



# Visual Analysis Approaches to Time Series Prediction

Visual Analytics – Interaktive Visualisierung sehr großer Datenmengen – Seminar SS 2018

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## 1. Introduction

## 2. Abstract Time Series

- An Early Approach
- A Trendy Approach
- A Popular Approach
- A Selective Approach
- A Specialized Approach

## 3. Spatial Time Series

- Predicting and Detecting Hotspots
- Mapping between Time and Space





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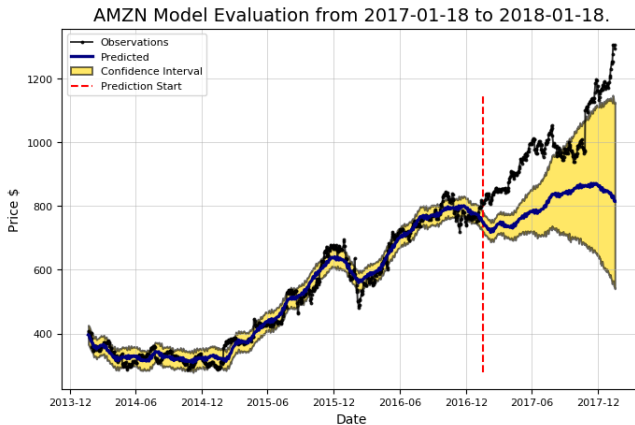


Figure 1: Amazon stock price prediction

[<https://towardsdatascience.com/stock-prediction-in-python-b66555171a2>]



## Abstract Time Series:

- ▶ What is the overall global trend?
- ▶ Do I have cyclic patterns?
- ▶ What are important periods of time?
- ▶ What is the best model?

## Spatial Time Series:

- ▶ What are regions with unusually high occurrences of events?
- ▶ How are these regions developing?
- ▶ Where are new hotspots occurring?





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# An Early Approach



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Figure 2: Workplace environment  
[Ichikawa et al., 2002]

- ▶ Goal: Trend detection, correlation detection
- ▶ Compare multiple variables and different time series
- ▶ External simulations

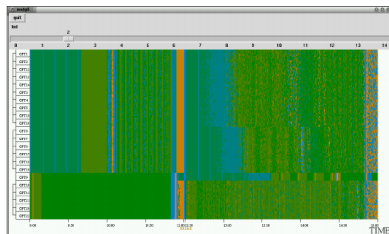


Figure 3: Color band display  
[Ichikawa et al., 2002]



# A Trendy Approach

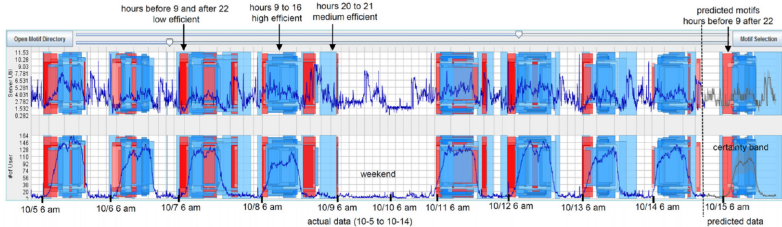


Figure 4: Peak preserving prediction [Hao et al., 2012]

- ▶ Goal: Trend detection
- ▶ Ensures Peak preservation





# A Popular Approach

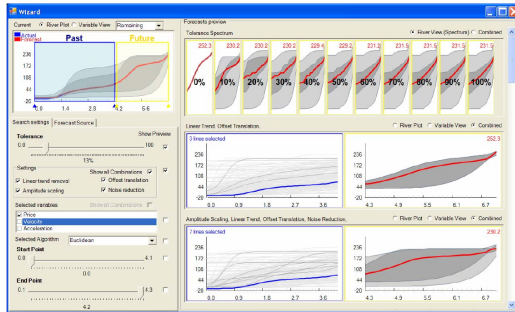


Figure 5: TimeSearcher3 simultaneous preview interface [Buono et al., 2007]

- ▶ Goal: Model selection
- ▶ Similarity based model and prediction
- ▶ Compare different parameters and subsets of data



# A Selective Approach



Figure 6: TiMoVA User Interface [Bögel et al., 2013; Bögel et al., 2014]

- ▶ Goal: Model selection
- ▶ Follows Box-Jenkins-Method

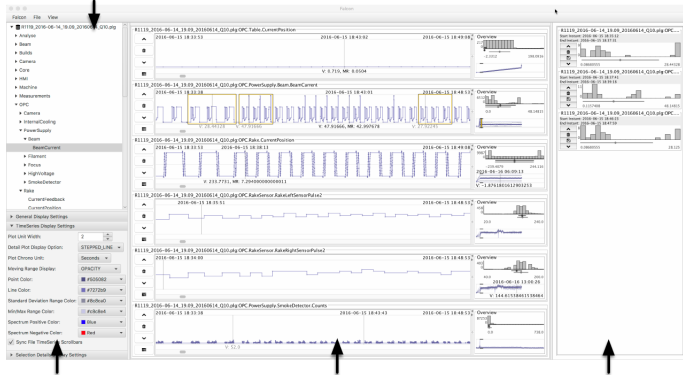


# A Specialized Approach



## Main Analysis Window

### File / Variables Tree View



Settings Panel

Variable Visualization Panel  
(Left: detailed time series, Right: overview)

Selection Details Panel

Figure 7: Falcon main window visualization [Steed et al., 2017]



# A Specialized Approach

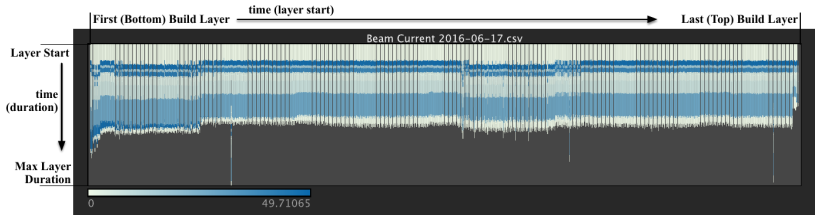


Figure 8: Falcon waterfall visualization [Steed et al., 2017]

- ▶ Goal: Correlation detection
- ▶ Supports high dimensional time series
- ▶ Application areas: predictive maintenance, quality assurance



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# Predicting and Detecting Hotspots

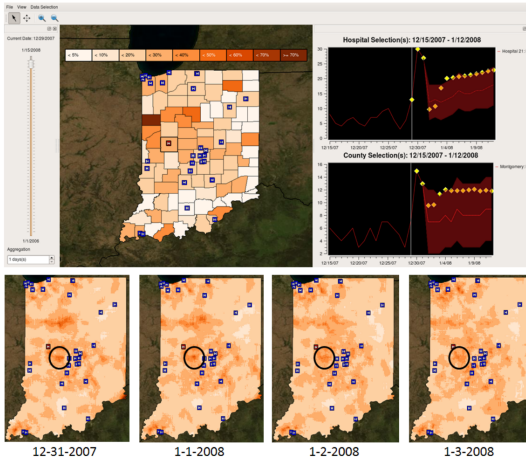


Figure 9: Forecasting Hotspots [Maciejewski et al., 2011]

- ▶ Modeling spatial approximation of time series prediction
- ▶ Combined visualization of time series prediction and spatial information
- ▶ Main focus: Hotspot detection and prediction



# Mapping between Time and Space

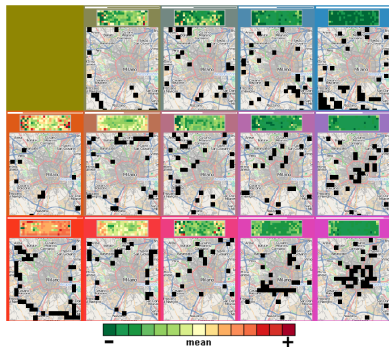


Figure 10: Time-in-space matrix  
[Andrienko et al., 2010]

- Clustering on spatial or temporal level and linkage to other dimension

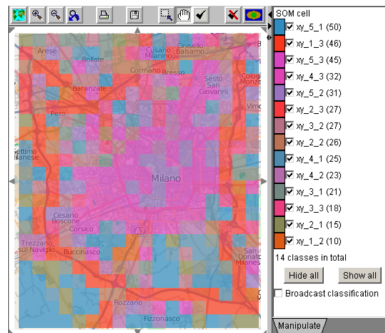


Figure 11: Spatial mapping of  
Time-in-space matrix [Andrienko et  
al., 2010]



## Three different goals:

- ▶ Trend detection
- ▶ Model selection
- ▶ Correlation detection

## Problems and open questions:

- ▶ Turning points, seasonality and outliers
- ▶ Applications are often specifically designed for one task
- ▶ Freedom vs. usability
- ▶ Multiple predictions and uncertainty
- ▶ How to deal with large amounts of predictions?
- ▶ What about sparse data?
- ▶ Preserve or remove peaks?







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Thank you for listening



**Questions?**