



Zero-shot Time Series Forecasting in Financial Data Analysis: Prospects and Challenges

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Data Characteristics and Problem Statements

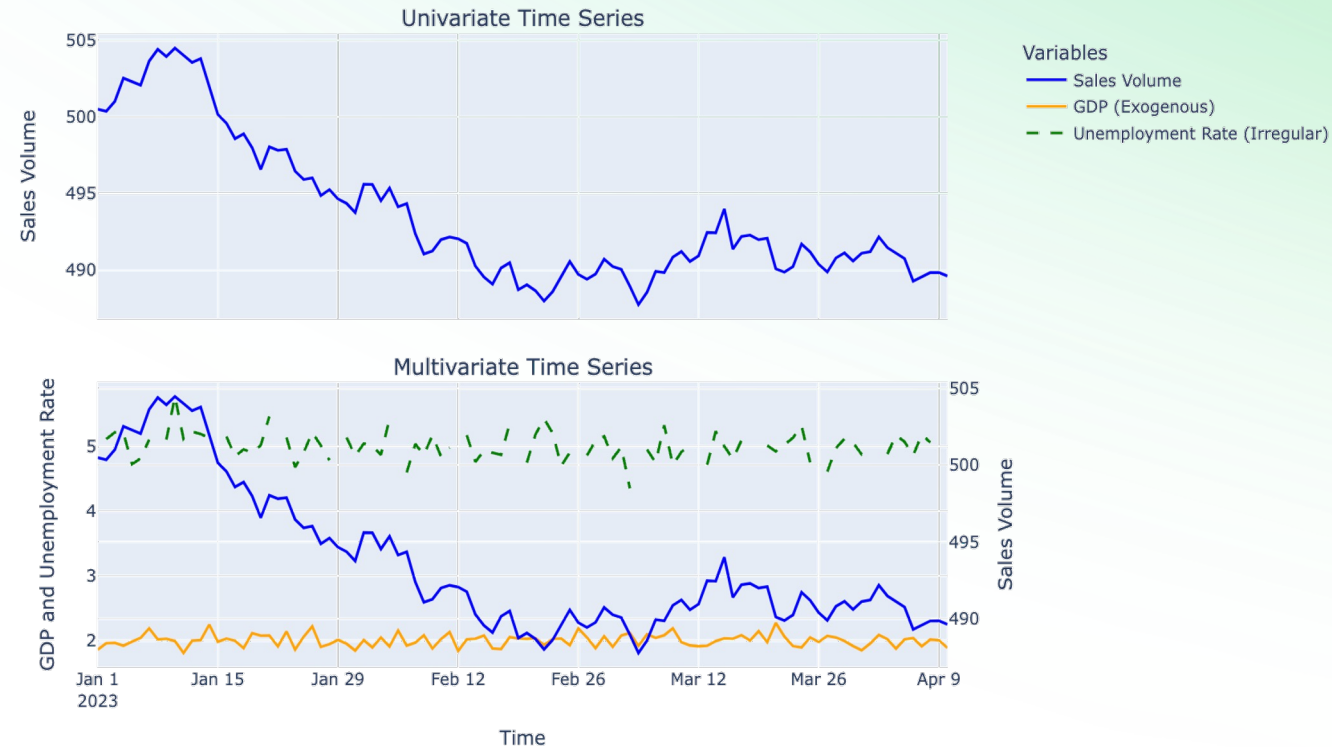
Forecasting instruments

Zero-shot models

Time Series Forecasting

Let's define the terminology

- **A time series** is a sequence of values ordered by time.
- **A multivariate time series** is a structure where multiple individual time series are considered simultaneously.
- **A regular time series** is a time series with evenly spaced time intervals between data points.
- **Additional features (exogenous variables, covariates)** are external factors not generated by the system but influencing the target variables.

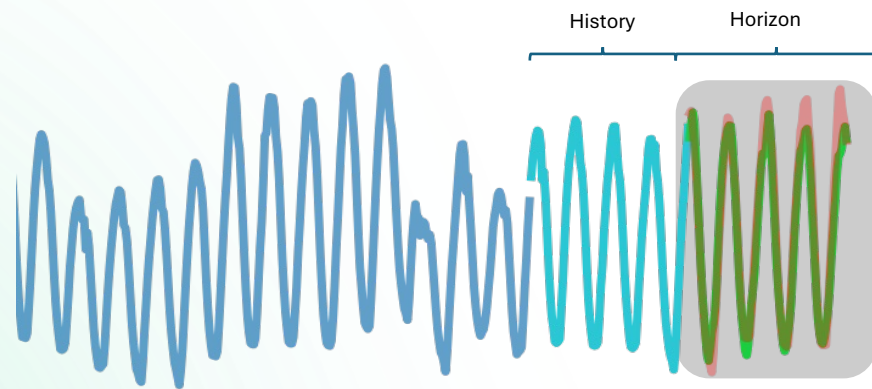


The top subplot represents the **univariate** TS, while the bottom subplot — the **multivariate** TS. Both `Sales Volume` and `GDP` are **regular**, whereas the `Unemployment Rate` is **irregular**. `GDP` and `Unemployment Rate` are **exogenous features** relative to `Sales Volume`.

Time Series Forecasting

Let's define the terminology

- **Forecasting Task:** given a history, the goal is to forecast over a horizon.



The history is not all the available data points, but rather the context length currently considered for generating predictions.

- **Forecasting Metrics** (between the prediction and the actual series):

- Mean Squared Error (MSE)

$$MSE = \frac{1}{n} \sum_{t=1}^n (y_t - \hat{y}_t)^2$$

- Mean Absolute Error (MAE)

$$MAE = \frac{1}{n} \sum_{t=1}^n |y_t - \hat{y}_t|$$

- Mean Absolute Percentage Error (MAPE)

$$MAPE = 100 \times \frac{1}{n} \sum_{t=1}^n \left| \frac{y_t - \hat{y}_t}{y_t} \right|$$

- Etc.

Characteristics of Financial Data and Tasks

which affect how they need to be analyzed



- We are not considering stock market data (**asset prices, market indices**, etc.).



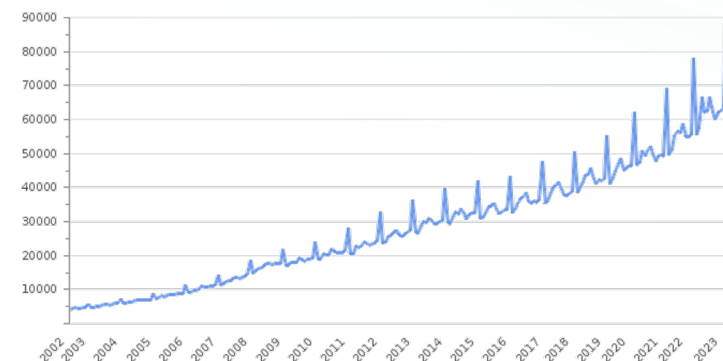
- We are also not considering **irregular** time series.



- We are focusing on macro- and microeconomic indicators, as well as financial data that reflect the activities of agents (such as **data on loans and deposits, labor market indicators**, and so on).



Interest rate in Russia (%)



Average wage in Russia (Rub / Month)

Characteristics of Financial Data and Tasks

which affect how they need to be analyzed

What properties of the data do we face?

- ① Short time series
- ② Data Instabilities and External Disruptions
- ③ Numerous exogenous variables and contextual information

What do we expect from the model?

- ① The model should work efficiently under data scarcity.
- ② The model should be flexible and adapt to changing macro-level distributions.
- ③ The model should be able to select relevant factors and account for their dynamics.



Data Characteristics and Problem Statements

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How to generate a forecast?

Groups of methods commonly used in forecasting tasks



Naive methods

- (Seasonal) Naive
- Mean, Median



Statistical methods

- ETS
- Theta
- ARIMA



ML methods

- ElasticNet
- GBMs



DL methods

- DLinear
- NBEATS
- PatchTST
- GPT2

How these methods meet our requirements?

Methods Group	Data scarcity	Flexibility and adaptability	Exogenous variables
Naive methods	+	-	-
Statistical methods	+	-	+ -
ML methods	+ -	+ -	+
DL methods	-	+	+



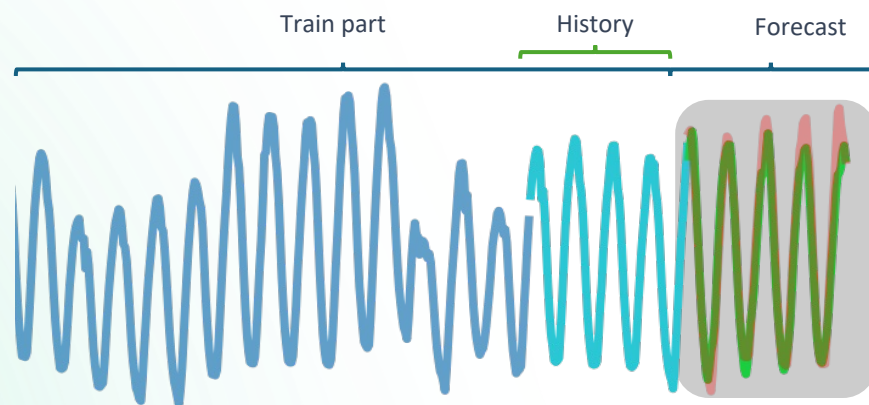
Data Characteristics and Problem Statements

Forecasting instruments

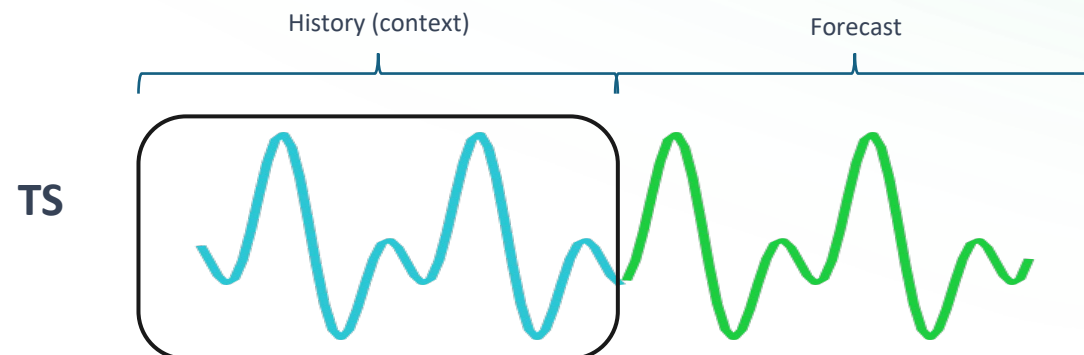
Zero-shot models

Supervised vs Zero-shot models

Supervised: Train the model to predict labels for new data based on patterns identified in the **training data**.



Zero-shot: Train the model to predict labels for new data **without training on the target dataset**, based on patterns identified in the unrelated data.



The variety of Zero-shot models

Zero-shot models for time series are an actively developing area

LLM

Non-adapted LLM

- [LSTPrompt](#)
- [PromptCast](#)
- [LLMTime](#)

Adapted LLM

- [Time-LLM](#)
- [FPT](#)
- [Chronos](#)
- [UniTS](#)
- [DAM](#)

Specialized

Using synthetic data

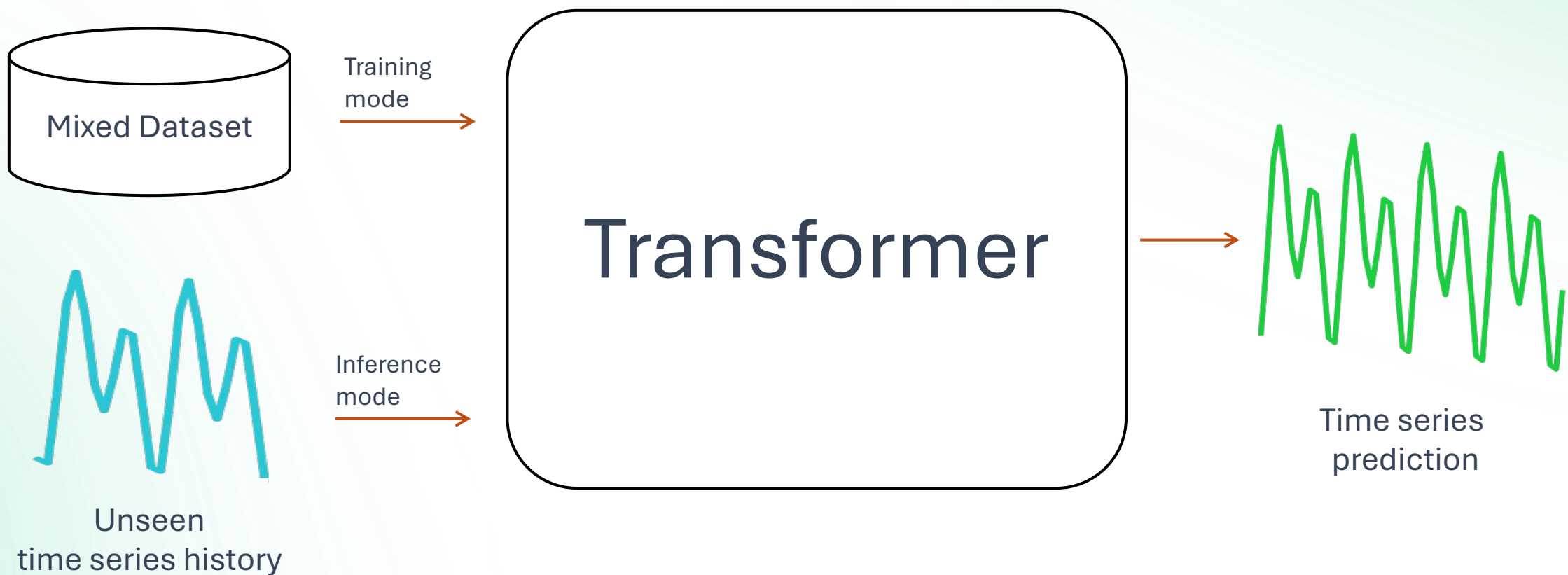
- [ForecastPFN](#)

Using real data

- [GPHT](#)
- [MOIRAI](#)
- [Moment](#)

Zero-shot models

- Zero-shot models for time series are mainly Transformers
- They require a large and diverse dataset for pretraining



Zero-shot for business case

We can adapt the training data properties to a specific task.

- In this case, we are forecasting **macroeconomic** time series
- Dozens to hundreds of **short** time series, which are mostly **independent**.

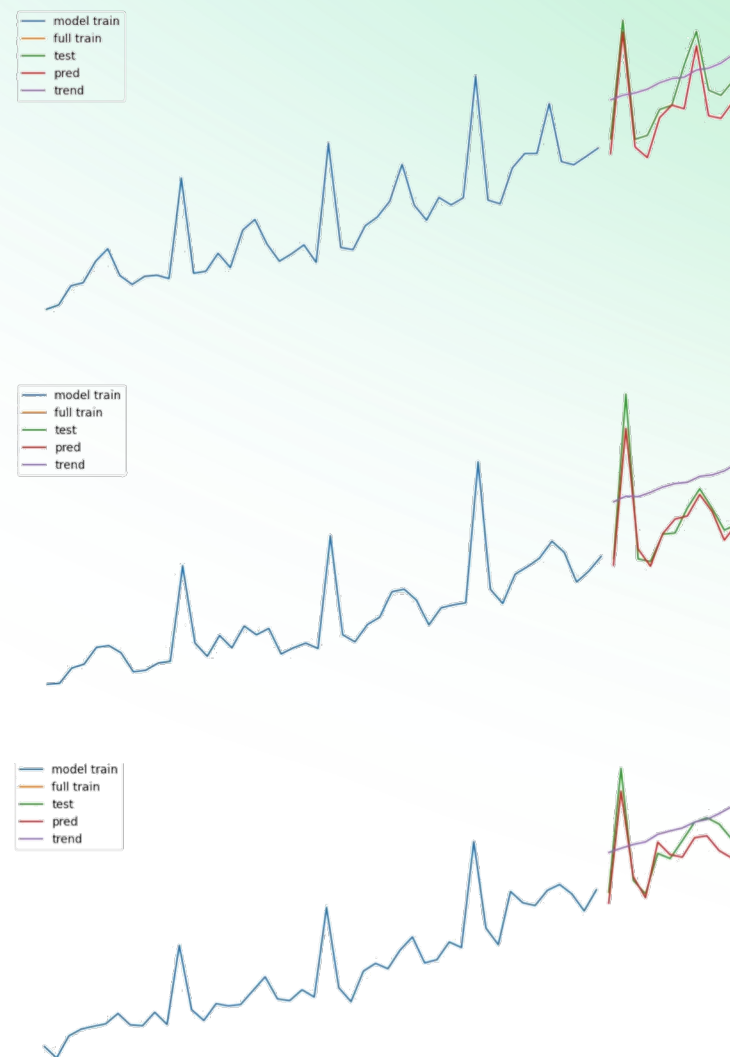
MAPE:

- Zero-shot – x%
- GBDT – **↑0.7%**
- Prophet x~100 – **↓0.2%**

Time (train + inference):

- Zero-shot – **< 1sec**
- GBDT – **30 sec**
- Prophet x~100 – **5 min**

Great quality with a speed advantage!



Predictions of a Zero-shot model trained not only to forecast the continuation of the time series but also to decompose it into trend and seasonality.

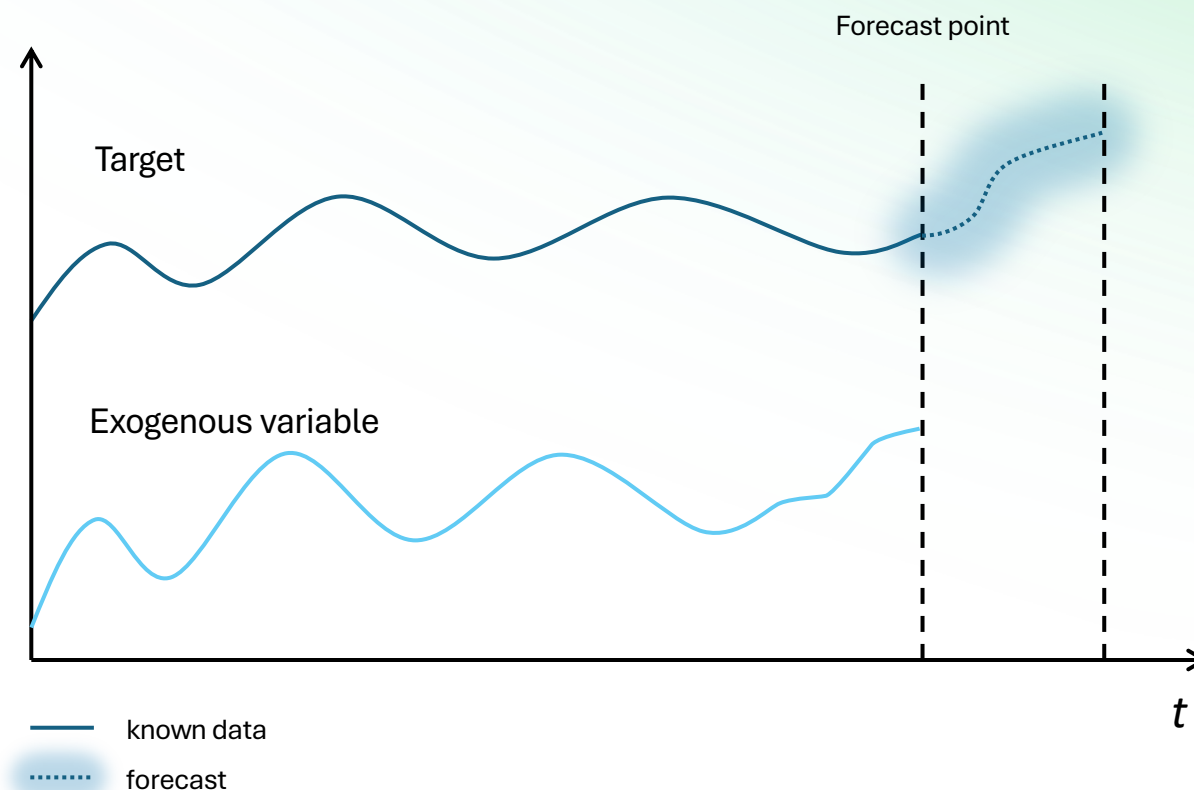
Zero-shot with exogenous data

The functionality of Zero-shot model can be extended to handle additional information.

Zero-shot with additional features is an unexplored scientific topic.

Possible solution:

- There are additional time series. The task is to transform these series into a forecast.
- Use TabPFN (Tabular Prior-Data Fitted Network).
- There is no open-source TabPFN for regression tasks yet, but there is a paper from the NeurIPS 2024 Workshop*.



* Hoo, S. B., Müller, S., Salinas, D., & Hutter, F. The Tabular Foundation Model TabPFN Outperforms Specialized Time Series Forecasting Models Based on Simple Features. In *NeurIPS 2024 Third Table Representation Learning Workshop*.

How does Zero-shot fit the requirements from data?

Let's return to our table

Methods Group	Data scarcity	Flexibility and adaptability	Exogenous variables
Naive methods	+	-	-
Statistical methods	+	-	+ -
ML methods	+ -	+ -	+
DL methods	-	+	+
Zero-shot	+	+	+

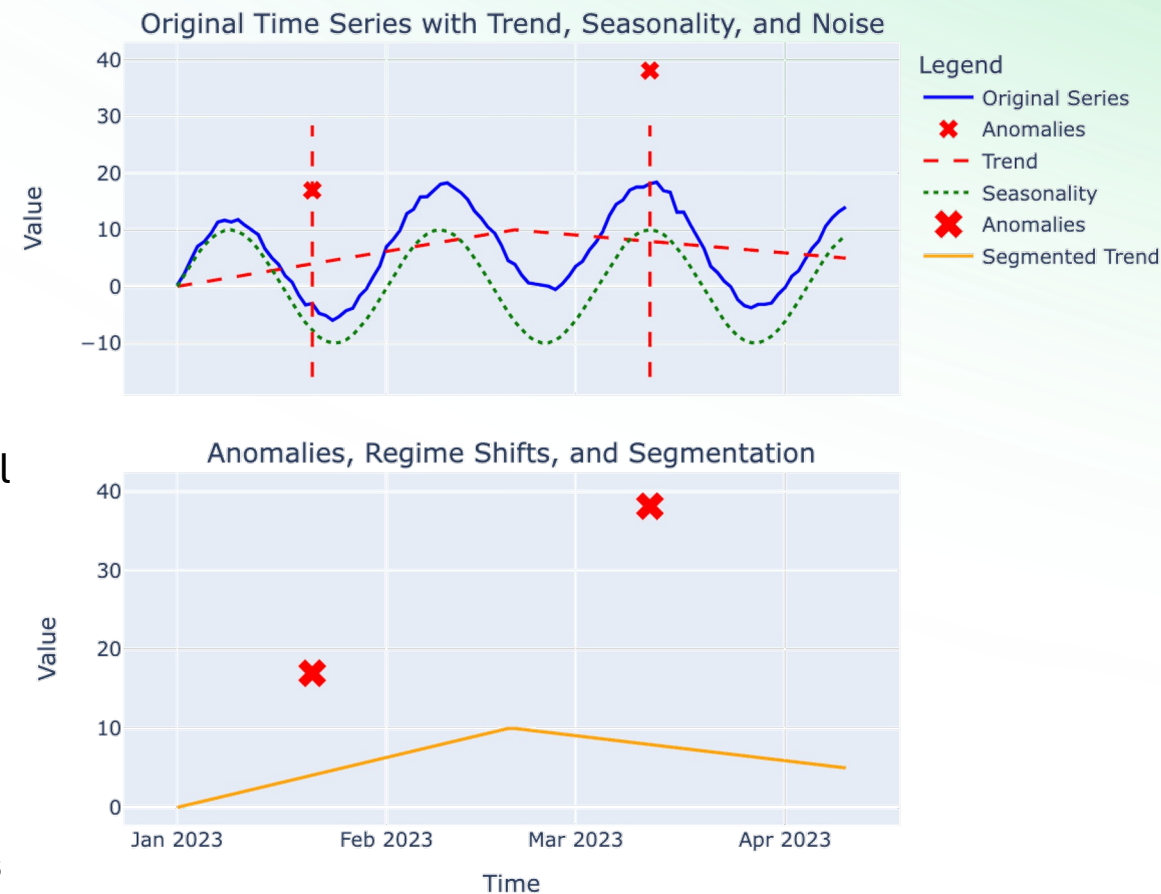
Possible directions for the development and challenges to overcome

1. Development of Multitasking:

- **Decomposition** into trend, seasonality, and noise, with tasks such as:
 - removal of trend, seasonality, or noise
 - **segmentation** based on trend
 - predicting the **trend type**
 - forecasting Fourier coefficients; **seasonality** period and **type**
- Detection of **anomalies** and **regime shifts**
- Generative tasks such as predicting stochastic differential equation (**SDE**) parameters, **interval forecasting**, and **filling in missing values**

2. Challenges:

- Integrating forecasting and regression tasks in one model
- Developing effective methods for generating synthetic data
- Finding efficient approaches to utilize exogenous features



Representation of different tasks: decomposition into trend, seasonality, and noise, along with detection of anomalies and regime shifts.



Thank you for your attention!

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