

- 1 TSFX: A Python package for time series feature
- <sub>2</sub> extraction
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#### **Software**

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**Summary** 

TSFX is a Python (Van Rossum & Drake, 2009) library for extracting features from time series data. It is inspired by the tsfresh (Christ et al., 2018) Python package with a special focus on performance on large time series datasets. To this end, it utilizes Polars (Vink et al., 2023) which is a fast DataFrame library written in Rust (*The Rust Programming Language*, n.d.) with Python bindings facilitated through PyO3 (Project & Contributors, n.d.). The feature extraction functions are implemented in Rust for even faster execution. To benchmark, the "1 billion row challenge" (Morling, n.d.) was used. The benchmark shows that compared to tsfresh, TSFX offers approximately 10 times higher performance, using the same set of time series features.

TSFX can be installed using pip:

pip install tsfx

TSFX can also be configured using a TOML (*TOML*, n.d.) configuration file (default name .tsfx-config.toml).

Below is a simple example of extracting features from a time series dataset:

```
import polars as pl
from tsfx import (
    ExtractionSettings,
    FeatureSetting,
    extract features,
df = pl.DataFrame(
        "id": ["a", "a", "a", "b", "b", "b", "c", "c", "c"],
        "val": [1.0, 2.0, 3.0, 1.0, 2.0, 3.0, 1.0, 2.0, 3.0],
        "value": [4.0, 5.0, 6.0, 6.0, 5.0, 4.0, 4.0, 5.0, 6.0],
    },
).lazy()
settings = ExtractionSettings(
    grouping_cols=["id"],
    feature_setting=FeatureSetting.Efficient,
    value_cols=["val", "value"],
gdf = extract_features(df, settings)
gdf = gdf.sort(by="id")
with pl.Config(set_tbl_width_chars=80):
    print(gdf)
```



19 Running the code above generates a new DataFrame with the extracted features:

shape: (3, 314)

	valsu   m_value   s     f64	:	number_ peaks	number_ peaks	number_	number_
a b c	6.0 6.0 6.0	2.0 2.0 2.0	 0.0		0.0 0.0 0.0	0.0

If the DataFrame has a time column, it is also possible to extract over a time window by passing DynamicGroupBySettings into the feature extraction settings, like so: ExtractionSettings(..., dynamic\_settings=DynamicGroupBySettings(...)).

## Statement of need

Time series is a ubiquitous data modality, present in many domains such as finance, industry, meteorology, and medicine, to mention a few. As hardware to collect and store time series data is becoming increasingly affordable, the amount of available time series data is increasing in many domains. A common preprocessing step when dealing with time series is feature extraction. This involves calculating representative features such as mean, variance, skewness, etc. from the time series to be used in downstream tasks such as classification, regression or clustering. For large time series datasets, performance is important for enabling timely data preprocessing. TSFX is made for this purpose: extracting features from large time series datasets.

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