

Ventilator_Manufacturer_Portfolio

September 29, 2021

1 Ventilator Manufacturer Portfolio Risk and Returns

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

import warnings
warnings.filterwarnings("ignore")

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

```
[2]: # input
# Ventilator Manufacturer
symbols = ['MDT', 'RMD', 'AHPI']
start = '2019-12-01'
end = '2020-04-14'
```

```
[3]: df = pd.DataFrame()
for s in symbols:
    df[s] = yf.download(s, start, end)['Adj Close']
```

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

```
[4]: from datetime import datetime
from dateutil import relativedelta

d1 = datetime.strptime(start, "%Y-%m-%d")
d2 = datetime.strptime(end, "%Y-%m-%d")
delta = relativedelta.relativedelta(d2, d1)
print('How many years of investing?')
print('%s years' % delta.years)
```

How many years of investing?
0 years

```
[5]: number_of_years = delta.years
```

```
[6]: days = (df.index[-1] - df.index[0]).days  
days
```

```
[6]: 133
```

```
[7]: df.head()
```

```
[7]:
```

	MDT	RMD	AHPI
Date			
2019-12-02	109.697014	148.676666	1.01
2019-12-03	109.776138	149.135651	1.08
2019-12-04	110.715752	149.814117	1.05
2019-12-05	110.972908	149.844055	1.05
2019-12-06	111.853165	150.253143	1.06

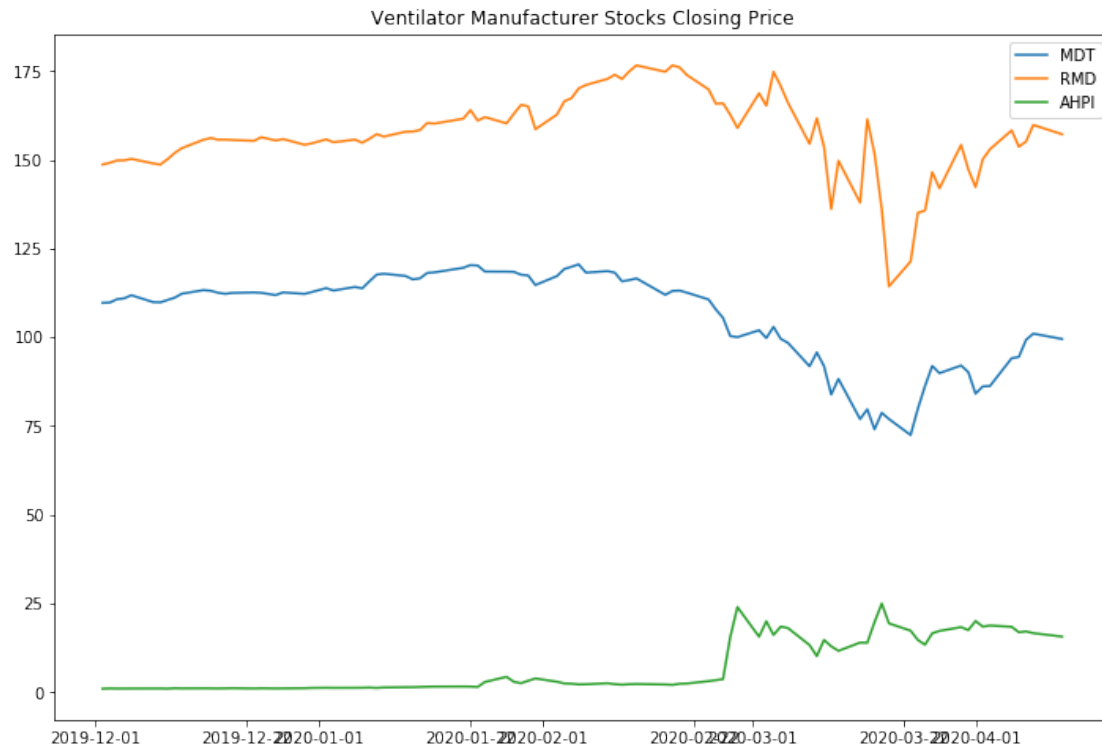
```
[8]: df.tail()
```

```
[8]:
```

	MDT	RMD	AHPI
Date			
2020-04-06	94.089996	158.300003	18.440001
2020-04-07	94.459999	153.679993	16.900000
2020-04-08	99.279999	155.160004	17.150000
2020-04-09	101.029999	159.820007	16.650000
2020-04-13	99.519997	157.190002	15.680000

```
[9]: plt.figure(figsize=(12,8))  
plt.plot(df)  
plt.title('Ventilator Manufacturer Stocks Closing Price')  
plt.legend(labels=df.columns)
```

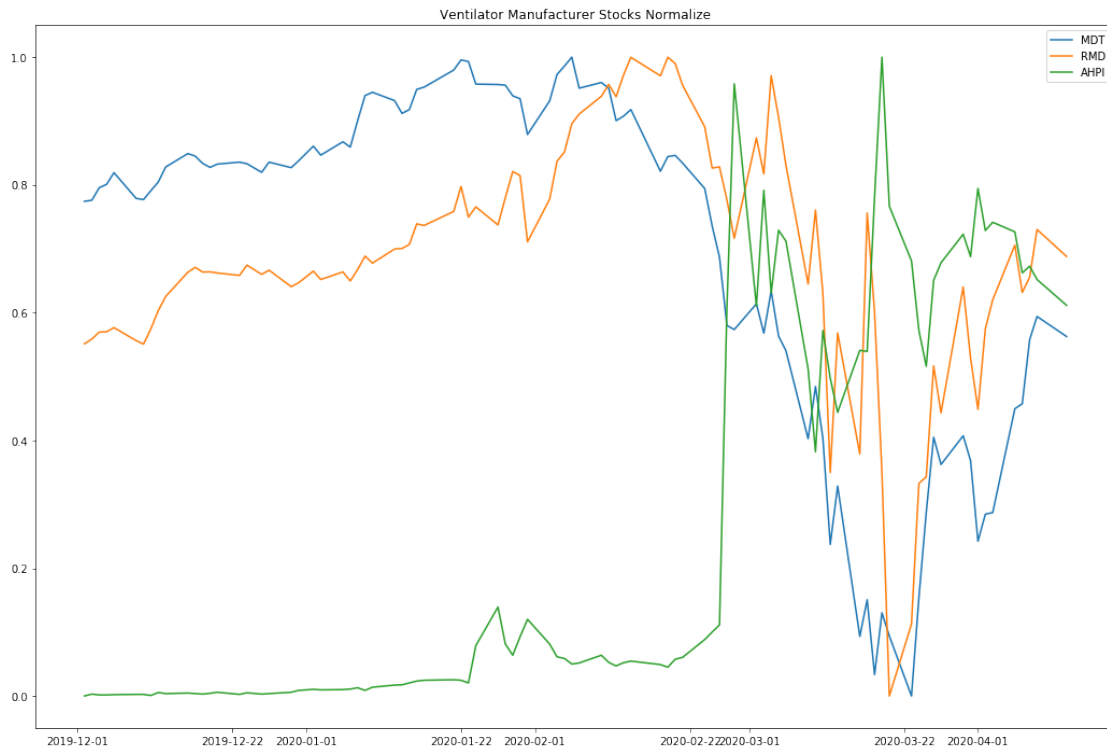
```
[9]: <matplotlib.legend.Legend at 0x2179a63d550>
```



```
[10]: # Normalize the data
normalize = (df - df.min()) / (df.max() - df.min())
```

```
[11]: plt.figure(figsize=(18,12))
plt.plot(normalize)
plt.title('Ventilator Manufacturer Stocks Normalize')
plt.legend(labels=normalize.columns)
```

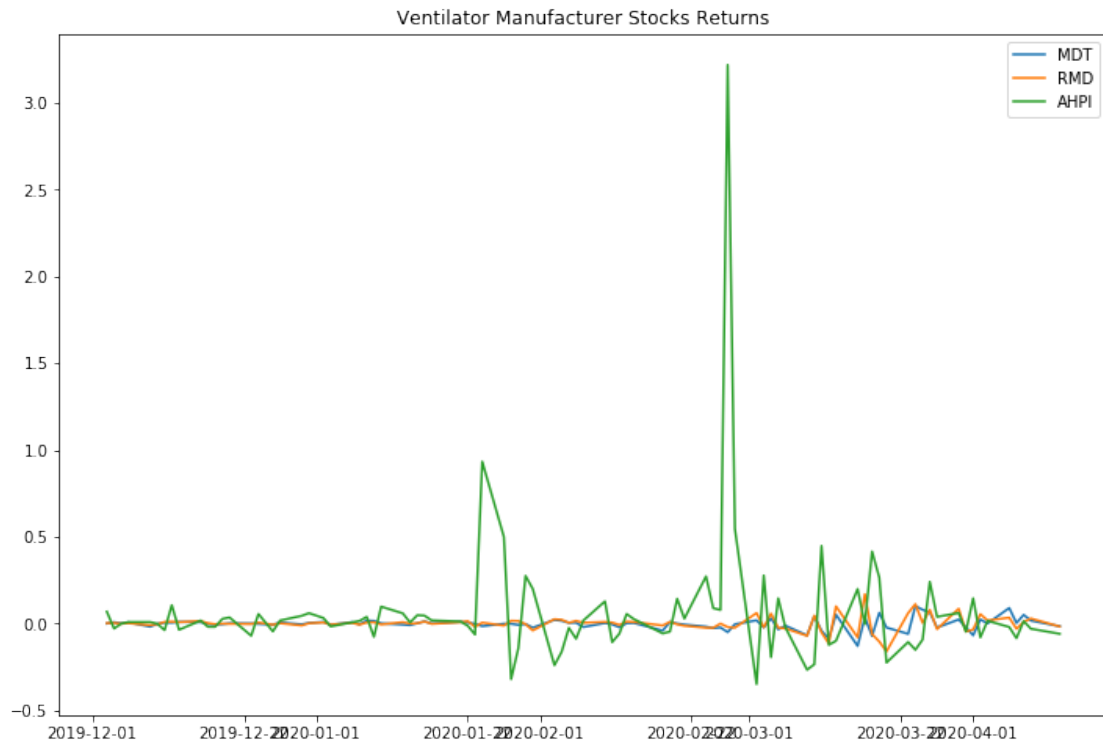
```
[11]: <matplotlib.legend.Legend at 0x2179aa10940>
```



```
[12]: stock_rets = df.pct_change().dropna()
```

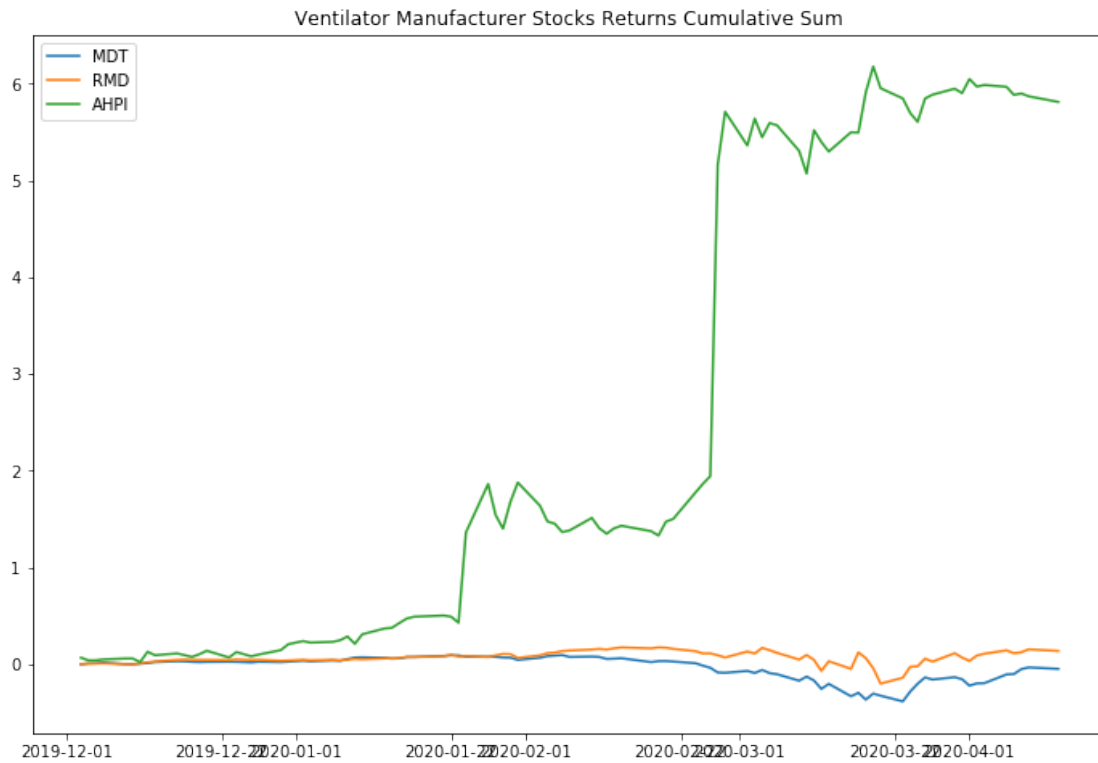
```
[13]: plt.figure(figsize=(12,8))
plt.plot(stock_rets)
plt.title('Ventilator Manufacturer Stocks Returns')
plt.legend(labels=stock_rets.columns)
```

```
[13]: <matplotlib.legend.Legend at 0x2179a6975f8>
```



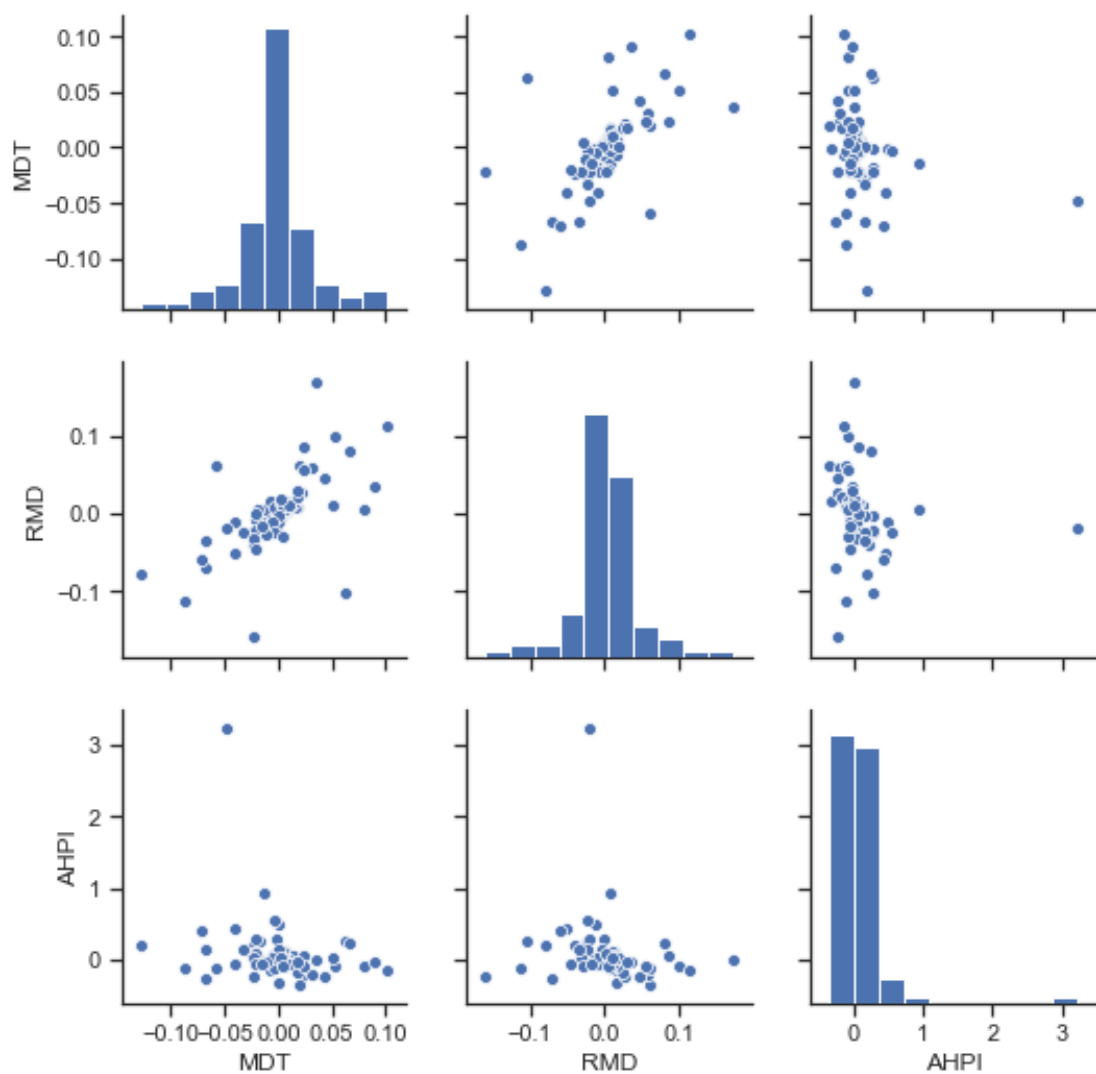
```
[14]: plt.figure(figsize=(12,8))
plt.plot(stock_rets.cumsum())
plt.title('Ventilator Manufacturer Stocks Returns Cumulative Sum')
plt.legend(labels=stock_rets.columns)
```

[14]: <matplotlib.legend.Legend at 0x2179a679400>

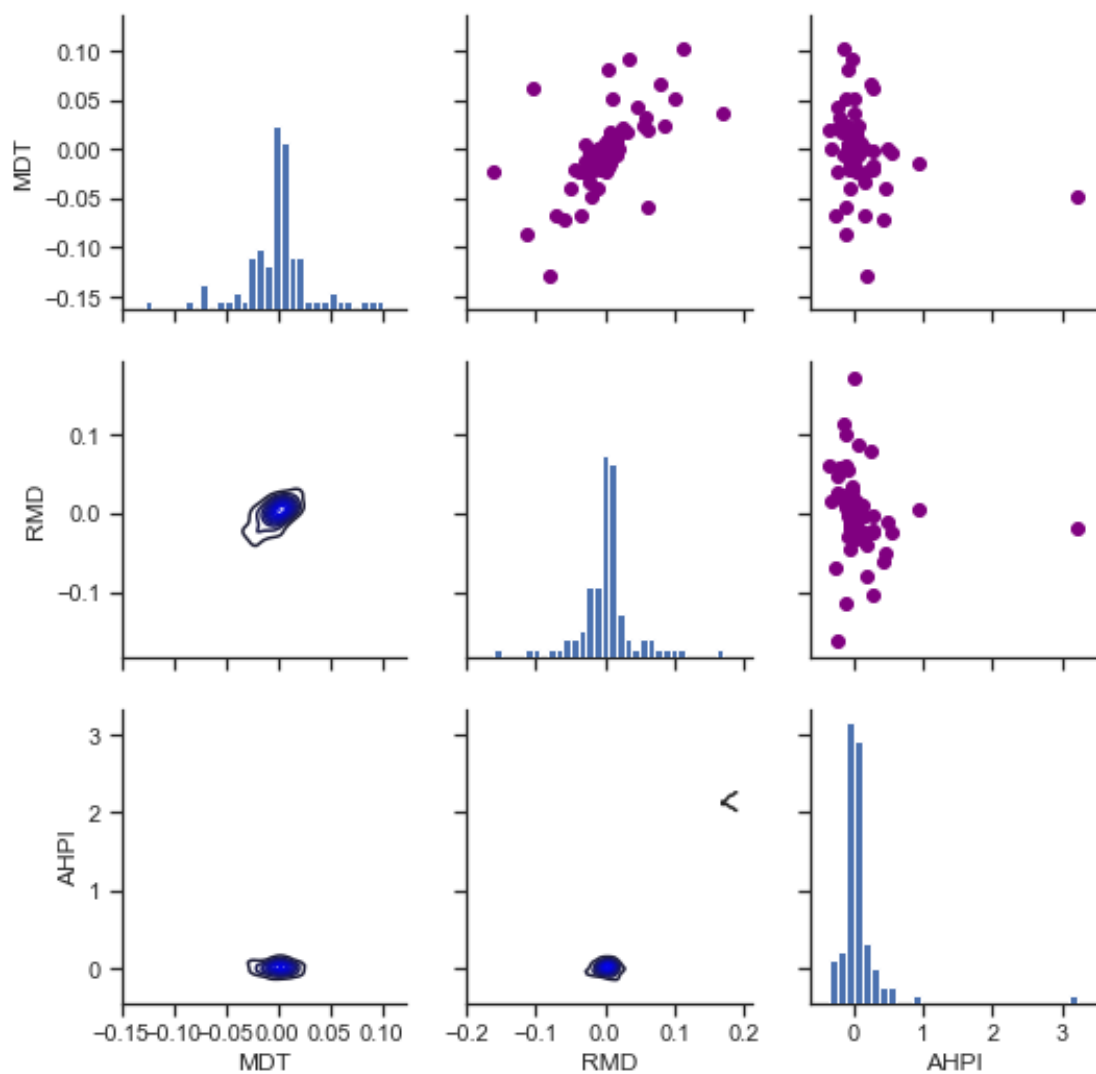


```
[15]: sns.set(style='ticks')
ax = sns.pairplot(stock_rets, diag_kind='hist')

nplot = len(stock_rets.columns)
for i in range(nplot) :
    for j in range(nplot) :
        ax.axes[i, j].locator_params(axis='x', nbins=6, tight=True)
```



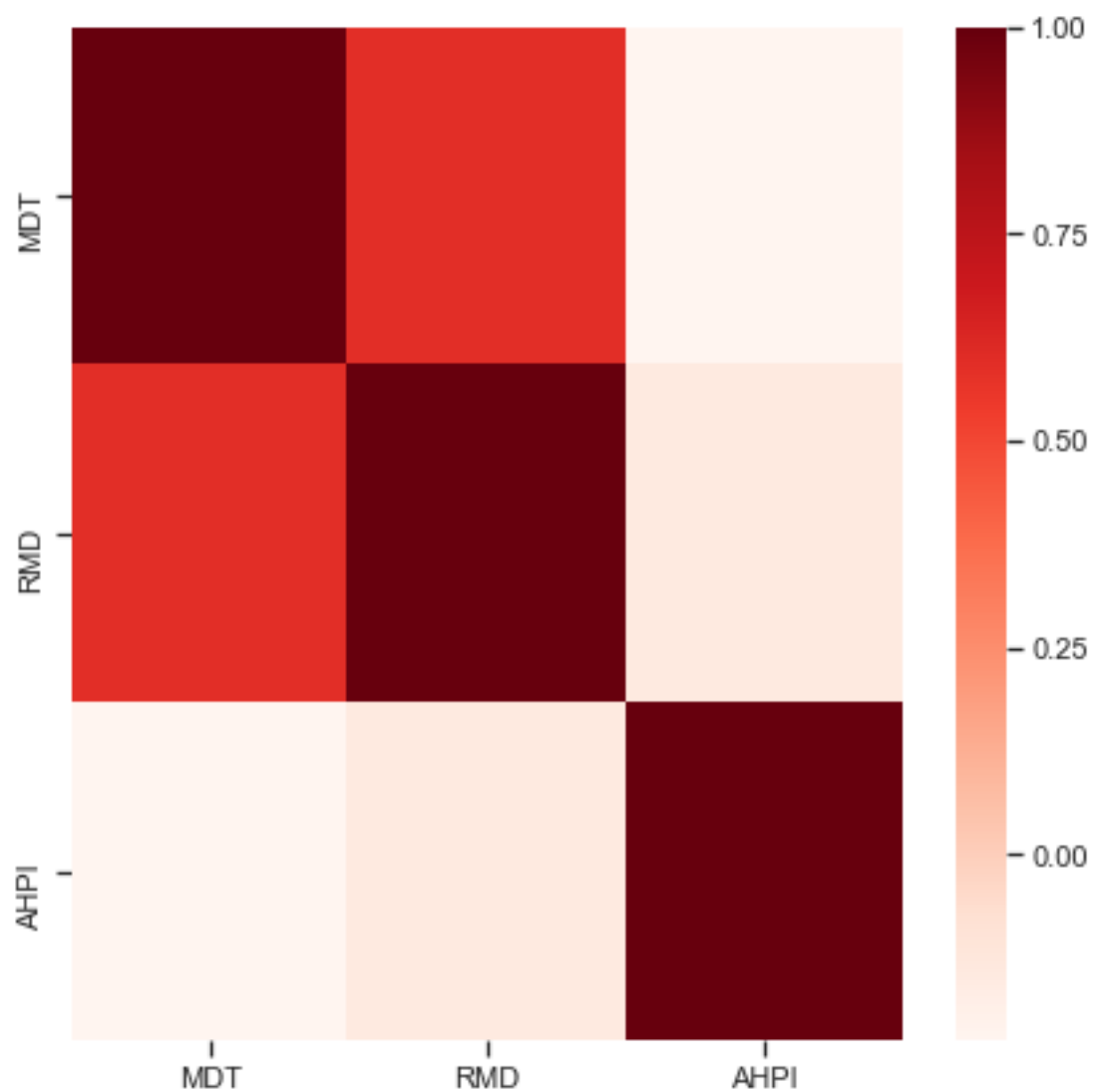
```
[16]: ax = sns.PairGrid(stock_rets)
ax.map_upper(plt.scatter, color='purple')
ax.map_lower(sns.kdeplot, color='blue')
ax.map_diag(plt.hist, bins=30)
for i in range(nplot) :
    for j in range(nplot) :
        ax.axes[i, j].locator_params(axis='x', nbins=6, tight=True)
```



```
[17]: plt.figure(figsize=(7,7))
      corr = stock_rets.corr()

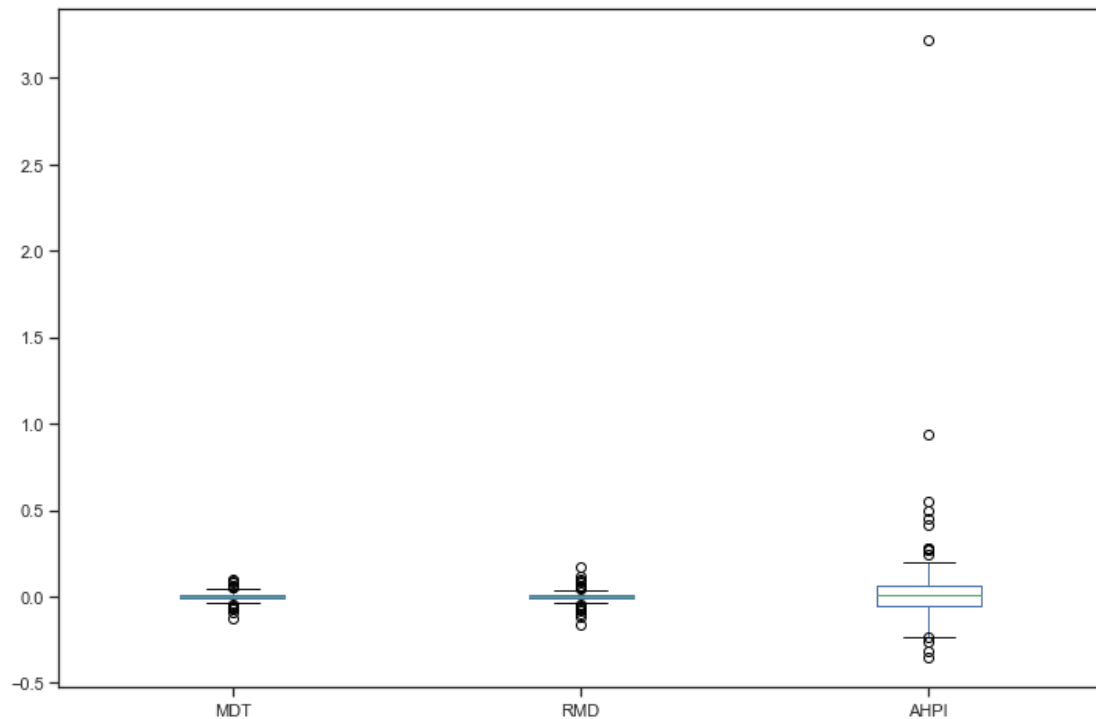
      # plot the heatmap
      sns.heatmap(corr,
                  xticklabels=corr.columns,
                  yticklabels=corr.columns,
                  cmap="Reds")
```

```
[17]: <matplotlib.axes._subplots.AxesSubplot at 0x2179adeb2b0>
```

```
[18]: # Box plot  
stock_rets.plot(kind='box',figsize=(12,8))
```

```
[18]: <matplotlib.axes._subplots.AxesSubplot at 0x2179c7ab2e8>
```

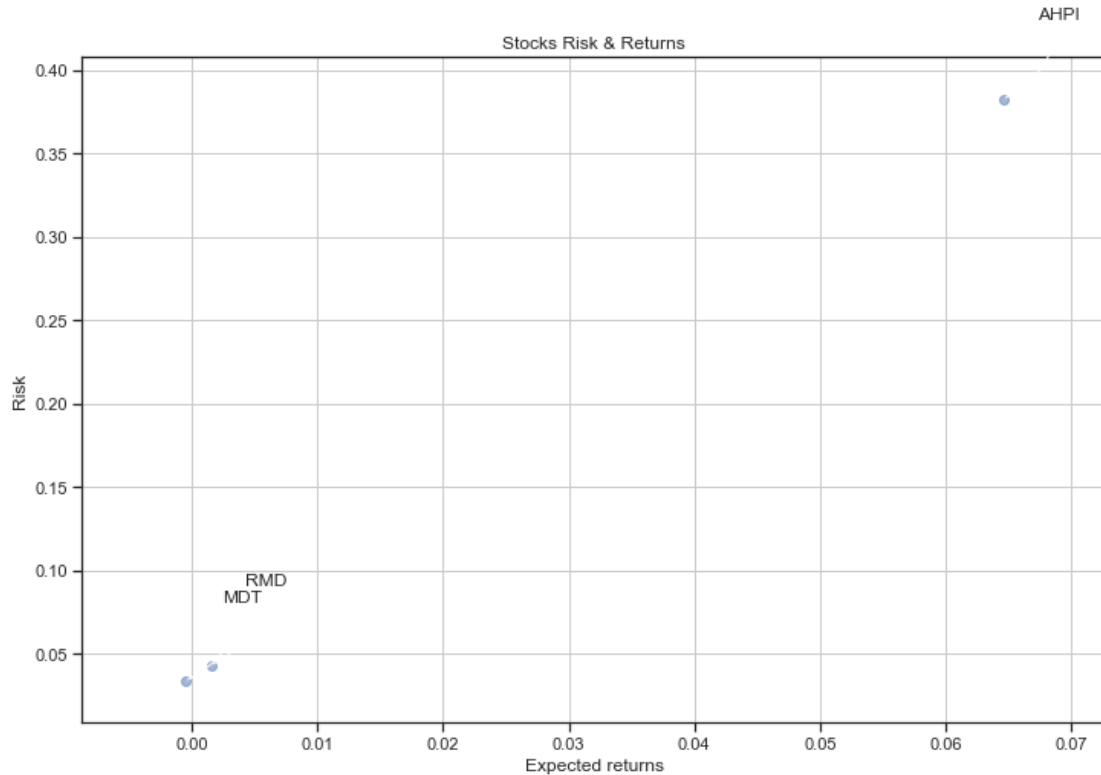


```
[19]: rets = stock_rets.dropna()

plt.figure(figsize=(12,8))
plt.scatter(rets.mean(), rets.std(),alpha = 0.5)

plt.title('Stocks Risk & Returns')
plt.xlabel('Expected returns')
plt.ylabel('Risk')
plt.grid(which='major')

for label, x, y in zip(rets.columns, rets.mean(), rets.std()):
    plt.annotate(
        label,
        xy = (x, y), xytext = (50, 50),
        textcoords = 'offset points', ha = 'right', va = 'bottom',
        arrowprops = dict(arrowstyle = '-', connectionstyle = 'arc3,rad=-0.3'))
```



```
[20]: rets = stock_rets.dropna()
area = np.pi*20.0

sns.set(style='darkgrid')
plt.figure(figsize=(12,8))
plt.scatter(rets.mean(), rets.std(), s=area)
plt.xlabel("Expected Return", fontsize=15)
plt.ylabel("Risk", fontsize=15)
plt.title("Return vs. Risk for Stocks", fontsize=20)

for label, x, y in zip(rets.columns, rets.mean(), rets.std()) :
    plt.annotate(label, xy=(x,y), xytext=(50, 0), textcoords='offset points',
        arrowprops=dict(arrowstyle='-',
        ↪connectionstyle='bar,angle=180,fraction=-0.2'),
        bbox=dict(boxstyle="round", fc="w"))
```



```
[21]: rest_rets = rets.corr()
pair_value = rest_rets.abs().unstack()
pair_value.sort_values(ascending = False)
```

```
[21]: AHPI  AHPI    1.000000
      RMD   RMD    1.000000
      MDT   MDT    1.000000
      RMD   MDT    0.594789
      MDT   RMD    0.594789
      AHPI  MDT    0.225555
      MDT   AHPI    0.225555
      AHPI  RMD    0.139522
      RMD   AHPI    0.139522
      dtype: float64
```

```
[22]: # Normalized Returns Data
Normalized_Value = ((rets[:] - rets[:].min()) / (rets[:].max() - rets[:].min()))
Normalized_Value.head()
```

```
[22]:           MDT      RMD      AHPI
Date
2019-12-03  0.560373  0.493460  0.117023
2019-12-04  0.594432  0.497888  0.089793
```

```

2019-12-05  0.567331  0.484718  0.097584
2019-12-06  0.591707  0.492379  0.100255
2019-12-09  0.481927  0.458577  0.100230

```

```
[23]: Normalized_Value.corr()
```

```

[23]:          MDT          RMD          AHPI
MDT    1.000000  0.594789 -0.225555
RMD    0.594789  1.000000 -0.139522
AHPI  -0.225555 -0.139522  1.000000

```

```

[24]: normalized_rets = Normalized_Value.corr()
normalized_pair_value = normalized_rets.abs().unstack()
normalized_pair_value.sort_values(ascending = False)

```

```

[24]: AHPI  AHPI    1.000000
      RMD   RMD    1.000000
      MDT   MDT    1.000000
      RMD   MDT    0.594789
      MDT   RMD    0.594789
      AHPI  MDT    0.225555
      MDT   AHPI    0.225555
      AHPI  RMD    0.139522
      RMD   AHPI    0.139522
      dtype: float64

```

```

[25]: print("Stock returns: ")
      print(rets.mean())
      print('-' * 50)
      print("Stock risks:")
      print(rets.std())

```

```

Stock returns:
MDT    -0.000514
RMD     0.001543
AHPI     0.064549
dtype: float64

```

```

-----
Stock risks:
MDT     0.033714
RMD     0.043191
AHPI     0.382719
dtype: float64

```

```

[26]: table = pd.DataFrame()
      table['Returns'] = rets.mean()
      table['Risk'] = rets.std()

```

```
table.sort_values(by='Returns')
```

```
[26]:
```

	Returns	Risk
MDT	-0.000514	0.033714
RMD	0.001543	0.043191
AHPI	0.064549	0.382719

```
[27]: table.sort_values(by='Risk')
```

```
[27]:
```

	Returns	Risk
MDT	-0.000514	0.033714
RMD	0.001543	0.043191
AHPI	0.064549	0.382719

```
[28]: rf = 0.01
table['Sharpe Ratio'] = (table['Returns'] - rf) / table['Risk']
table
```

```
[28]:
```

	Returns	Risk	Sharpe Ratio
MDT	-0.000514	0.033714	-0.311864
RMD	0.001543	0.043191	-0.195799
AHPI	0.064549	0.382719	0.142529

```
[29]: table['Max Returns'] = rets.max()
```

```
[30]: table['Min Returns'] = rets.min()
```

```
[31]: table['Median Returns'] = rets.median()
```

```
[32]: total_return = stock_rets[-1:].transpose()
table['Total Return'] = 100 * total_return
table
```

```
[32]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
MDT	-0.000514	0.033714	-0.311864	0.101893	-0.128237	
RMD	0.001543	0.043191	-0.195799	0.170376	-0.159883	
AHPI	0.064549	0.382719	0.142529	3.217391	-0.347917	

	Median Returns	Total Return
MDT	0.001055	-1.494608
RMD	0.002681	-1.645604
AHPI	0.011012	-5.825822

```
[33]: table['Average Return Days'] = (1 + total_return)**(1 / days) - 1
table
```

```
[33]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
MDT	-0.000514	0.033714	-0.311864	0.101893	-0.128237	
RMD	0.001543	0.043191	-0.195799	0.170376	-0.159883	
AHPI	0.064549	0.382719	0.142529	3.217391	-0.347917	

	Median Returns	Total Return	Average Return Days
MDT	0.001055	-1.494608	-0.000113
RMD	0.002681	-1.645604	-0.000125
AHPI	0.011012	-5.825822	-0.000451

```
[34]: initial_value = df.iloc[0]
ending_value = df.iloc[-1]
table['CAGR'] = ((ending_value / initial_value) ** (252.0 / days)) -1
table
```

```
[34]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
MDT	-0.000514	0.033714	-0.311864	0.101893	-0.128237	
RMD	0.001543	0.043191	-0.195799	0.170376	-0.159883	
AHPI	0.064549	0.382719	0.142529	3.217391	-0.347917	

	Median Returns	Total Return	Average Return Days	CAGR
MDT	0.001055	-1.494608	-0.000113	-0.168462
RMD	0.002681	-1.645604	-0.000125	0.111268
AHPI	0.011012	-5.825822	-0.000451	179.583628

```
[35]: table.sort_values(by='Average Return Days')
```

```
[35]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
AHPI	0.064549	0.382719	0.142529	3.217391	-0.347917	
RMD	0.001543	0.043191	-0.195799	0.170376	-0.159883	
MDT	-0.000514	0.033714	-0.311864	0.101893	-0.128237	

	Median Returns	Total Return	Average Return Days	CAGR
AHPI	0.011012	-5.825822	-0.000451	179.583628
RMD	0.002681	-1.645604	-0.000125	0.111268
MDT	0.001055	-1.494608	-0.000113	-0.168462