

# Assignment 1

Develop a forecasting model of the bitcoin price that performs better than chance. The model can be either a regression or classification model.

## Regression

The model should predict the price at the next timepoint

## Classification

The model should predict whether the price will go, for example, up or down. However, it doesn't have to be a binary prediction (i.e. up/down) but may also be a multi-class prediction (e.g. up/down/no\_change). However many classes you chose, you will need to define the class boundaries.

Example of class boundaries:

class	interval
0	-infinity to -0.5%
1	-0.5% to 0.5%
2	0.5% to +infinity

## Dataset

The dataset is available in **data.zip** on Canvas. The dataset consists of multiple CSV files with ohlcv (open, high, low, close, volume) data at different frequencies (e.g. every 30 min, 1 hour, 1 day or 1 week). You have to chose which frequency you want to use and whether to predict the opening, high, low or close price.

## Validation

Partition the dataset into training and validation sets. Use a 20-point moving average as the baseline prediction to beat.

# Regression

Regression will be evaluated with the Mean Squared Error (MSE) and you need to get lower MSE than the baseline prediction.

# Classification

Classification will be evaluated with the F1-score and you will need to be a higher F1-score than the baseline prediction (partitioned according to the class boundaries you have defined).

# Submission & test

Once submitted, your model will be tested against previously unseen data. Your goal is to beat the 20-point moving average on the test set. In your submission, you will need to specify with data frequency you used to train your model (5 min, 15 min, 1 hour, ...) and whether it predicts the open, high, low or close price. If you decide on a classification model, then you will also need to submit the class boundaries that you used.

The submission should be a **PY** file and the trained model weights. The **PY** file should have to functions: **preprocess** and **model**. If you're using Keras, then you can find information about how to save model weights in their [FAQ](#).

# Functions

- **preprocess()**
  - Description: loads and preprocesses the data.
  - Arguments: a list of names of the data (**CSV**) files (the list can have length 1).
  - Returns: Numpy arrays **X** and **y**. Inputs and targets, respectively.
- **model()**
  - Description: loads the model weights and returns predictions given some inputs.
  - Arguments: filename of the model weights, and **X** (returned by **preprocess()**).
  - Returns: A numpy array of predictions.

The assignment is due at 2:30 pm on February 3.