

PivotPoint

September 29, 2021

1 Pivot Point Indicators

[https://en.wikipedia.org/wiki/Pivot_point_\(technical_analysis\)](https://en.wikipedia.org/wiki/Pivot_point_(technical_analysis))

https://stockcharts.com/school/doku.php?id=chart_school:technical_indicators:pivot_points

```
[1]: import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd

import warnings
warnings.filterwarnings("ignore")

# fix_yahoo_finance is used to fetch data
import fix_yahoo_finance as yf
yf.pdr_override()
```

```
[2]: # input
symbol = 'AMD'
market = 'SPY'
start = '2017-01-01'
end = '2019-01-01'

# Read data
dataset = yf.download(symbol,start,end)
benchmark = yf.download(market,start,end)

# View Columns
dataset.head()
benchmark.head()
```

```
[*****100%*****] 1 of 1 downloaded
```

```
[*****100%*****] 1 of 1 downloaded
```

```
[2]:          Open          High          Low          Close  Adj Close  \
Date
2017-01-03  225.039993  225.830002  223.880005  225.240005  215.875137
```

2017-01-04	225.619995	226.750000	225.610001	226.580002	217.159409
2017-01-05	226.270004	226.580002	225.479996	226.399994	216.986893
2017-01-06	226.529999	227.750000	225.899994	227.210007	217.763229
2017-01-09	226.910004	227.070007	226.419998	226.460007	217.044403

Date	Volume
2017-01-03	91366500
2017-01-04	78744400
2017-01-05	78379000
2017-01-06	71559900
2017-01-09	46265300

```
[3]: dataset['Returns'] = dataset['Adj Close'].pct_change().dropna()
```

2 Stock Pivot Points

2.1 Standard Pivot Points

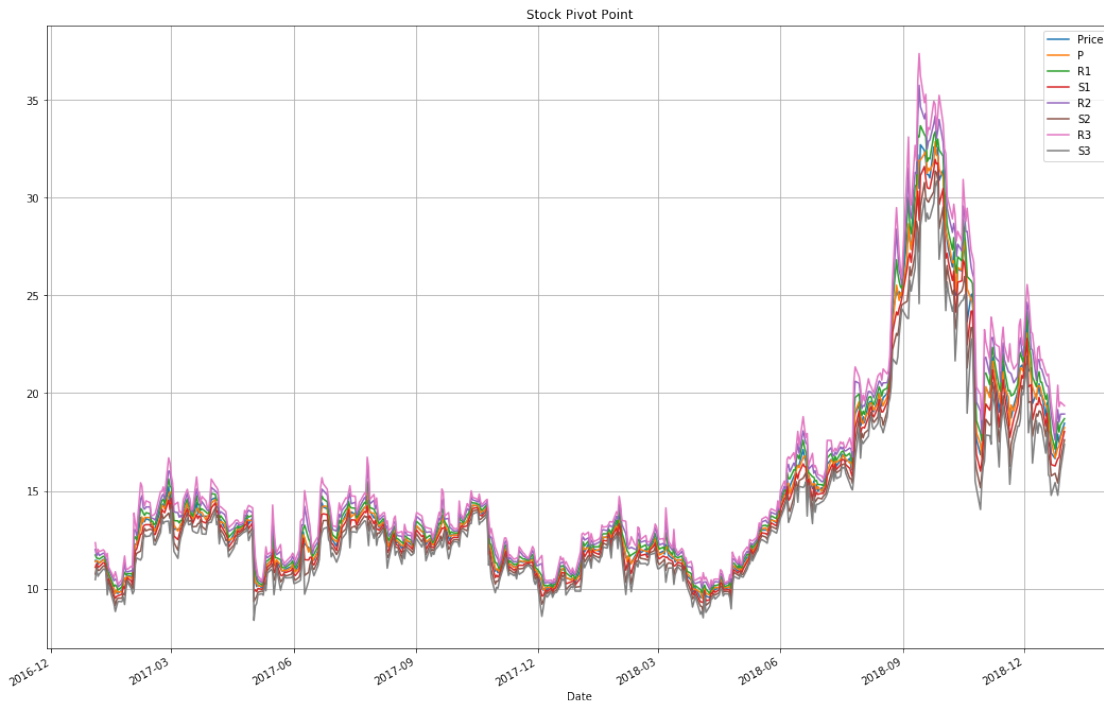
```
[4]: # Floor Pivot Points (Basic Pivot Points) - Support and Resistance
# https://www.investopedia.com/trading/using-pivot-points-for-predictions/
PP = pd.Series((dataset['High'] + dataset['Low'] + dataset['Close']) / 3)
R1 = pd.Series(2 * PP - dataset['Low'])
S1 = pd.Series(2 * PP - dataset['High'])
R2 = pd.Series(PP + dataset['High'] - dataset['Low'])
S2 = pd.Series(PP - dataset['High'] + dataset['Low'])
R3 = pd.Series(dataset['High'] + 2 * (PP - dataset['Low']))
S3 = pd.Series(dataset['Low'] - 2 * (dataset['High'] - PP))
R4 = pd.Series(dataset['High'] + 3 * (PP - dataset['Low']))
S4 = pd.Series(dataset['Low'] - 3 * (dataset['High'] - PP))
R5 = pd.Series(dataset['High'] + 4 * (PP - dataset['Low']))
S5 = pd.Series(dataset['Low'] - 4 * (dataset['High'] - PP))
P = pd.Series((dataset['Open'] + (dataset['High'] + dataset['Low'] +
↪ dataset['Close']))) / 4) # Opening Price Formula
psr = {'P':P, 'R1':R1, 'S1':S1, 'R2':R2, 'S2':S2, 'R3':R3, 'S3':S3, 'R4':R4,
↪ 'S4':S4, 'R5':R5, 'S5':S5}
PSR = pd.DataFrame(psr)
dataset = dataset.join(PSR)
print(dataset.head())
```

Date	Open	High	Low	Close	Adj Close	Volume	Returns	\
2017-01-03	11.42	11.65	11.02	11.43	11.43	55182000	NaN	
2017-01-04	11.45	11.52	11.24	11.43	11.43	40781200	0.000000	
2017-01-05	11.43	11.69	11.23	11.24	11.24	38855200	-0.016623	
2017-01-06	11.29	11.49	11.11	11.32	11.32	34453500	0.007117	
2017-01-09	11.37	11.64	11.31	11.49	11.49	37128000	0.015018	

	P	R1	R2	R3	R4	R5 \
Date						
2017-01-03	11.3800	11.713333	11.996667	12.343333	12.69	13.036667
2017-01-04	11.4100	11.553333	11.676667	11.833333	11.99	12.146667
2017-01-05	11.3975	11.543333	11.846667	12.003333	12.16	12.316667
2017-01-06	11.3025	11.503333	11.686667	11.883333	12.08	12.276667
2017-01-09	11.4525	11.650000	11.810000	11.980000	12.15	12.320000

	S1	S2	S3	S4	S5
Date					
2017-01-03	11.083333	10.736667	10.453333	10.17	9.886667
2017-01-04	11.273333	11.116667	10.993333	10.87	10.746667
2017-01-05	11.083333	10.926667	10.623333	10.32	10.016667
2017-01-06	11.123333	10.926667	10.743333	10.56	10.376667
2017-01-09	11.320000	11.150000	10.990000	10.83	10.670000

```
[5]: # labels = ['Price', 'P', 'R1', 'S1', 'R2', 'S2', 'R3', 'S3']
pivot_point = pd.concat([dataset['Adj Close'], P, R1, S1, R2, S2, R3, S3], axis=1)
    ↪ plot(figsize=(18,12), grid=True)
plt.title('Stock Pivot Point')
plt.legend(['Price', 'P', 'R1', 'S1', 'R2', 'S2', 'R3', 'S3'], loc=0)
plt.show()
```

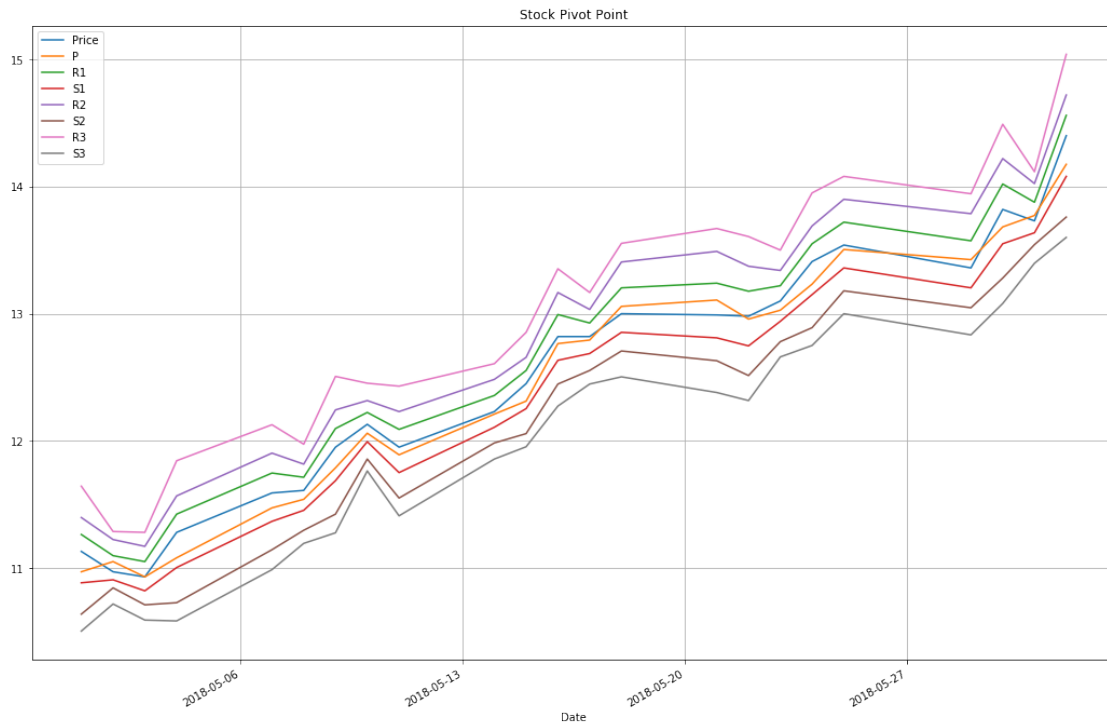


```
[6]: dataset['Adj Close']['2018-05-01':'2018-06-01']
```

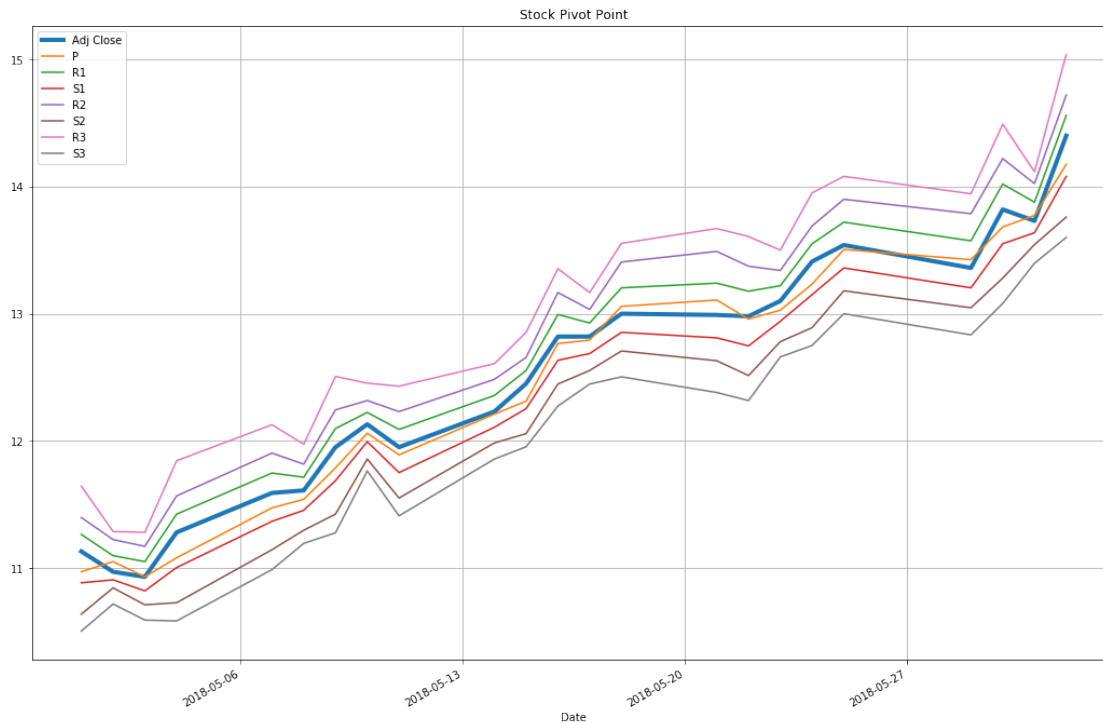
```
[6]: Date
2018-05-01    11.13
2018-05-02    10.97
2018-05-03    10.93
2018-05-04    11.28
2018-05-07    11.59
2018-05-08    11.61
2018-05-09    11.95
2018-05-10    12.13
2018-05-11    11.95
2018-05-14    12.23
2018-05-15    12.45
2018-05-16    12.82
2018-05-17    12.82
2018-05-18    13.00
2018-05-21    12.99
2018-05-22    12.98
2018-05-23    13.10
2018-05-24    13.41
2018-05-25    13.54
2018-05-29    13.36
2018-05-30    13.82
2018-05-31    13.73
2018-06-01    14.40
Name: Adj Close, dtype: float64
```

```
[7]: date_range = dataset[['Adj_
    ↳Close', 'P', 'R1', 'S1', 'R2', 'S2', 'R3', 'S3']] ['2018-05-01': '2018-06-01'] # Pick_
    ↳Date Ranges
```

```
[8]: date_range.plot(figsize=(18,12),grid=True)
plt.title('Stock Pivot Point')
plt.legend(['Price', 'P', 'R1', 'S1', 'R2', 'S2', 'R3', 'S3'], loc=0)
plt.show()
```



```
[9]: ax = date_range.plot(figsize=(18,12), grid=True)
ax.lines[0].set_linewidth(4) # Plot Specific Line
plt.title('Stock Pivot Point')
plt.legend()
plt.show()
```



2.2 Woodie's Pivot Points

```
[10]: # Woodie's Pivot Points
P = pd.Series((dataset['High'] + dataset['Low'] + 2*dataset['Close']) / 4)
R1 = pd.Series(2 * P - dataset['Low'])
S1 = pd.Series(2 * P - dataset['High'])
R2 = pd.Series(P + dataset['High'] - dataset['Low'])
S2 = pd.Series(P - dataset['High'] + dataset['Low'])
wpp = {'P':P, 'R1':R1, 'S1':S1, 'R2':R2, 'S2':S2}
WPP = pd.DataFrame(wpp)
# dataset = dataset.join(WPP)
```

```
[11]: WPP.head()
```

```
[11]:
```

	P	R1	R2	S1	S2
Date					
2017-01-03	11.3825	11.745	12.0125	11.115	10.7525
2017-01-04	11.4050	11.570	11.6850	11.290	11.1250
2017-01-05	11.3500	11.470	11.8100	11.010	10.8900
2017-01-06	11.3100	11.510	11.6900	11.130	10.9300
2017-01-09	11.4825	11.655	11.8125	11.325	11.1525

2.3 Camarilla's Pivot Points

```
[12]: # Camarilla's Pivot Points
R1 = pd.Series((dataset['High'] - dataset['Low']) * 1.1 / (2+dataset['Close']))
R2 = pd.Series((dataset['High'] - dataset['Low']) * 1.1 / (4+dataset['Close']))
R3 = pd.Series((dataset['High'] - dataset['Low']) * 1.1 / (6+dataset['Close']))
R4 = pd.Series((dataset['High'] - dataset['Low']) * 1.1 / (12+dataset['Close']))
S1 = pd.Series((dataset['Close'] - (dataset['High']-dataset['Low']) * 1.1)/12)
S2 = pd.Series((dataset['Close'] - (dataset['High']-dataset['Low']) * 1.1)/6)
S3 = pd.Series((dataset['Close'] - (dataset['High']-dataset['Low']) * 1.1)/4)
S4 = pd.Series((dataset['Close'] - (dataset['High']-dataset['Low']) * 1.1)/2)
cpp = {'R1':R1, 'S1':S1, 'R2':R2, 'S2':S2, 'R3':R3, 'S3':S3, 'R4':R4, 'S4':S4}
CPP = pd.DataFrame(cpp)
# dataset = dataset.join(CPP)
```

```
[13]: CPP.head()
```

```
[13]:
```

	R1	R2	R3	R4	S1	S2 \
Date						
2017-01-03	0.051601	0.044913	0.039759	0.029577	0.894750	1.789500
2017-01-04	0.022934	0.019961	0.017671	0.013146	0.926833	1.853667
2017-01-05	0.038218	0.033202	0.029350	0.021773	0.894500	1.789000
2017-01-06	0.031381	0.027285	0.024134	0.017925	0.908500	1.817000
2017-01-09	0.026909	0.023434	0.020755	0.015453	0.927250	1.854500

	S3	S4
Date		
2017-01-03	2.68425	5.3685
2017-01-04	2.78050	5.5610
2017-01-05	2.68350	5.3670
2017-01-06	2.72550	5.4510
2017-01-09	2.78175	5.5635

2.4 Tom DeMark's

```
[14]: # Tom DeMark's
dataset = yf.download(symbol,start,end)

h_l_c = dataset['Close'] < dataset['Open']
h_lc = dataset['Close'] > dataset['Open']
hl_c = dataset['Close'] == dataset['Open']
P = np.zeros(len(dataset['Close']))
P[h_l_c] = dataset['High'][h_l_c] + 2.0 * dataset['Low'][h_l_c] +
    dataset['Close'][h_l_c]
P[h_lc] = 2.0 * dataset['High'][h_lc] + dataset['Low'][h_lc] +
    dataset['Close'][h_lc]
```

```

P[h1_c] = dataset['High'][h1_c] + dataset['Low'][h1_c] + 2.0 *
↳dataset['Close'][h1_c]
S1 = P / 2.0 - dataset['High']
R1 = P / 2.0 - dataset['Low']
P = P / 4.0
tdm = {'P': P, 'S1': S1, 'R1': R1}
TDM = pd.DataFrame(tdm)

```

[*****100%*****] 1 of 1 downloaded

[15]: TDM.head()

```

[15]:
          P      R1      S1
Date
2017-01-03  11.4375  11.855  11.225
2017-01-04  11.3575  11.475  11.195
2017-01-05  11.3475  11.465  11.005
2017-01-06  11.3525  11.595  11.215
2017-01-09  11.5200  11.730  11.400

```

2.5 Fibonacci's Pivot Point

```

[16]: # Fibonacci's Pivot Points
PP = pd.Series((dataset['High'] + dataset['Low'] + dataset['Close']) / 3)
R1 = pd.Series((PP + (dataset['High'] - dataset['Low']) * 0.382))
R2 = pd.Series((PP + (dataset['High'] - dataset['Low']) * 0.618))
R3 = pd.Series((PP + (dataset['High'] - dataset['Low']) * 1.000))
S1 = pd.Series((PP - (dataset['High'] - dataset['Low']) * 0.382))
S2 = pd.Series((PP - (dataset['High'] - dataset['Low']) * 0.618))
S3 = pd.Series((PP - (dataset['High'] - dataset['Low']) * 1.000))
fpp = {'PP':PP, 'R1':R1, 'S1':S1, 'R2':R2, 'S2':S2, 'R3':R3, 'S3':S3}
FPP = pd.DataFrame(fpp)
# dataset = dataset.join(CPP)

```

[17]: FPP.head()

```

[17]:
          PP      R1      R2      R3      S1      S2 \
Date
2017-01-03  11.366667  11.607327  11.756007  11.996667  11.126007  10.977327
2017-01-04  11.396667  11.503627  11.569707  11.676667  11.289707  11.223627
2017-01-05  11.386667  11.562387  11.670947  11.846667  11.210947  11.102387
2017-01-06  11.306667  11.451827  11.541507  11.686667  11.161507  11.071827
2017-01-09  11.480000  11.606060  11.683940  11.810000  11.353940  11.276060

          S3
Date
2017-01-03  10.736667

```



```

2017-01-04  11.116667
2017-01-05  10.926667
2017-01-06  10.926667
2017-01-09  11.150000

```

2.6 Chicago Floor Trading Pivotal Point

<https://www.fmlabs.com/reference/default.htm>

```

[18]: PP = pd.Series((dataset['High'] + dataset['Low'] + dataset['Close']) / 3)
      R1 = pd.Series(PP * 2 - dataset['Low'].shift())
      R2 = pd.Series(PP + (dataset['High'].shift() - dataset['Low'].shift()))
      S1 = pd.Series(PP * 2 - dataset['High'].shift())
      S2 = pd.Series(PP - (dataset['High'].shift() - dataset['Low'].shift()))
      CFpp = {'PP':PP, 'R1':R1, 'S1':S1, 'R2':R2, 'S2':S2}
      CFPP = pd.DataFrame(CFpp)

```

```

[19]: CFPP.head()

```

```

[19]:
      PP      R1      R2      S1      S2
Date
2017-01-03  11.366667      NaN      NaN      NaN      NaN
2017-01-04  11.396667  11.773333  12.026667  11.143333  10.766667
2017-01-05  11.386667  11.533333  11.666667  11.253333  11.106667
2017-01-06  11.306667  11.383333  11.766667  10.923333  10.846667
2017-01-09  11.480000  11.850000  11.860000  11.470000  11.100000

```