

# Deep Learning for Time Series

Becoming gurus of time series together

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# Agenda

1. Time
2. Time series
3. Time series analysis
4. Limitation of traditional time series methods
5. Deep learning for time series



1.

# Why is “time” important?

Let's start with reaching a consensus of time



*Albert Einstein's considered "Time" is the fourth dimension in the universe in Special Relativity.*

# Time “Series”

A time series is a **sequence** taken at **successive** **equally spaced** data points in time.



# Three main types of data science

Regression

Classification

Time series

The background features abstract geometric shapes in various shades of green and teal. On the right side, there are several white icons: a planet with rings, a rocket ship, a globe, and several stars.

# Time Series Analysis

Methods for analyzing time series data in order to extract meaningful statistics and other characteristics of the data.



2.

# Concept Recap

Let's start with the time series analysis sharing two weeks ago



# Basic components of time series

## **Trend**

Long term **movement** of data

## **Cycle**

Long term (1 year+) wavelike variations

## **Seasonality**

Short term, **regular** variation

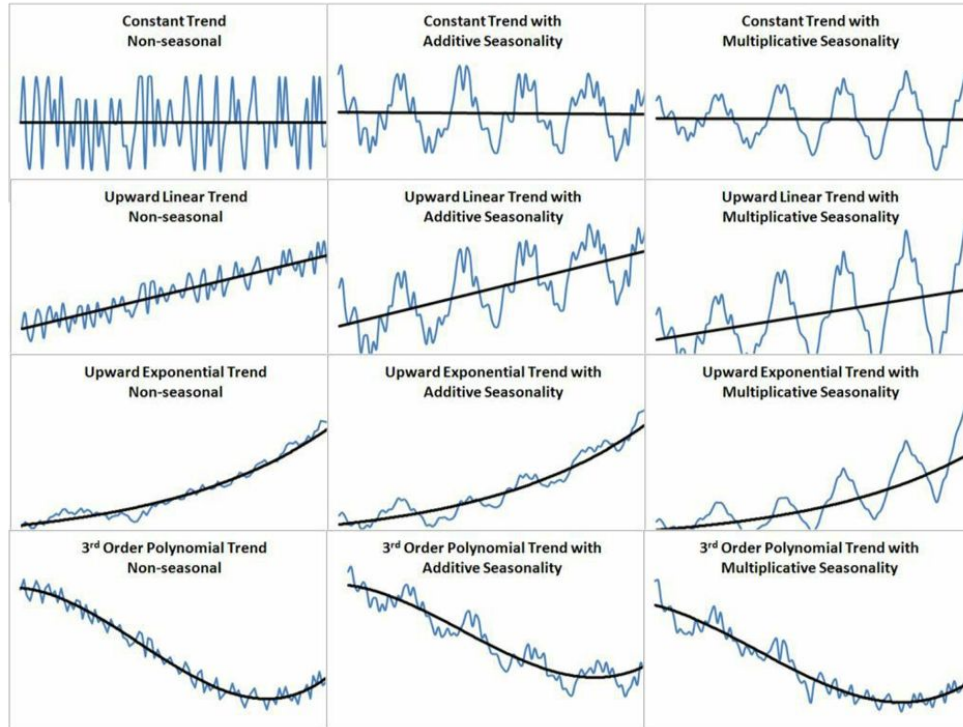
## **Random Variations**

**Residual** variations after all other elements

## **Irregular Variations**

**Unusual circumstance** and did not reflect typical behavior

# Common trend and seasonality patterns



# How to forecast?

The background of the slide features a vibrant green and teal geometric pattern. On the right side, there are white icons of a planet with rings, a rocket ship, and several stars, set against a light green background that resembles a starry sky.

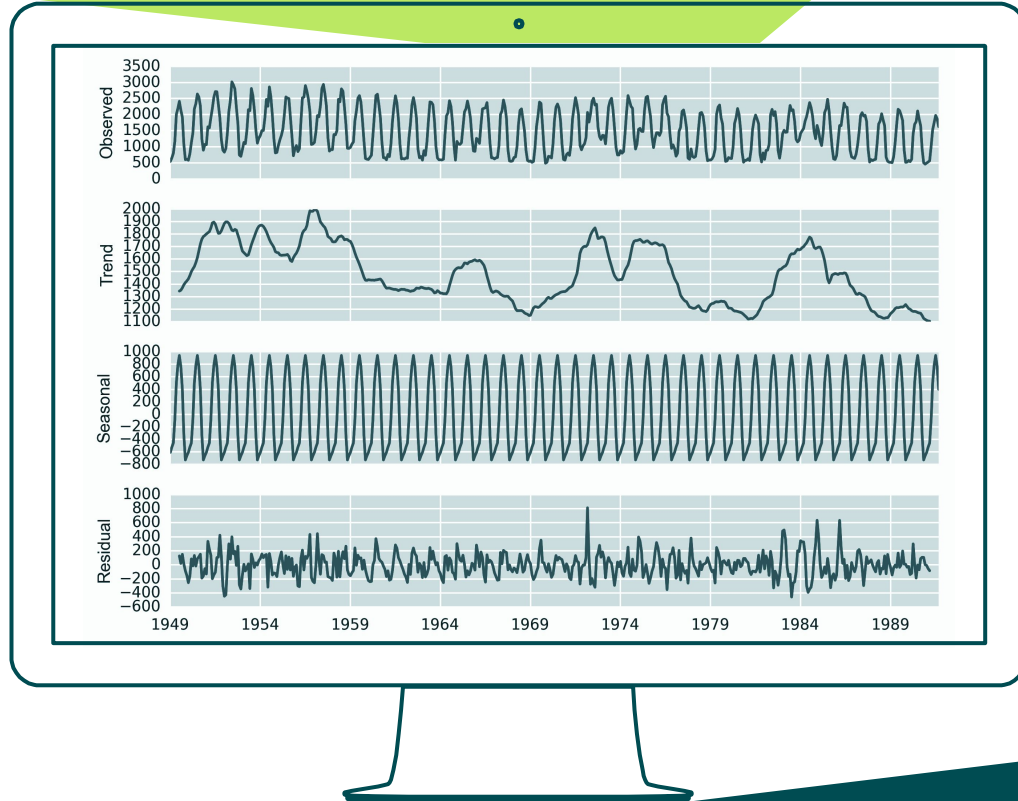
Basically, classic time series forecasting models are trying to **decompose** and **learn** the series from the five elements.

E.g.

$$F = T + S + C + I$$

$$F = T * S * C * I$$

# Decomposition of time series





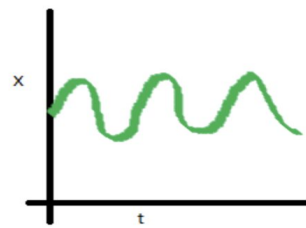
# But

Nonstationary and time-varying volatility make time series be hard to analyze.

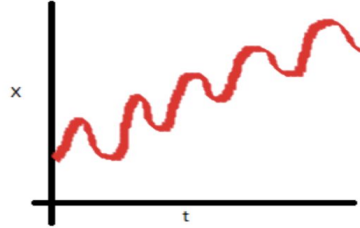
We need to make series stationary to forecast easier.



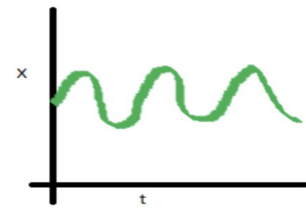
# What is nonstationary?



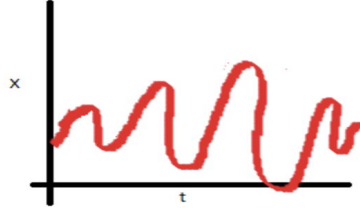
Stationary series



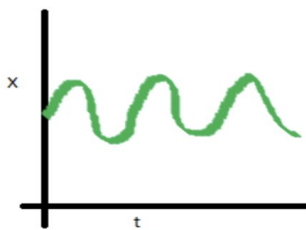
Non-Stationary series



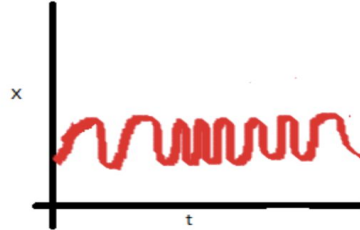
Stationary series



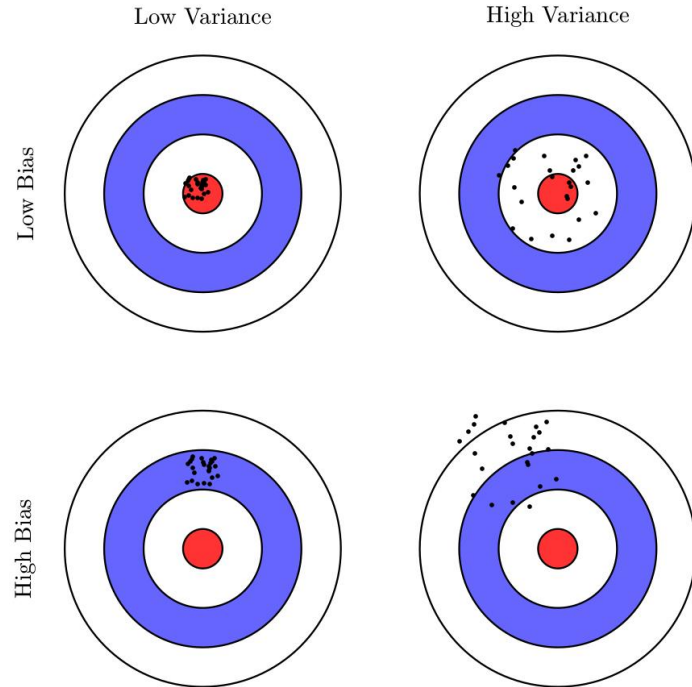
Non-Stationary series



Stationary series



Non-Stationary series



# How

N-order differences or linear combination make series tend to be a constant or a linear function.

-Sir William Granger

N-order diff. =  $Y_t - Y_{t-n}$

Linear com. =  $Y_t - \alpha X_t$

# Traditional time series methods

## **ARIMA family**

AR, MA, ARMA, ARIMA, SARIMA, vecARIMA...

## **Exponential smoothing family**

Single exp smoothing, double exp smoothing, Holt exp smoothing...

## **ARCH family**

ARCH, GARCH, EGARCH...



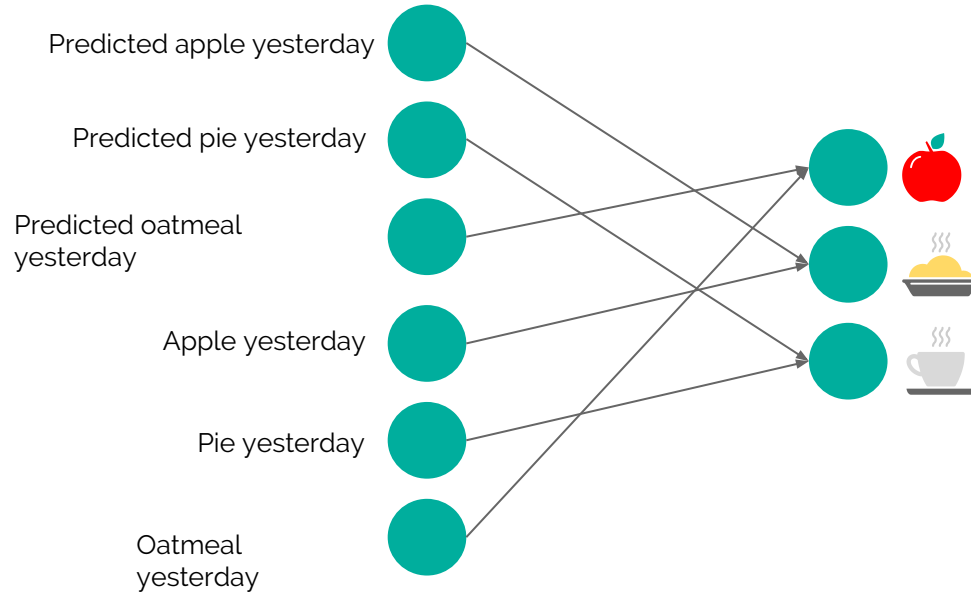
# Limitation of traditional forecast models

- ◆ Focus on complete data
- ◆ Focus on linear relationships
- ◆ Focus on fixed temporal dependence
- ◆ Focus on univariate data
- ◆ Focus on one-step forecasts

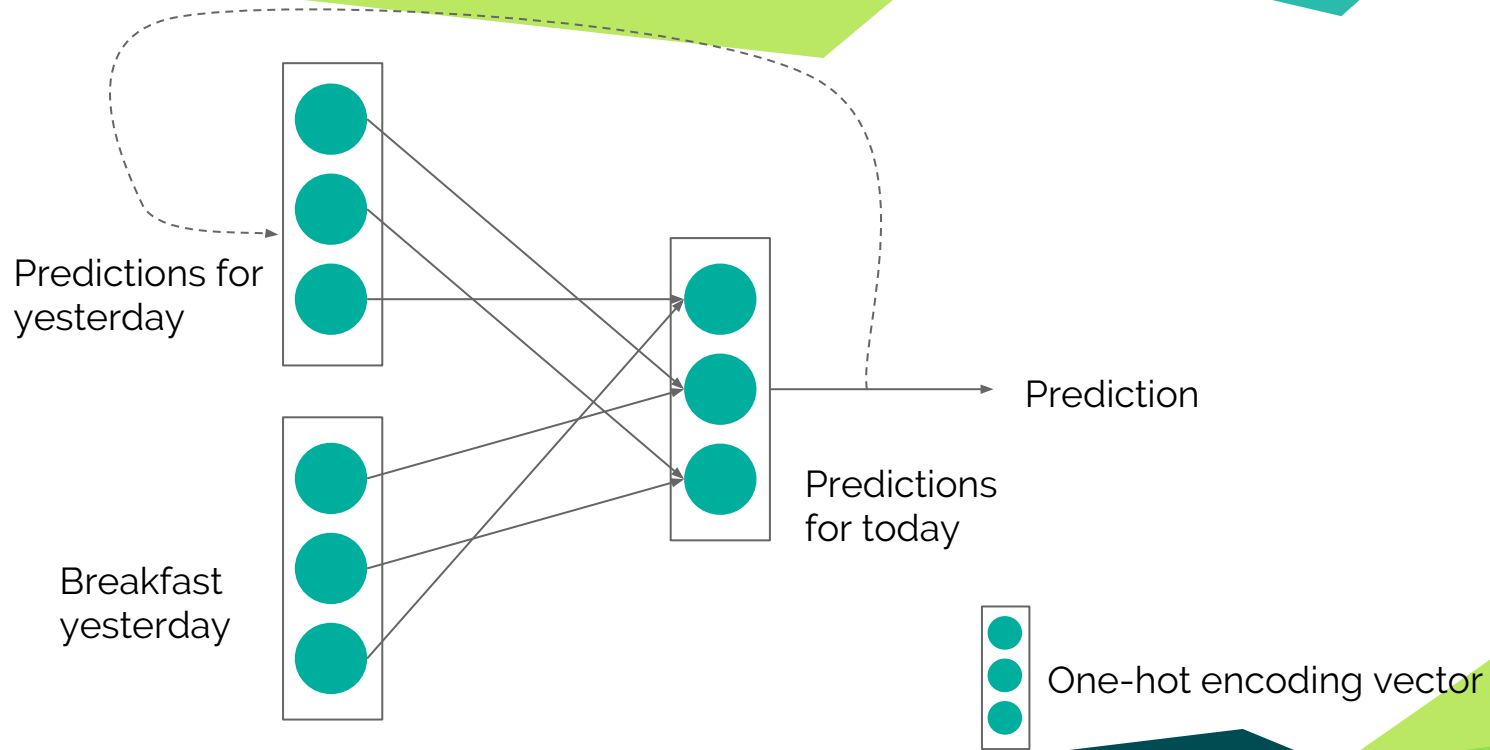


# *Deep Learning for Time Series*

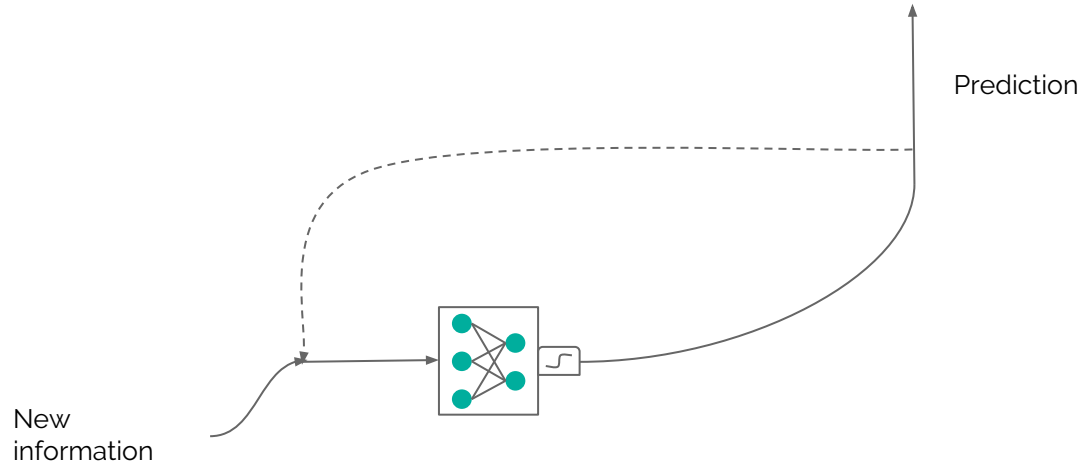
# Start from what's for breakfast tomorrow?



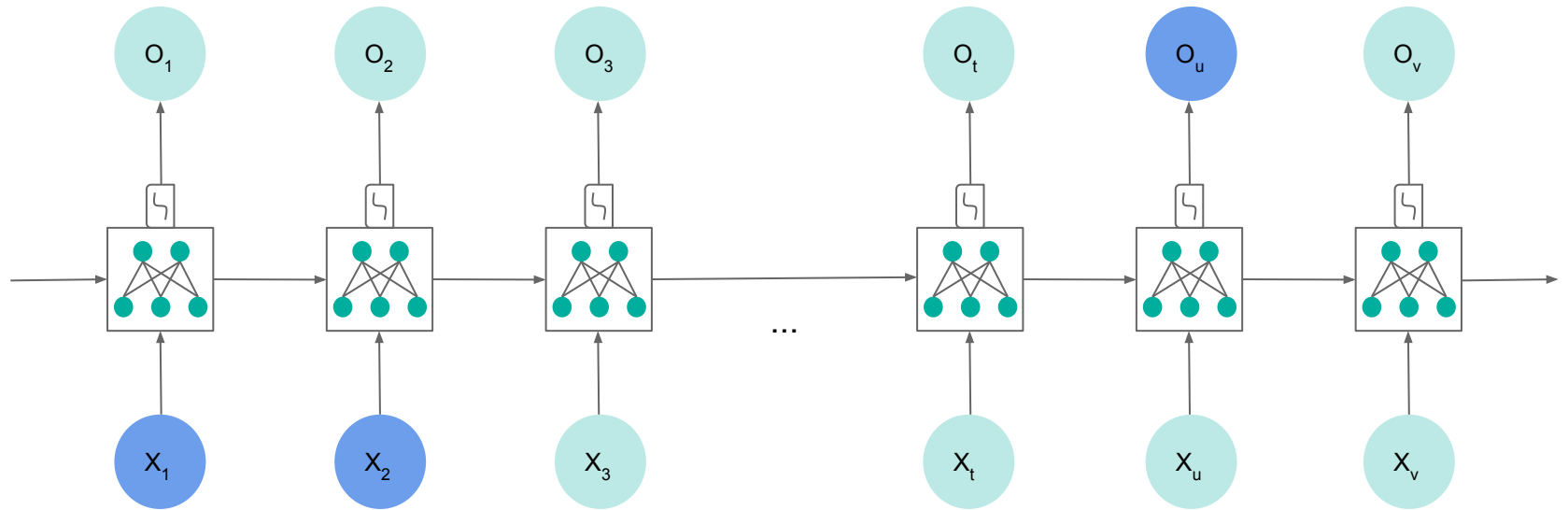
# RNN



# Simplified RNN



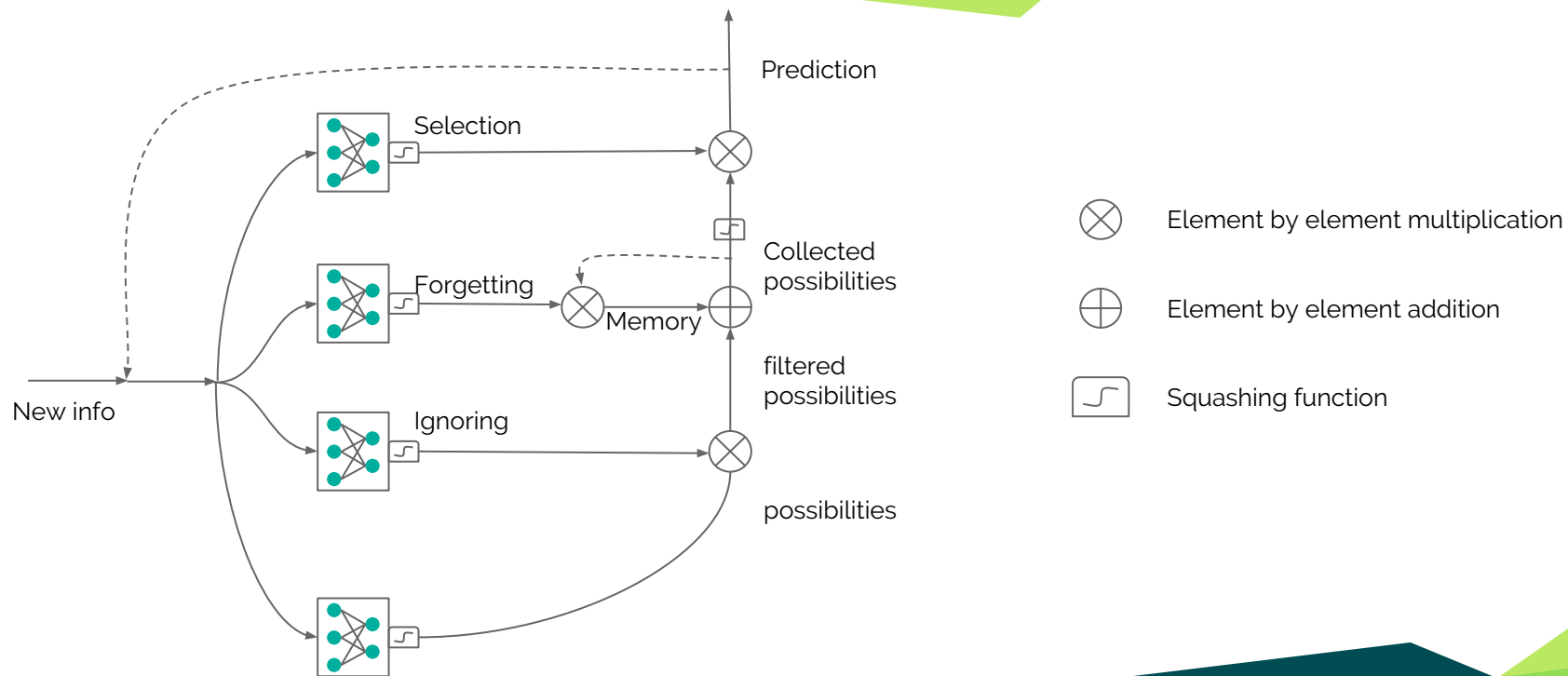
# Expend RNN



# Limitation of RNN

- ◆ Short term memory
- ◆ Vanishing/ Exploding gradients

# LSTM





# References

1. [A beginner's guide to RNN and LSTM](#)
2. [遞歸神經網路\(RNN\)和長短期記憶模型\(LSTM\)的運作原理](#)
3. [Deep learning for time series analysis](#)
4. [Time Series Prediction with LSTM Recurrent Neural Networks in Python with Keras](#)



Hands on