

# Growth\_Stock\_Portfolio

September 29, 2021

## 1 Growth Stock Portfolio

```
[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")

# yahoo finance data
import yfinance as yf
yf.pdr_override()
```

```
[2]: # input
# Growth Stock
title = "Growth Stock for July"
symbols = ['STNE', 'CRWD', 'ZNGA']
start = '2020-04-01'
end = '2020-06-26'
```

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

```
[*****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]: 85
```

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

```
[7]:
```

	STNE	CRWD	ZNGA
Date			
2020-04-01	19.530001	54.580002	6.76
2020-04-02	19.879999	56.689999	6.68
2020-04-03	17.820000	57.820000	6.64
2020-04-06	22.240000	59.070000	6.81
2020-04-07	23.639999	59.599998	6.56

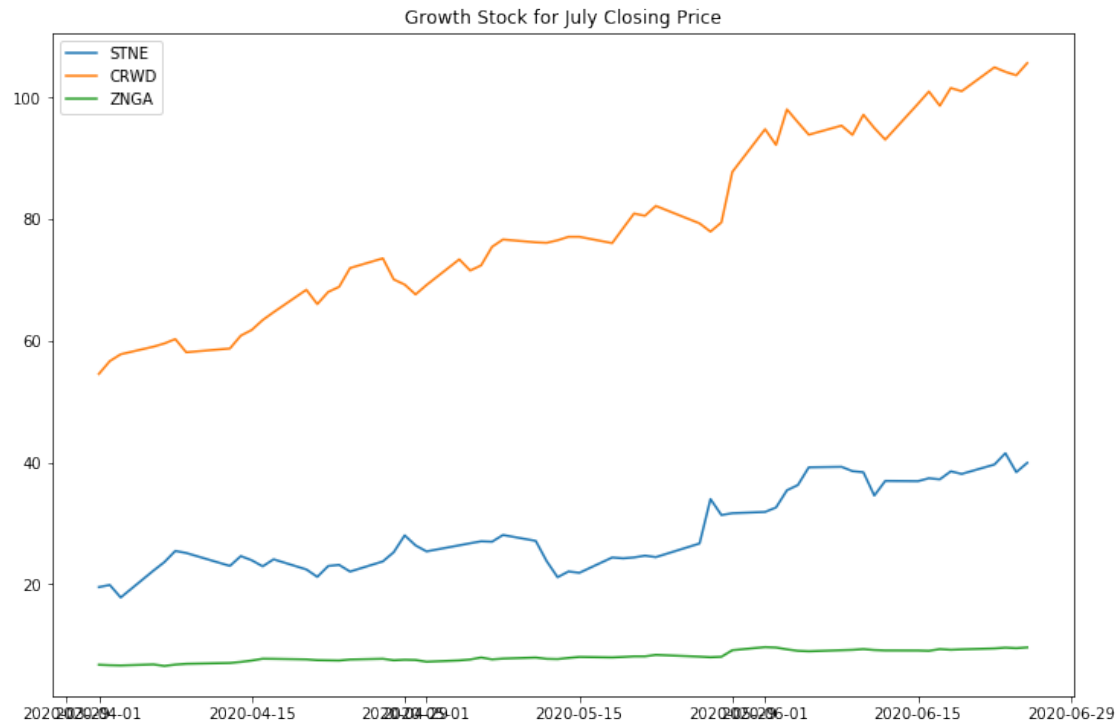
```
[8]: df.tail()
```

```
[8]:
```

	STNE	CRWD	ZNGA
Date			
2020-06-19	38.139999	101.080002	9.29
2020-06-22	39.689999	105.019997	9.44
2020-06-23	41.549999	104.269997	9.57
2020-06-24	38.430000	103.699997	9.48
2020-06-25	39.959999	105.709999	9.61

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

```
[9]: <matplotlib.legend.Legend at 0x1d5fce1a6a0>
```



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

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

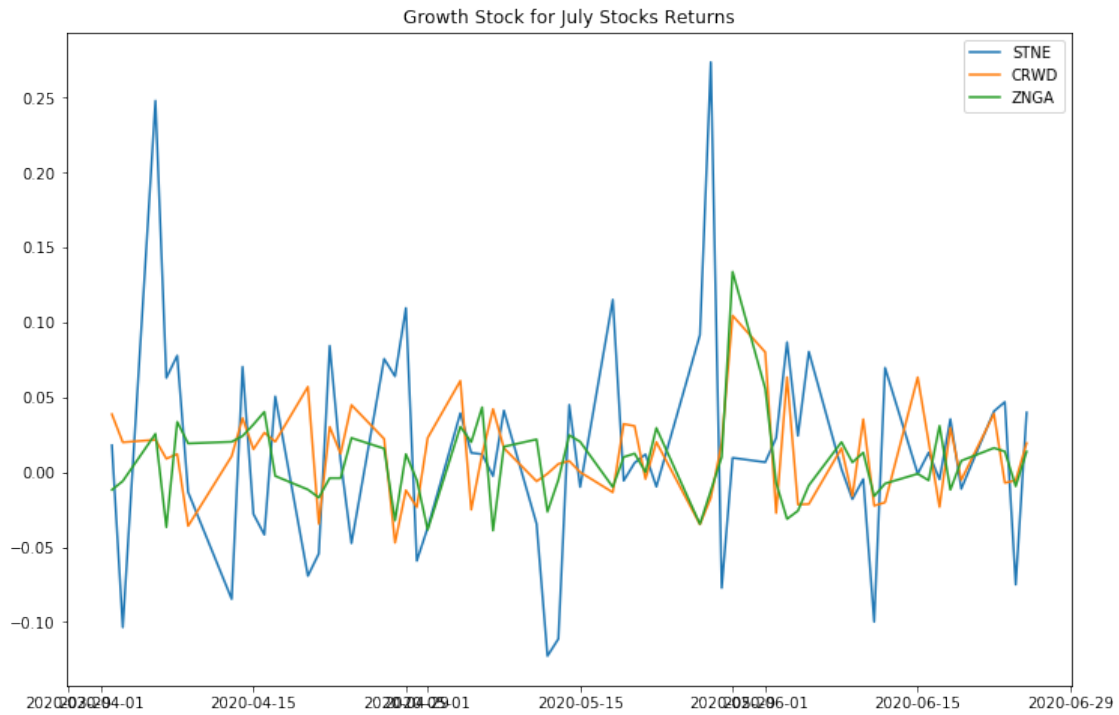
```
[11]: <matplotlib.legend.Legend at 0x1d5fd1f8b38>
```



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

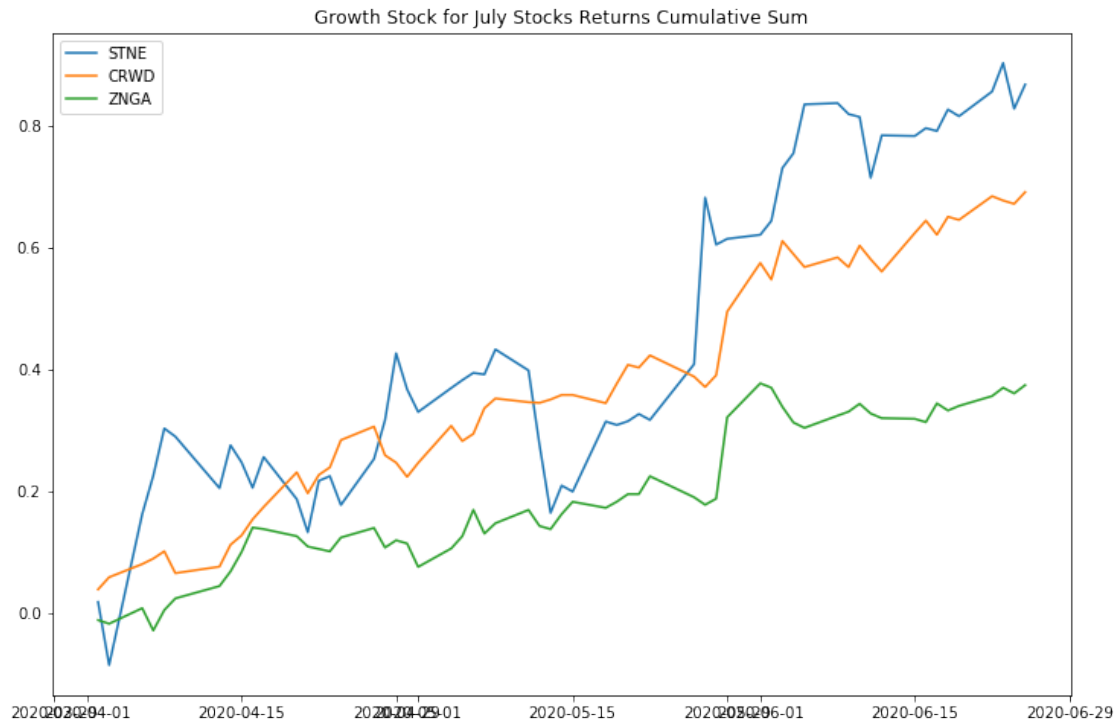
```
[13]: plt.figure(figsize=(12,8))
plt.plot(stock_rets)
plt.title(title + ' Stocks Returns')
plt.legend(labels=stock_rets.columns)
```

```
[13]: <matplotlib.legend.Legend at 0x1d5fce84d30>
```



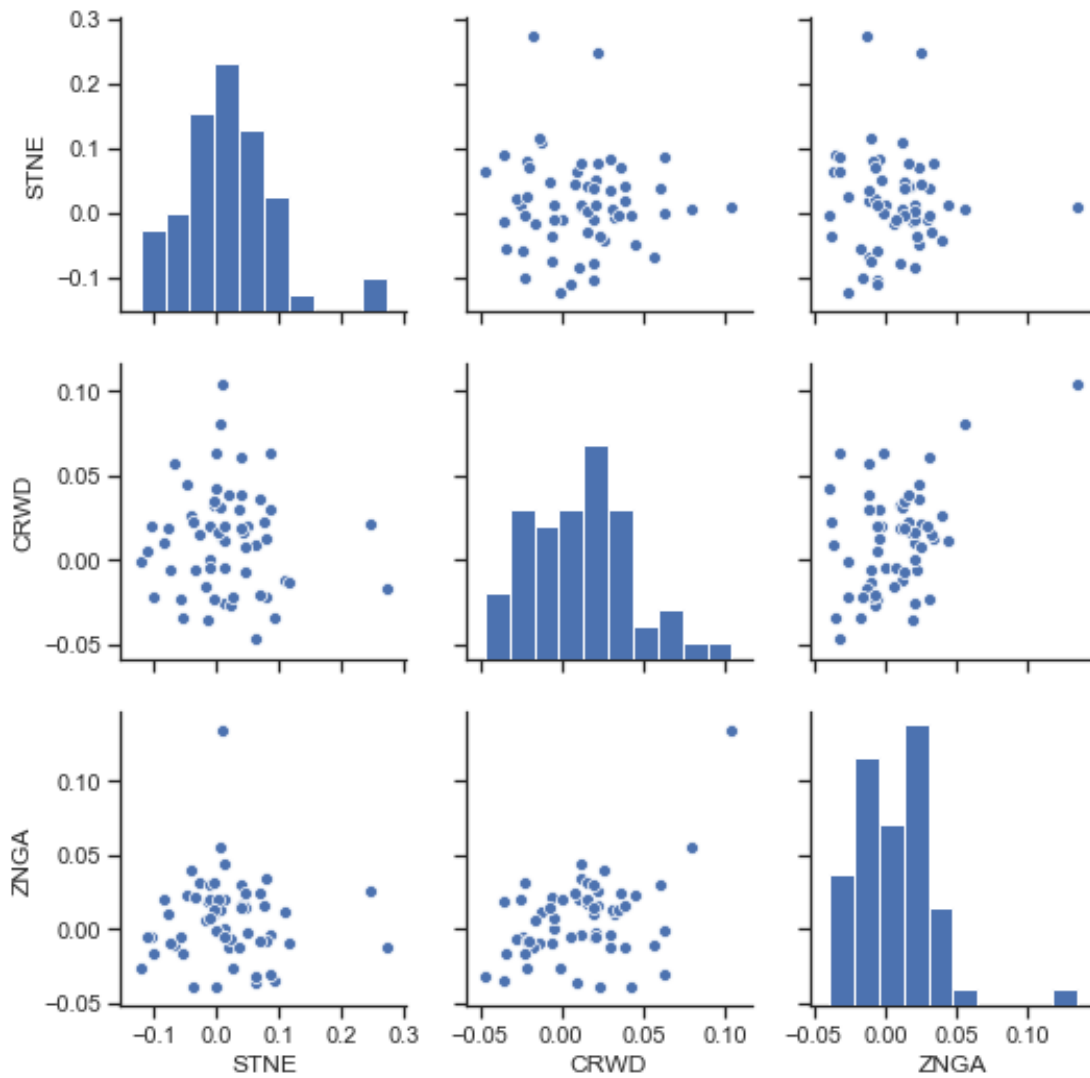
```
[14]: plt.figure(figsize=(12,8))
plt.plot(stock_rets.cumsum())
plt.title(title + ' Stocks Returns Cumulative Sum')
plt.legend(labels=stock_rets.columns)
```

[14]: <matplotlib.legend.Legend at 0x1d5fd225940>

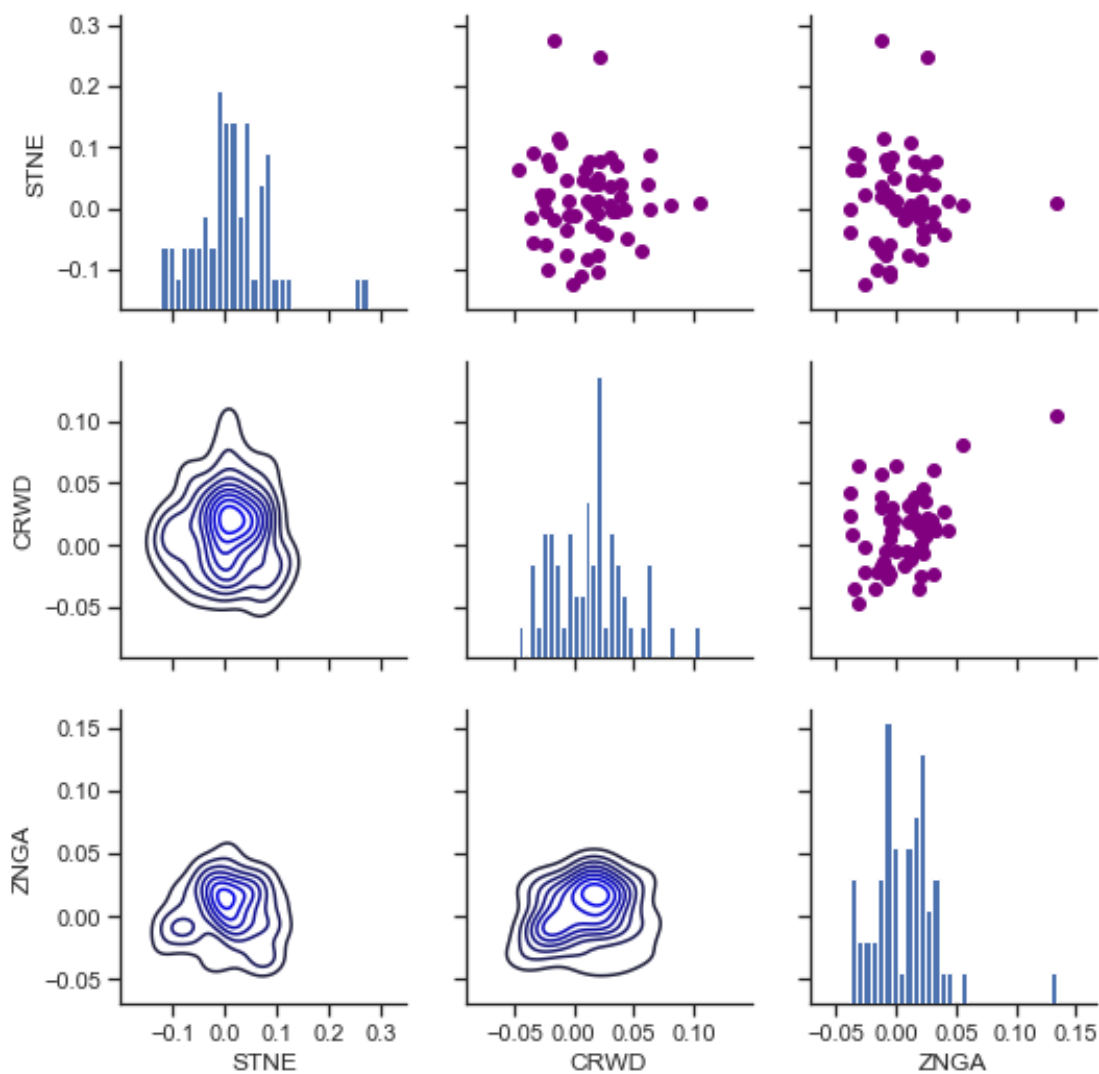


```
[15]: sns.set(style='ticks')
ax = sns.pairplot(stock_rets, diag_kind='hist')

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)
```

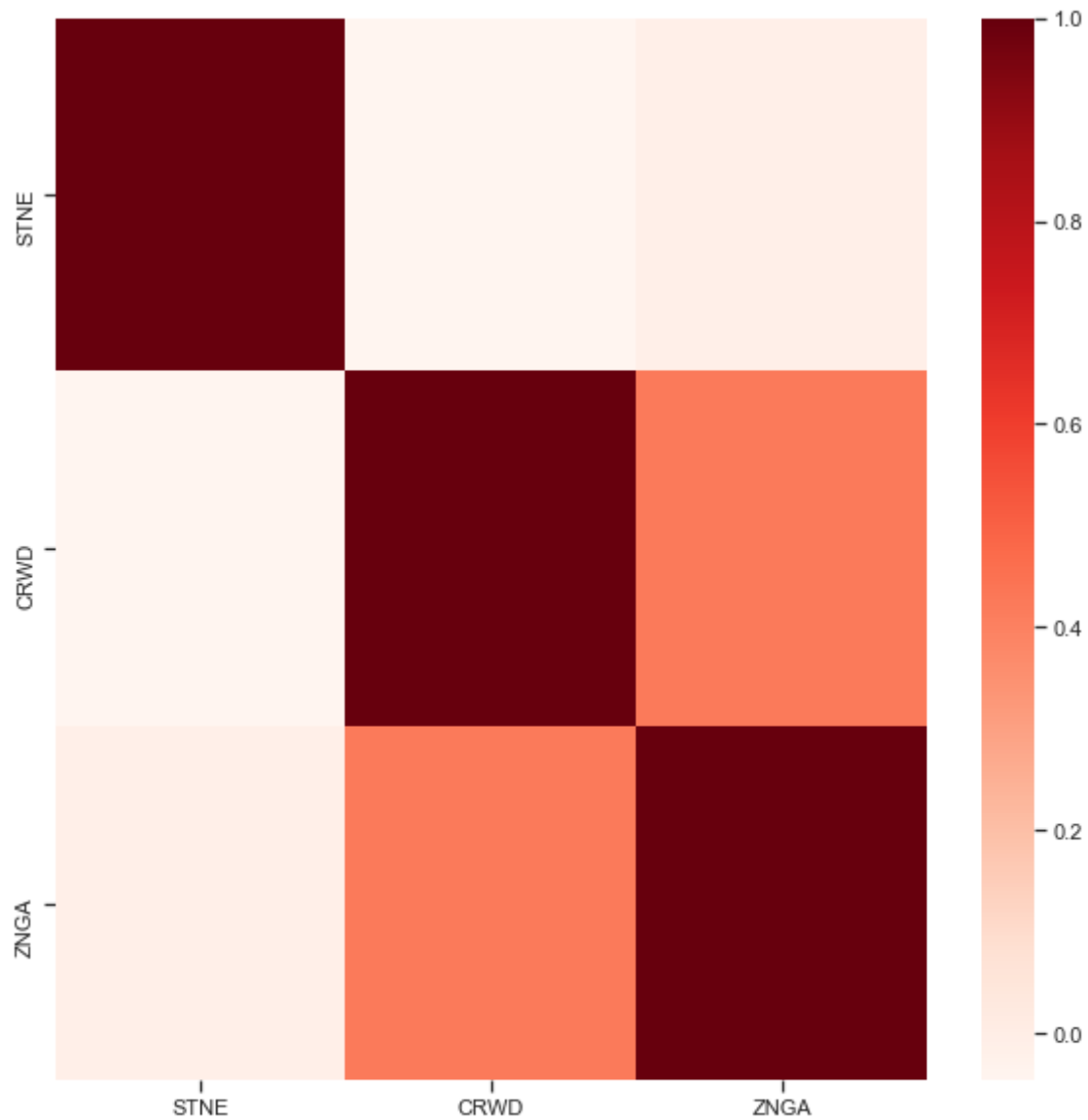


```
[17]: plt.figure(figsize=(10,10))
      corr = stock_rets.corr()

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

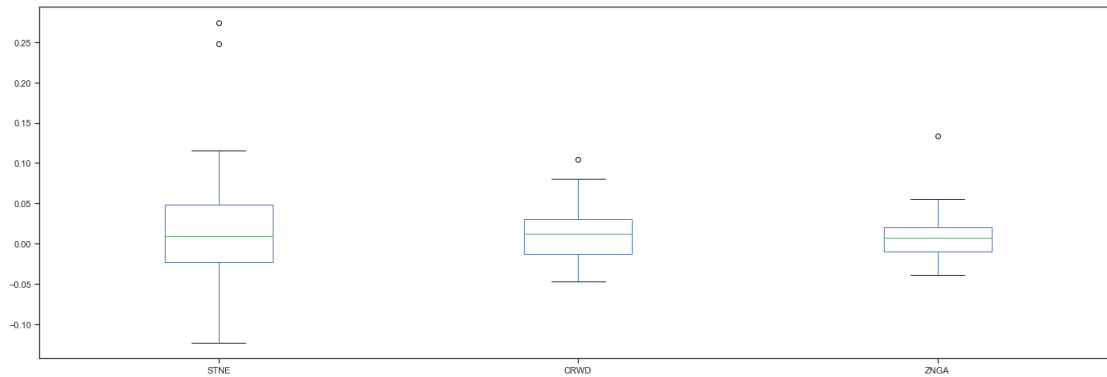
```
[17]: <matplotlib.axes._subplots.AxesSubplot at 0x1d5fce7b3c8>
```





```
[18]: # Box plot  
stock_rets.plot(kind='box',figsize=(24,8))
```

```
[18]: <matplotlib.axes._subplots.AxesSubplot at 0x1d5ff074a58>
```

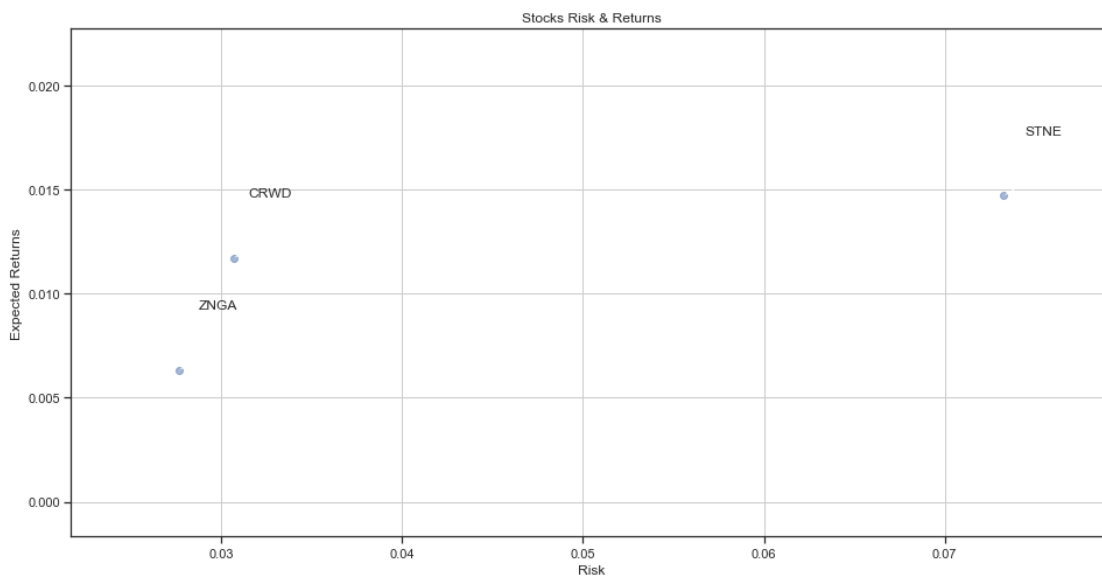


```
[19]: rets = stock_rets.dropna()

plt.figure(figsize=(16,8))
plt.scatter(rets.std(), rets.mean(),alpha = 0.5)

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

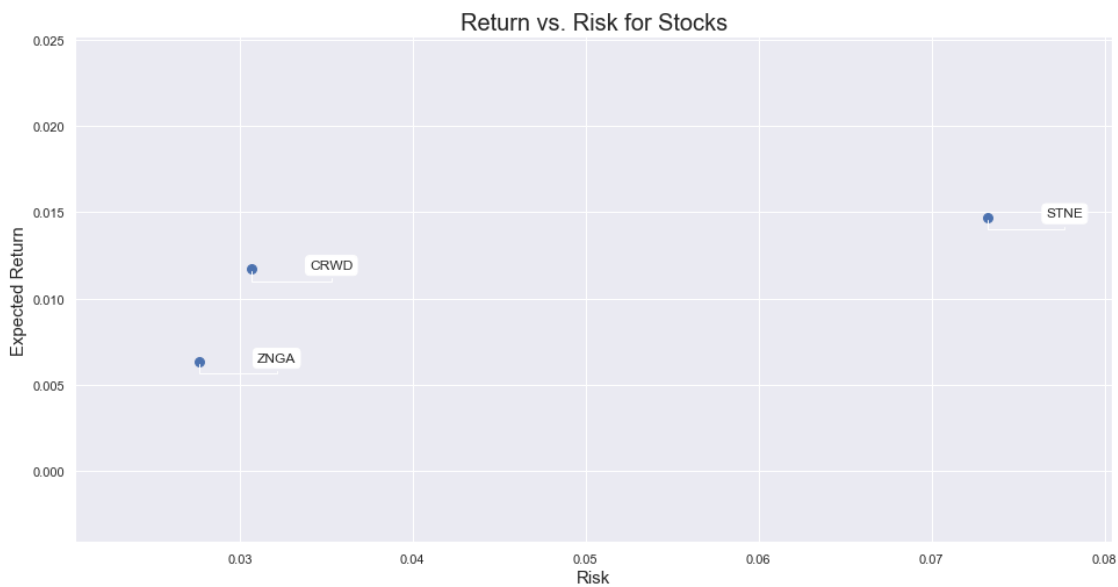
for label, x, y in zip(rets.columns, rets.std(), rets.mean()):
    plt.annotate(
        label,
        xy = (x, y), xytext = (50, 50),
        textcoords = 'offset points', ha = 'right', va = 'bottom',
        arrowprops = dict(arrowstyle = '-', connectionstyle = 'arc3,rad=-0.3'))
```



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

sns.set(style='darkgrid')
plt.figure(figsize=(16,8))
plt.scatter(rets.std(), rets.mean(), s=area)
plt.xlabel("Risk", fontsize=15)
plt.ylabel("Expected Return", fontsize=15)
plt.title("Return vs. Risk for Stocks", fontsize=20)

for label, x, y in zip(rets.columns, rets.std(), rets.mean()) :
    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]: def annual_risk_return(stock_rets):
tradeoff = stock_rets.agg(["mean", "std"]).T
tradeoff.columns = ["Return", "Risk"]
tradeoff.Return = tradeoff.Return*252
tradeoff.Risk = tradeoff.Risk * np.sqrt(252)
return tradeoff
```

```
[22]: tradeoff = annual_risk_return(stock_rets)
tradeoff
```

```
[22]:
```

	Return	Risk
STNE	3.708787	1.163257
CRWD	2.953634	0.487298
ZNGA	1.598953	0.439222

```
[23]: import itertools

colors = itertools.cycle(["r", "b", "g"])

tradeoff.plot(x = "Risk", y = "Return", kind = "scatter", figsize = (13,9), s = 20,
              fontsize = 15, c='g')
for i in tradeoff.index:
    plt.annotate(i, xy=(tradeoff.loc[i, "Risk"]+0.002, tradeoff.loc[i, "Return"]+0.002), size = 15)
plt.xlabel("Annual Risk", fontsize = 15)
plt.ylabel("Annual Return", fontsize = 15)
plt.title("Return vs. Risk for " + title + " Stocks", fontsize = 20)
plt.show()
```



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

```
[24]: ZNGA  ZNGA    1.000000
      CRWD  CRWD    1.000000
      STNE  STNE    1.000000
      ZNGA  CRWD    0.427355
      CRWD  ZNGA    0.427355
           STNE    0.046759
      STNE  CRWD    0.046759
      ZNGA  STNE    0.006471
      STNE  ZNGA    0.006471
      dtype: float64
```

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

```
[25]:           STNE      CRWD      ZNGA
Date
2020-04-02  0.354740  0.565720  0.157151
2020-04-03  0.048221  0.442260  0.190980
2020-04-06  0.935069  0.453374  0.373771
2020-04-07  0.468298  0.369996  0.013210
2020-04-08  0.505836  0.390488  0.419681
```

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

```
[26]:           STNE      CRWD      ZNGA
STNE  1.000000 -0.046759 -0.006471
CRWD -0.046759  1.000000  0.427355
ZNGA -0.006471  0.427355  1.000000
```

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

```
[27]: ZNGA  ZNGA    1.000000
      CRWD  CRWD    1.000000
      STNE  STNE    1.000000
      ZNGA  CRWD    0.427355
      CRWD  ZNGA    0.427355
           STNE    0.046759
      STNE  CRWD    0.046759
      ZNGA  STNE    0.006471
      STNE  ZNGA    0.006471
      dtype: float64
```

```
[28]: print("Stock returns: ")
      print(rets.mean())
```

```
print('-' * 50)
print("Stock risks:")
print(rets.std())
```

```
Stock returns:
STNE    0.014717
CRWD    0.011721
ZNGA    0.006345
dtype: float64
```

```
-----
Stock risks:
STNE    0.073278
CRWD    0.030697
ZNGA    0.027668
dtype: float64
```

```
[29]: table = pd.DataFrame()
      table['Returns'] = rets.mean()
      table['Risk'] = rets.std()
      table.sort_values(by='Returns')
```

```
[29]:      Returns      Risk
      ZNGA  0.006345  0.027668
      CRWD  0.011721  0.030697
      STNE  0.014717  0.073278
```

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

```
[30]:      Returns      Risk
      ZNGA  0.006345  0.027668
      CRWD  0.011721  0.030697
      STNE  0.014717  0.073278
```

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

```
[31]:      Returns      Risk  Sharpe Ratio
      STNE  0.014717  0.073278      0.064377
      CRWD  0.011721  0.030697      0.056057
      ZNGA  0.006345  0.027668     -0.132098
```

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

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

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

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

```
[35]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
STNE	0.014717	0.073278	0.064377	0.273783	-0.122742	
CRWD	0.011721	0.030697	0.056057	0.104528	-0.047147	
ZNGA	0.006345	0.027668	-0.132098	0.133829	-0.038994	

	Median Returns	Total Return
STNE	0.009560	3.981261
CRWD	0.012340	1.938286
ZNGA	0.007592	1.371309

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

```
[36]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
STNE	0.014717	0.073278	0.064377	0.273783	-0.122742	
CRWD	0.011721	0.030697	0.056057	0.104528	-0.047147	
ZNGA	0.006345	0.027668	-0.132098	0.133829	-0.038994	

	Median Returns	Total Return	Average Return Days
STNE	0.009560	3.981261	0.000459
CRWD	0.012340	1.938286	0.000226
ZNGA	0.007592	1.371309	0.000160

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

```
[37]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
STNE	0.014717	0.073278	0.064377	0.273783	-0.122742	
CRWD	0.011721	0.030697	0.056057	0.104528	-0.047147	
ZNGA	0.006345	0.027668	-0.132098	0.133829	-0.038994	

	Median Returns	Total Return	Average Return Days	CAGR
STNE	0.009560	3.981261	0.000459	7.352104
CRWD	0.012340	1.938286	0.000226	6.097661
ZNGA	0.007592	1.371309	0.000160	1.837513

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

```
[38]:
```

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
ZNGA	0.006345	0.027668	-0.132098	0.133829	-0.038994	
CRWD	0.011721	0.030697	0.056057	0.104528	-0.047147	

STNE	0.014717	0.073278	0.064377	0.273783	-0.122742
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	Median Returns	Total Return	Average Return Days	CAGR
ZNGA	0.007592	1.371309	0.000160	1.837513
CRWD	0.012340	1.938286	0.000226	6.097661
STNE	0.009560	3.981261	0.000459	7.352104