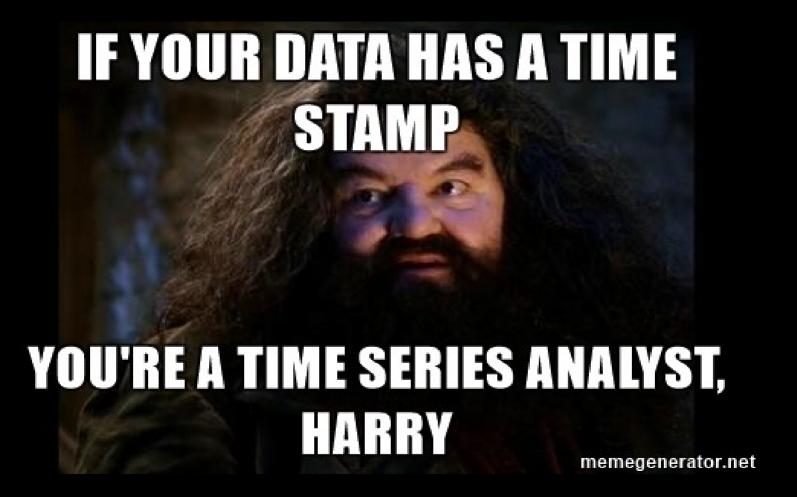
# Mining Time Series

#### Mining Massive Datasets

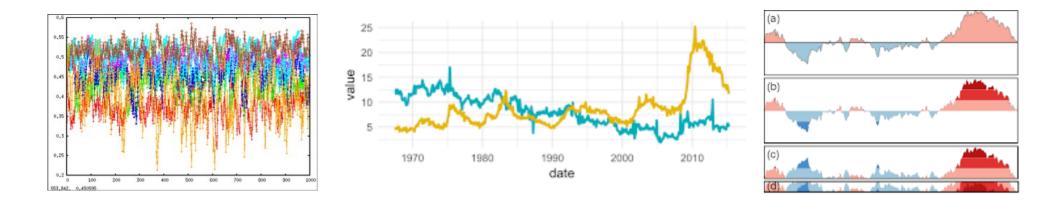
Prof. Carlos Castillo — <a href="https://chato.cl/teach">https://chato.cl/teach</a>



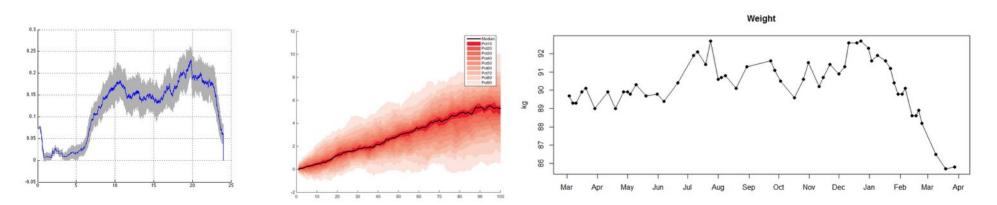


#### Sources

- 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

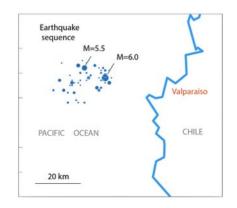


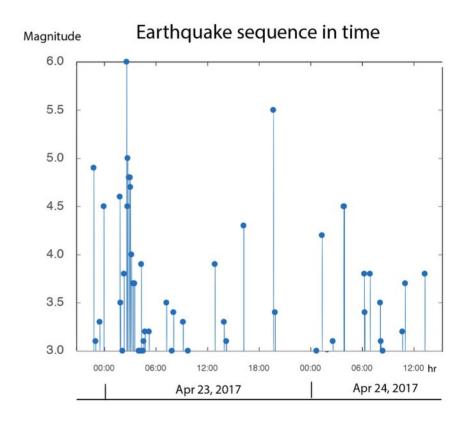
#### Why do we mine time series? Examples



#### Seismic data

- Observations = earthquakes
- Goal: characterize when peaks occur

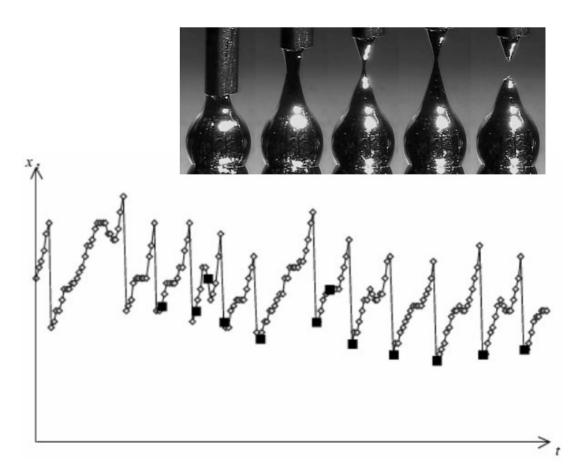




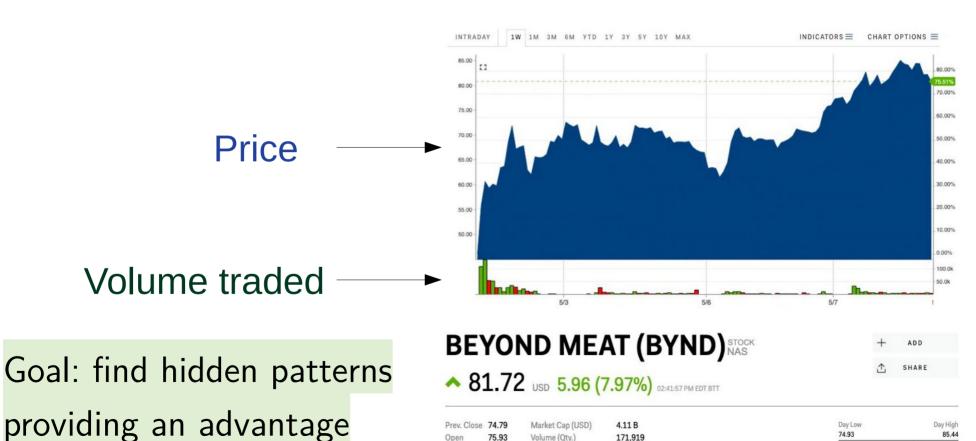
#### Liquid metal droplets

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

Goal: prediction of release



### Stock prices



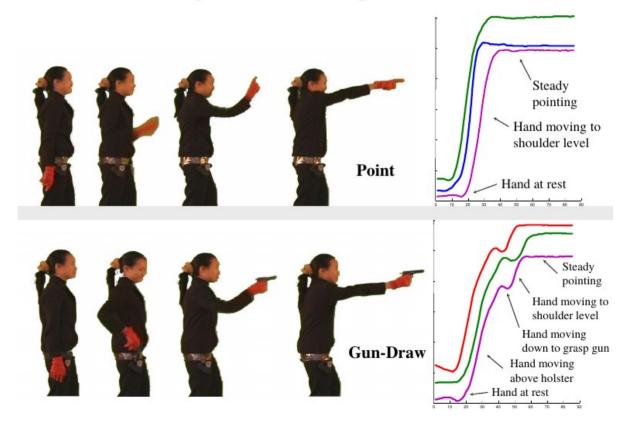
Volume (Qty.)

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
- •

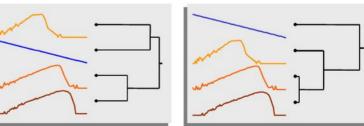
- 1. All require a reasonable definition of the **similarity** between two time series
- 2. All can be done in **real-time** or **retrospectively**

#### 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

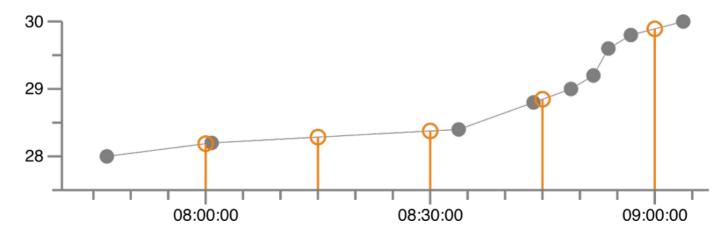
- Length n, timestamps  $t_1$ ,  $t_2$ , ...,  $t_n$
- Values at time  $t_i$ :  $(y_i^1, y_i^2, ..., y_i^d)$
- If series is univariate we drop the superscript

## Missing values: linear interpolation

• Let  $t_i < t_x < t_j$ 

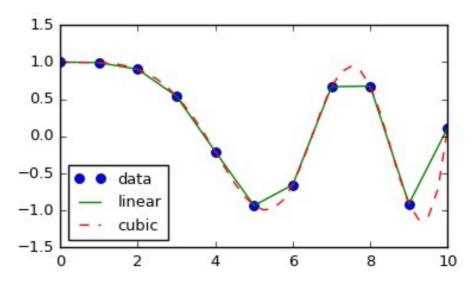
$$y_x = y_i + \left(\frac{t_x - t_i}{t_i - t_i}\right) \cdot (y_j - y_i)$$

Example: make an irregular series regular



## Missing values: splines

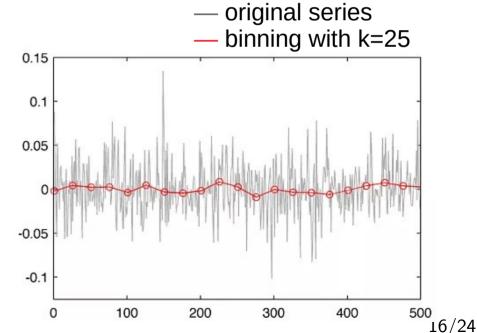
Cubic polynomials between  $y_i$ ,  $y_{i+1}$  that have the same slope at those points as the original curve.



## 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-guide/sma

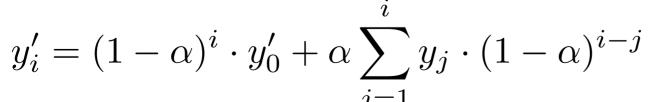
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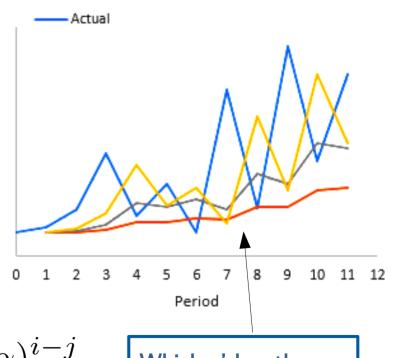
## Noise removal: exponential smoothing

 Combine previously smoothed point with current point

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

Recursively substituting





Which y' has the larger alpha?

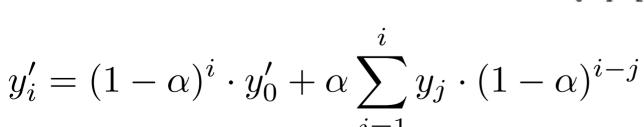
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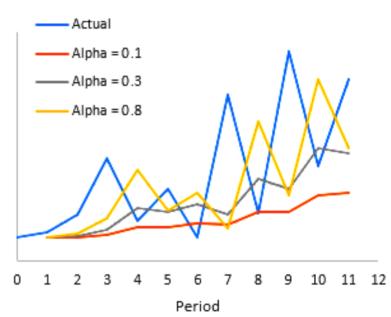
## Noise removal: exponential smoothing

 Combine previously smoothed point with current point

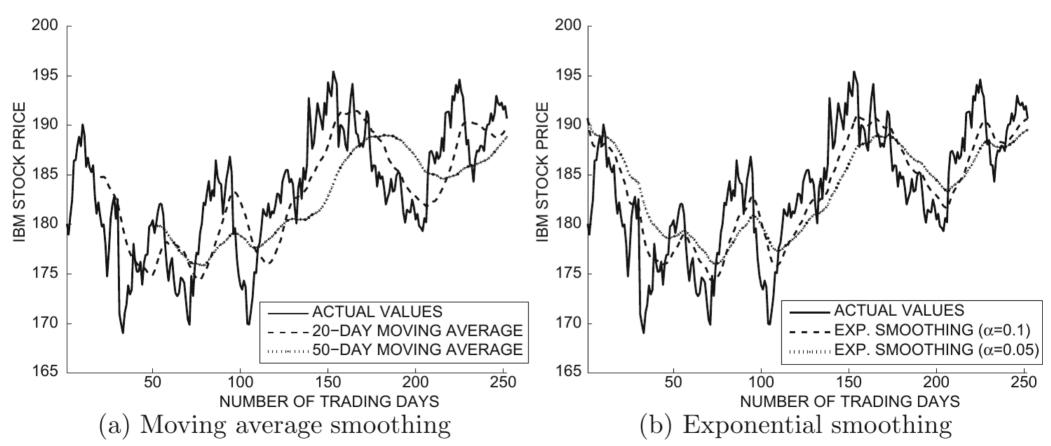
$$y_i' = \alpha \cdot y_i + (1 - \alpha) \cdot y_{i-1}'$$

Recursively substituting





# Moving average vs exponential smoothing



#### Exercise: smooth a time series

• 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  $14.10 \rightarrow 1-6$