Multidimensional Pair Trading in China's Stock Market

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Abstract

Pair trading is a classic approach to algorithmically trading financial assets such as equities and commodities. Traditional pair trading uses only 2 non-stationary time series to form cointegrated pair with stationarity, where mean-reversion method can be applied. This study extends the pair trading strategy to multi-dimensional cointegrated set of financial time series and test the effectiveness of cointegration on China's stock market, particularly, CSI 300 Index component stocks.

Introduction

Linear regression is a commonly used method to find the relationship between several time series processes. However, Granger and Newbold^[1] argued that linear regression analysis might lead to spurious regression, a situation in which the regression relationship appears to be statistically significant when variables are unrelated. Many economic time series exhibit random walk and other nonstationary behavior, leading to a higher possibility of finding spurious correlation under regression analysis. On the contrast, cointegration theory provides an alternative approach to study non-stationary time series. When a time series is not stationary, there is no tendency for its level to return to a constant mean over time; moreover, the volatility of the process is expected to grow boundlessly in the long run and any prediction based on the historical observations is not possible. Cointegration theory allows the identification of integration among time series that have similar dynamics in the long run and the estimation of their relationships.

Gregory-Hansen Cointegration Test with Structural Breaks: structural breaks or regime shifts are common in financial time series due to market emergencies or major changes in policies and international relationships. Normal cointegration test assumes the cointegration to be both stationary in long-run mean and volatility. Gregory-Hansen's Cointegration Test allows one unknown structural breaks in:

(i) Level shift;

(ii) Level shift with trend;

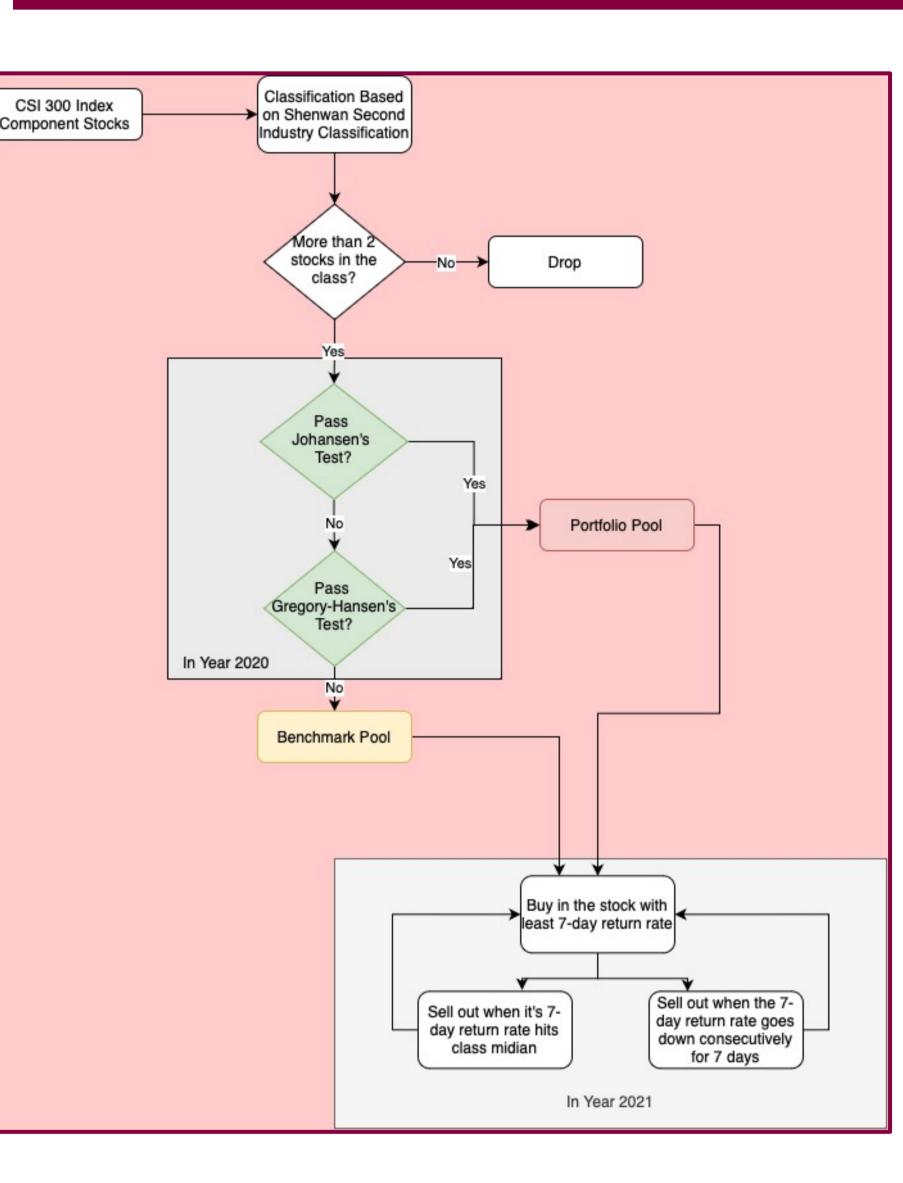
(iii) Regime shift (both level and slope coefficient can change)



In this study, we will be using Johansen's approach to exam cointegration relationship among multiple time series of stock prices. For cointegrated stock groups, we expect the stock with least 7-day return rate to rebound and thus, we buy in and hold the stock until its bounce.

Methodology

Integration: y_t is I(d) if its $(d-1)^{th}$ difference is I(0). That is, $\Delta^d y_t$ is stationary.



Portfolio Pool: 300 stocks are categorized into 62 industrial classes and 37 of them have more than 2 stocks within each class. Among these 37 groups, 18 groups have passed either Johansen's Cointegration Test or Gregory-Hansen's Cointegration Test. **Johansen's Cointegration**

Test: the cointegration test result of industry group

'Insurance'

		-		Johansen	
Number of obs = 240	rend: Constant				
Number of lags = 3	mple : 05jan2020 thru 31aug2020				
Critical					
Trace value				Maximum	
statistic 5%	Eigenvalue	LL	Params	rank	
75.1468 68.52		-922.80642	55	0	
33.2131* 47.21	0.16031	-901.83957	64	1	
19.3380 29.68	0.05617	-894.90204	71	2	
8.8329 15.41	0.04283	-889.64947	76	3	
3.0336 3.76	0.02387	-886.74984	79	4	
	0.01256	-885.23304	80	5	

* selected rank

Augmented Dickey-Fuller Test: ADF test is a common way to assess the integration of a time series. Assume time series y_t follows autoregressive process of order p (AR(p)):

$$\Delta y_t = C + \rho y_{t-1} + \sum_{i=2}^p \phi_i \Delta y_{t-(i-1)} + \varepsilon_t$$

where:

$$\begin{cases} \rho = \sum_{i=1}^{p} \alpha_i - 1 \\ \phi_i = -\sum_{j=i}^{p} \alpha_j \end{cases}$$

The null and alternative hypothesis are:

$$H_0: \rho = 0 \qquad H_1: \rho < 0$$

ADF test statistic is:

$$F_{\tau} = \frac{\hat{\rho}}{SE(\hat{\rho})}$$

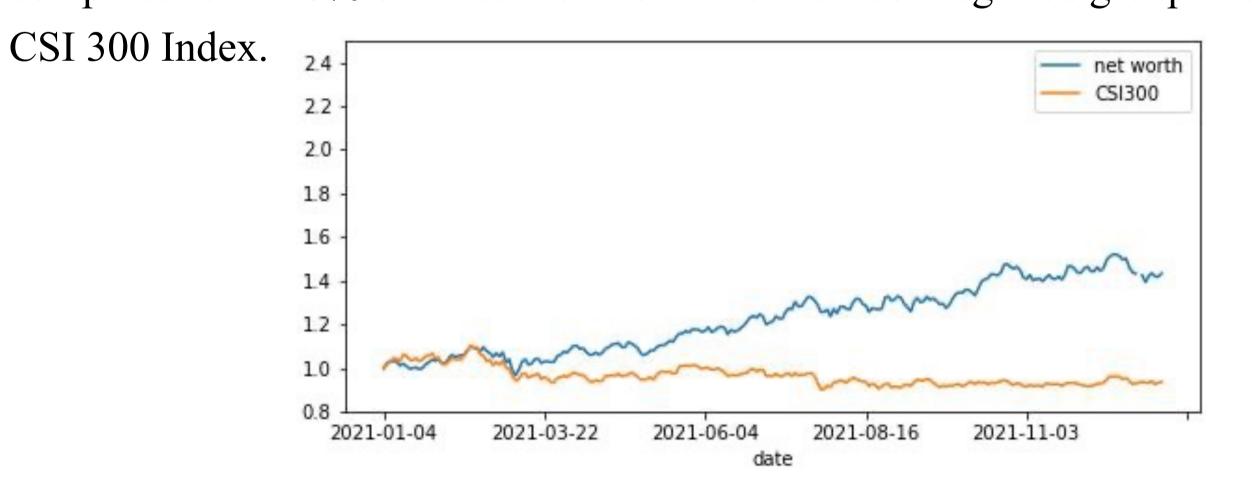
Cointegration: An m * 1 vector time series Y_t is said to be cointegrated of order (d,b), CI(d,b) where $0 < b \le d$, if each of its component series Y_{it} is I(d) but some linear combination $\alpha^T Y_t$ is I(d-b) for some constant vector $\alpha \neq 0$.

Johansen's Cointegration Test: Estimate y_t by vector autoregressive (VAR) process with order p:

Gregory-Hansen's Cointegration Test: the test result of industry group

'Insurance'	Model: Cha	nge in Regime	Cointegratio and Trend kaike criteri	-	ne Shifts Number Maximum		243 8
		Test Statistic	Breakpoint	Date	Asymptot 1%	ic Critical 5%	Values 10%
	ADF Zt	-6.93 -7.17	115 116	25apr2020 26apr2020	-7.31 -7.31	-6.84 -6.84	-6.58 -6.58
	Za	-82.94	116	26apr2020	-100.69	-88.47	-82.30

Investment Result: The annual return of investment for year 2021 is 43.4%, compared to 11.27% of investment return for non-cointegrated group and -7% for



Discussion

• The cointegration filtration exhibits outstanding performance for gaining excess return compared to benchmark (CSI 300 Index)

 $y_t = \Phi_1 y_{t-1} + \cdots + \Phi_p y_{t-p} + a_t + u_t$

The error correction model is:

$$\Delta y_{t} = \Pi y_{t-1} + \sum_{i=1}^{p-1} \Gamma_{i} \Delta y_{t-i} + a_{t} + u_{t}$$
$$\Pi = \sum_{i=1}^{p} \Phi_{i} - I \qquad \Gamma_{i} = -\sum_{j=i+1}^{p} \Phi_{j}$$

Let the rank of be r:

where:

if r = 0: there is no cointegration relationships

if 0 < r < k: there are r cointegration relationships

• Overrejection of Johansen's Cointegration Test when dimensionality is **high:** it has been pointed out that the Johansen's cointegration test tends to overreject the null hypothesis (no cointegration relationship) when either the cross-sectional dimension (N) or time duration dimension (T) is high. Our study result generally supports this, but further research is expected.

Ν	3	4	5	6	7	8	9	10
Equally Weighted Return (no cointegrated class)	16.69%	20.86%	17.92%	5.90%	-23.42%	5.61%	-	12.20%
Equally Weighted Return (cointegrated class)	-17.89%	35.87%	-15.21%	25.31%	50.84%	79.55%	4.63%	2.01%

References: [1] Granger, C. W., & Newbold, P. (1974). Spurious regressions in econometrics. Journal of econometrics, 2(2), 111-120. [2] Caldeira, J., & Moura, G. V. (2013). Selection of a portfolio of pairs based on cointegration: A statistical arbitrage strategy. Available at SSRN 2196391. [3] Chiu, M. C., & Wong, H. Y. (2015). Dynamic cointegrated pairs trading: Mean–variance time-consistent strategies. Journal of Computational and Applied Mathematics, 290, 516-534. [4] Johansen, S. (1988). Statistical analysis of cointegration vectors. Journal of economic dynamics and control, 12(2-3), 231-254. [5] Bykhovskaya, A., & Gorin, V. (2022). Cointegration in large VARs. The Annals of Statistics, 50(3), 1593-1617.