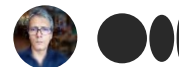


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ALGORITHMIC TRADING SIMPLIFIED

Evaluate Price Oscillator Stock Trading Strategy

Algorithmic Trading and PnL Evaluation



Sarit Maitra Apr 14 · 10 min read ★

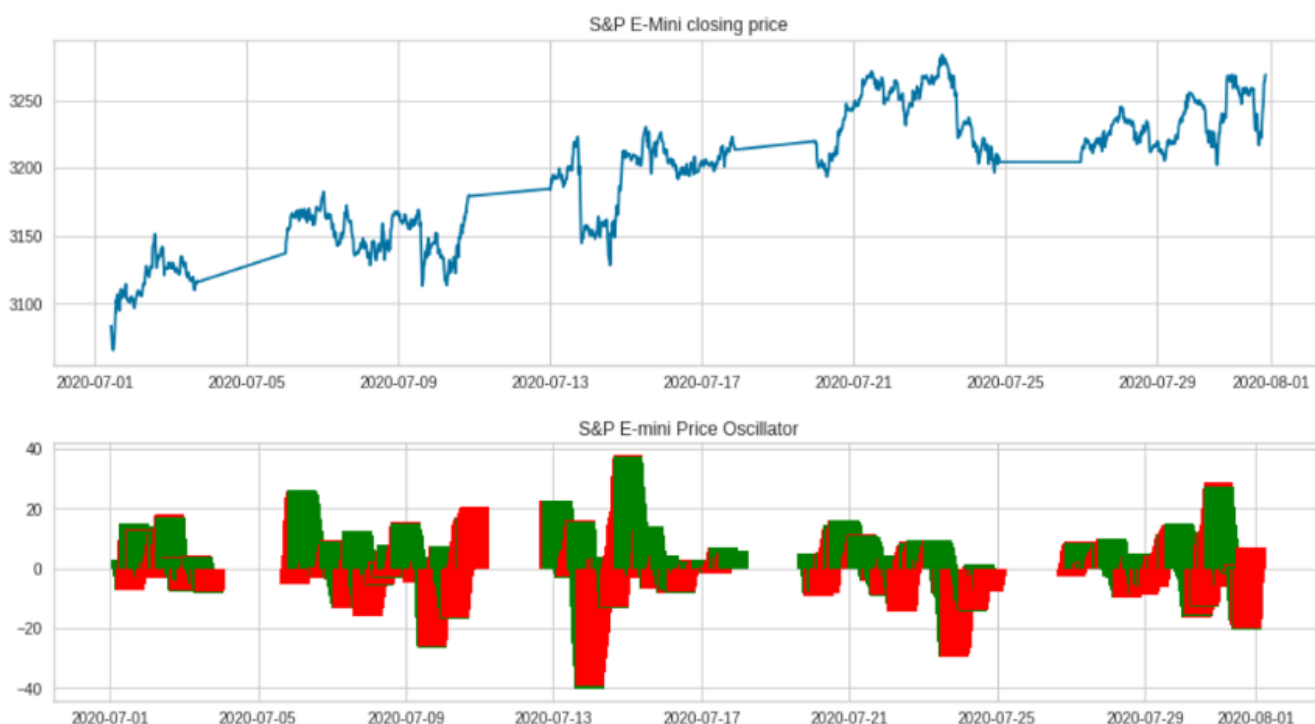


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Price Oscillator is a technical indicator calculating the difference between two price ^{*} moving averages. It is quite similar to MACD indicator. Crossovers of two moving

averages correspond to crossovers of price oscillator [PO (MACD)] and zero central signal line around it oscillates.

To understand price oscillator, we need to know the exponential moving averages (EMA) concept. In simple term, EMA is the average price over a certain number of days, with more recent days weighted more heavily i.e. exponentially. EMA needs to be calculated over a period of appropriate length to maximize meaningful data while minimizing random movement. There are some basic thumb rule as 26 days for long EMA and 13 days (well between 9–13 days) for short EMA. However, these are arbitrary and can be adjusted based on strategic movement.

Now, price oscillator is: $(\text{ema}(\text{short}) - \text{ema}(\text{long})) / \text{ema}(\text{long})$. So, essentially, we're looking at the short-term average as a fraction of the long-term average; hence, price oscillator.

Also, because the price oscillator compares two EMAs, we are able to compare movements through different time frames. Here, we will experiment with absolute price oscillator and develop a PnL program.

We will experiment with SP e-mini futures index with 15 min frequency data set.

```
def SPFutures():
    df = pd.read_csv("July_Futures_15min.csv")
    df.set_index('timestamp', inplace=True)
    df.sort_index(ascending=True, inplace=True)
    df.index = pd.to_datetime(df.index)
    df = df[df['volume'] > 0]
    df.sort_index(ascending=True, inplace=True)
    df.drop_duplicates(inplace=True) # dropping duplicates if any
    df.fillna(method='pad', inplace=True)
    return (df)

df = SPFutures()
print(df.head())
```

timestamp	es_close	es_open	es_high	es_low	volume
2020-07-01 00:00:00	3081.169643	3081.116071	3081.660714	3080.625000	147.857143
2020-07-01 00:15:00	3081.633333	3081.533333	3082.000000	3081.158333	72.133333
2020-07-01 00:30:00	3084.941667	3084.741667	3085.308333	3084.391667	85.733333
2020-07-01 00:45:00	3084.750000	3084.850000	3085.191667	3084.366667	69.400000
2020-07-01 01:00:00	3085.016667	3084.801667	3085.541667	3084.435000	80.133333

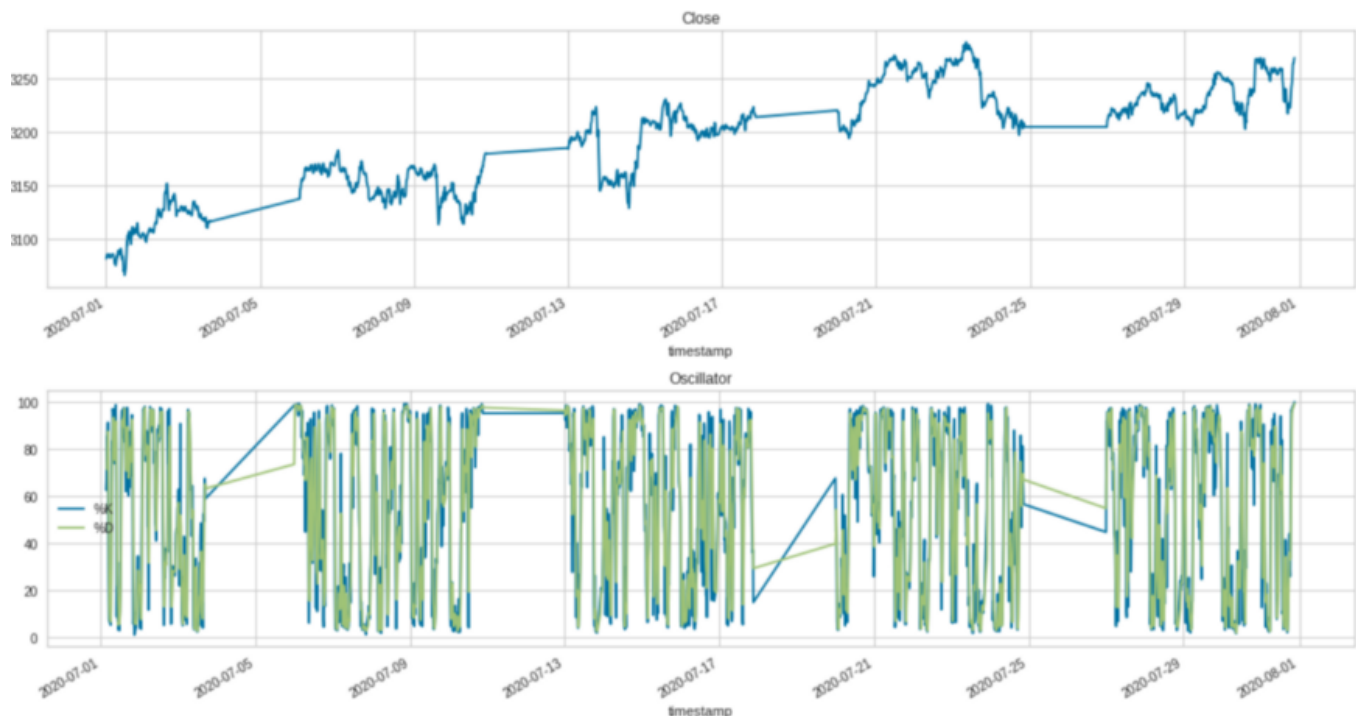
Stochastic Oscillator:

Let us first see how a simple Stochastic Oscillator based strategy works with the given data set and plot an equity curve to display cumulative returns.

```
d = df.copy()

# Create the "L14" column in the DataFrame
d['L_14'] = d['es_low'].rolling(window=14).min()
# Create the "H14" column in the DataFrame
d['H_14'] = d['es_high'].rolling(window=14).max()
# Create the "%K" column in the DataFrame
d['%K'] = 100 * ((d['es_close'] - d['L_14']) / (d['H_14'] - d['L_14']))
# Create the "%D" column in the DataFrame
d['%D'] = d['%K'].rolling(window=3).mean()

# plot
fig, axes = plt.subplots(nrows=2, ncols=1, figsize=(15,10))
d['es_close'].plot(ax=axes[0]); axes[0].set_title('Close')
d[['%K', '%D']].plot(ax=axes[1]); axes[1].set_title('Oscillator')
plt.tight_layout(); plt.show()
```



Let us run a simple back test and draw an equity curve to check the performance of the strategy. Here, our strategy is:

- a sell is initiated when the %K line crosses down through the %D line and the value of the oscillator is above 75.
- exit from sell signal is given when the %K line crosses back up through the %D line

Likewise:

- buy is initiated when the %K line crosses up through the %D line and the value of the oscillator is below 25
- exit from buy signal is given when the %K line crosses back down through the %D line

Let us back-test the goodness-of-fit of our strategy.

```
d['enter_sell'] = ((d['%K'] < d['%D']) & (d['%K'].shift(1) >
d['%D'].shift(1))) & (d['%D'] > 75)

d['exit_sell'] = ((d['%K'] > d['%D']) & (d['%K'].shift(1) <
d['%D'].shift(1)))

d['short'] = np.nan
d.loc[d['enter_sell'], 'short'] = -1
d.loc[d['exit_sell'], 'short'] = 0

#Set initial position on day 1 to flat
d['short'][0] = 0
d['short'] = d['short'].fillna(method='pad')

d['enter_buy'] = ((d['%K'] > d['%D']) & (d['%K'].shift(1) <
d['%D'].shift(1))) & (d['%D'] < 25)

d['exit_buy'] = ((d['%K'] < d['%D']) & (d['%K'].shift(1) >
d['%D'].shift(1)))

d['long'] = np.nan
d.loc[d['enter_buy'], 'long'] = 1
d.loc[d['exit_buy'], 'long'] = 0

d['long'][0] = 0
d['long'] = d['long'].fillna(method='pad')
```

```
# Add Long and Short positions together to get final strategy
position (1 for long, -1 for short and 0 for flat)

d['Position'] = d['long'] + d['short']
```

Visualization:

```
d['market_returns'] = d['es_close'].pct_change()
d['strategy_returns'] = d['market_returns'] * d['Position'].shift(1)
d[['strategy_returns', 'market_returns']].cumsum().plot()
plt.show()
```



So, here we have experimented with Stochastic Oscillator (SO) to find that out that, simple SO based trading strategy shows positive return for the month. However, looking at the market return plot, a simple buy and hold strategy would have been profitable too.

Let us experiment with absolute price oscillator on the same data set and explore more details by doing a back test. As discussed earlier, APO needs fast and slow moving EMAs to go forward.

Calculating Price Oscillator:

The Price Oscillator is simply the nine-day EMA, decreased and then divided by the 26-day EMA $\{[EMA(9day) - EMA(26day)] / EMA(26day)\}$.

Tradition and convention have deemed 26 days to be the dividing line between the short term and the “minor intermediate” term in the stock market, with the “very short” term lasting between five and 13 days.(Investopedia)

Parameters for EMA Calculation:

Let us consider 10 (fast EMA) and 40 (slow ema) parameters.

When we calculate MA, we sum up the closing prices over the past n-times and then divide by n. The EMA calculation uses a smoothing factor to place a higher weight on recent data points and is considered as much more efficient than the linear weighted average.

1. We calculate the smoothing factor: $multiplier = 2 / (EMA\ length + 1)$.
2. $EMA[0] = (close[0] - EMA[1]) * multiplier + EMA[1]$.

```
fastPeriod = 10
fastSmooth = 2 / (fastPeriod + 1)
fastEma = 0
fastEma_val = []

slowPeriod = 40
slowSmooth = 2 / (slowPeriod + 1)
slowEma = 0
slowEma_val = []
poValues = []
```

We are also storing the price oscillator values to track computed Price Oscillator value signals. Let us calculate the values of the Price Oscillator and visualize the histogram.

```
def ema(price, period):
    ema = price.rolling(period).mean()
    return ema

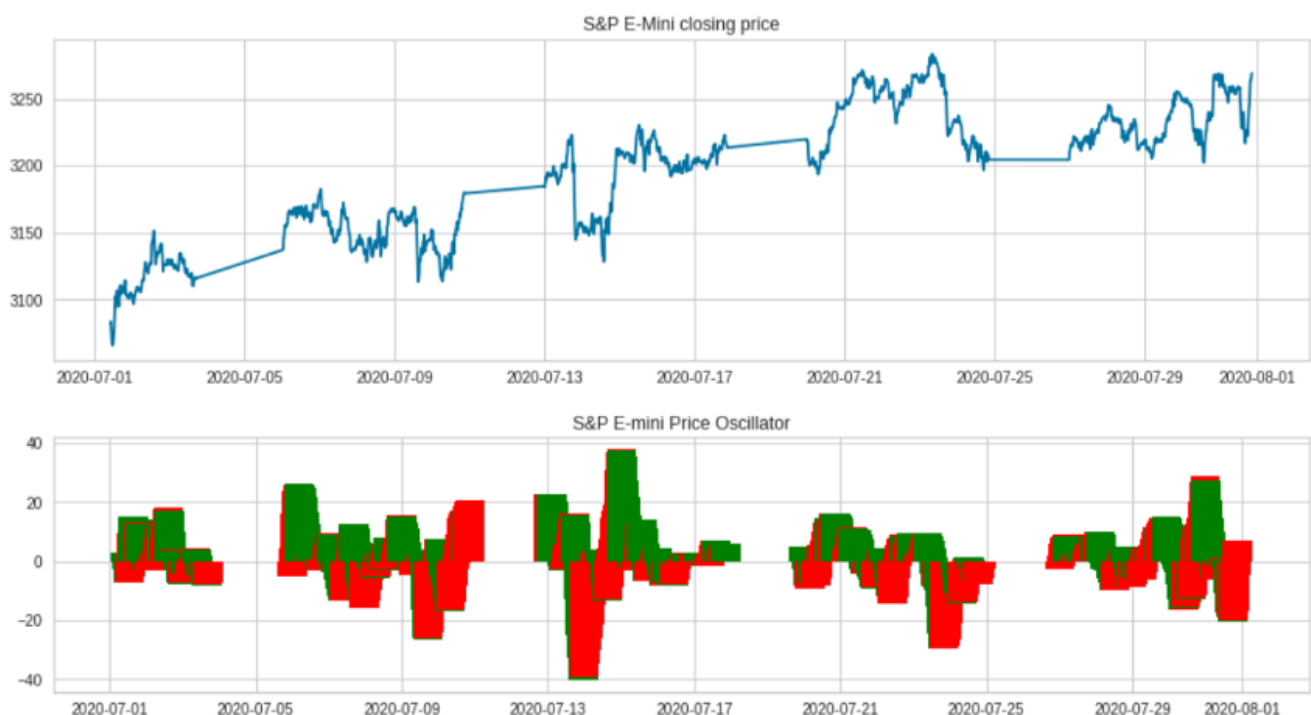
def po(price, period1, period2):
    median = price.rolling(2).median()
    short = ema(median, period1)
    long = ema(median, period2)
    po = short - long
    po_df = DataFrame(po).rename(columns = {'Close':'po'})
    return po_df

df['po'] = po(df['es_close'], 10, 40)
df.dropna(inplace=True)
plt.figure(figsize=(15,8))

ax1 = plt.subplot2grid((10,1), (0,0), rowspan = 5, colspan = 1)
ax2 = plt.subplot2grid((10,1), (6,0), rowspan = 4, colspan = 1)
ax1.plot(df['es_close'])
ax1.set_title('S&P E-Mini closing price')

for i in range(len(df)):
    if df['po'][i-1] > df['po'][i]:
        ax2.bar(df.index[i], df['po'][i], color = 'green')
    else:
        ax2.bar(df.index[i], df['po'][i], color = 'red')

ax2.set_title('S&P E-mini Price Oscillator')
plt.show()
```



We can see, whenever the current bar > previous bar, green bar is plotted on the chart, and, whenever the current bar < previous bar, red bar is plotted on the chart.

Trading strategy:

Let us create variables for trading strategy, position & pnl management.

- orders [] and positions[] are two containers we have created to track buy/sell order, +1 for buy order, -1 for sell order, 0 for no-action and track positions, +ve for long positions, -ve for short positions, 0 for flat/no position respectively.
- pnls[] for the sum of closed_pnl i.e. pnls already locked in and open_pnl i.e. pnls for open-position marked to market price
- last_buyPrice and last_sellPrice are the prices at which last buy or sell was made; these will be used to prevent over-trading at/around the same price.
- position is the current position of the trading strategy
- buy_sumPrice & sell_sumPrice are the summation of products of buy_trade_price and buy_trade_qty for every buy Trade made since last time being flat

```
orders = []
positions = []
pnls = []
last_buyPrice = 0
last_sellPrice = 0
position = 0
buy_sumPrice_qty = 0
buy_sum_qty = 0
sell_sumPrice_qty = 0
sell_sum_qty = 0
open_pnl = 0
closed_pnl = 0
```

Let is define some constants to define strategy behavior/thresholds.

```
po_buyEnter = -10
po_sellEnter = 10

min_price_movement_from_last_trade = 10
num_stocks_per_trade = 10
```



```
min_profit_to_close = 10 * num_stocks_per_trade
```

Let us define constants/variables that are used to compute standard deviation as a volatility measure.

Here our strategy is simple:

We will perform a sell trade at close_price if the following conditions are met:

- The PO trading signal value is above Sell-Entry threshold and the difference between last trade-price and current-price is different enough.
- We are long(+ve position) and either APO trading signal value is at or above 0 or current position is profitable enough to lock profit.

We will perform a buy trade at close_price if the following conditions are met:

- The PO trading signal value is below Buy-Entry threshold and the difference between last trade-price and current-price is different enough.
- We are short(-ve position) and either APO trading signal value is at or below 0 or current position is profitable enough to lock profit.

```
maPeriods = 20
priceHist = []

close = df['es_close'].copy()

for close_price in close:
    priceHist.append(close_price)
    if len(priceHist) > maPeriods:
        del (priceHist[0])

sma = stats.mean(priceHist)
variance = 0
for histPrice in priceHist:
    variance = variance + ((histPrice - sma) ** 2)

stdev = math.sqrt(variance / len(priceHist))
stdev_factor = stdev/15
if stdev_factor == 0:
    stdev_factor = 1
```

```

if (fastEma == 0):
    fastEma = close_price
    slowEma = close_price
else:
    fastEma = (close_price - fastEma) * smooth_fast * stdev_factor + fastEma
    slowEma = (close_price - slowEma) * smooth_slow * stdev_factor + slowEma

fastEma_val.append(fastEma)
slowEma_val.append(slowEma)

po = fastEma - slowEma
poValues.append(po)

if ((po > po_sellEnter * stdev_factor and abs(close_price - last_sellPrice) > min_price_movement_from_last_trade * stdev_factor)

or
(position > 0 and (po >= 0 or open_pnl > min_profit_to_close / stdev_factor))):
    orders.append(-1) # sell trade
    last_sellPrice = close_price
    position -= num_stocks_per_trade
    sell_sumPrice_qty += (close_price * num_stocks_per_trade) # update vwap sell-price
    sell_sum_qty += num_stocks_per_trade
    print( "SELL ", num_stocks_per_trade, " @ ", close_price, "Position => ", position)

elif ((po < po_buyEnter * stdev_factor and abs(close_price - last_buyPrice) > min_price_movement_from_last_trade * stdev_factor)
or
(position < 0 and (po <= 0 or open_pnl > min_profit_to_close / stdev_factor))):
    orders.append(+1) # buy trade
    last_buyPrice = close_price
    position += num_stocks_per_trade # increase position by the size of this trade

    buy_sumPrice_qty += (close_price * num_stocks_per_trade) # update the vwap buy-price
    buy_sum_qty += num_stocks_per_trade
    print( "BUY ", num_stocks_per_trade, " @ ", close_price, "POSITION: ", position ); print()
else:
    orders.append(0)
    positions.append(position)

open_pnl = 0
if position > 0:
    if sell_sum_qty > 0:

```

```

    open_pnl = abs(sell_sum_qty) * (sell_sumPrice_qty/sell_sum_qty
    - buy_sumPrice_qty/buy_sum_qty)

    open_pnl += abs(sell_sum_qty - position) * (close_price -
    buy_sumPrice_qty / buy_sum_qty)
    elif position < 0:
        if buy_sum_qty > 0:
            open_pnl = abs(buy_sum_qty) * (sell_sumPrice_qty/sell_sum_qty -
            buy_sumPrice_qty/buy_sum_qty)

            open_pnl += abs(buy_sum_qty - position) *
            (sell_sumPrice_qty/sell_sum_qty - close_price)

        else:
            closed_pnl += (sell_sumPrice_qty - buy_sumPrice_qty)
            buy_sumPrice_qty = 0
            buy_sum_qty = 0
            sell_sumPrice_qty = 0
            sell_sum_qty = 0
            last_buyPrice = 0
            last_sellPrice = 0

    print( "Open/PnL-> ", open_pnl, " Closed/PnL-> ", closed_pnl, "
    Total/PnL-> ", (open_pnl + closed_pnl) );
    pnls.append(closed_pnl + open_pnl)

```

```

Open/PnL-> 10656.333333333187 Closed/PnL-> 56315.41666666452 Total/PnL-> 66971.7499999977
BUY 10 @ 3222.258333333333 POSITION: -10

Open/PnL-> 11177.708333333121 Closed/PnL-> 56315.41666666452 Total/PnL-> 67493.12499999764
BUY 10 @ 3227.575 POSITION: 0

Open/PnL-> 0 Closed/PnL-> 61333.83333333103 Total/PnL-> 61333.83333333103
BUY 10 @ 3234.675 POSITION: 10

Open/PnL-> 0.0 Closed/PnL-> 61333.83333333103 Total/PnL-> 61333.83333333103
BUY 10 @ 3240.808333333333 POSITION: 20

Open/PnL-> 61.3333333333212 Closed/PnL-> 61333.83333333103 Total/PnL-> 61395.16666666436
Open/PnL-> 130.6666666667516 Closed/PnL-> 61333.83333333103 Total/PnL-> 61464.4999999977
BUY 10 @ 3246.766666666667 POSITION: 30

Open/PnL-> 180.50000000000637 Closed/PnL-> 61333.83333333103 Total/PnL-> 61514.33333333104
SELL 10 @ 3254.6583333333333 Position => 20
Open/PnL-> 278.1666666666606 Closed/PnL-> 61333.83333333103 Total/PnL-> 61611.99999999769
SELL 10 @ 3263.5916666666667 Position => 10
Open/PnL-> 595.9166666666667 Closed/PnL-> 61333.83333333103 Total/PnL-> 61929.74999999769
SELL 10 @ 3264.1 Position => 0
Open/PnL-> 0 Closed/PnL-> 61934.83333333103 Total/PnL-> 61934.83333333103
Open/PnL-> 0 Closed/PnL-> 61934.83333333103 Total/PnL-> 61934.83333333103
Open/PnL-> 0 Closed/PnL-> 61934.83333333103 Total/PnL-> 61934.83333333103

```

Preparing the data frame from the trading strategy results and visualizes the results:

```

data = df.copy()
data = data.assign(closePrice = pd.Series(close, index =df.index))
data = data.assign(short = pd.Series(fastEma_val, index = df.index))
data = data.assign(long = pd.Series(slowEma_val, index = df.index))
data = data.assign(priceOs = pd.Series(poValues, index = df.index))
data = data.assign(trades = pd.Series(orders, index = data.index))
data = data.assign(position = pd.Series(positions, index =
data.index))
data = data.assign(pnl = pd.Series(pnls, index=data.index))
data =
data[['closePrice','short','long','priceOs','trades','position','pnl'
]]
data.head(2)

```

	closePrice	short	long	priceOs	trades	position	pnl
timestamp							
2020-07-01 00:00:00	3081.169643	3081.169643	3081.169643	0.000000	0	0	0.0
2020-07-01 00:15:00	3081.633333	3081.170946	3081.169992	0.000953	0	0	0.0

```

plt.figure(figsize = (10, 6))
data['closePrice'].plot(color = 'gray',lw = 1., legend=True)
data['short'].plot(color = 'green', lw = 1., legend=True)
data['long'].plot(color = 'red', lw = 1., legend=True)
plt.title("E-Mini S &P 500 Futures - Market price")
plt.show()

```



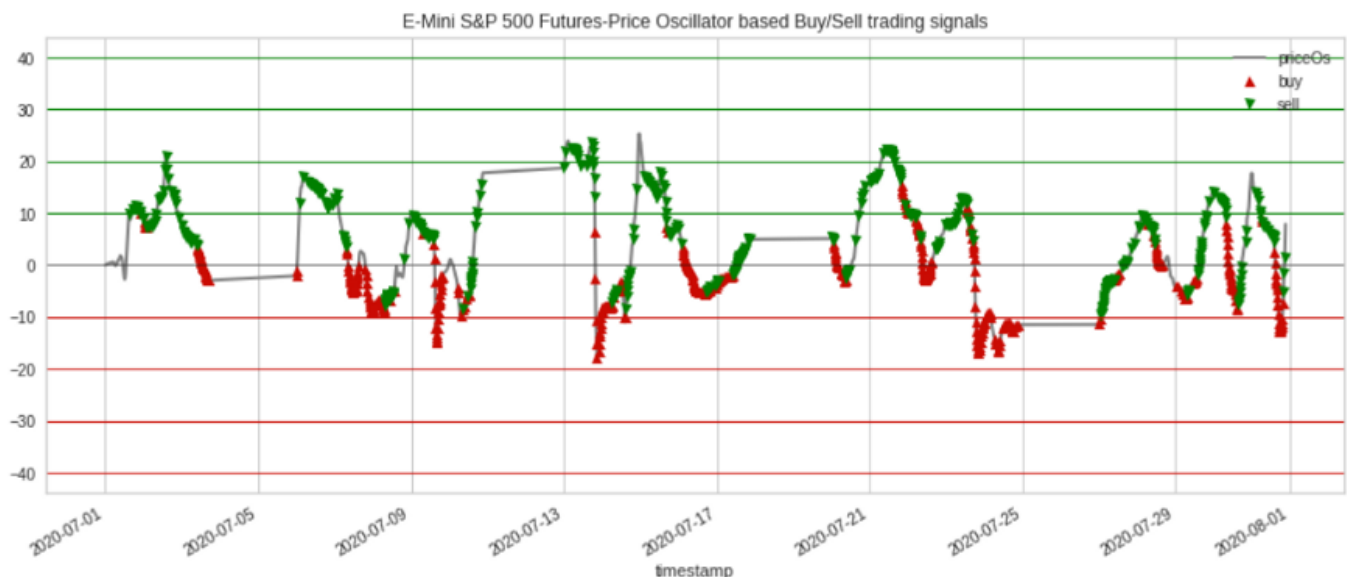


```
plt.figure(figsize = (15, 6))
data['priceOs'].plot(color='gray', lw=2.,
legend=True)plt.plot(data.loc[data.trades == 1 ].index,
data.priceOs[data.trades == 1 ], color='r', lw=0, marker='^',
markersize=7, label='buy')
plt.plot(data.loc[data.trades == -1 ].index,
data.priceOs[data.trades == -1 ], color='green', lw=0, marker='v',
markersize=7, label='sell')
plt.axhline(y=0, lw=0.5, color='k')

for i in range(po_buyEnter, po_buyEnter * 5, po_buyEnter ):
    plt.axhline(y=i, lw=1., color='r')

for i in range(po_sellEnter, po_sellEnter * 5, po_sellEnter ):
    plt.axhline(y=i, lw=1., color='green')

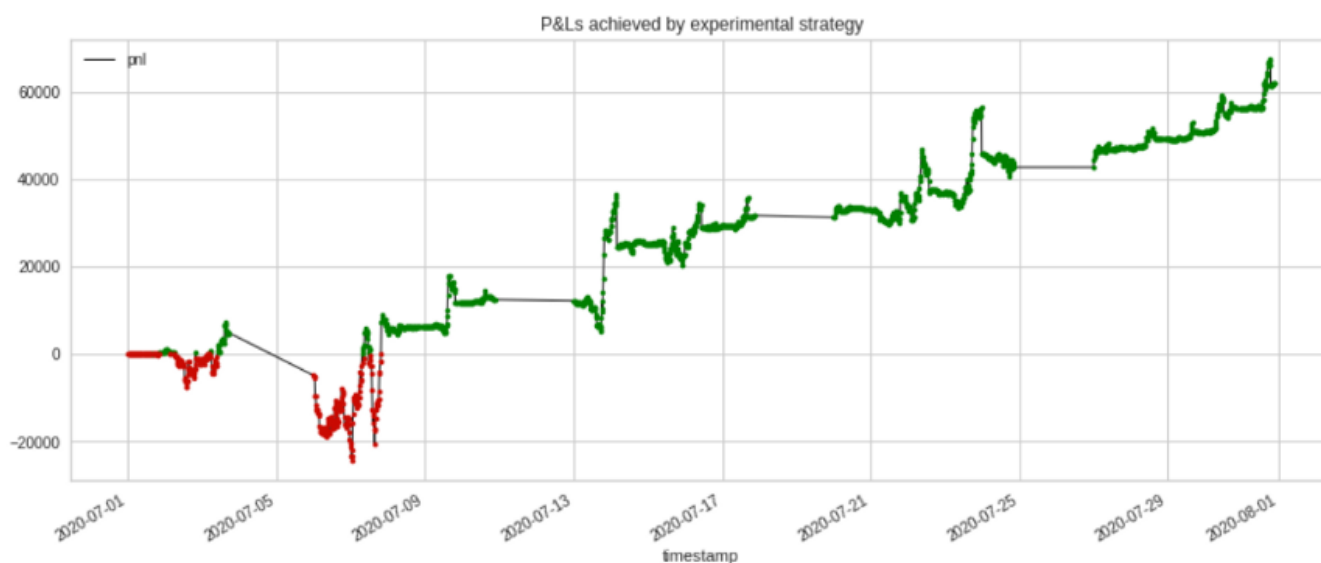
plt.legend()
plt.title('E-Mini S&P 500 Futures-Price Oscillator based Buy/Sell
trading signals')
plt.show()
```



All signals are generated by the signal line crossing above zero-level which is considered bullish while crossing below zero-level is deemed bearish. As short-term momentum

increases or decreases and eclipses long-term momentum the signal is generated. As is common with most oscillators, divergence in the price and indicator can also alert investors to early turnarounds.

```
plt.figure(figsize = (15, 6))
data['pnl'].plot(color='k', lw=1., legend=True)
plt.plot(data.loc[data.pnl > 1 ].index, data.pnl[data.pnl > 1 ],
color='green', lw=0, marker='.')
plt.plot(data.loc[data.pnl < 1 ].index, data.pnl[data.pnl < 1 ],
color='r', lw=0, marker='.')
plt.legend()
plt.title('P&Ls achieved by experimental strategy')
plt.show()
```



Key takeaways:

It is always advisable to experiment with other technical indicators prior to any official decision. As we have seen so far that, Price Oscillator is a momentum oscillator and measures the difference between two moving averages. Almost similar with MACD, the Price Oscillator can be shown with a signal line, a histogram and a centerline. Along with Price Oscillator, strategies e.g. candlestick patterns, triple EMA, volume indicator etc. are good to experiment. Volume Indicator is preferred by many because of the simple fact that, high volumes confirm well the entry signals generated by the price oscillator indicator, low volumes hint that the trend is being exhausted.

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