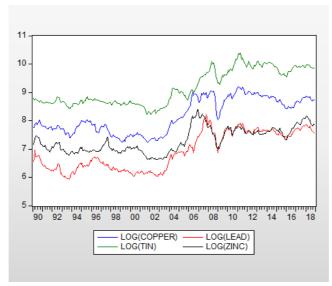
### DSC5211C Quantitative Risk Management Workshop 5

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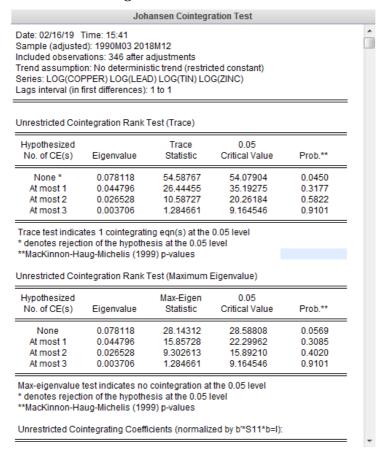
### **LI LIPING (A0186040M)**

#### 1. First, check they are non-stationary!



The prices are not stationary from the graph above. There is not a long-term mean and we see an obvious increase in 05.

#### 2. Then, use the Johansen - Cointegration Test.



LOG(COPPER) -2.664133 4.599743 3.798540 -0.254843	LOG(LEAD) 6.755345 -1.138564 1.556980 0.728597	LOG(TIN) -4.381520 -3.601509 -2.014232 0.615509	LOG(ZINC) -0.442444 0.888597 -5.279514 0.310015	C 18.44987 -3.114668 15.14970 -11.42763				
Unrestricted Adjustment Coefficients (alpha):								
D(LOG(COPP D(LOG(LEAD)) D(LOG(TIN)) D(LOG(ZINC))	0.006838 -0.004979 0.009695 0.001629	-0.004264 0.004102 0.006532 -0.000571	-0.001509 -0.001740 0.000283 0.005745	-0.002849 -0.003780 -0.001506 -0.002782				
1 Cointegrating Equation(s): Log likelihood 2192.584								
Normalized coint	egrating coeffic	ients (standard err	or in parenthese	es)				
LOG(COPPER) 1.000000	LOG(LEAD) -2.535663 (0.45645)	LOG(TIN) 1.644632 (0.40814)	LOG(ZINC) 0.166074 (0.34076)	C -6.925281 (1.93706)				
Adjustment coeffi D(LOG(COPP	Adjustment coefficients (standard error in parentheses) D(LOG(COPP0.018218							
D(LOG(LEAD))	(0.00820) 0.013266 (0.00982)							
D(LOG(TIN))	-0.025830 (0.00746)							
D(LOG(ZINC))	-0.004341 (0.00838)							
2 Cointegrating Equation(s): Log likelihood 2200.512								
		ients (standard er	•	•				
LOG(COPPER) 1.000000	LOG(LEAD) 0.000000	LOG(TIN) -1.045597	LOG(ZINC) 0.196117	C -0.001223				
0.000000	1.000000	(0.17043) -1.060957	(0.24422) 0.011848	(1.03506) 2.730669				
		(0.10856)	(0.15556)	(0.65931)				
-		d error in parenthe	eses)					
D(LOG(COPP	-0.037830 (0.01632)	0.051048 (0.02104)						
D(LOG(LEAD))	0.032134	-0.038308						
	(0.01957)	(0.02522)						
D(LOG(TIN))	0.004214	0.058059						
D(LOG(ZINC))	(0.01477) -0.006969	(0.01904) 0.011657						
D(E0G(ZINO))	(0.01671)	(0.02154)						
3 Cointegrating Equation(s): Log likelihood 2205.164								
Normalized coint		ients (standard en	•	•				
1.000000	LOG(LEAD) 0.000000	LOG(TIN) 0.000000	LOG(ZINC) -1.554440	C 3.157166				
	0.00000	5.00000	(0.22633)	(1.65523)				
0.000000	1.000000	0.000000	-1.764425	5.935456				
0.000000	0.000000	1.000000	(0.25745) -1.674218	(1.88281) 3.020657				
0.000000	0.000000	1.000000	(0.27570)	(2.01632)				
Adjustment coeffic	cients (standar	d error in parenthe	eses)					
D(LOG(COPP	-0.043561	0.048699	-0.011566					
	(0.02006)	(0.02157)	(0.01848)					
D(LOG(LEAD))	0.025526	-0.041016	0.010547					
D/I 00/77111	(0.02404)	(0.02585)	(0.02215)					
D(LOG(TIN))	0.005290 (0.01816)	0.058500 (0.01953)	-0.066575 (0.01673)					
D(LOG(ZINC))	0.014855	0.020603	-0.016654					
	(0.02044)	(0.02198)	(0.01883)					

#### What is the conclusion?

From the result, there is not cointegration equation. Because in the first row, p=0.045 is lower than the significance level so we reject the null hypothesis.

# 3. You decided to proceed and estimate a cointegration model with 1 cointegration equation.

Vector Error Correction Estimates

Vector Error Correction Estimates

Date: 02/16/19 Time: 15:46

Sample (adjusted): 1990M03 2018M12

Included observations: 346 after adjustments

Standard errors in ( ) & t-statistics in [ ]

Cointegrating Eq:	CointEq1	
LOG(COPPER(-1))	1.000000	
LOG(LEAD(-1))	-2.535663 (0.45645) [-5.55520]	
LOG(TIN(-1))	1.644632 (0.40814) [4.02953]	
LOG(ZINC(-1))	0.166074 (0.34076) [0.48736]	
С	-6.925281 (1.93706) [-3.57515]	

From the result, we can see that LOG(ZINC(-1)) is not significant for the long-term relationship so we exclude it from our equation. The equation is shown below:

$$LOG(COPPER(-1)) = 2.54 \cdot LOG(LEAD(-1)) - 1.64 \cdot LOG(TIN(-1)) + 6.93$$

However, the sum of coefficients is not 1.

The system of equations estimated is as follows. Interpret the result. Which prices are the main drivers? How can you improve the model?

Error Correction:	D(LOG(COP	. D(LOG(LEAD))	D(LOG(TIN))	D(LOG(ZINC))
CointEq1	-0.018218 (0.00820)	0.013266 (0.00982)	-0.025830 (0.00746)	-0.004341 (0.00838)
	[-2.22041]	[1.35031]	[-3.46049]	[-0.51821]
D(LOG(COPPER(-1)))	0.420624	0.089441	0.091521	0.191225
	(0.06844)	(0.08195)	(0.06226)	(0.06987)
	[ 6.14621]	[ 1.09147]	[ 1.46997]	[2.73686]
D(LOG(LEAD(-1)))	-0.089978	0.227445	0.016599	-0.052085
	(0.05860)	(0.07017)	(0.05331)	(0.05983)
	[-1.53538]	[3.24128]	[ 0.31135]	[-0.87054]
D(LOG(TIN(-1)))	-0.021968	-0.153571	0.200727	-0.221108
	(0.06506)	(0.07791)	(0.05919)	(0.06643)
	[-0.33765]	[-1.97125]	[3.39117]	[-3.32867]
D(LOG(ZINC(-1)))	0.032673	0.024900	-0.016137	0.303236
	(0.06797)	(0.08139)	(0.06184)	(0.06939)
	[ 0.48070]	[ 0.30595]	[-0.26096]	[ 4.36983]

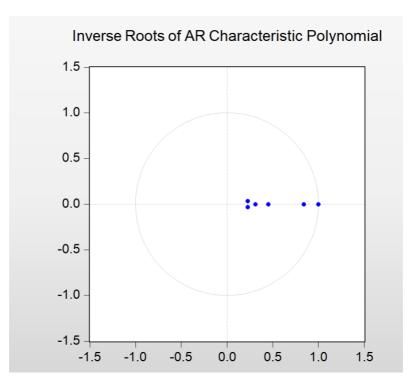
From the result above, we think the variables D(LOG(TIN(-1))) and D(LOG(COPPER(-1))) are more important. Both the prices of TIN and COPPER influence the price of ZINC. Since

the ZINC will not influence and will not be influenced by the long-term relationship so we will exclude ZINC from our model. We need to include the other two variables.

Error Correction:	D(LOG(COP	D(LOG(LEAD))	D(LOG(TIN))	D(LOG(ZINC))
CointEq1	-0.022916	0.020034	-0.025251	-0.004988
	(0.01031)	(0.01214)	(0.00947)	(0.01062)
	[-2.22252]	[1.65062]	[-2.66535]	[-0.46962]
D(LOG(COPPER(-1)))	0.448480	0.106298	0.073327	0.177540
	(0.07206)	(0.08482)	(0.06621)	(0.07423)
D(LOG(COPPER(-2)))	[ 6.22411] -0.143017	[ 1.25327] -0.111641	[1.10754]	[ 2.39167] 0.001596
_((	(0.07228)	(0.08509)	(0.06642)	(0.07447)
	[-1.97855]	[-1.31210]	[-0.05701]	[ 0.02143]
D(LOG(LEAD(-1)))	-0.097823	0.219139	0.026915	-0.054667
	(0.05929)	(0.06979)	(0.05448)	(0.06108)
	[-1.64992]	[3.13996]	[ 0.49406]	[-0.89499]
D(LOG(LEAD(-2)))	0.008355	-0.008847	0.047613	0.011454
	(0.06056)	(0.07128)	(0.05564)	(0.06239)
	[ 0.13797]	[-0.12412]	[ 0.85571]	[ 0.18359]
D(LOG(TIN(-1)))	-0.059336	-0.174132	0.165232	-0.248940
	(0.06734)	(0.07927)	(0.06188)	(0.06938)
	[-0.88113]	[-2.19676]	[2.67041]	[-3.58829]
D(LOG(TIN(-2)))	0.191401	0.200644	0.095623	0.135224
	(0.06828)	(0.08037)	(0.06273)	(0.07034)
	[ 2.80331]	[2.49652]	[1.52424]	[1.92244]
D(LOG(ZINC(-1)))	0.061559	0.029801	-0.014521	0.335260
	(0.07163)	(0.08432)	(0.06582)	(0.07380)
	[ 0.85937]	[ 0.35343]	[-0.22062]	[4.54301]
D(LOG(ZINC(-2)))	0.014934	0.037207	0.041792	-0.061302
	(0.07033)	(0.08279)	(0.06463)	(0.07246)
	[ 0.21232]	[ 0.44941]	[0.64667]	[-0.84601]
С	0.001272	0.001077	0.001946	0.000945
	(0.00306)	(0.00360)	(0.00281)	(0.00315)
	[0.41617]	[0.29930]	[0.69297]	[0.30017]

In the new model, TIN affects the prices of COPPER and LEAD. So TIN is the most important driver among the four prices.

You finally need to analyze the stability conditions to test if any of the parameters or relationship between them is explosive.

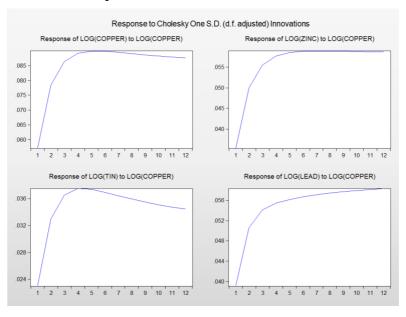


As shown in the graph above, the inverse unit roots of Z lie inside the unit circle, which means the value of Z is higher than 1. Therefore the time series does converge.

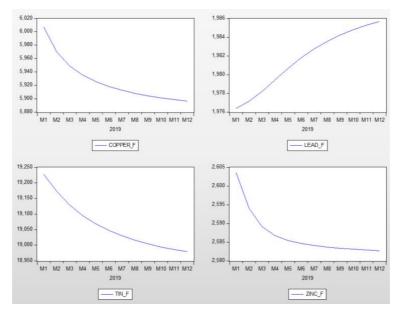
You can test the impact of an "innovation" or "shock" in one of the prices on the long-term impact of the other prices.

The vertical axis is expressed in units of the LOG(COPPER) variable. The solid line is a point estimate for the amount LOG(COPPER) is expected to change following a unit impulse after the number of periods on the horizontal axis.

For example, in the lower left graph, a one-unit price of TIN is expected to lead to a change in the price of COPPER after 5 periods.



Finally, you can test the model's forecasts. In the next figure you find the out-of-sample forecasts for 2019. What can you conclude?



From the long run, the price of COPPER, ZINC and TIN will decrease in 2019. The price of LEAD will increase.