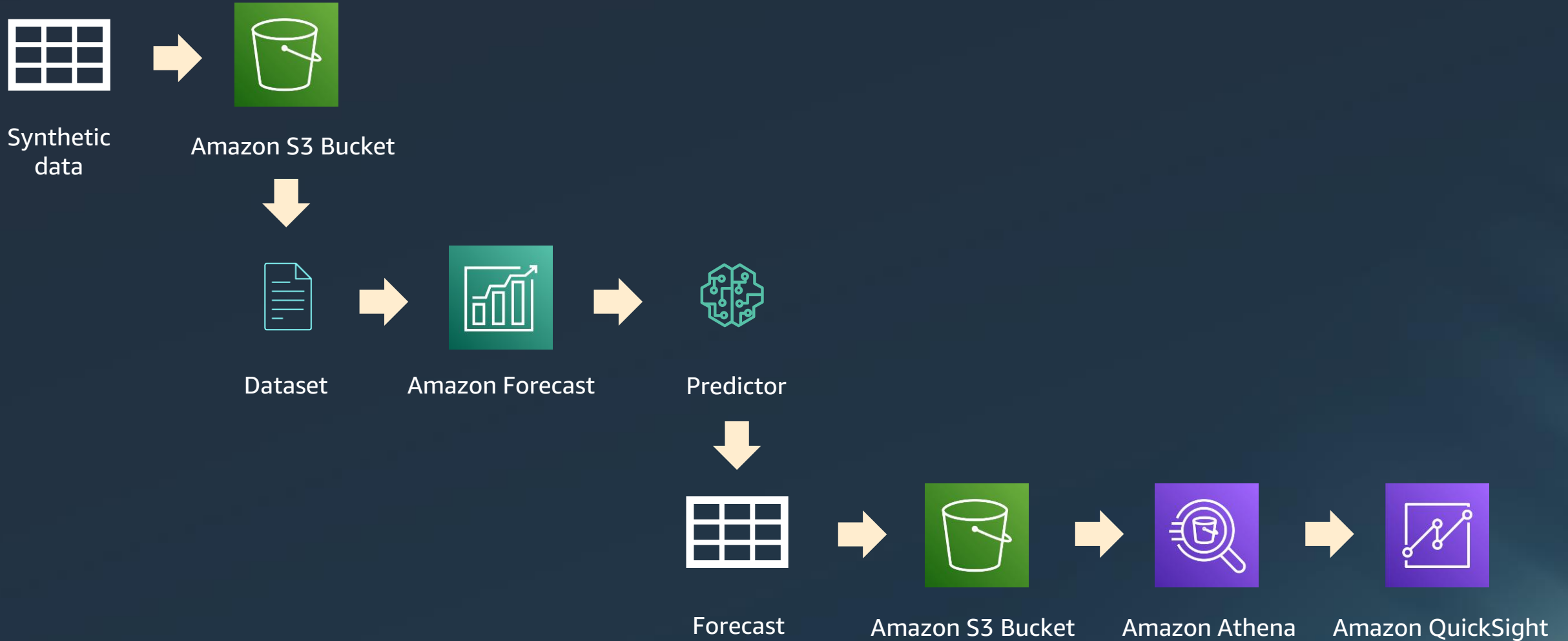


Amazon S3 Bucket

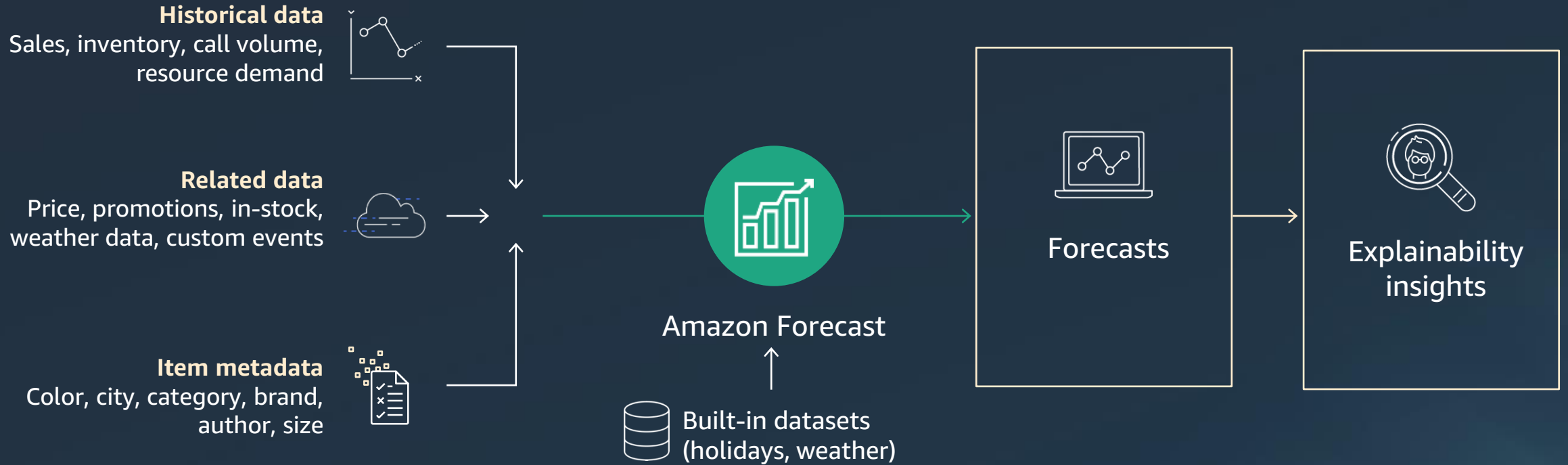
Architecture components



Architecture components



Amazon Forecast – key inputs and outputs



Target time-series (TTS)

REQUIRED FILE THAT CONTAINS WHAT YOU ARE TRYING TO PREDICT

up to 10
dimensions

referred to as
y-value
or target value

location_id	item_id	target_value	timestamp
A55	1885B	177	2016-11-08
A55	1993B	270	2016-11-08
A55	2539B	189	2016-11-08
A55	2139B	54	2016-11-08
A55	2631B	40	2016-11-08

Related time-series (RTS)

OPTIONAL FILE THAT HELPS EXPLAIN AND INFORM THE PREDICTIONS

up to 10
dimensions

numeric features (x-value multi-variates)

location_id	item_id	checkout_price	base_price	emailer for promotion	homepage featured	timestamp
A10	1062	131.95	182.36	0	1	2019-01-29
A10	1062	157.14	157.14	0	0	2018-10-16
A10	1062	158.17	156.17	0	0	2018-10-30
A10	1062	159.08	182.39	0	0	2017-12-05
A10	1062	159.08	183.33	0	0	2017-03-21

Item metadata (ITEM)

OPTIONAL FILE PROVIDING CATEGORICAL FEATURES

no dimensions
only key

up to 10 categorical (string)

item_id	food_category	food_cuisine
2826	Sandwich	Italian
2664	Salad	Italian
2569	Salad	Italian
1230	Beverages	Continental
1207	Beverages	Continental

Next steps



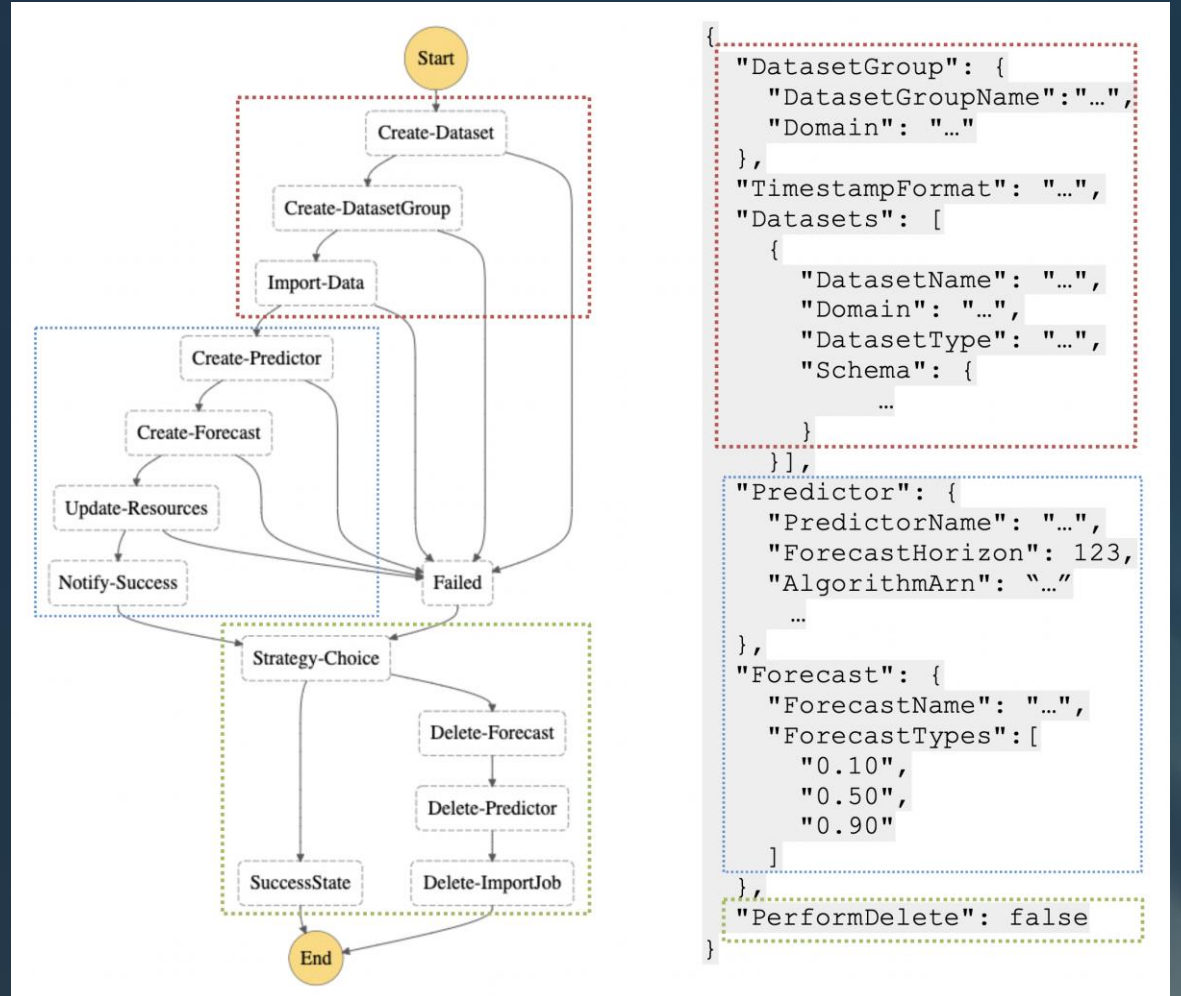
Next steps



Automated
retraining



Create
managed
workflows
using
AWS Step
Functions



Next steps



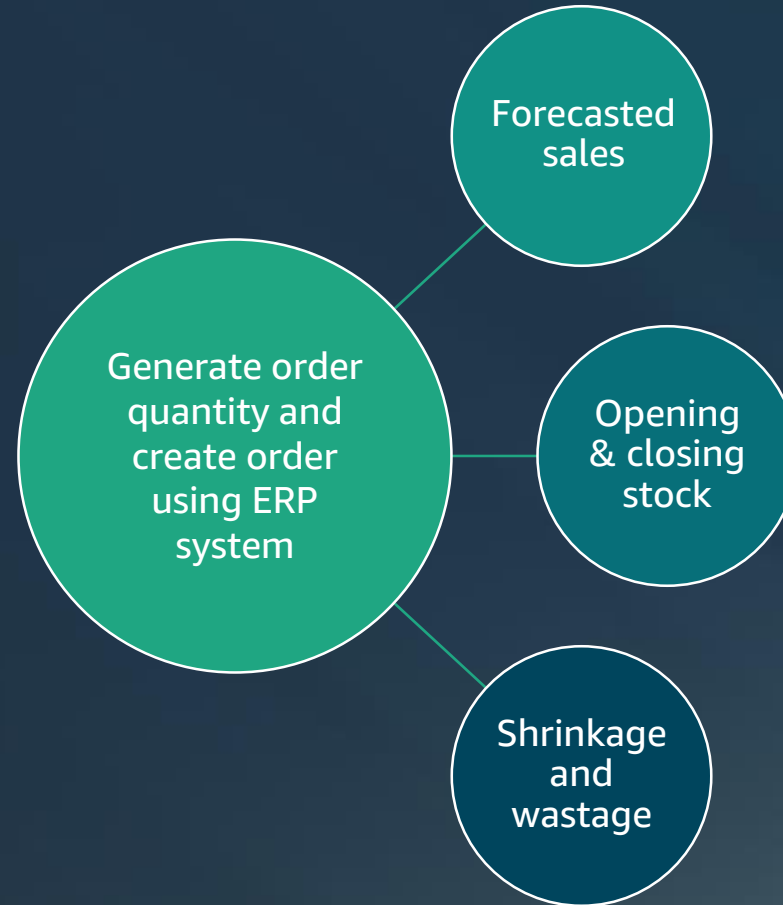
Automated
retraining



Integrated
replenishment



Use jobs to
make
forecasts &
integrate
with ERP
system



Next steps



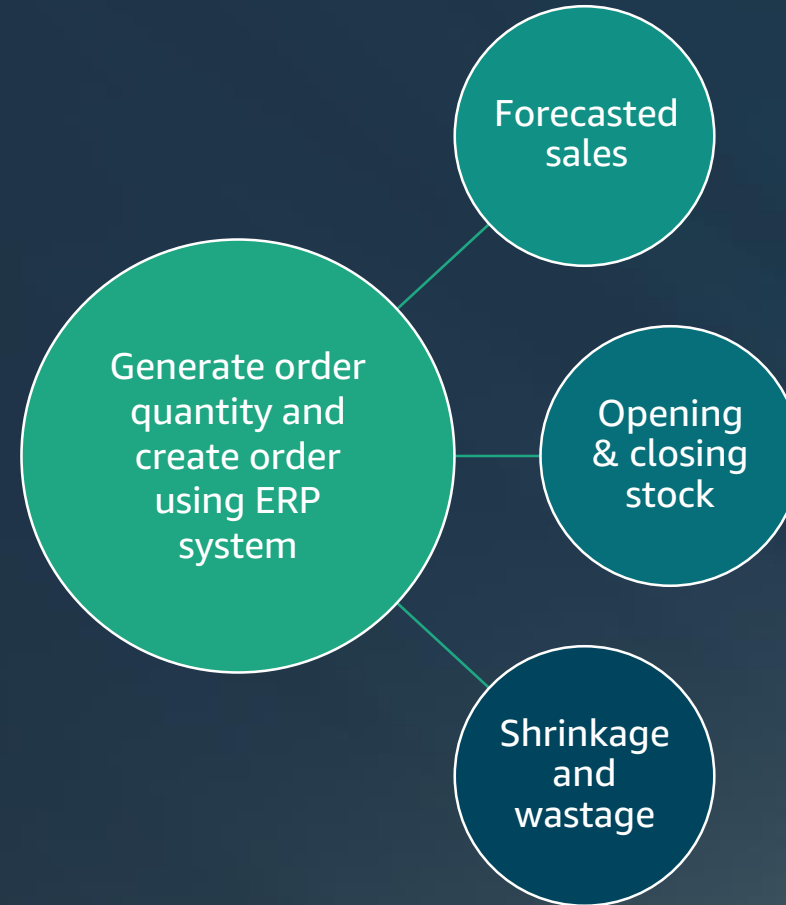
Automated
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Use jobs to
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Select
quantiles
based on
stockouts
and
wastage

Next steps



Automated
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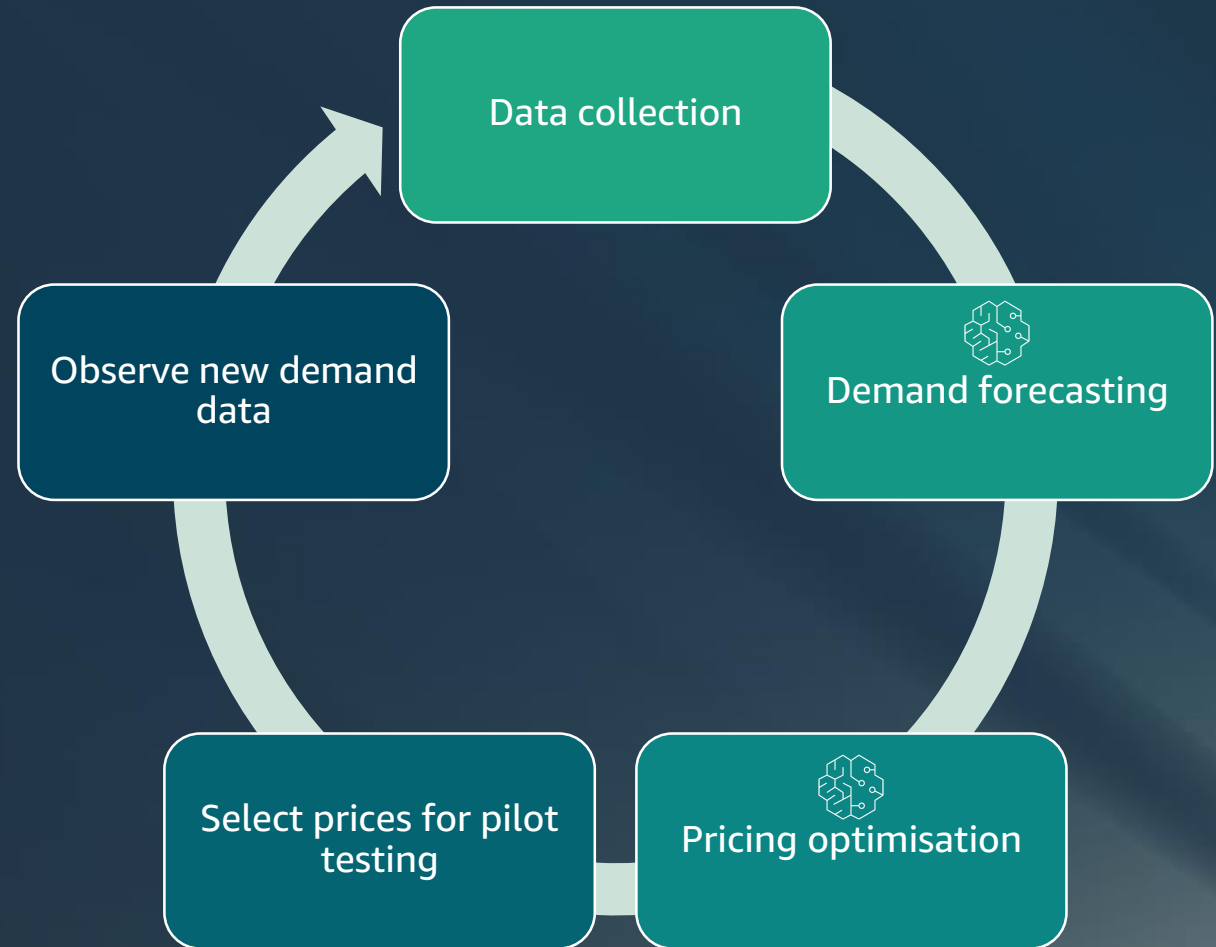
Integrated
replenishment



Dynamic pricing



Optimise
using
iterative
workflows
and tests



Recap

- Using machine learning based forecasting has brought about many tangible benefits to AWS customers in use cases such as inventory planning
- Amazon SageMaker Canvas and Amazon Forecast allow customers to perform forecasting without having a background in machine learning
- Scoping a successful PoC involves measuring business metrics such as the cost of stockouts and/or excess inventory

Additional resources



[Measuring demand forecast benefits repository](#)

Learn how to translate forecasts to quantifiable business metrics



[Amazon Forecast MLOps Github repository](#)

An easily adjustable template for deploying multiple parallel workloads



[Amazon Forecast AWS solution](#)

One-click deployment using an AWS CloudFormation template



[Implementing Amazon Forecast in the retail industry - blog post](#)

Creating and productionizing a solution for demand forecasting in 8 weeks, saving 16 labor hours monthly



[AWS machine learning blog](#)

From forecasting demand to ordering – An automated machine learning approach with Amazon Forecast to decrease stockouts, excess inventory, and costs



[Bosch blog post](#)

Large-scale revenue forecasting at Bosch with Amazon Forecast and Amazon SageMaker custom models

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