



# Time Series Forecasting

*project by:*



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# Summary

Goal

*Forecast the next quarter of  
US Cocoa Bean Imports*



# Models

	Versions:	Tuned Models:
Triple Exponential Smoothing	Raw Data vs. Log Transformed	Multiplicative* = 2
SARIMA	Raw Data vs. Log Transformed	Manual, Short vs. Full Grid Search = 6
Prophet	Raw Data vs. Log Transformed	Grid Search* = 2

\* The "untuned" versions of these models (e.g., "additive") for TES shown on upcoming slides but excluded from final evaluation

# Results

**Triple Exponential Smoothing** model on raw data performed best with lowest MAE & (tied for) lowest MAPE:

Model	Log Data	Tuning	MAE	MAPE	MAE_H1	MAE_H2	MAE_H3	MAPE_H1	MAPE_H2	MAPE_H3
ExponentialSmoothing	N	Manual	9,559,131	27%	8,232,452	9,662,805	10,782,135	24%	29%	30%
ExponentialSmoothing	Y	Manual	10,529,673	29%	8,454,657	10,832,715	12,301,647	24%	30%	33%
Sarima	N	Short Grid Search	17,625,899	67%	11,563,515	18,363,968	22,950,214	40%	72%	88%
Sarima	N	Grid Search	9,843,491	33%	8,015,458	10,640,699	10,874,315	29%	37%	33%
Sarima	N	Manual	22,289,323	92%	16,481,080	23,139,193	27,247,696	72%	99%	105%
Sarima	Y	Short Grid Search	9,876,090	27%	8,183,549	9,969,859	11,474,863	22%	27%	31%
Sarima	Y	Grid Search	9,886,487	29%	7,851,162	10,110,756	11,697,541	23%	31%	32%
Sarima	Y	Manual	10,271,880	29%	8,358,623	10,539,407	11,917,611	24%	30%	33%
Prophet	N	Grid Search	11,510,718	46%	12,449,653	11,544,815	10,537,687	54%	47%	37%
Prophet	Y	Grid Search	11,333,605	44%	13,120,463	11,264,117	9,616,237	59%	44%	29%

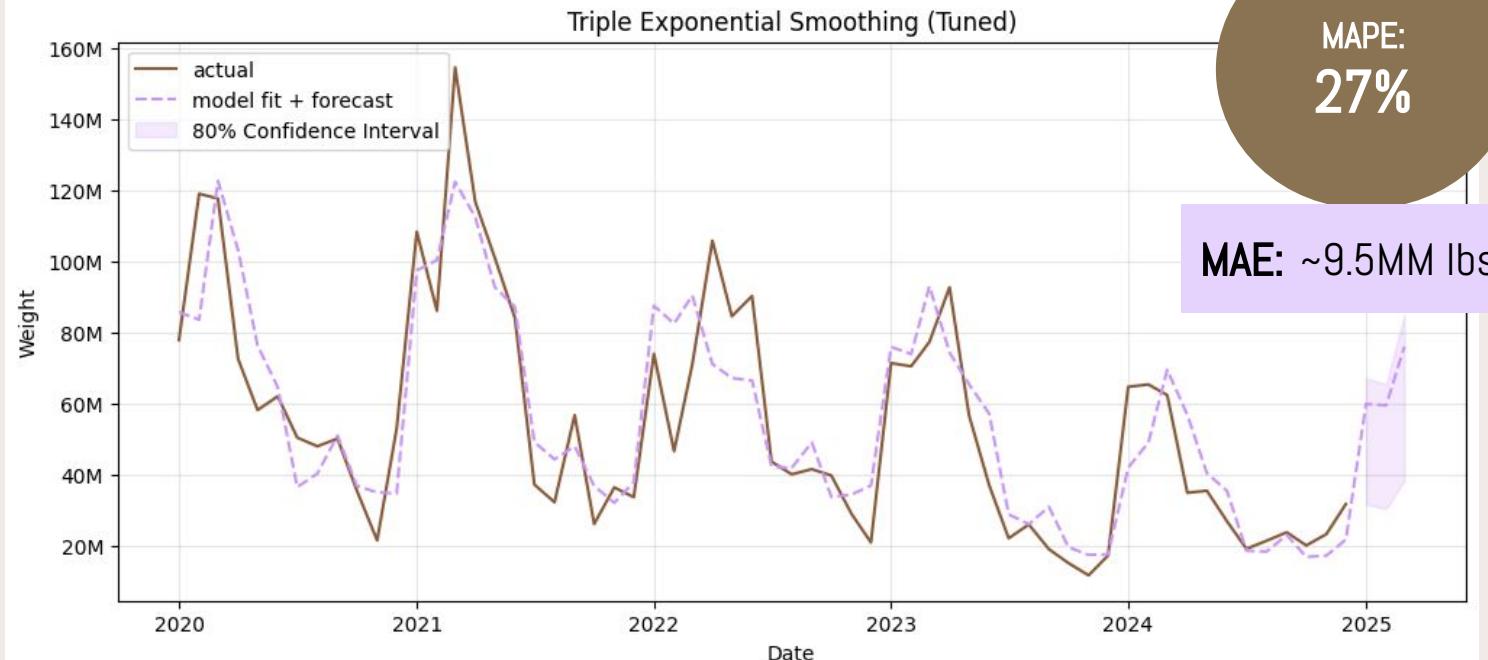
# Results

## Forecast

Date	Forecast	Lower_CI	Upper_CI
2025-01-01	59,915,879	31,650,287	67,112,957
2025-02-01	59,477,191	30,411,927	65,506,481
2025-03-01	75,993,848	38,341,516	84,996,332

```
triple_tuned = ExponentialSmoothing(  
    data['Weight'],  
    trend="additive",  
    damped_trend=True,  
    seasonal="multiplicative",  
    seasonal_periods=12  
).fit(optimized=True)
```

## Model



# MAPE Levels

- ❖ <20% = excellent
- ❖ **20-30% = good**
- ❖ <40% = acceptable for monthly forecasting

Good for seasonal / volatile, high volume, monthly data – a difficult combination to forecast.

It may benefit from additions/further fine-tuning.

*Note, MAPE is higher than it would be with non-seasonal data where it can be lower (up to <5%)*



*notebook link:*



**GitHub** Pages



# Technical Setup

## Data

**Source:** Census US imports of Cocoa Beans 2020-2024

**Type:** Structured

**Observations:** 60 (months)

**Features:** Month/Year

**Target:** Weight & Spend

*Note, N is higher originally as the grain was by Country, but Country was not used in this forecast*

## Setup

**Language:** Python

**Packages:** prophet, seaborn, matplotlib, numpy, pandas

**Compute:** Python3 CPU in Google Colab

## Evaluation Metrics

- Mean Absolute Error (MAE) & Mean Absolute Percentage Error (MAPE) on forecast



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