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Many Fintech activities involve the financial market

Rationale

This includes, for example:

- robo-advisory
- robo-for-advisory
- trading platforms

And speaking of the financial market means speaking of risks

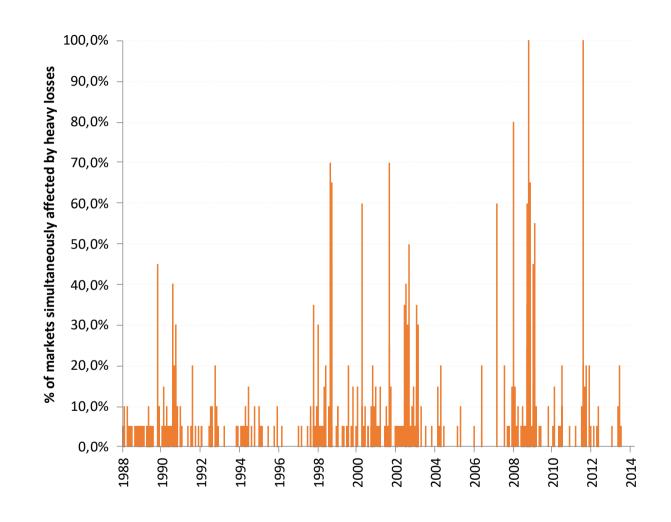
Market crashes do happen

| Larg         | jest Real         | Declines in | U.S. Stock I | Market Hist |                 |                |   |
|--------------|-------------------|-------------|--------------|-------------|-----------------|----------------|---|
| Pain<br>Rank | Pain Index<br>(%) | Peak        | Trough       | Recovery    | Decline<br>Rank | Decline<br>(%) | Event(s)  |
| 1            | 100.00            | Aug 1929    | May 1932     | Nov 1936    | 1               | 79.00          | 1929 Crash & Great Depression                       |
| 2            | 89.34             | Jun 1911    | Dec 1920     | Dec 1924    | 4               | 50.96          | WWI & Influenza                                     |
| 3            | 85.51             | Aug 2000    | Feb 2009     | May 2013    | 2               | 54.00          | Lost Decade (Dot-Com Bust & Global Financial Crisis |
| 4            | 80.41             | Dec 1972    | Sep 1974     | Jun 1983    | 3               | 51.87          | Inflation, Vietnam, & Watergate                     |
| 5            | 59.57             | Feb 1937    | Mar 1938     | Feb 1945    | 5               | 49.93          | Great Depression & WWII                             |
| 6            | 29.06             | May 1946    | Feb 1948     | Oct 1950    | 6               | 37.18          | Postwar Bear Market                                 |
| 7            | 14.22             | Nov 1968    | Jun 1970     | Nov 1972    | 7               | 35.54          | nflationary Bear Market                             |
| 8            | 8.23              | Jan 1906    | Oct 1907     | Aug 1908    | 8               | 34.22          | Panic of 1907                                       |
| 9            | 8.18              | Apr 1899    | Jun 1900     | Mar 1901    | 9               | 30.41          | Cornering of Northern Pacific Stock                 |
| 10           | 7.73              | Aug 1987    | Nov 1987     | Jul 1989    | 10              | 30.21          | Black Monday  |
| 11           | 6.25              | Nov 1886    | Mar 1888     | May 1889    | 13              | 22.04          | Depression & Railroad Strikes                       |
| 12           | 5.00              | Apr 1903    | Sep 1903     | Nov 1904    | 14              | 21.67          | Rich Man's Panic                                    |
| 13           | 4.80              | May 1890    | Jul 1891     | Feb 1892    | 17              | 20.11          | Baring Brothers Crisis                              |
| 14           | 3.55              | Dec 1961    | Jun 1962     | Apr 1963    | 12              | 22.80          | Height of Cold War & Cuban Missile Crisis           |
| 15           | 3.20              | Aug 1897    | Mar 1898     | Aug 1898    | 15              | 21.13          | Outbreak of Boer War                                |
| 16           | 3.14              | Oct 1892    | Jul 1893     | Mar 1894    | 11              | 27.32          | Silver Agitation                                    |
| 17           | 3.11              | Sep 1909    | Jul 1910     | Feb 1911    | 16              | 20.55          | Enforcement of Sherman Antitrust Act                |
| 18           | 1.00              | Dec 2019    | Mar 2020     | Jul 2020    | 18              | 20.00          | COVID-19 Pandemic                                   |
|              |                   |             |              |             |                 | \ /            |   |

Data as of Feb. 28, 2021. Sources: Kaplan et al. (2009); Ibbotson (2020); Morningstar Direct; Goetzmann/Ibbotson, and Peng (2000); Pierce (1982); www.econ.yale.edu/^shiller/data.htm, Ibbotson Associates SBBI US Large-Cap Stock Inflation Adjusted Total Return Extended Index.

The frequency of simultaneous "tail" events is increasing over time

% of markets (all the 20 main world Stock Exchanges) simultaneously affected by large weekly losses (2.5%-tail of the empirical Copula estimated on weekly data since 1/1988)



Source: Zenti, R. (2014) «Volatility, decision models and complexity in financial market», Artificial Intelligence and Cognitive Science, Il Mulino

# The market's worst days had a big impact on investment returns

#### The market's worst days have had a large effect on returns

Growth of \$1 in the S&P 500 Index from Dec. 31, 1927, to Dec. 31, 2015

| Days                           | Ending value (\$) | Cumulative return (%)  |
|--------------------------------|-------------------|------------------------|
| Total cumulative return        | 115.40            | 11,440.46              |
| Miss 10 best                   | 38.28             | ( 3,728.33 ) <b>3X</b> |
| Miss 10 worst                  | 362.01            | 36,100.79              |
| Miss 10 best and miss 10 worst | 120.09            | 11,908.91              |
| Cash                           | 19.33             | 1,832.88               |

Sources: Bloomberg L.P., Invesco, Morningstar.

# The problem

- One of the biggest problems of financial market is its annoying tendency to crash
- Market crises correspond to "risk-off" situations, in which risk premia and financial assets exhibit anomalous behavior
- There are big gains in detecting such crashes early on: risk prevention and improved financial performance
- Rather than predicting risk-off situations, it is sufficient to recognize them at their dawn
- It's more <u>nowcasting</u> than forecasting
- The large amount of financial data available invites us to solve the problem using data science

## Solution: Early Warning Systems

- Main goal: detecting crises before most damage has been made; and reducing false alarms
- Data Science provides us with many methods that can be successfully applied (also used in combination)

Business case:
Let's look at the data



## Data overview

### Weekly data from Bloomberg

- Key equity indices
- Bond indices (Global, Corporate IG/HY, Inflation-linked, Municipals, Mortgages)
- Short/medium/long term interest rates
- Key exchange rates
- Commodities
- Leading indicators (Economic surprise, Baltic Dry Index)
- VIX (option implied volatility)

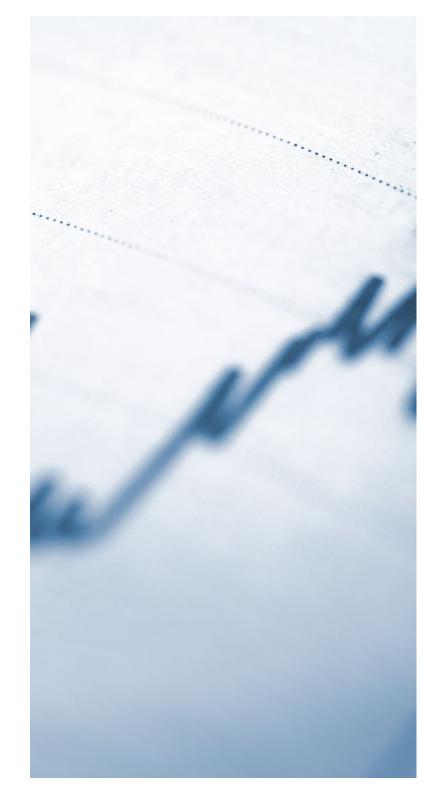
A label «abnormal/normal»

• What would you do?

Brainstorming

(don't be shy)

 What kind of algorithms would you use? Why?



## My focus: anomaly detection

- Anomaly: an observation which deviates so much from the other observations as to create suspicion that it was generated by a different mechanism
- Often indicative of something interesting
- Any deviations from the normal behavior that is unusual and significant is of special interest and may require action
- Detecting anomalies is increasingly becoming the core of many business operations, including finance
- Anomaly detection is also used for data cleaning removing outliers from a dataset before training another model – and for unbalanced classification

### Some basic notions

- **Abnormal** instances = *anomalies* = *outliers*
- Normal instances = inliers

- **Novelty Detection** = the Anomaly Detection algorithm is trained on a clean dataset *without outliers* in order to identify whether a new instance is an outlier or not
- Outlier Detection = the Anomaly Detection algorithm is trained on a dataset with outliers

Some applications in Financial Services

Early Warning Systems / Risk Modeling

Statistical Arbitrage (investments)

**Transactional Frauds** 

Anti-Money Laundering

Customer Behavior Analytics (in general)



Matlab session starts

## Take home on Early Warning Systems & Anomaly Detection

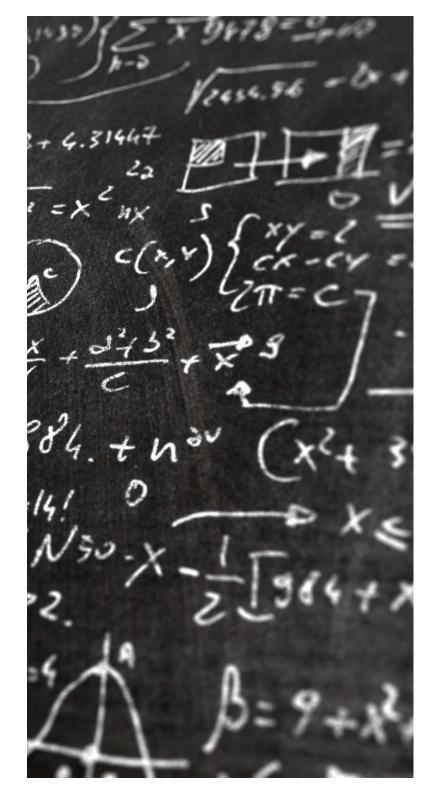
- Increasingly popular class of algorithms in Data Science
- You might use:
  - Probabilistic models (multivariate Gaussian, Copulas, whatever)
  - Classifiers
  - Isolation Forest
  - Clustering methods
  - Latent variable models like State-Space, PCA, etc
  - Simple proximity models, ie, distancebased model, like k-NN
  - ...
- You might also imagine an investment strategy based on that (buy/sell = f(risk-on/off)

#### **Now YOU**

It's your turn: use your favorite techniques to estimate customers' needs and recommend products, write the code (use my code, or start from scratch, or whatever...as you like), and we'll talk about it next time

Some useful links on Early Warning Systems for market crises

- https://www.federalreservehistory.org/essays/s tock-market-crash-of-1929
- <a href="https://www.federalreservehistory.org/essays/g">https://www.federalreservehistory.org/essays/g</a> reat-recession-and-its-aftermath
- https://arxiv.org/abs/0905.0220
- https://www.treasury.gov/initiatives/wsr/ofr/D ocuments/OFRwp0001\_BisiasFloodLoValavanis\_ ASurveyOfSystemicRiskAnalytics.pdf.



### Next time (as usual)

- Each group will present ideas, results, doubts, code snippet, etc
- Be short and concise VERY CONCISE, you are... many: 5'-10' each group
- Prepare plots, charts, tables, commented code snippets
- Get ready to share your screen and your work be it little or a lot – don't be shy
- I will schedule «Office Hours»: time outside of class to meet with you, to discuss the material being presented in class or some ideas you have



### Next «Office hours»

- Thursday 14 April, h 17:30-18:30
- We will use Webex (my room = same virtual room used for the lectures)
- Please book if you want to talk to me; write to me at this email address (NOT the Politecnico email):

raffaele.zenti@virtualb.it