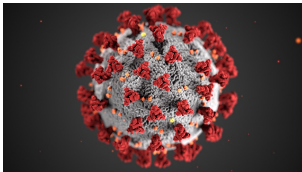


Pandemic Forecasting via Stock Market Indicators

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MIT PRIMES

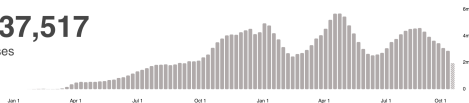
October 16 2021



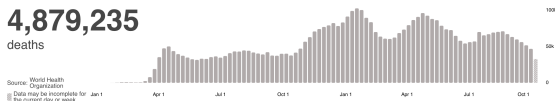
Importance

People around the world have been struggling with CoVID-19

239,437,517
confirmed cases

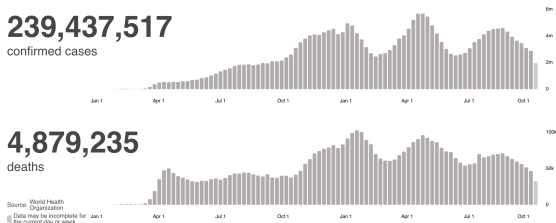


4,879,235
deaths



Importance

People around the world have been struggling with CoVID-19



We turned to stock markets in order to find new forecasting tools.

Stock Markets

Stock Market Attributes

- Random time series
- Move in trends



Technical indicators can predict trend continuation and reversals.

Technical Analysis

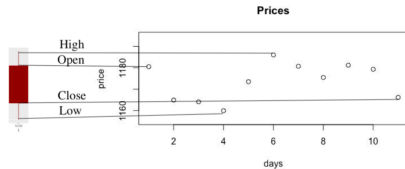
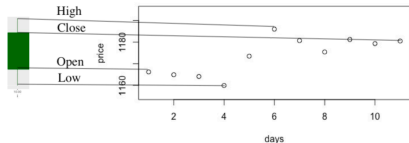
Tools for identifying trend reversals which we explore:

- Candlesticks
- Candlestick patterns
- MACD analysis
- RSI analysis

Candlesticks

The first tool we adapt from finance to pandemics are candlesticks. This collects the time series into larger units and presents the information as follows:

$$\text{Candlestick}_t = (O_t, C_t, H_t, L_t).$$



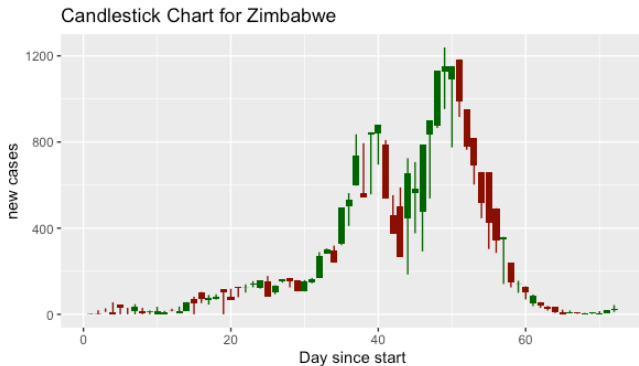
Example

Line plot for Apple's share price can be translated into candlesticks:



Application to the Pandemic

Organize daily new cases into weekly candlesticks.



Candlestick Patterns

Candlestick patterns – combinations of specific consecutive candlesticks – used as stock market forecasters.

Prominent examples that we shall study for COVID are:



Bullish
Engulfing



Bearish
Engulfing



Dark Cloud
Cover



Hammer



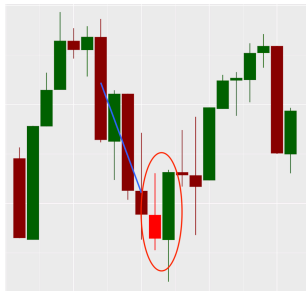
Hanging Man

Bullish (Bearish) pattern predicts reversal to Uptrend (Downtrend).

Candlestick Patterns

Example: Bullish Engulfing

- Pattern shown in the centre of chart (right).
- Preceding trend is downtrend.
- Predicts reversal of trend.



MACD: Moving Average Convergence Divergence

MACD indicator claims predictive power in stock markets.

MACD is based on observations on moving averages.

Example from *S&P 500 Index*.



RSI: Relative Strength Index

RSI is another indicator claiming predictive power in stock markets.

RSI identifies asset prices that have move “too quickly”.

Example from *S&P* 500 Index.



Application to the Pandemic

We explore the application of these indicators to COVID data.

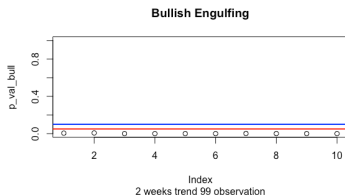
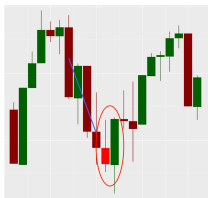
Our R code scans WHO COVID data of 237 countries and identifies different signals and patterns.

Signal Name	Number of observations
Bullish Engulfing	99
Bearish Engulfing	123
Hammer	127
Hanging Man	156
Dark Cloud Cover	30
Bullish MACD	217
Bearish MACD	245
Bullish RSI	46
Bearish RSI	1057

Application to the Pandemic

Our R code identified 99 occurrences of Bullish Engulfing Pattern.

We then perform statistical test: p -value (y -axis) is calculated n days after the pattern (x -axis).



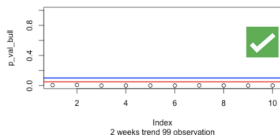
Circles below blue (red) line imply p -value of 0.1(0.05).

Bullish Engulfing is a significant forecaster for COVID data.

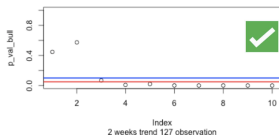
Results

Repeat for other indicators, here we show selection.

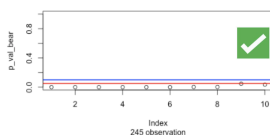
Bullish Engulfing



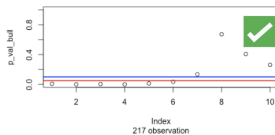
Hammer



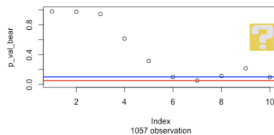
Bearish MACD



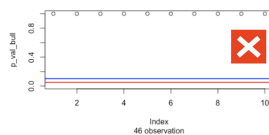
Bullish MACD



Bearish RSI



Bullish RSI



Many are statistically significant forecasters of future COVID cases.

Summary

- We have repurposed technical indicators of stock market for use in pandemics.
- Intuition due to fact pandemic and stock market both can be modeled as random walks.
- We showed that technical indicators do have statistically significant forecasters of near term COVID cases.
- Forecasting can aid in allocating medical resources.

Acknowledgment

I would like to thank **Prof. James Unwin** for his guidance along the way and **MIT PRIMES-USA** for making this collaboration possible.

Thank You!

Exponentially Moving Average

Consider a dataset of length n , normally the closing prices, $\{C_i\}$

$$V_i[C_i] = \begin{cases} C_1 & i = 1 \\ \alpha C_i + (1 - \alpha)V_{i-1} & i > 1 \end{cases},$$

where α is the smoothing factor. V_n is the **Exponentially Moving Average (EMA)** of $\{C_i\}$. In particular,

$$V_n = \alpha[C_n + (1 - \alpha)C_{n-1} \cdots (1 - \alpha)^{n-1}C_1].$$

- Places greater weight on C_n

Moving Average Convergence Divergence

The signals are from two intertwining lines. Common values for $(n_1, n_2, n_3) = (12, 26, 9)$, $\alpha = \frac{2}{n+1}$.

$$\text{MACD}(n_1, n_2) = V_{n_1} - V_{n_2}$$

- the strength of up trend

$$S = V_{n_3}[\text{MACD}(n_1, n_2)]$$

- the strength of change in trend

Relative Strength Index

- Divide the dataset $\{C_n\}$ into two sets, Gain and Drop.

$$\{G_t\} = \frac{C_t - C_{t-1}}{C_t}, \quad C_t > C_{t-1}$$

$$\{D_t\} = \frac{C_{t-1} - C_t}{C_t}, \quad C_t < C_{t-1}$$

Calculate the Average,

$$\bar{G}_t = V_n[G_t],$$

$$\bar{D}_t = V_n[D_t].$$

The index is given by

$$\text{RSI} = 100 - \frac{100}{1 + \frac{\bar{G}_t}{\bar{D}_t}}.$$

- Measures the relative strength of gains and drops.

Wilcoxon Signed Rank Test

Consider dataset $\{x_i\}$ and hypothesized median \tilde{x}

To compare the distribution on both side, rank the difference and compute the sum. Formally, the statistic W is

$$W = \sum \text{sgn}(d_t)R(d_t) ,$$

where $R(\cdot)$ is the rank and

$$\text{sgn}(d_t) = \begin{cases} 1, & d_t > 0 \\ -1, & d_t < 0 \\ 0, & d_t = 0 \end{cases} .$$