2. Carry Trade Implementation

November 19, 2019

1 FX Leverage Carry-Trade

1.1 1. Load data

```
[2]: data_path_new = 'Data/to_send.csv'
final_data_new = pd.read_csv(data_path_new)

[3]: final_data_new = final_data_new.set_index('Date')
final_data_new.index = pd.to_datetime(final_data_new.index)

[4]: #final_data_new.iloc[324:]
```

1.2 2. Implement Strategy

```
[5]: # if you want to test single currency or single trading period, just modify the following list

fx_list = ['USD', 'AUD', 'GBP']

period_list = [7, 30, 60]

[6]: # idx, row_row = next(final_data.iterrows())

# ct.find_max_signal(row_row, period_list, fx_list)
```

```
[7]: results_07 = ct.algo_loop(final_data_new, fx_list, period_list, leverage = 2.0)
     results 09 = ct.algo loop(final data new.iloc[324:], fx list, period list,
      \rightarrowleverage = 2.0)
    2019-11-19 22:17:09:476546: Beginning Carry-Trade Strategy run
    2019-11-19 22:17:23:462677: Algo run complete.
    2019-11-19 22:17:23:463248: Beginning Carry-Trade Strategy run
    2019-11-19 22:17:36:658169: Algo run complete.
 [9]: results 07.to csv('Results/results 07.csv')
     results_09.to_csv('Results/results_09.csv')
 [8]: results 09
 [8]:
                      Signal FX_name Period Foreign_IR Domestic_IR FX_Rate \
    Date
     2009-01-02 0.000484649
                                 GBP
                                          2M
                                                0.02505
                                                          0.0075625
                                                                     133.563
     2009-01-05 0.000484649
                                 GBP
                                          2M
                                                0.02445
                                                                     137.348
                                                          0.0073875
     2009-01-06 0.000484649
                                 GBP
                                          2M 0.0240875
                                                          0.0071875
                                                                      139.69
     2009-01-07 0.000484649
                                 GBP
                                          2M
                                               0.023675
                                                           0.007075
                                                                      139.85
     2009-01-08 0.000484649
                                 GBP
                                          2M 0.0230375
                                                          0.0069375 138.773
                                                                      75.247
     2019-11-05 5.13588e-05
                                 AUD
                                          2M
                                                 0.0092
                                                          -0.000965
                                          2M
     2019-11-06 5.13588e-05
                                 AUD
                                                 0.0093
                                                          -0.000935
                                                                      75.024
     2019-11-07 5.13588e-05
                                 AUD
                                          2M
                                               0.009183
                                                         -0.0010383
                                                                      75.385
     2019-11-08 5.13588e-05
                                 AUD
                                          2M
                                               0.009216
                                                           -0.00109
                                                                      74.938
     2019-11-11 5.13588e-05
                                 AUD
                                          2M
                                                 0.0091
                                                          -0.001135
                                                                      74.704
                  Equity Asset Pos Unreal_Return Real_Return Drawdown
    Date
     2009-01-02
                   10000
                             20000
                                       0.0014587
                                                                      0
                                                            0
     2009-01-05
                   10000
                             20000
                                                            0
                                                                      0
                                       0.0591682
     2009-01-06
                   10000
                             20000
                                       0.0946591
                                                            0
                                                                      0
     2009-01-07
                   10000
                             20000
                                        0.096636
                                                            0
                                                                      0
     2009-01-08
                   10000
                             20000
                                       0.0805256
                                                            0
                                                                      0
                           33623.7
                                       0.0857553
                                                     0.681185 -0.301316
     2019-11-05 16811.8
     2019-11-06 16811.8
                           33623.7
                                       0.0795475
                                                     0.681185 -0.301316
                           33623.7
     2019-11-07 16811.8
                                       0.0895318
                                                     0.681185 -0.301316
     2019-11-08 16811.8
                                                     0.681185 -0.301316
                           33623.7
                                       0.0746929
     2019-11-11 16811.8
                           33623.7
                                       0.0705299
                                                     0.681185 -0.301316
     [2713 rows x 11 columns]
[21]: print('Cumulative Return, after crisis:', results_09['Real_Return'][-1])
     print('Cumulative Return, before crisis:', results 07['Real Return'][-1])
     print('Max Drawdown, after crisis:', results_09['Drawdown'].min())
     print('Max Drawdown, before crisis:', results_07['Drawdown'].min())
```

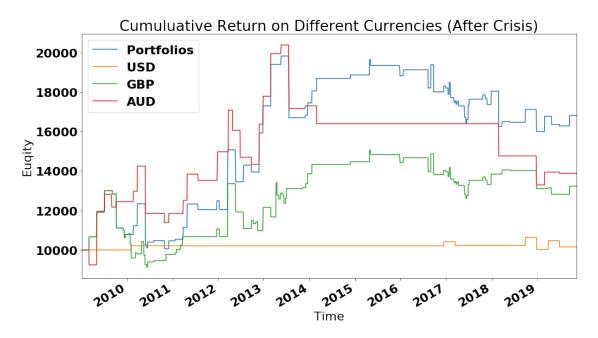
```
Cumulative Return, before crisis: -0.3585947561379189
    Max Drawdown, after crisis: -0.3831904787844942
    Max Drawdown, before crisis: -0.7247113114426323
[11]: # if you want to test single currency or single trading period, just modify the
     → following list
     usd list = ['USD']
     aud_list = ['AUD']
     gbp list = ['GBP']
[12]: results_usd = ct.algo_loop(final_data_new.iloc[324:], usd_list, period_list,_u
     \rightarrowleverage = 2.0)
     results_aud = ct.algo_loop(final_data_new.iloc[324:], aud_list, period_list,_u
      \rightarrowleverage = 2.0)
     results_gbp = ct.algo_loop(final_data_new.iloc[324:], gbp_list, period_list,_u
      \rightarrowleverage = 2.0)
    2019-11-19 22:19:26:896856: Beginning Carry-Trade Strategy run
    2019-11-19 22:19:38:339608: Algo run complete.
    2019-11-19 22:19:38:340064: Beginning Carry-Trade Strategy run
    2019-11-19 22:19:45:937960: Algo run complete.
    2019-11-19 22:19:45:938454: Beginning Carry-Trade Strategy run
    2019-11-19 22:19:54:804058: Algo run complete.
[13]: results usd.to csv('Results/results usd 09.csv')
     print('Cumulative Return, USD only:', results_usd['Real_Return'][-1])
    Cumulative Return, USD only: 0.015332769349171382
[14]: results_aud.to_csv('Results/results_aud_09.csv')
     print('Cumulative Return, AUD only:', results aud['Real Return'][-1])
    Cumulative Return, AUD only: 0.3883647339032503
[15]: results_gbp.to_csv('Results/results_gbp_09.csv')
     print('Cumulative Return, GBP only:', results_gbp['Real_Return'][-1])
    Cumulative Return, GBP only: 0.32343472268958107
    1.3 3. Analysis
    1.4 3.1 After Crisis
```

Cumulative Return, after crisis: 0.6811848552276105

[USD, GBP, AUD]

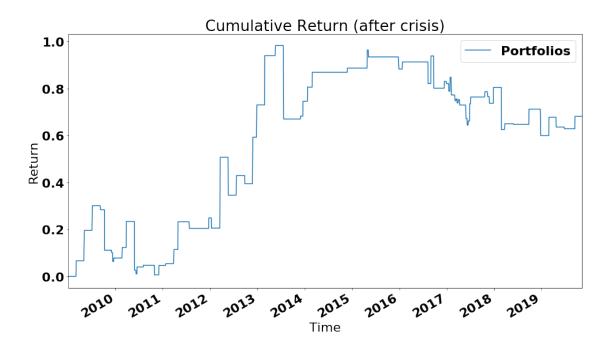
put in one time series graph: [USD], [GBP], [AUD]

```
[29]: results_09['Equity'].plot(label='Portfolios')
    results_usd['Equity'].plot(label='USD')
    results_gbp['Equity'].plot(label='GBP')
    results_aud['Equity'].plot(label='AUD')
    plt.legend()
    plt.xlabel('Time')
    plt.ylabel('Euqity')
    plt.title('Cumuluative Return on Different Currencies (After Crisis)')
    plt.savefig('Results/Real_Return_4comb_post_crisis.jpg')
```



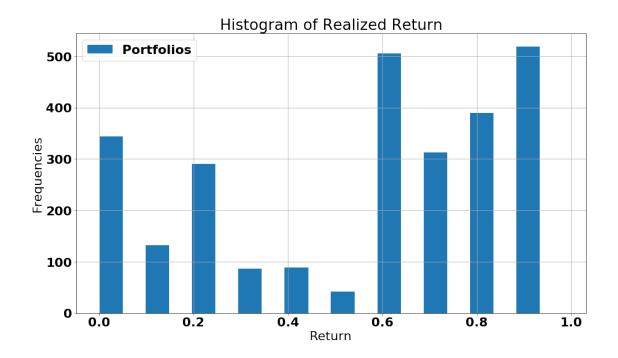
1.4.1 3.1.1 Realized Return Time Series

```
[22]: results_09['Real_Return'].plot(label='Portfolios')
plt.legend()
plt.xlabel('Time')
plt.ylabel('Return')
plt.title('Cumulative Return (after crisis)')
plt.savefig('Results/Real_Return_post_crisis.jpg')
```



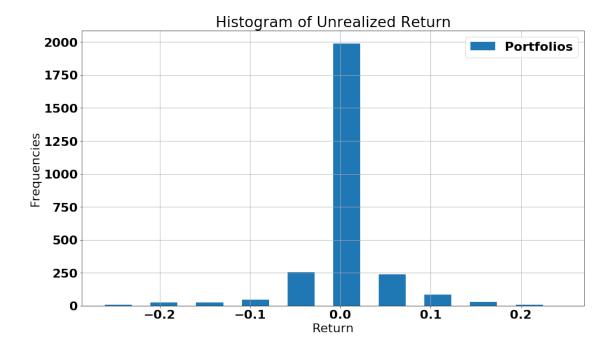
1.4.2 3.1.2 Realized Return Histogram

```
[23]: results_09['Real_Return'].hist(width=0.05, label='Portfolios')
   plt.legend()
   plt.xlabel('Return')
   plt.ylabel('Frequencies')
   plt.title('Histogram of Realized Return')
   plt.savefig('Results/Real_Return_hist_post_crisis.jpg')
```



1.4.3 3.1.3 Unrealized Return Histogram

```
[24]: results_09['Unreal_Return'].hist(width=0.03, label='Portfolios')
plt.legend()
plt.xlabel('Return')
plt.ylabel('Frequencies')
plt.title('Histogram of Unrealized Return')
plt.savefig('Results/Unreal_Return_post_crisis.jpg')
```



1.4.4 3.1.4 Value at Risk

- sort return from smallest to largest
- calculate quantile(0.05) = 95%

```
[25]: return_09 = results_09['Unreal_Return'].sort_values()
    print('VaR at 90%:', return_09.quantile(0.1))
    print('VaR at 95%:', return_09.quantile(0.05))
    print('VaR at 99%:', return_09.quantile(0.01))
    #return_09
```

VaR at 90%: -0.017924809982412096 VaR at 95%: -0.04625538554631749 VaR at 99%: -0.1626349052729469

1.4.5 3.1.5 Sharpe Ratio

mu / sigma

Sharpe Ratio each day: 0.15823949102082854 Sharpe Ratio each year: 2.5069850151429405

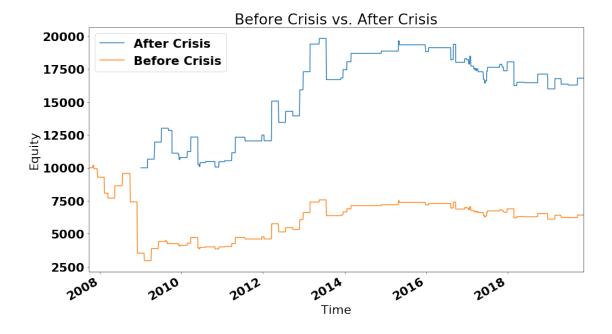
```
[27]: volatility_09 = results_09['Real_Return'].std()
print('Volatility, after crisis:', volatility_09)
```

Volatility, after crisis: 0.31550667123808457

1.5 3.2 Before Crisis vs. After Crisis

time series graph; sharpe ratio

```
[30]: results_09['Equity'].plot(label='After Crisis')
    results_07['Equity'].plot(label='Before Crisis')
    plt.legend()
    plt.xlabel('Time')
    plt.ylabel('Equity')
    plt.title('Before Crisis vs. After Crisis')
    plt.savefig('Results/Real_Return_before&post_crisis.jpg')
```



```
[37]: sharpe_07 = results_07['Unreal_Return'].mean()/results_07['Unreal_Return'].

⇒std()

print('Sharpe Ratio each day, before crisis:', sharpe_07)

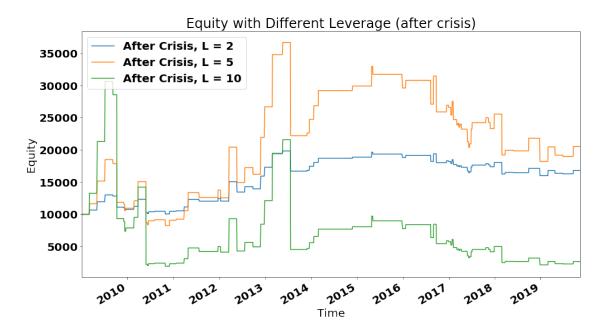
print('Sharpe Ratio each year, before crisis:', sharpe_07 * np.sqrt(251))
```

Sharpe Ratio each day, before crisis: 0.011846622873341955 Sharpe Ratio each year, before crisis: 0.18768580353692282

```
[28]: volatility_07 = results_07['Real_Return'].std()
print('Volatility, after crisis:', volatility_09)
```

```
print('Volatility, before crisis:', volatility_07)
    Volatility, after crisis: 0.31550667123808457
    Volatility, before crisis: 0.1516022133265221
    1.5.1 3.2.1 Value at Risk
[38]: return_09 = results_09['Unreal_Return'].sort_values()
     print('VaR at 90%, after crisis:', return_09.quantile(0.1))
     print('VaR at 95%, after crisis:', return_09.quantile(0.05))
     print('VaR at 99%, after crisis:', return_09.quantile(0.01))
     #return 09
    VaR at 90%, after crisis: -0.017924809982412096
    VaR at 95%, after crisis: -0.04625538554631749
    VaR at 99%, after crisis: -0.1626349052729469
[39]: return_07 = results_07['Unreal_Return'].sort_values()
     print('VaR at 90%, before crisis:', return_07.quantile(0.1))
     print('VaR at 95%, before crisis:', return_07.quantile(0.05))
     print('VaR at 99%, before crisis:', return_07.quantile(0.01))
     #return 09
    VaR at 90%, before crisis: -0.03720340479002022
    VaR at 95%, before crisis: -0.09568580250763162
    VaR at 99%, before crisis: -0.23524271427766016
    1.6 3.3 Leverage Analysis
[31]: results_09_15 = ct.algo_loop(final_data_new.iloc[324:], fx_list, period_list,__
      \rightarrowleverage = 5.0)
     results_09_110 = ct.algo_loop(final_data_new.iloc[324:], fx_list, period_list,__
      \rightarrowleverage = 10.0)
    2019-11-19 22:29:35:299234: Beginning Carry-Trade Strategy run
    2019-11-19 22:29:49:360151: Algo run complete.
    2019-11-19 22:29:49:360672: Beginning Carry-Trade Strategy run
    2019-11-19 22:30:03:264973: Algo run complete.
[44]: results_09['Equity'].plot(label='After Crisis, L = 2')
     results_09_15['Equity'].plot(label='After Crisis, L = 5')
     results_09_l10['Equity'].plot(label='After Crisis, L = 10')
     plt.legend()
     plt.xlabel('Time')
     plt.ylabel('Equity')
     plt.title('Equity with Different Leverage (after crisis)')
```

```
plt.savefig('Results/Real_Return_leverages_post_crisis.jpg')
```



```
[34]: sharpe_09 = results_09['Unreal_Return'].mean()/results_09['Unreal_Return'].

⇒std()
sharpe_09_15 = results_09_15['Unreal_Return'].mean()/

⇒results_09_15['Unreal_Return'].std()
sharpe_09_110 = results_09_110['Unreal_Return'].mean()/

⇒results_09_110['Unreal_Return'].std()
print('L2, Sharpe Ratio each year:', sharpe_09 * np.sqrt(251))
print('L5, Sharpe Ratio each year:', sharpe_09_15 * np.sqrt(251))
print('L10, Sharpe Ratio each year:', sharpe_09_110 * np.sqrt(251))
```

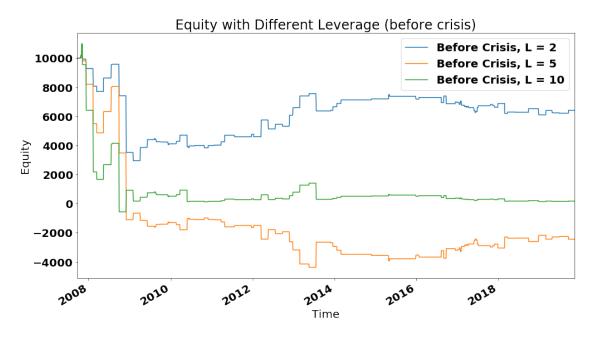
L2, Sharpe Ratio each year: 2.5069850151429405 L5, Sharpe Ratio each year: 2.4878586178373836 L10, Sharpe Ratio each year: 2.481475446063007

```
[33]: results_07_15 = ct.algo_loop(final_data_new, fx_list, period_list, leverage = 5.

→0)
results_07_110 = ct.algo_loop(final_data_new, fx_list, period_list, leverage = 
→10.0)
```

2019-11-19 22:30:29:470253: Beginning Carry-Trade Strategy run 2019-11-19 22:30:44:457826: Algo run complete. 2019-11-19 22:30:44:458072: Beginning Carry-Trade Strategy run 2019-11-19 22:30:58:631277: Algo run complete.

```
[45]: results_07['Equity'].plot(label='Before Crisis, L = 2')
    results_07_15['Equity'].plot(label='Before Crisis, L = 5')
    results_07_110['Equity'].plot(label='Before Crisis, L = 10')
    plt.legend()
    plt.xlabel('Time')
    plt.ylabel('Equity')
    plt.title('Equity with Different Leverage (before crisis)')
    plt.savefig('Results/Real_Return_leverage_before_crisis.jpg')
```



```
[36]: sharpe_07 = results_07['Unreal_Return'].mean()/results_07['Unreal_Return'].

⇒std()
sharpe_07_15 = results_07_15['Unreal_Return'].mean()/
⇒results_07_15['Unreal_Return'].std()
sharpe_07_110 = results_07_110['Unreal_Return'].mean()/
⇒results_07_110['Unreal_Return'].std()
print('L2, Sharpe Ratio each year:', sharpe_07 * np.sqrt(251))
print('L5, Sharpe Ratio each year:', sharpe_07_15 * np.sqrt(251))
print('L10, Sharpe Ratio each year:', sharpe_07_110 * np.sqrt(251))
```

L2, Sharpe Ratio each year: 0.18768580353692282 L5, Sharpe Ratio each year: 0.15795741508622746 L10, Sharpe Ratio each year: 0.14805875981358932