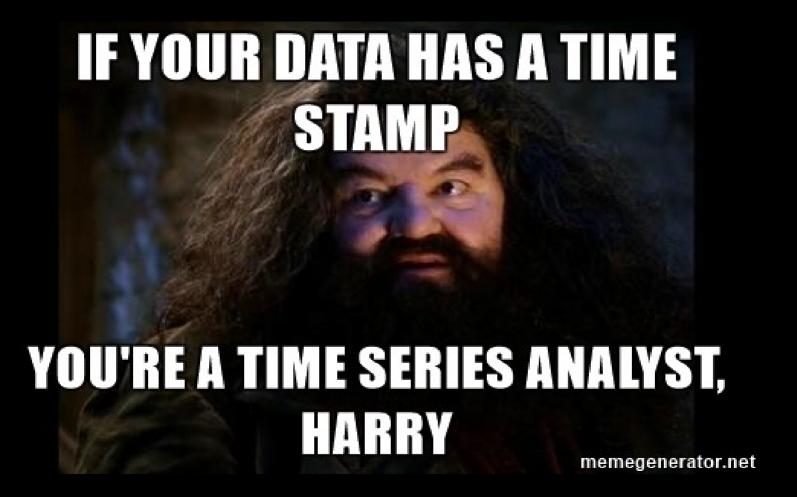
Mining Time Series

Mining Massive Datasets

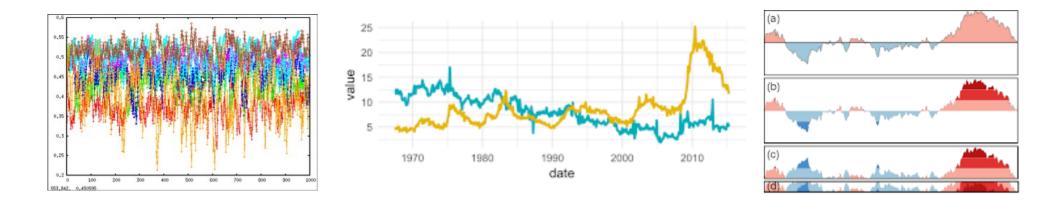
Prof. Carlos Castillo — https://chato.cl/teach



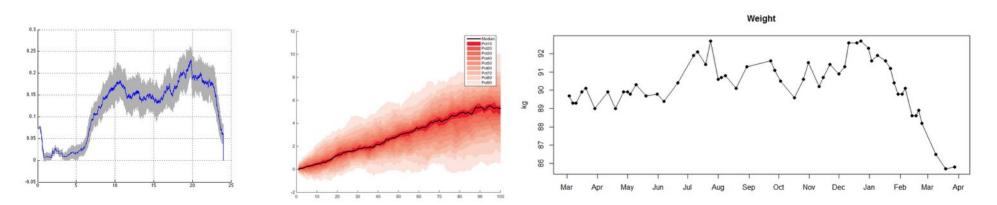


Sources

- Data Mining, The Textbook (2015) by Charu Aggarwal (chapter 14)
- Introduction to Time Series Mining (2006) tutorial by Keogh Eamonn [alt. link]
- Time Series Data Mining (2006) slides by Hung Son Nguyen

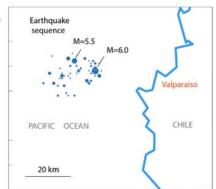


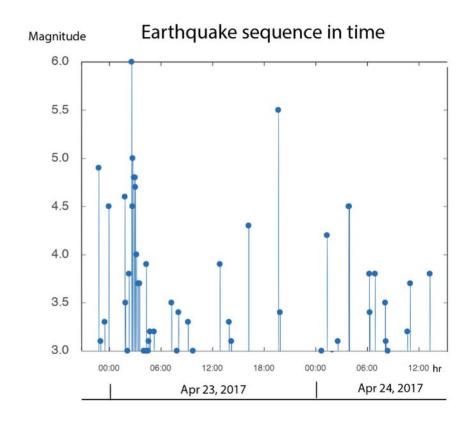
Why do we mine time series? Examples



Seismic data

- Observations = earthquakes
- Goal: characterize when peeks Earthquake sequence

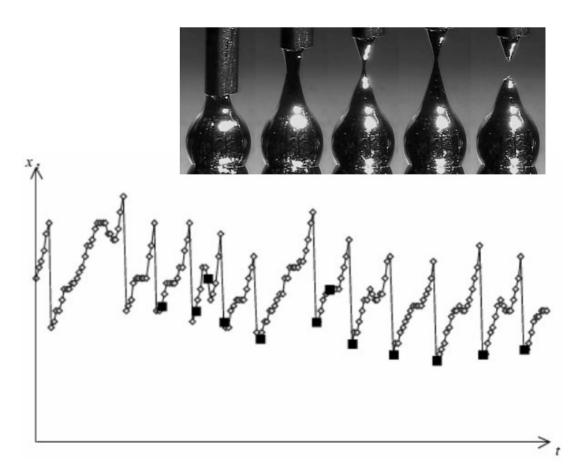




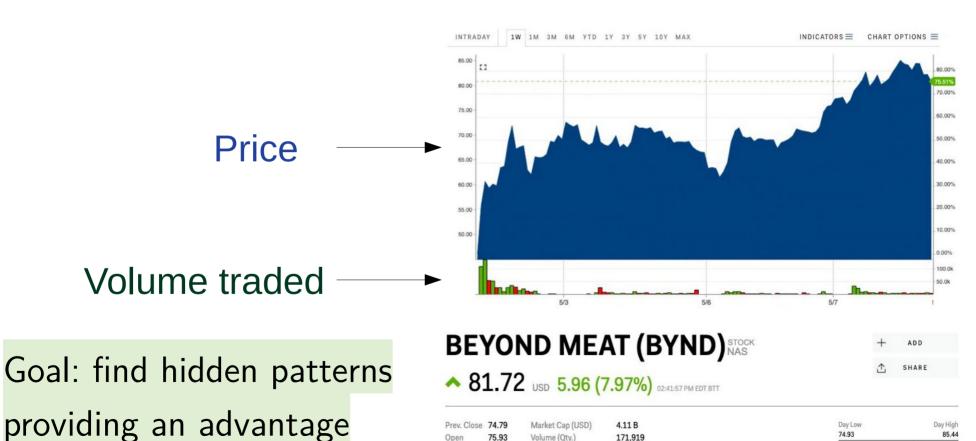
Liquid metal droplets

- ♦ = length of hot metal droplet
 - = droplet release (chaotic, noisy)

Goal: prediction of release



Stock prices



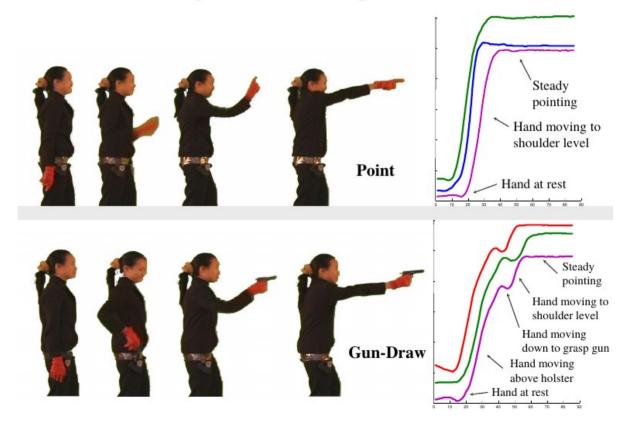
Volume (Qty.)

171,919

81.78

Video data / gestures

- Series of angles of articulations in the body
- Temporal patterns can reveal gestures



Applications

- Clustering
- Classification
- Motif discovery
- Event detection
- •

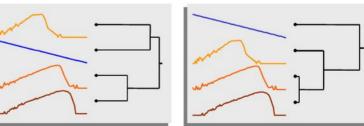
- 1)All require a reasonable definition of the **similarity** between two time series
- 2)All can be done in **real-time** or **retrospectively**

Context vs Behavior

- Contextual attribute(s)
 - $-x(i) = t_i = timestamp$ is the typical one
 - Sometimes other attributes providing context
- Behavioral attribute(s)
 - $^ y^{j}(i)=$ temperature, angle, price, sensor reading, ... $j \in 1 \dots d$

What are the difficulties?

- High sampling rate of many series over extended periods of time means ...
 - Tons of data
 - Things are bound to fail at several points (missing data, noisy data)
- Subjectivity



Preparing a time series

Notation: multivariate time series

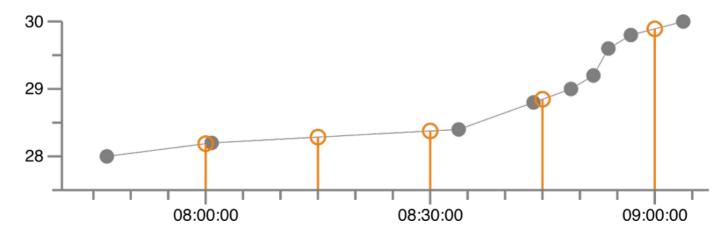
- Length n, timestamps $t_1, t_2, ..., t_n$
- Values at time $t_i : (y_i^1, y_i^2, ..., y_i^d)$
- If series is univariate we drop the superscript

Missing values: linear interpolation

• Let $t_i < t_x < t_j$

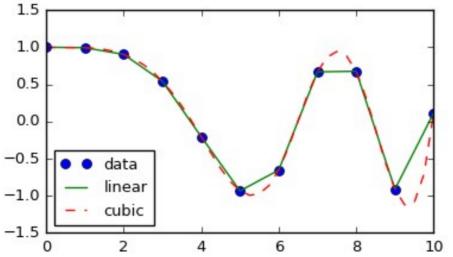
$$y_x = y_i + \left(\frac{t_x - t_i}{t_i - t_i}\right) \cdot (y_j - y_i)$$

Example: make an irregular series regular



Missing values: splines

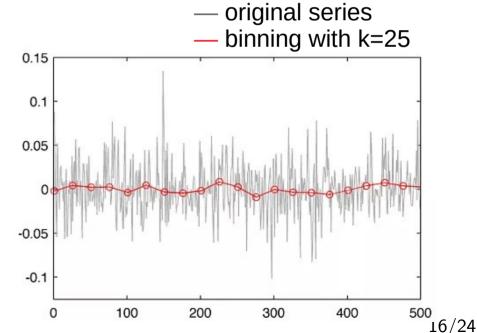
Cubic polynomials between y_i , y_{i+1} that have the same slope at those points as the original curve.



Noise removal: binning

• Replace series by average of values in bins (subsequences) of length $k = \frac{1}{2}$

$$y'_{i+1} = \frac{1}{k} \sum_{r=1}^{k} y_{i \cdot k + r}$$



Noise removal:

moving average smoothing

• Equivalent to overlapping bins

$$y_i' = \frac{1}{k} \sum_{r=1}^k y_{i-r+1}$$

- Larger k leads to smoother series, but losses more information
- Use smaller k for first k-1 items



https://www.fidelity.com/learning-center/trading-investing/technical-analysis/technical-indicator-quide/sma

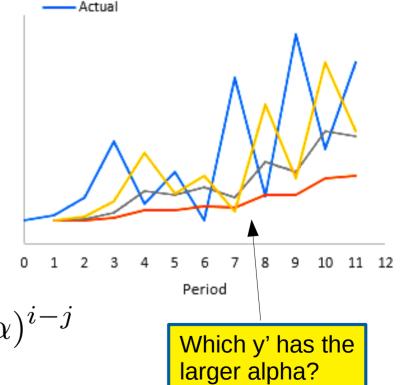
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Noise removal: exponential smoothing

 Combine previously smoothed point with

$$\mathbf{c}_{y_i'} = \alpha \cdot y_i + (1 - \alpha) \cdot y_{i-1}'$$

• $Fy_i' = (1 - \alpha)^i \cdot y_0' + \alpha \sum_{i=1}^i y_i \cdot (1 - \alpha)^{i-j}$

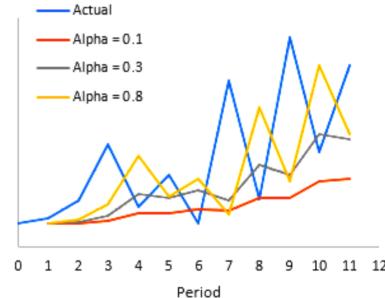


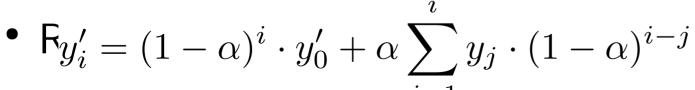
Noise removal:

exponential smoothing

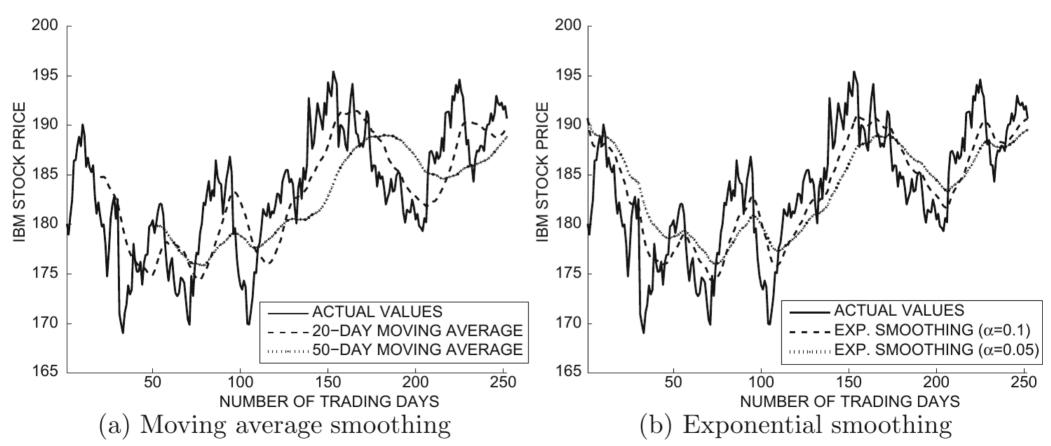
 Combine previously smoothed point with

$$\mathbf{y}_{i}' = \alpha \cdot y_{i} + (1 - \alpha) \cdot y_{i-1}'$$





Moving average vs exponential smoothing



Exercise

Answer in Google Spreadsheet

• Given the following series:

t	1	2	3	4	5	6	7	8	9	10
y(t)	2	4	12	2	1	-2	0	15	3	3
1. y'(t)										
2. y'(t)										

- 1. Moving average with k=3
- 2. Exponential average with alpha=0.5

Summary

Things to remember

- Series preparation
 - Interpolation
 - Smoothing

Exercises for TT27-TT29

- Data Mining, The Textbook (2015) by Charu Aggarwal
 - Exercises $14.10 \rightarrow 1-6$