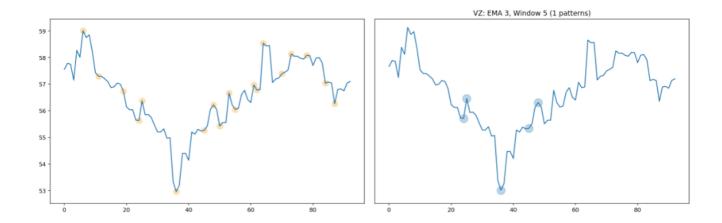
Algorithmically Detecting (and Trading) Technical Chart Patterns with Python





Defining Technical Chart Patterns Programmatically

Ever wondered how to programmatically define technical patterns in price data?

At the fundamental level, technical patterns come from local minimum and maximum points in price. From there, the technical patterns may be defined by relative comparisons in these min/max points.

Let's see if we could have played this by algorithmically identifying any inverse head & shoulders patterns!

Follow along with the notebook here.

samchaaa/alpaca_tech_screener

Contribute to samchaaa/alpaca_tech_screener development by creating an account on GitHub.

```
github.com
```

The following code can easily be retooled to work as a screener, backtester, or trading algo, with any timeframe or patterns you define.

Disclaimer: this code is intended as a starting point for finding technical patterns, it is for educational purposes only. <u>The framework for this code came from here.</u>

An Empirical Algorithmic Evaluation of Technical Analysis

At a recent meeting of the Quantopian staff journal club, I presented a paper by Andrew Lo, Harry Mamaysky, and Jiang...

www.quantopian.com

Step 1.) Read in data

I'm reading in data using the <u>Alpaca</u> API (which I'll also use to place trades later).

I wrote this function to grab data beyond the one request limit of 2,000 minute bars. Later we'll resample to our timeframe of choice.

```
import pandas as pd
 2
     from datetime import timedelta
 3
     import alpaca trade api as tradeapi
 4
 5
 6
     api = tradeapi.REST('YOUR API KEY HERE',
                          'YOUR API SECRET CODE HERE',
                          'https://paper-api.alpaca.markets')
 8
 9
     def get_data(symbol, lookback):
10
11
         all data = pd.DataFrame()
         for x in range(lookback):
12
             if x == 0:
13
                 data = api.polygon.historic_agg('minute', symbol, limit=None).df
             else:
                 data = api.polygon.historic_agg('minute', symbol, _from = (data.index.min() - timede
             start = data.index.min().strftime('%x %X')
```

We'll resample data separately, in case we want to try out different timeframes later.

Step 2.) Find minima and maxima

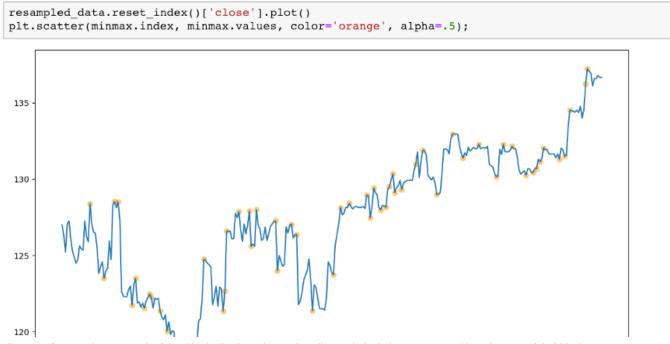
For this step we'll use a function from scipy's signal processing library to find peaks in the data.

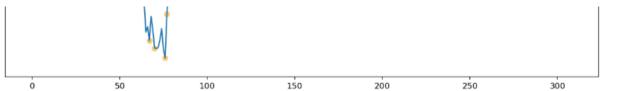
This code looks complicated, but the point is to return the integer index values with price, for each min/max point.

```
import numpy as np
from scipy.signal import argrelextrema

def get_max_min(prices, smoothing, window_range):
    smooth_prices = prices['close'].rolling(window=smoothing).mean().dropna()
    local_max = argrelextrema(smooth_prices.values, np.greater)[0]
    local_min = argrelextrema(smooth_prices.values, np.less)[0]
    price_local_max_dt = []
    for i in local_max:
        if (i>window_range) and (i<len(prices)-window_range):</pre>
```

Let's plot it with the resampled price data to visually confirm we're on the right track.





Step 3.) Find patterns

To find patterns, we simply iterate over all our min max points, and find windows where the points meet some pattern criteria.



For example, an inverse head and shoulders can roughly be defined as:

C < A, B, D, E

A, E < B, D

To filter for head and shoulders with even necklines:

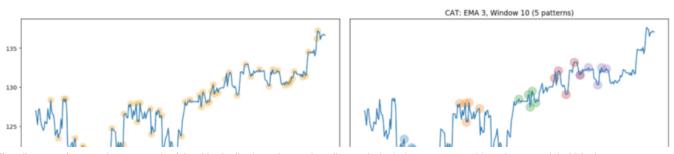
```
abs(B-D) < np.mean([B, D])*0.05
```

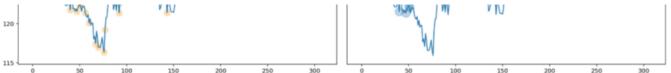
(The difference between the necklines must not be more than 5%.)

Here's the code:

```
from collections import defaultdict
 2
 3
     def find_patterns(max_min):
         patterns = defaultdict(list)
 4
         # Window range is 5 units
         for i in range(5, len(max_min)):
             window = max min.iloc[i-5:i]
             # Pattern must play out in less than n units
             if window.index[-1] - window.index[0] > 100:
                 continue
13
             a, b, c, d, e = window.iloc[0:5]
15
             # IHS
             if a<b and c<a and c<e and c<d and e<d and abs(b-d)<=np.mean([b,d])*0.02:
17
                    patterns['IHS'].append((window.index[0], window.index[-1]))
19
20
         return patterns
21
22
     patterns = find_patterns(minmax)
23
     patterns
get patterns hosted with ♥ by GitHub
                                                                                                view raw
```

And a plot for visual confirmation:





As you can see, we are getting more patterns than we need. Our params (smoothing and window range) are too sensitive for this timeframe (60 minutes).

Step 4.) Reorganize and iterate to find best params

In order to find the best params, I reorganized my code into functions and iterated through multiple stocks, smoothing, and window parameters.

```
1
 2
     def plot_minmax_patterns(prices, max_min, patterns, stock, window, ema):
         incr = str((prices.index[1] - prices.index[0]).seconds/60)
 4
         if len(patterns) == 0:
             pass
 8
         else:
             num pat = len([x for x in patterns.items()][0][1])
             f, axes = plt.subplots(1, 2, figsize=(16, 5))
10
11
             axes = axes.flatten()
             prices = prices.reset index()['close']
             axes[0].plot(prices )
13
             axes[0].scatter(max min.index, max min, s=100, alpha=.3, color='orange')
             axes[1].plot(prices )
15
             for name, end day nums in patterns.items():
                 for i, tup in enumerate(end day nums):
17
                     sd = tup[0]
19
                     ed = tup[1]
                     axes[1].scatter(max min.loc[sd:ed].index,
21
                                    max min.loc[sd:ed].values,
                                    s=200, alpha=.3)
23
                     plt.yticks([])
             plt.tight layout()
             plt.title('{}: {}: EMA {}, Window {} ({} patterns)'.format(stock, incr, ema, window, num
25
27
     def get_results(prices, max_min, pat, stock, ema_, window_):
28
         incr = str((prices.index[1] - prices.index[0]).seconds/60)
         #fw list = [1, 12, 24, 36]
```

```
6/20/2021
                fw_list = [1, 2, 3]
      32
      33
                results = []
       34
                if len(pat.items()) > 0:
                    end_dates = [v for k, v in pat.items()][0]
                    for date in end_dates:
       36
      37
                        param_res = {'stock': stock,
                                      'increment': incr,
       38
       39
                                      'ema': ema ,
                                      'window': window_,
      40
                                      'date': date}
      41
      42
                        for x in fw list:
                            returns = (prices['close'].pct_change(x).shift(-x).reset_index(drop=True).dropna
      43
      44
                            try:
                                 param_res['fw_ret_{}'.format(x)] = returns.loc[date[1]]
      45
      46
                            except Exception as e:
                                 param_res['fw_ret_{}'.format(x)] = e
      47
                        results.append(param_res)
      48
                else:
      49
      50
                    param_res = {'stock': stock,
                                  'increment': incr,
      51
      52
                                  'ema': ema_,
                                  'window': window_,
      53
      54
                                  'date': None}
                    for x in fw list:
      55
                        param_res['fw_ret_{}'.format(x)] = None
      56
      57
                    results.append(param_res)
      58
                return pd.DataFrame(results)
      59
      60
            def screener(stock_data, ema_list, window_list, plot, results):
      61
                all results = pd.DataFrame()
      63
      64
                for stock in stock_data:
                    prices = stock_data[stock]
      65
      66
                    for ema_ in ema_list:
      67
                        for window_ in window_list:
      68
                            max_min = get_max_min(prices, smoothing=ema_, window_range=window_)
      69
       70
                            pat = find patterns(max min)
      71
                            if plot == True:
      72
      73
                                 plot_minmax_patterns(prices, max_min, pat, stock, window_, ema_)
      74
                            if results == True:
       75
                                 all necults - nd concat/[all necults got necults/nnices
```

Run the above like so:

```
'GOOG', 'GS', 'LNKD',
stock_data = get_stock_data(stocklist, 5)
Getting stock data: 100% | 14/14 [00:20<00:00, 1.36s/it]
# Run different timeframes here without requesting new data. '5T' = 5 minutes, '60T' = 1 hour, '12
OT' = 2 hours, etc.
resampled_stock_data = resample(stock_data, '360T')
# Run the screener on ema_list and window_list, plotting results.
ema_list = [3, 10, 20, 30, ]
window_list = [3, 10, 20, 30, ]
results = screener(resampled_stock_data, ema_list, window_list, plot=True, results=True)
                                                                   MSFT: 360.0: EMA 3, Window 3 (1 patterns)
108
104
 102
 98
                                                                   MMM: 360.0: EMA 3, Window 3 (1 patterns)
197.5
195.0
192.5
190.0
187.5
185.0
182.5
180.0
177.5
```

Now we can see how our timeframes, patterns, and params are playing out!

Step 5.) Go live!

To use this live, I made the following changes to screener():

And ran like so:

```
stocklist = ['AA', 'AAL', 'AAPL', 'AMZN'] # Long list of stocks here
stock_data = get_stock_data(stocklist, 2)

resampled_stock_data = resample(stock_data, '360T')

ema_list = [5]
window_list = [5]

results = screener(resampled_stock_data, ema_list, window_list)

for x in results:
    api.submit_order(x, 100, 'buy', 'market', 'day')
```

Finding the right params for your pattern to play out may take experimentation. See the results() function in the notebook to confirm whether your patterns have a positive edge or not.



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