

Technical Analysis using Bollinger Bands

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Introduction to Bollinger Bands

- Useful tool for analyzing trend strength & monitoring when a reversal may be occurring
- Three bands:
 - Moving average of the price
 - Two standard deviations above moving average
 - Two standard deviations below moving average
- Movement above and below these bands signal buy or sell strategies to traders
- Bands contract and widen with volatility



- Upper band - level that is statistically high or expensive
- Lower band - level that is statistically low or cheap
- Bollinger bandwidth correlates to the volatility of the market

Data Prep

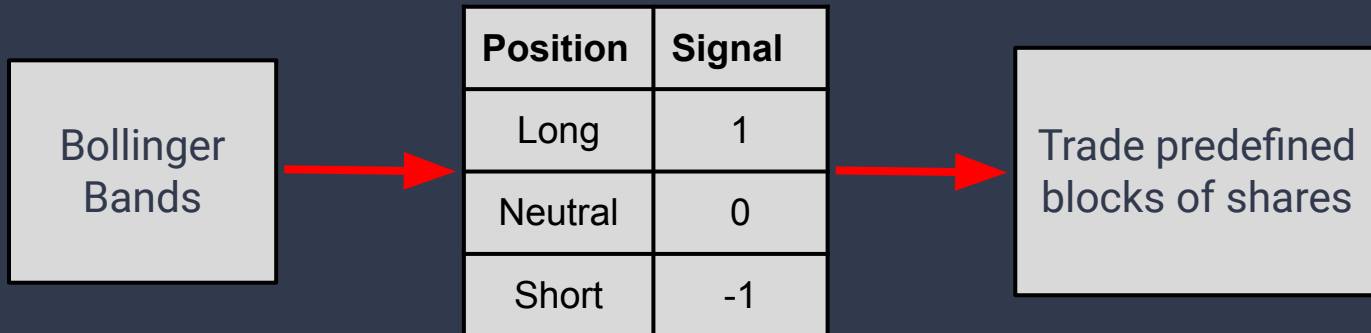
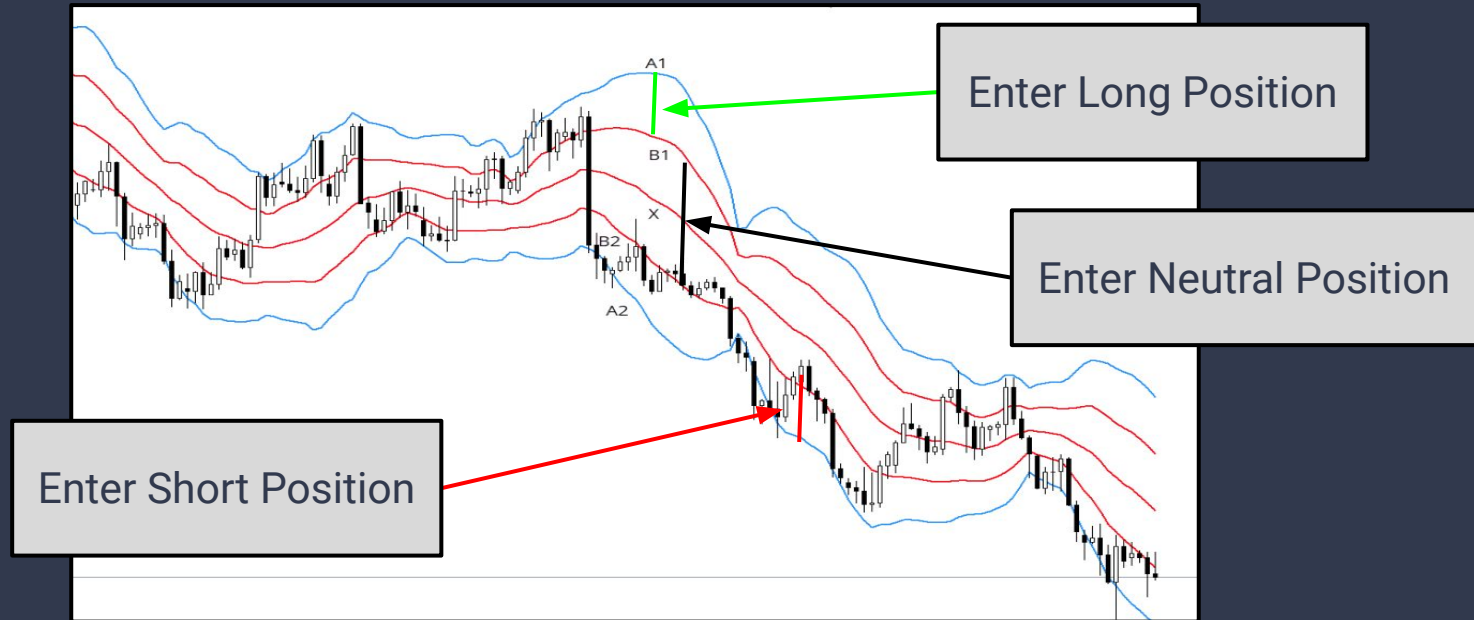
- Price data imported from SHIFT for a basket of 5 stocks
 - AAPL, AMZN, BP, COP, XOM
- Bollinger bands are calculated using a window of 10 periods
 - 10-period Moving Average of stock prices
 - 10-period Standard Deviation of stock prices
 - Upper and Lower Bands = $MA \pm 2 * StDev$
 - Upper and Lower Mid Bands = $MA \pm StDev$



Trading Strategy

Utilizing the Bollinger bands created, we are able to formulate buy/sell signals for a momentum strategy that our algorithm will trade off of.

- Long:
 - Last Price \in [Upper Mid, Upper Band]
- Neutral:
 - Last Price \in [Lower Mid, Upper Mid]
- Short:
 - Last Price \in [Lower Mid, LowerBand]



```
# Define variables to track the portfolio's P&L and portfolio value
```

```
portfolio_pnl = 0
```

```
portfolio_value = 1000000
```

```
# For each ticker
```

```
for ticker in stocks:
```

```
    # Filter the initial data to the current ticker
```

```
    fltr = BS_spread["Ticker"] == ticker
```

```
    # Calculate the bollinger bands for that stock with a window of 10 periods
```

```
    df = get_bollinger_bands(BS_spread[fltr], 10).reset_index()
```

```
    # Plot the bollinger bands
```

```
    plot_bollinger_bands(df, ticker)
```

```
    # Calculate the number of shares to buy/sell at a time
```

```
    n_shares = round((portfolio_value / len(stocks)) / df.loc[0, "Last"], 0)
```

```
    # Calculate the signal from the previously defined logic
```

```
    df = get_signal(df)
```

```
    # Trade based on that signal and record the point-in-time P&L
```

```
    df = get_pnl(df, n_shares)
```

```
    # Calculate the total P&L for this stock
```

```
    total_pnl = round(df.loc[df.shape[0] - 1, "PnL"], 2)
```

```
    # Add that P&L to the total portfolio
```

```
    portfolio_pnl += total_pnl
```

```
    # Plot that ish
```

```
    cols1 = ["PnL"]
```

```
    cols2 = ["Signal", "signal change"]
```

```
    # visualize $$$
```

```
    ax1 = df[cols1].plot(title = ticker + " Strategy P&L = $" + str(total_pnl), figsize=(12, 7))
```

```
    ax2 = df[cols2].plot(title = "Signal and Signal Change", figsize=(12, 7))
```

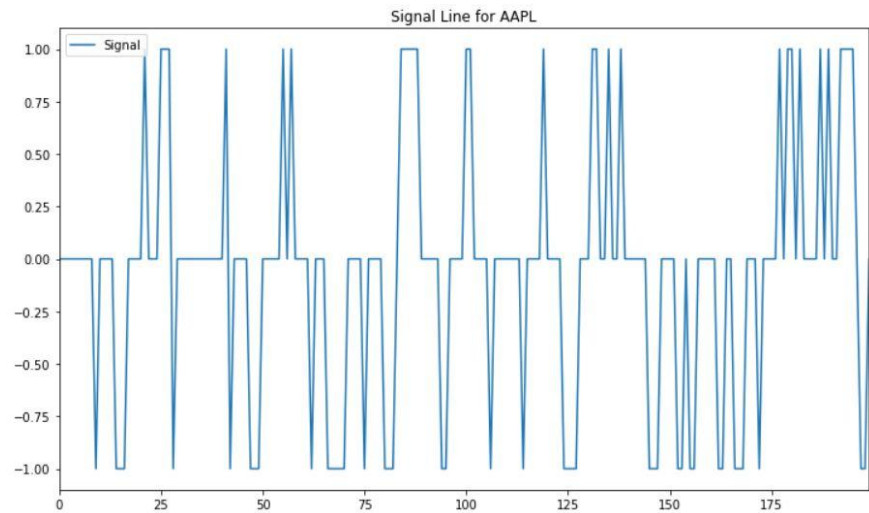
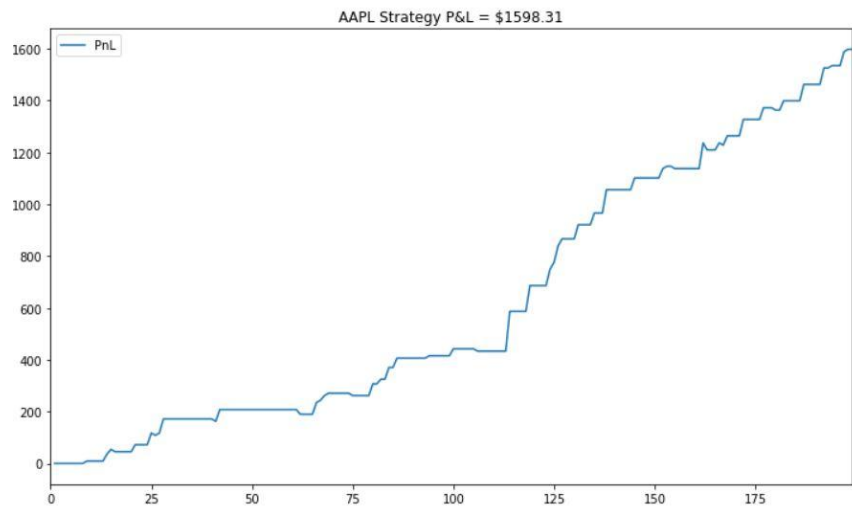
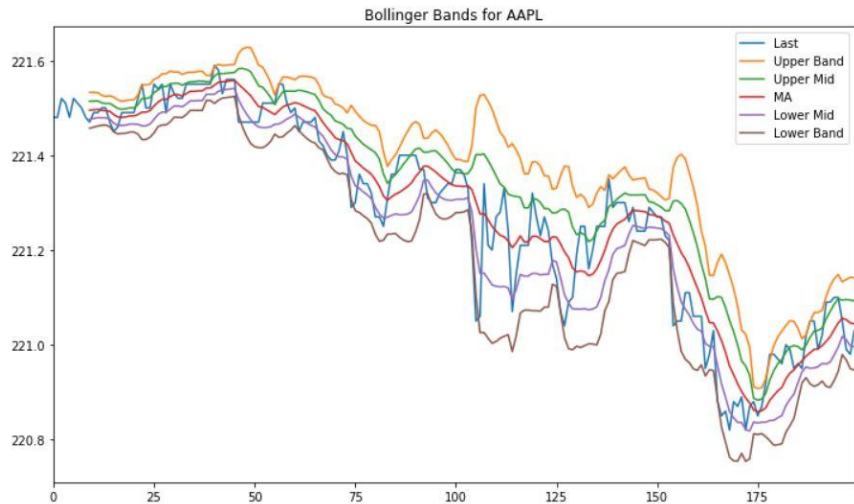
Strategy Execution

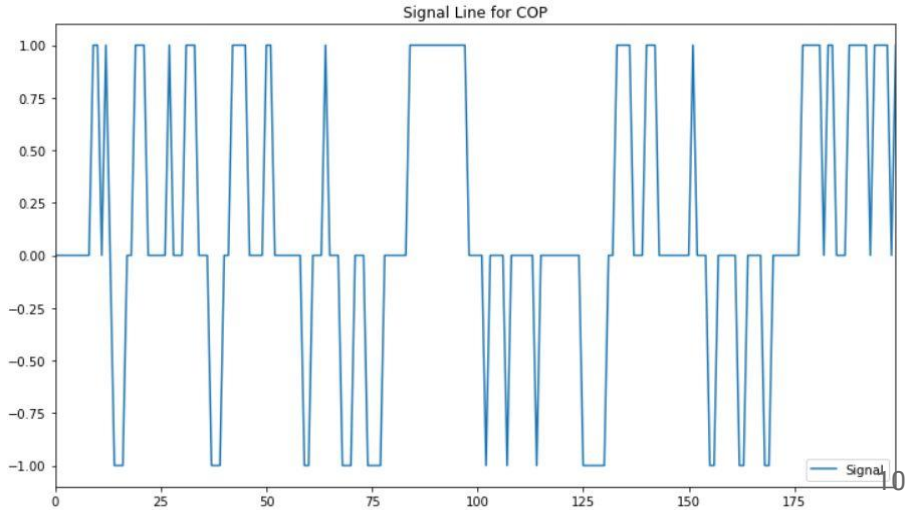
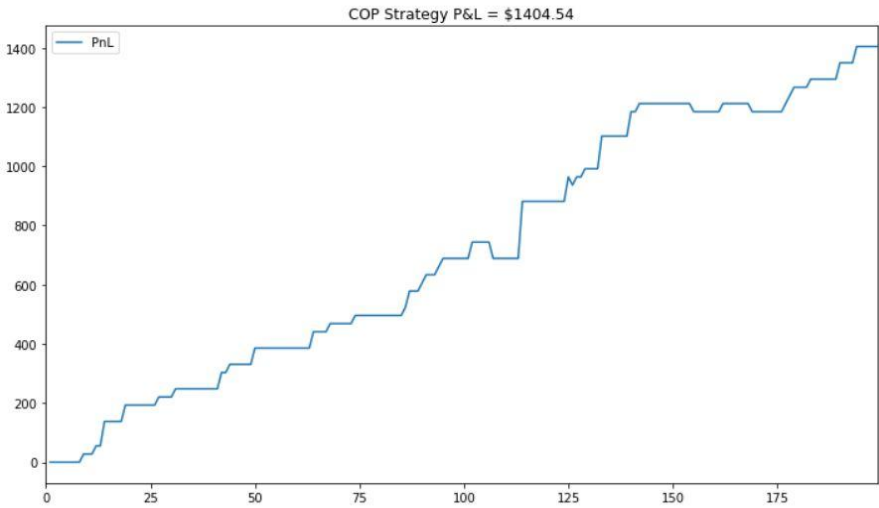
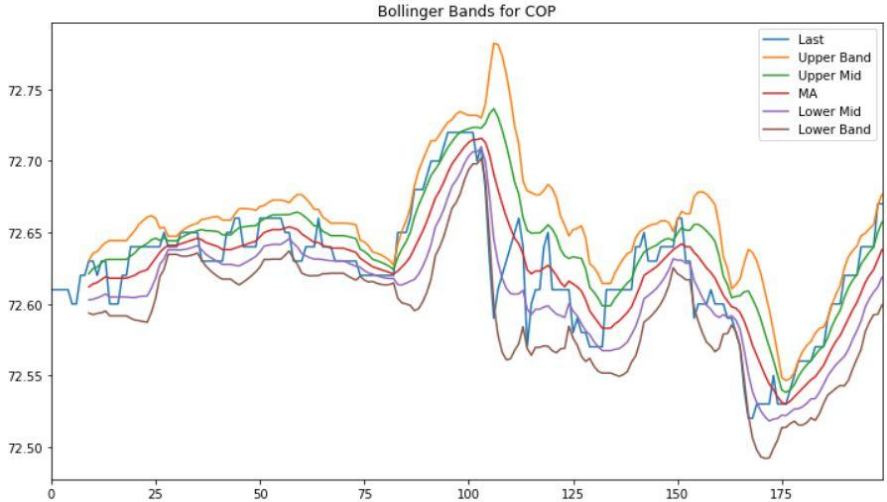
1. Calculate Bollinger Bands
2. Estimate # shares to trade
3. Calculate the signal
4. Execute trades using the signal and produce P&L
5. Summarize data

Results

- Assume a portfolio of \$1,000,000
 - Divide equally across all stocks
- Long/Short blocks of shares at a time
 - Makes decision every 3 seconds
- 0.63% return in 10 minutes

	Profit/Loss
AAPL	\$1,598.31
AMZN	\$1,763.62
BP	\$361.28
COP	\$1,404.54
XOM	\$1,156.67
Total	\$6,284.42





Conclusion

- Our approach to a Bollinger band trading strategy proves to be profitable in a market simulation for our basket of stocks
- Our strategy is very active in the market throughout the trading period
- Future work:
 - Expand to a larger basket of stocks or a more diverse portfolio
 - Test strategy over longer time frame
 - Test/Deploy strategy in a real market environment

Questions?