

Time Series Features and Statistical tests

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Features extraction

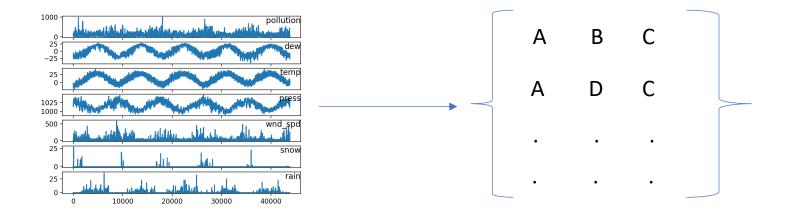
- Features are typically statistical values or associated concepts that summarize a given time series.
- Typically used to: accelerate computations, have a set of common features for series of different lengths, etc.
- Examples: mean, maximum, minimum, median, kurtosis, number of points above the mean, symmetry looking, etc.
- Existing libraries: Khiva, tsfresh, etc.





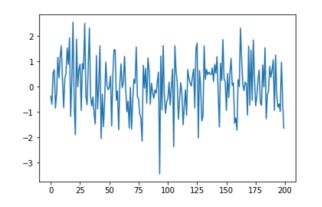
What can we use features for?

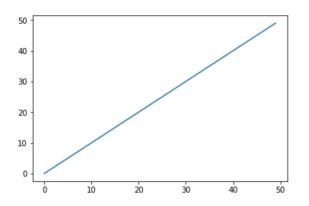
 To apply classical machine learning algorithms mainly for clustering and prediction/forecasting

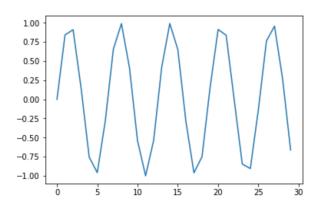




Features. Example





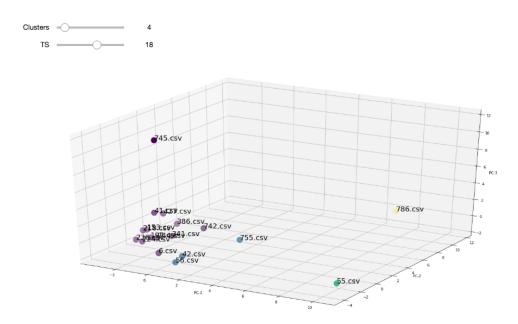


	mean	kurtosis	skewness	symmetry_looking
0	0.039324	0.106709	-0.113297	True
1	24.499994	-1.200000	0.000001	True
2	0.042269	-1.509307	-0.082693	False



Features extraction. Use cases

- Based on the features matrix, classic machine learning techniques are able to clusterize or classify the time series.
- If the feature vector is too long, PCA or equivalent algorithms might be needed.



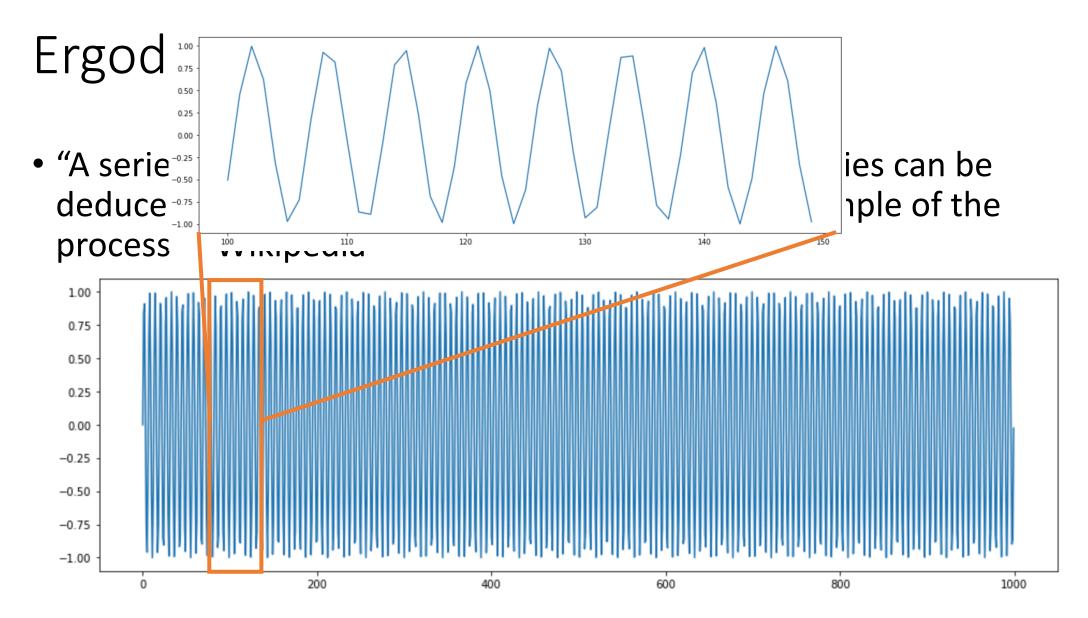


Statistical tests

• Statistical tests check statistical properties of the series

TIME SERIES PROPERTY	STATISTICAL TEST	
Regular/Irregular	Time axis check.	
Trend/Non-Trend	Pearson correlation trend test and its p-value.	
	Daniels trend and its p-value.	
	Mann-Kendall trend test and its p-value.	
Ergodic/Non-Ergodic	Means of autocorrelation function evaluation. Augmented Dickey Fuller test.	
Periodic/Non-Periodic	Fisher's test for periodicity.	
Seasonal/Non-Seasonal	Chi-Square Goodness-of-fit test.	
	Kolmorov Simrnov type statistic.	
	Harmonic analyses based on the Edward type statistics.	







What can we use statistical tests for?

- To choose the most suitable algorithm in order to satisfy a determined problem
- To determine if an algorithm is not recommended to analyse a time series



Statistical tests. Use Cases.

- Discriminate non applicable time series algorithms
- More suitable selection of algorithms: filtering, distances, forecasting, clustering, etc.
- Examples:

ALGORITHM	RECOMMENDATION	
ARIMA - FORECASTING	Most suitable for stationary time series.	
SIMPLE EXPONENTIAL SMOOTHING - FORECASTING	Most suitable for time series which have no trend and do not contain a seasonal component.	
SELF-ORGANIZING MAPS - CLUSTERING	Most suitable for periodic time series	
ROBUST ANOMALY DETECTION – ANOMALY DETECTION	Most suitable for stationary time series.	



Statistical tests. Use Cases.

ANALISYS	REGULAR	PERIODIC	
METHOD 1	YES/NO	YES/NO	
METHOD 2			
METHOD 3			
METHOD 4			



Exercise

- <u>Dataset:</u> 100 energy consumption time series of commercial sites during 2012. The series are tagged by subindustry.
- <u>Target:</u> Apply classical machine learning algorithms in order to predict the subindustry of the time series.
- Work to do: Clustering of time series using khiva's time series features algorithms.

