Cloud_Platform_Portfolio

September 29, 2021

1 Cloud Platform 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
# Cloud Business
title = "Cloud Stock"
symbols = ['MSFT', 'AMZN', 'BOX','BABA','TEAM','ECOM', 'AVLR', 'ZM']
start = '2019-12-01'
end = '2020-06-01'
```

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

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

[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]:
```
179
```

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

[7]:

Date	MSFT	AMZN	BOX	BABA	TEAM	ECOM	AVLR	ZM
2019-12-02	148.728424	1781.599976	18.100000	196.309998	120.839996	9.09	74.449997	68.930000
2019-12-03	148.489731	1769.959961	18.110001	194.899994	123.550003	9.12	75.470001	70.019997
2019-12-04	149.026764	1760.689941	17.600000	193.740005	123.470001	9.14	75.230003	69.959999
2019-12-05	149.106323	1740.479980	17.389999	200.000000	123.949997	9.11	75.760002	69.669998
2019-12-06	150.916321	1751.599976	17.340000	201.889999	123.360001	9.21	75.589996	62.740002

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

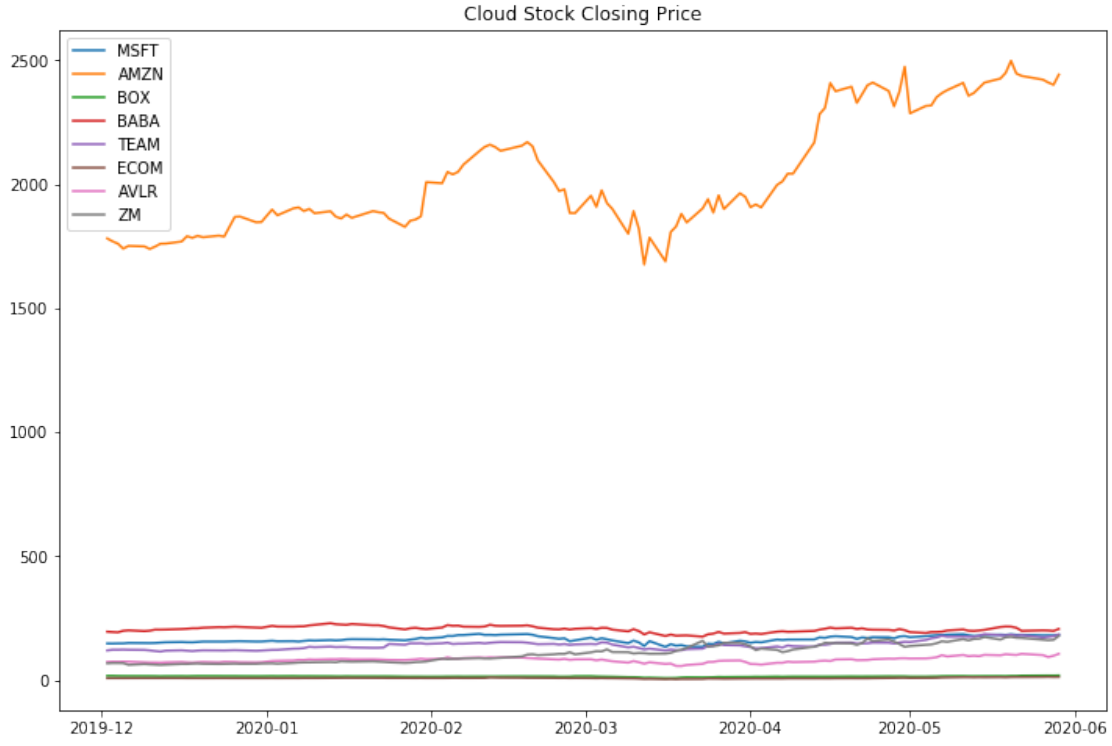
[8]:

Date	MSFT	AMZN	BOX	BABA	TEAM	ECOM	AVLR	ZM
2020-05-22	183.509995	2436.879883	19.090000	199.699997	182.600006	13.47	107.019997	171.059998
2020-05-26	181.570007	2421.860107	19.520000	201.720001	173.059998	13.82		
2020-05-27	181.809998	2410.389893	19.639999	201.179993	175.309998	14.27		
2020-05-28	181.399994	2401.100098	19.490000	199.490005	178.320007	13.71		
2020-05-29	183.250000	2442.370117	19.980000	207.389999	185.300003	13.95		

```
2020-05-26  102.000000  164.000000
2020-05-27   94.080002  161.970001
2020-05-28   99.279999  163.550003
2020-05-29  107.059998  179.479996
```

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

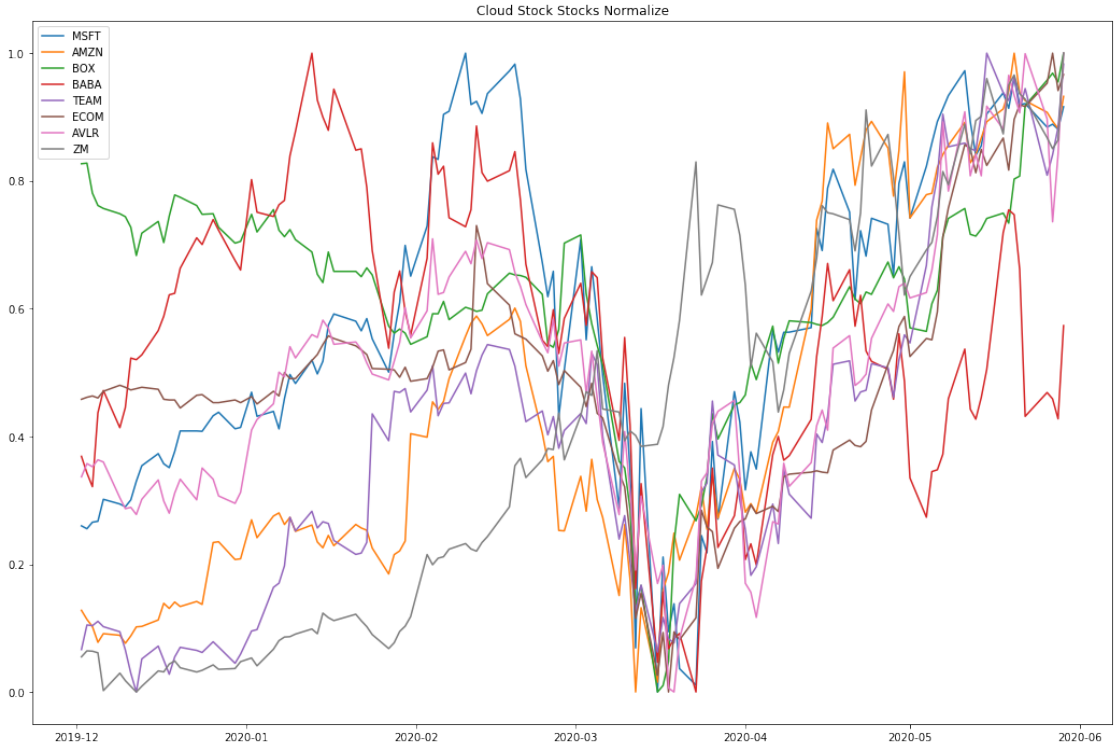
[9]: <matplotlib.legend.Legend at 0x1db125e6f28>



[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)
```

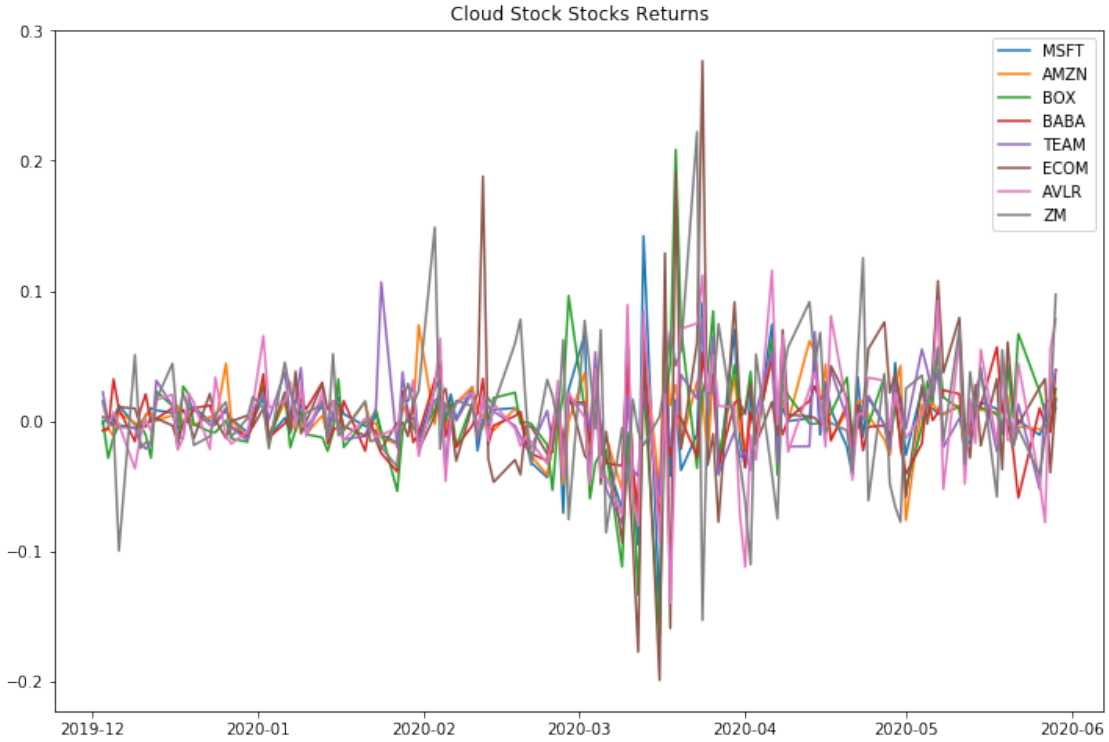
[11]: <matplotlib.legend.Legend at 0x1db1384bc18>



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

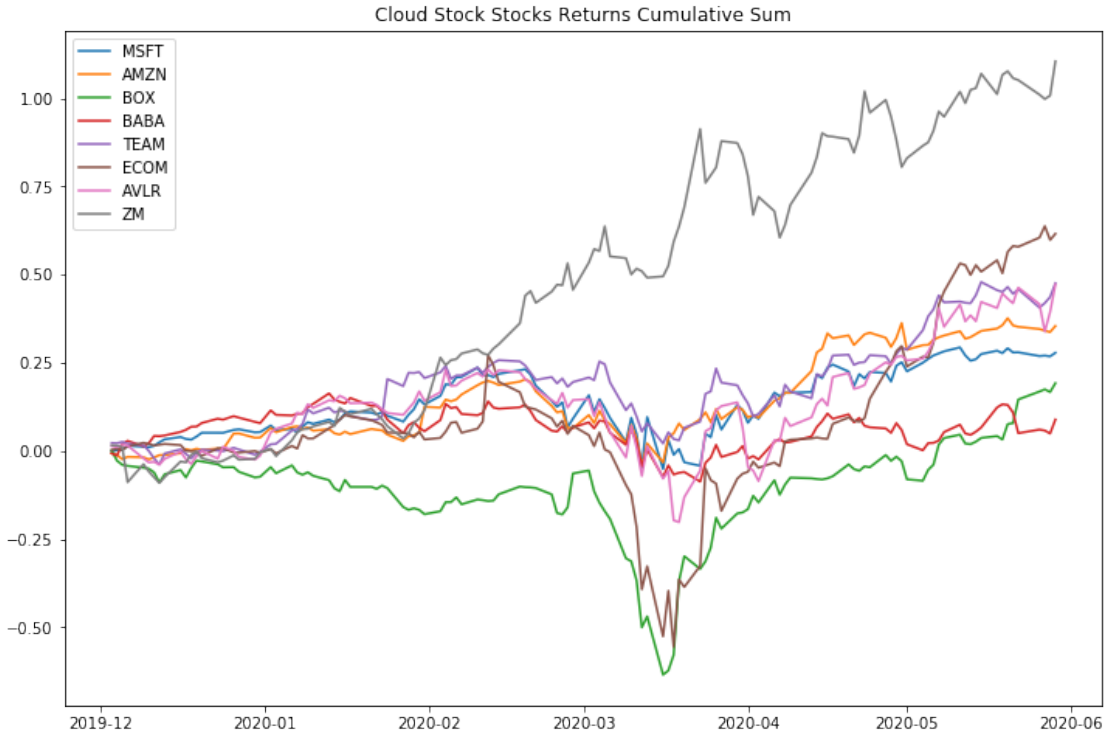
[13]:
```
plt.figure(figsize=(12,8))
plt.plot(stock_rets)
plt.title(title + ' Stocks Returns')
plt.legend(labels=stock_rets.columns)
```

[13]: <matplotlib.legend.Legend at 0x1db126b2780>

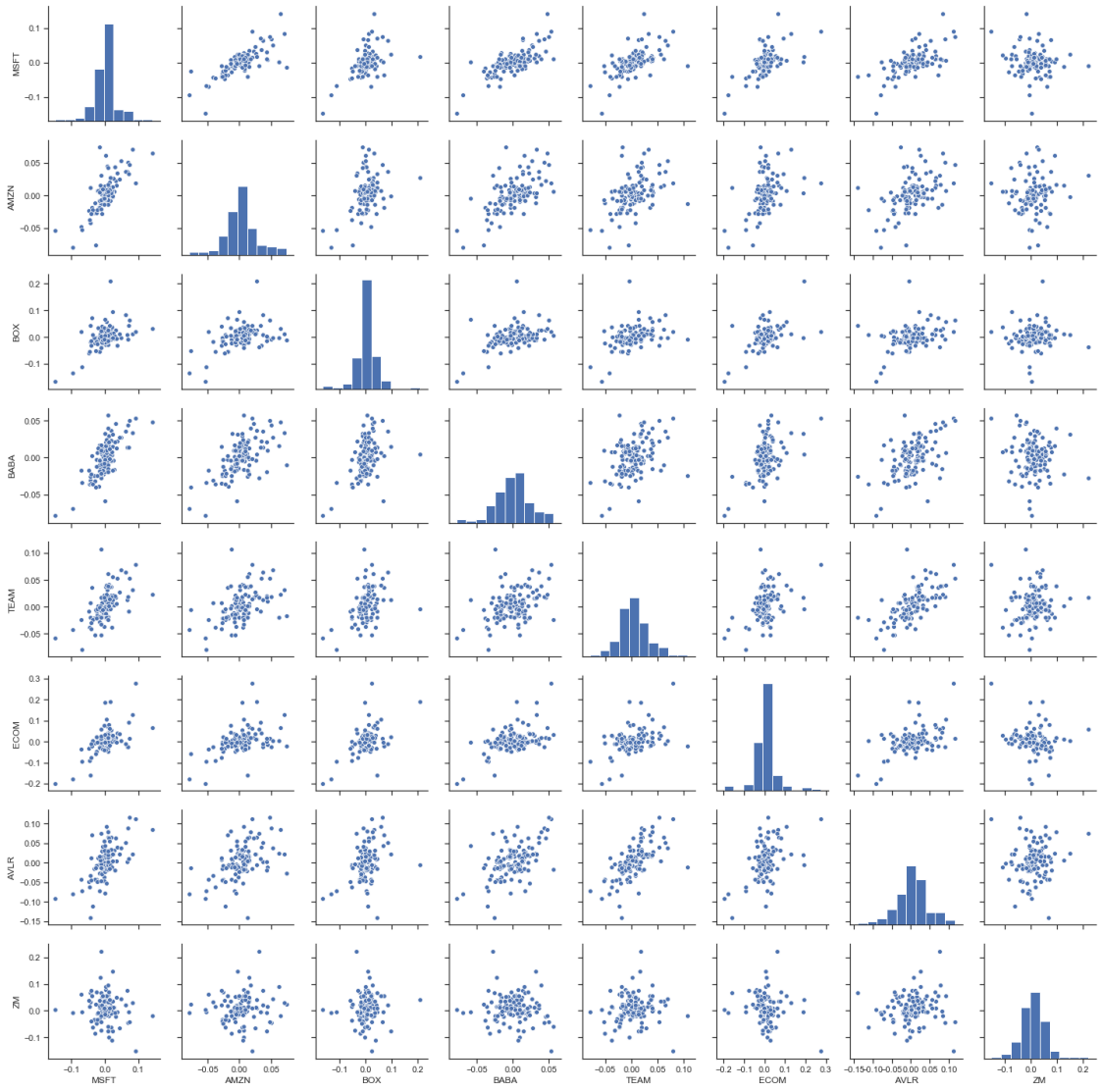


[14]:
```
plt.figure(figsize=(12,8))
plt.plot(stock_rets.cumsum())
plt.title(title + ' Stocks Returns Cumulative Sum')
plt.legend(labels=stock_rets.columns)
```

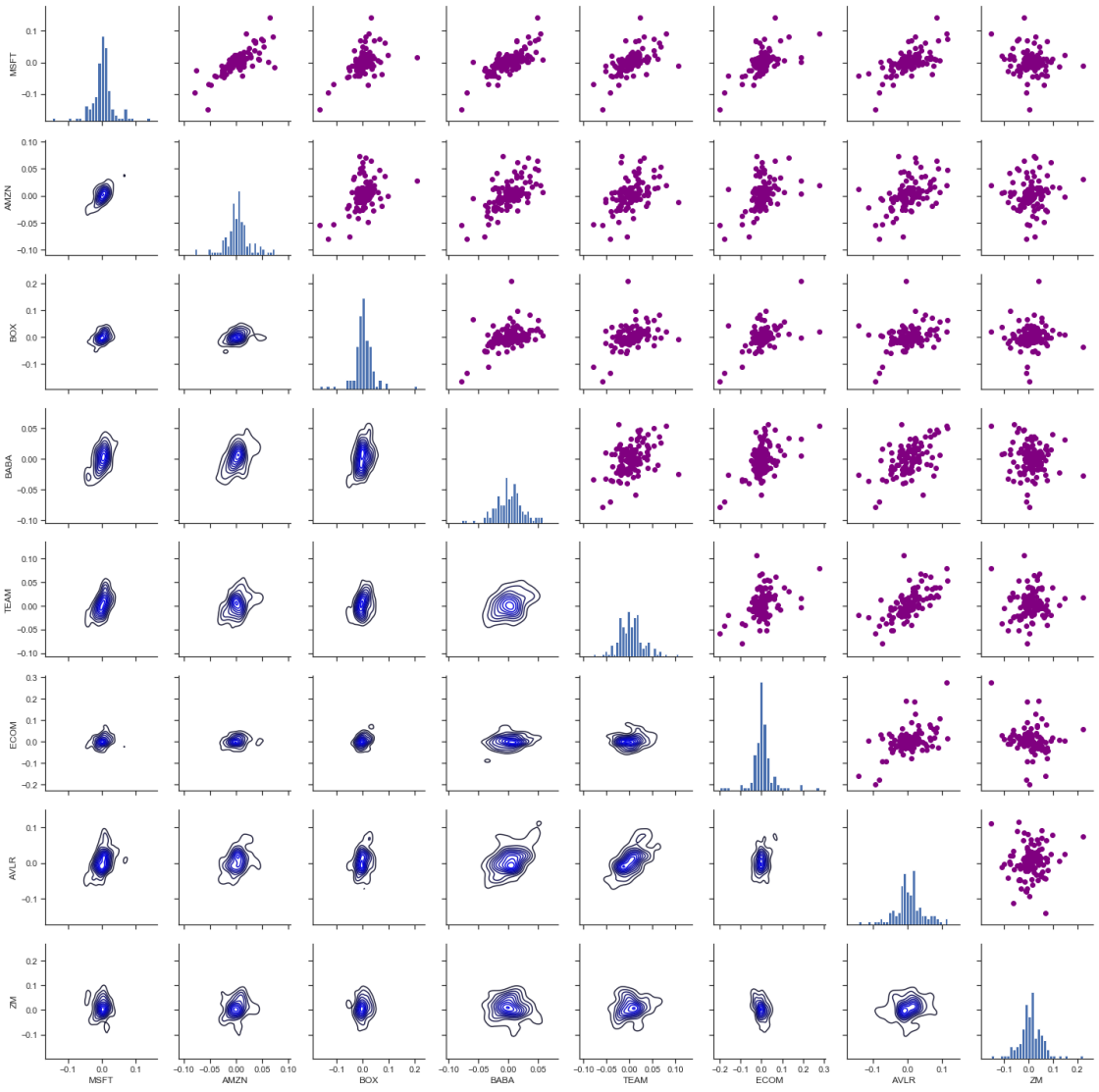
[14]:
```
<matplotlib.legend.Legend at 0x1db1270ec18>
```



[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)
```

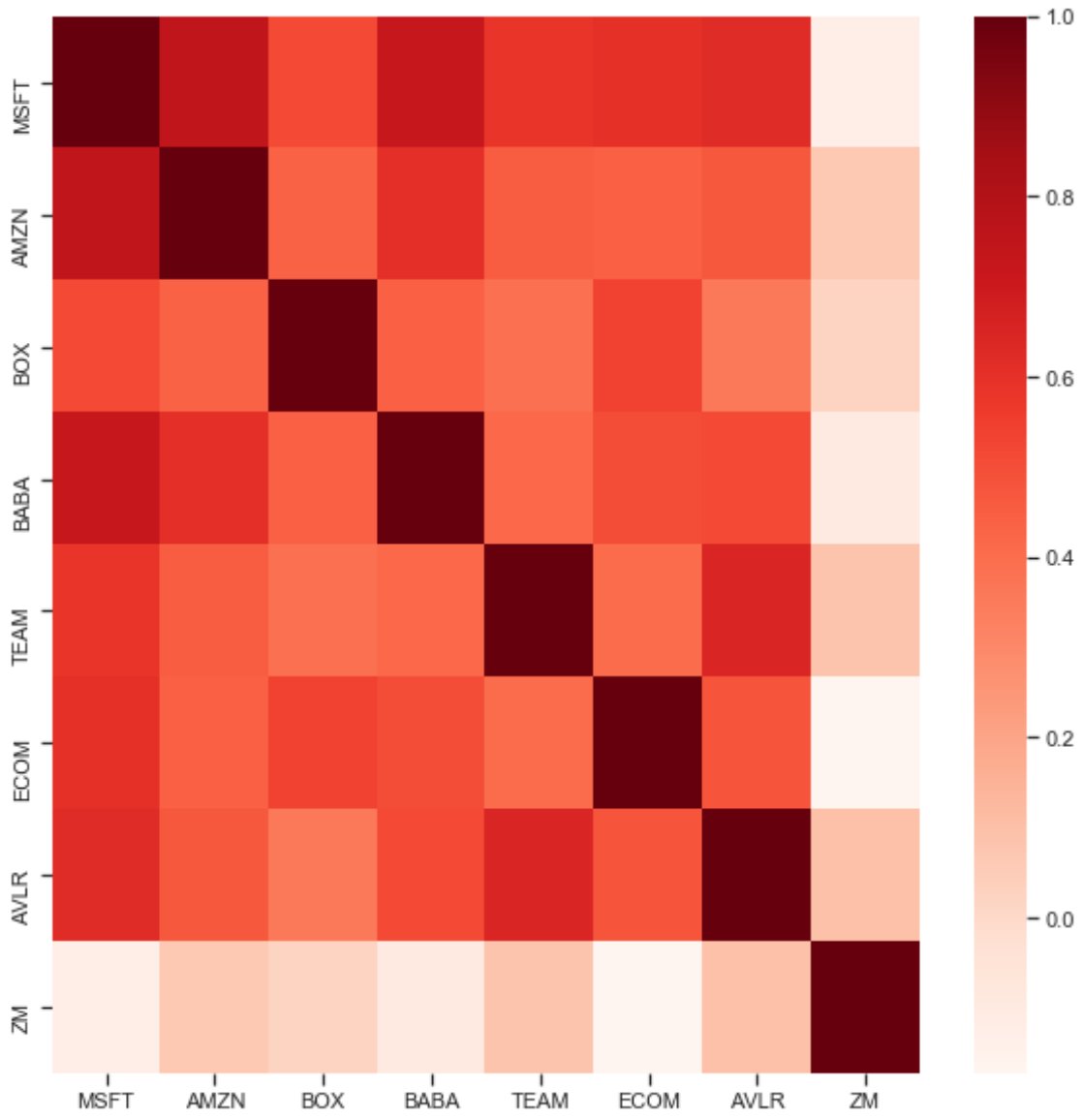


[16]:
```
ax = sns.PairGrid(stock_rets)
ax.map_upper(plt.scatter, color='purple')
ax.map_lower(sns.kdeplot, color='blue')
ax.map_diag(plt.hist, bins=30)
for i in range(nplot) :
    for j in range(nplot) :
        ax.axes[i, j].locator_params(axis='x', nbins=6, tight=True)
```



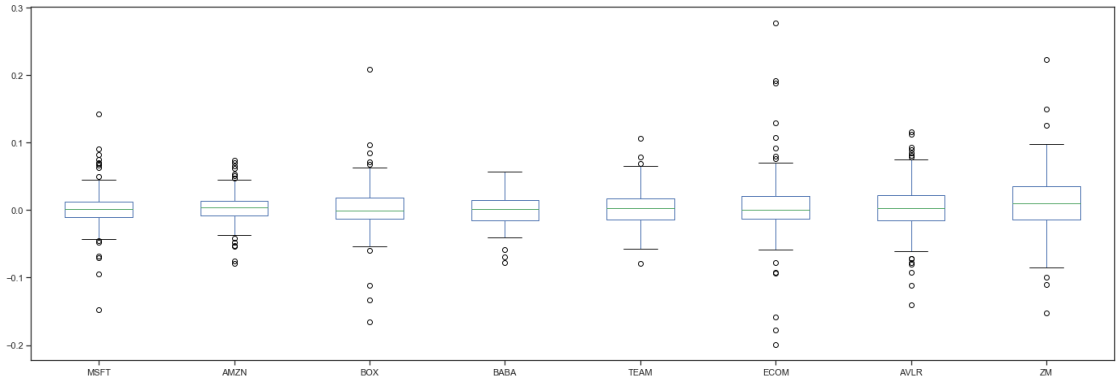
[17]:
```
plt.figure(figsize=(10,10))
corr = stock_rets.corr()

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

[17]: <matplotlib.axes._subplots.AxesSubplot at 0x1db1a2c4470>

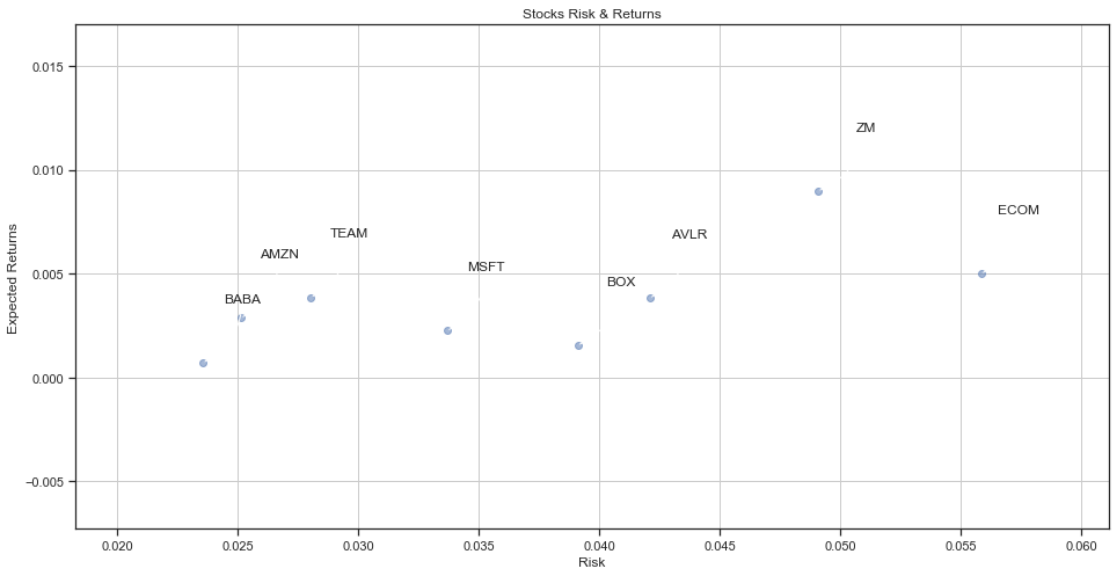


[18]:
```
# Box plot
stock_rets.plot(kind='box',figsize=(24,8))
```

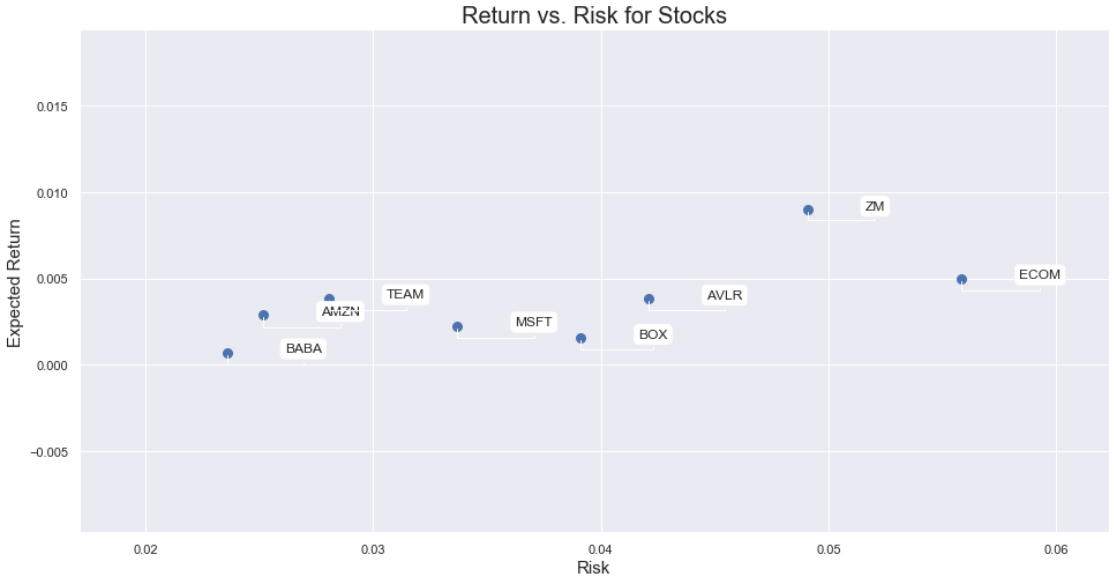
[18]: <matplotlib.axes._subplots.AxesSubplot at 0x1db1a1802b0>



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

[21]:
```
ZM    ZM      1.000000
AVLR  AVLR    1.000000
AMZN  AMZN    1.000000
BOX   BOX     1.000000
BABA  BABA    1.000000
```

```
TEAM  TEAM    1.000000
ECOM  ECOM    1.000000
MSFT  MSFT    1.000000
AMZN  MSFT    0.751667
MSFT  AMZN    0.751667
BABA  MSFT    0.727818
MSFT  BABA    0.727818
AVLR  TEAM    0.649127
TEAM  AVLR    0.649127
MSFT  AVLR    0.621726
AVLR  MSFT    0.621726
AMZN  BABA    0.610477
BABA  AMZN    0.610477
ECOM  MSFT    0.602911
MSFT  ECOM    0.602911
      TEAM    0.583097
TEAM  MSFT    0.583097
ECOM  BOX     0.537930
BOX   ECOM    0.537930
AVLR  BABA    0.518445
BABA  AVLR    0.518445
MSFT  BOX     0.514382
BOX   MSFT    0.514382
ECOM  BABA    0.508249
BABA  ECOM    0.508249
                …
AMZN  TEAM    0.450339
TEAM  AMZN    0.450339
BOX   BABA    0.445309
BABA  BOX     0.445309
AMZN  ECOM    0.442798
ECOM  AMZN    0.442798
BOX   AMZN    0.439672
AMZN  BOX     0.439672
BABA  TEAM    0.418024
TEAM  BABA    0.418024
ECOM  TEAM    0.407615
TEAM  ECOM    0.407615
BOX   TEAM    0.393006
TEAM  BOX     0.393006
BOX   AVLR    0.363039
AVLR  BOX     0.363039
ECOM  ZM      0.173027
ZM    ECOM    0.173027
      MSFT    0.127018
MSFT  ZM      0.127018
AVLR  ZM      0.100528
```

```
ZM    AVLR    0.100528
      BABA    0.092025
BABA  ZM      0.092025
ZM    TEAM    0.087617
TEAM  ZM      0.087617
AMZN  ZM      0.069392
ZM    AMZN    0.069392
BOX   ZM      0.027906
ZM    BOX     0.027906
Length: 64, dtype: float64
```

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

[22]:

Date	MSFT	AMZN	BOX	BABA	TEAM	ECOM	AVLR	ZM
2019-12-03	0.503474	0.475043	0.444131	0.525086	0.545490	0.424976	0.600548	0.449612
2019-12-04	0.521506	0.483513	0.367378	0.534201	0.421017	0.422646	0.534521	0.405159
2019-12-05	0.510860	0.442726	0.410760	0.817536	0.445481	0.411127	0.574515	0.396391
2019-12-06	0.550939	0.559497	0.434969	0.648249	0.398832	0.441126	0.538182	0.142200
2019-12-09	0.500141	0.509944	0.428780	0.462737	0.400021	0.438591	0.404661	0.543452

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

[23]:

	MSFT	AMZN	BOX	BABA	TEAM	ECOM	AVLR	ZM
MSFT	1.000000	0.751667	0.514382	0.727818	0.583097	0.602911	0.621726	-0.127018
AMZN	0.751667	1.000000	0.439672	0.610477	0.450339	0.442798	0.471472	0.069392
BOX	0.514382	0.439672	1.000000	0.445309	0.393006	0.537930	0.363039	0.027906
BABA	0.727818	0.610477	0.445309	1.000000	0.418024	0.508249	0.518445	-0.092025
TEAM	0.583097	0.450339	0.393006	0.418024	1.000000	0.407615	0.649127	0.087617
ECOM	0.602911	0.442798	0.537930	0.508249	0.407615	1.000000	0.480489	-0.173027
AVLR	0.621726	0.471472	0.363039	0.518445	0.649127	0.480489	1.000000	0.100528
ZM	-0.127018	0.069392	0.027906	-0.092025	0.087617	-0.173027	0.100528	1.000000

```
BABA -0.092025
TEAM  0.087617
ECOM -0.173027
AVLR  0.100528
ZM    1.000000
```

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

[24]:
```
ZM    ZM      1.000000
AVLR  AVLR    1.000000
AMZN  AMZN    1.000000
BOX   BOX     1.000000
BABA  BABA    1.000000
TEAM  TEAM    1.000000
ECOM  ECOM    1.000000
MSFT  MSFT    1.000000
AMZN  MSFT    0.751667
MSFT  AMZN    0.751667
BABA  MSFT    0.727818
MSFT  BABA    0.727818
AVLR  TEAM    0.649127
TEAM  AVLR    0.649127
MSFT  AVLR    0.621726
AVLR  MSFT    0.621726
AMZN  BABA    0.610477
BABA  AMZN    0.610477
ECOM  MSFT    0.602911
MSFT  ECOM    0.602911
      TEAM    0.583097
TEAM  MSFT    0.583097
ECOM  BOX     0.537930
BOX   ECOM    0.537930
AVLR  BABA    0.518445
BABA  AVLR    0.518445
MSFT  BOX     0.514382
BOX   MSFT    0.514382
ECOM  BABA    0.508249
BABA  ECOM    0.508249
                 …
AMZN  TEAM    0.450339
TEAM  AMZN    0.450339
BOX   BABA    0.445309
BABA  BOX     0.445309
AMZN  ECOM    0.442798
ECOM  AMZN    0.442798
```

```
BOX   AMZN    0.439672
AMZN  BOX     0.439672
BABA  TEAM    0.418024
TEAM  BABA    0.418024
ECOM  TEAM    0.407615
TEAM  ECOM    0.407615
BOX   TEAM    0.393006
TEAM  BOX     0.393006
BOX   AVLR    0.363039
AVLR  BOX     0.363039
ECOM  ZM      0.173027
ZM    ECOM    0.173027
      MSFT    0.127018
MSFT  ZM      0.127018
AVLR  ZM      0.100528
ZM    AVLR    0.100528
      BABA    0.092025
BABA  ZM      0.092025
ZM    TEAM    0.087617
TEAM  ZM      0.087617
AMZN  ZM      0.069392
ZM    AMZN    0.069392
BOX   ZM      0.027906
ZM    BOX     0.027906
Length: 64, dtype: float64
```

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

```
Stock returns:
MSFT    0.002263
AMZN    0.002881
BOX     0.001562
BABA    0.000723
TEAM    0.003868
ECOM    0.005007
AVLR    0.003840
ZM      0.008986
dtype: float64
--------------------------------------------------
Stock risks:
MSFT    0.033700
AMZN    0.025151
BOX     0.039116
```

```
BABA    0.023570
TEAM    0.028044
ECOM    0.055849
AVLR    0.042101
ZM      0.049077
dtype: float64
```

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

[26]:

	Returns	Risk
BABA	0.000723	0.023570
BOX	0.001562	0.039116
MSFT	0.002263	0.033700
AMZN	0.002881	0.025151
AVLR	0.003840	0.042101
TEAM	0.003868	0.028044
ECOM	0.005007	0.055849
ZM	0.008986	0.049077

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

[27]:

	Returns	Risk
BABA	0.000723	0.023570
AMZN	0.002881	0.025151
TEAM	0.003868	0.028044
MSFT	0.002263	0.033700
BOX	0.001562	0.039116
AVLR	0.003840	0.042101
ZM	0.008986	0.049077
ECOM	0.005007	0.055849

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

[28]:

	Returns	Risk	Sharpe Ratio
MSFT	0.002263	0.033700	-0.229579
AMZN	0.002881	0.025151	-0.283029
BOX	0.001562	0.039116	-0.215711
BABA	0.000723	0.023570	-0.393575
TEAM	0.003868	0.028044	-0.218661
ECOM	0.005007	0.055849	-0.089397
AVLR	0.003840	0.042101	-0.146316
ZM	0.008986	0.049077	-0.020652

[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	\
MSFT	0.002263	0.033700	-0.229579	0.142169	-0.147390	
AMZN	0.002881	0.025151	-0.283029	0.073791	-0.079221	
BOX	0.001562	0.039116	-0.215711	0.208506	-0.165599	
BABA	0.000723	0.023570	-0.393575	0.056952	-0.078093	
TEAM	0.003868	0.028044	-0.218661	0.106680	-0.078692	
ECOM	0.005007	0.055849	-0.089397	0.276632	-0.198708	
AVLR	0.003840	0.042101	-0.146316	0.115826	-0.139837	
ZM	0.008986	0.049077	-0.020652	0.222214	-0.152795	

	Median Returns	Total Return
MSFT	0.001828	1.019849
AMZN	0.003638	1.718796
BOX	-0.001301	2.514109
BABA	0.001700	3.960095
TEAM	0.002327	3.914309
ECOM	0.000000	1.750545
AVLR	0.002832	7.836421
ZM	0.009621	9.740136

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

[33]:

	Returns	Risk	Sharpe Ratio	Max Returns	Min Returns	\
MSFT	0.002263	0.033700	-0.229579	0.142169	-0.147390	
AMZN	0.002881	0.025151	-0.283029	0.073791	-0.079221	
BOX	0.001562	0.039116	-0.215711	0.208506	-0.165599	
BABA	0.000723	0.023570	-0.393575	0.056952	-0.078093	
TEAM	0.003868	0.028044	-0.218661	0.106680	-0.078692	
ECOM	0.005007	0.055849	-0.089397	0.276632	-0.198708	
AVLR	0.003840	0.042101	-0.146316	0.115826	-0.139837	
ZM	0.008986	0.049077	-0.020652	0.222214	-0.152795	

	Median Returns	Total Return	Average Return Days
MSFT	0.001828	1.019849	0.000057
AMZN	0.003638	1.718796	0.000095
BOX	-0.001301	2.514109	0.000139

```
BABA        0.001700      3.960095             0.000217
TEAM        0.002327      3.914309             0.000215
ECOM        0.000000      1.750545             0.000097
AVLR        0.002832      7.836421             0.000422
ZM          0.009621      9.740136             0.000519
```

[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	Median Returns	Total Return	Average Return Days	CAGR
MSFT	0.002263	0.033700	-0.229579	0.142169	-0.147390	0.001828	1.019849	0.000057	0.341588
AMZN	0.002881	0.025151	-0.283029	0.073791	-0.079221	0.003638	1.718796	0.000095	0.559098
BOX	0.001562	0.039116	-0.215711	0.208506	-0.165599	-0.001301	2.514109	0.000139	0.149263
BABA	0.000723	0.023570	-0.393575	0.056952	-0.078093	0.001700	3.960095	0.000217	0.080364
TEAM	0.003868	0.028044	-0.218661	0.106680	-0.078692	0.002327	3.914309	0.000215	0.825504
ECOM	0.005007	0.055849	-0.089397	0.276632	-0.198708	0.000000	1.750545	0.000097	0.827550
AVLR	0.003840	0.042101	-0.146316	0.115826	-0.139837	0.002832	7.836421	0.000422	0.667637
ZM	0.008986	0.049077	-0.020652	0.222214	-0.152795	0.009621	9.740136	0.000519	2.846816

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

[35]:

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TEAM	0.002327	3.914309	0.000215	0.825504
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