



# **An economic analysis on bitcoin**

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# MOTIVATION FOR CHOOSING THE TOPIC

- Novelty character (faculty recommendation).
- The agresivity with which it appears in international transactions.
- Vastness of the topic.
- The possibility of performing research in an exciting and unexplored topic.



# OBJECTIVES OF THE PAPER

- To understand what bitcoin is and how it's produced.
- To see if bitcoin is or not real money.
- To analyze the two exchange rate series: BTC/USD and BTC/EUR.
- Modelling an ARMA model and forecasting.
- Volatility analysis: ARCH, GARCH, EGARCH, TARARCH.

# IMPORTANCE OF THE THEME

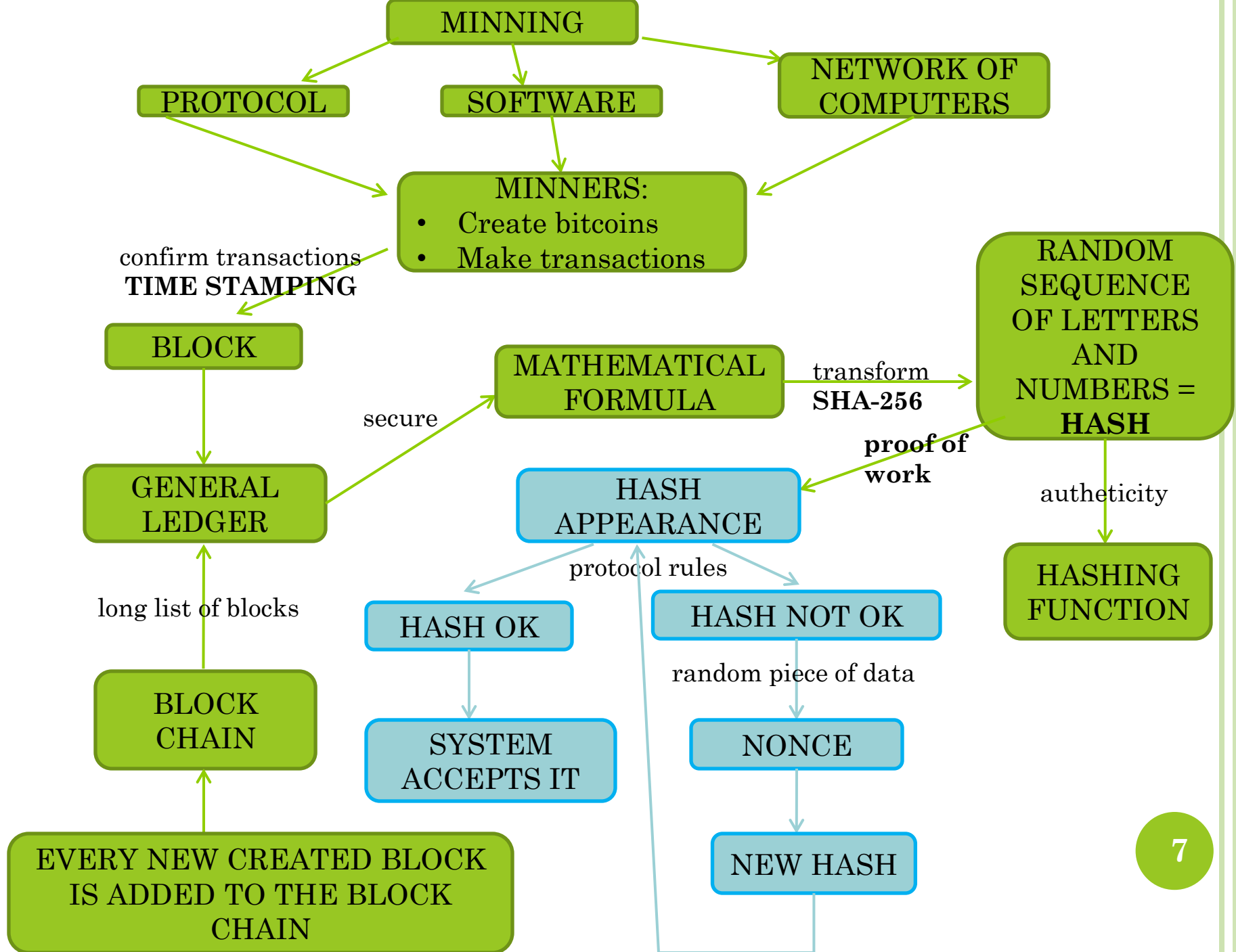
- The long history of virtual currencies dating to 1982, when David Chaum first described a cryptographic system for untraceable payments.
- Understanding bitcoin, how it works, what it is and its potential.
- Major international interest for this currency.

# WHAT IS BITCOIN?

- The European Central Bank (ECB) defines virtual currency as „a type of unregulated, digital money, which is issued and usually controlled by it's developers, and used and accepted among the members of a specific virtual community”
- Bitcoin is the first cryptocurrency and it uses SHA-256 (a set of cryptographic hash functions designed by the U.S. National Security Agency) as it's ***proof-of-work*** scheme.
- The FBI says that bitcoin is a „*decentralized, peer-to-peer* (P2P) network-based *virtual currency* that provides a venue for individuals to generate, transfer, launder, and steal illicit funds with some anonymity.”

# HOW ARE BITCOINS CREATED

- Bitcoin are created through a process called **mining** composed of three elements: protocol, software and network of computers.





# BITCOIN VS FIAT MONEY

## SIMILARITIES

- no intrinsic value
- depends on peoples trust in the currency
- inconvertibility

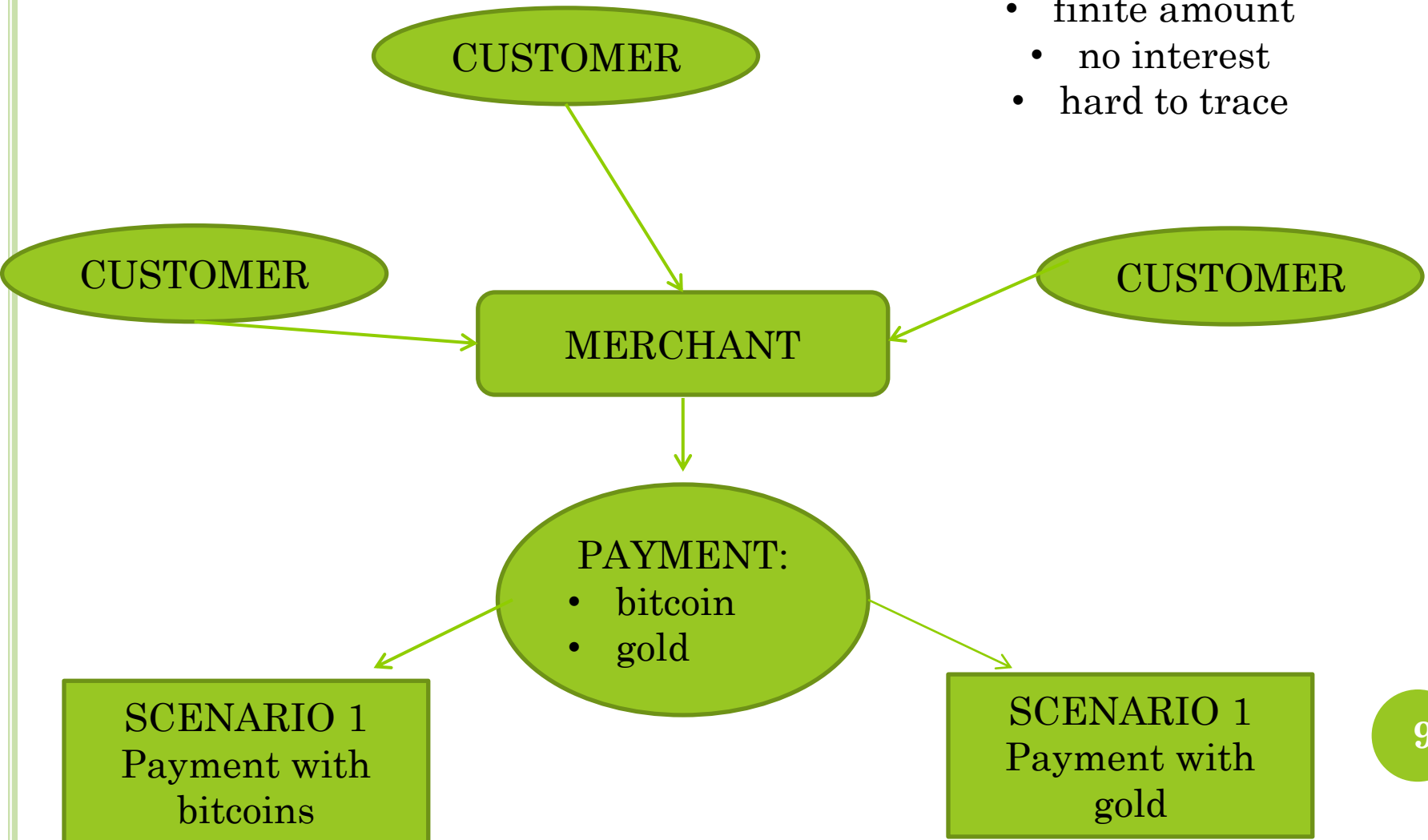
## DIFFERENCES

- legal tender
- not backed by a commodity (gold or silver)

# WHICH ONE IS BETTER - BITCOINS OR GOLD?

## SIMILARITIES

- finite amount
- no interest
- hard to trace



# MONEY FUNCTIONS

- According to the „Economics of Money, Banking and Financial Markets”, Miskin, money have three functions:
- *Medium of Exchange* – money are used to pay for goods and services;
- *Unit of Account* – money are used to measure the value of different goods and services;
- *Store of Value* – „used to save purchasing power from the time income is received until the time it is spent.”

# BITCOIN – MONEY OR COMMODITY?

The oppinions about bitcoin are divide throughout the world.

- ECB classifies it as a virtual currency scheme with bidirectional flow.
- Germany defined them as private money and financial instrument.
- Denmark doesn't consider bitcoin a currency nor asset.
- IRS stated that for tax purposes „virtual currency is treated as a property.- General tax principles applicable to property transactions apply to transactions using virtual currency. ”
- Countries like Iceland and Vietman, made bitcoin illegal due to lack of laws applicable for virtual currency.

# BITCOINS IN THE WORLD

## BITBILLS



## CASASCIUS BITCOIN POS



## CASASCIUS BITCOIN COINS



## BITCOIN ATM ROMANIA



## BITCOIN OPTIONS

Expires	Option	Profit(%)	Units	Price(€)	Return(€)	
15.04.14	Bitcoin/USD 701.4 <span style="color: green;">▲</span>	70 %	1	€ 50	€ 85	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go above 701.4 at 16:00 15.04.14						
15.04.14	Bitcoin/USD 651.3 <span style="color: green;">▲</span>	50 %	1	€ 50	€ 75	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go above 651.3 at 16:00 15.04.14						
15.04.14	Bitcoin/USD 350.5 <span style="color: red;">▼</span>	50 %	1	€ 50	€ 75	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go below 350.5 at 16:00 15.04.14						
15.04.14	Bitcoin/USD 301.10 <span style="color: red;">▼</span>	70 %	1	€ 50	€ 85	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go below 301.10 at 16:00 15.04.14						
30.04.14	Bitcoin/USD 701.4 <span style="color: green;">▲</span>	70 %	1	€ 50	€ 85	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go above 701.4 at 16:00 30.04.14						
30.04.14	Bitcoin/USD 651.3 <span style="color: green;">▲</span>	50 %	1	€ 50	€ 75	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go above 651.3 at 16:00 30.04.14						
30.04.14	Bitcoin/USD 370.4 <span style="color: red;">▼</span>	50 %	2	€ 100	€ 150	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go below 370.4 at 16:00 30.04.14						
30.04.14	Bitcoin/USD 301.10 <span style="color: red;">▼</span>	70 %	2	€ 100	€ 170	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go below 301.10 at 16:00 30.04.14						
15.05.14	Bitcoin/USD 701.4 <span style="color: green;">▲</span>	70 %	2	€ 100	€ 170	<a href="#">BUY »</a>
Will bitcoin price (as defined in the general terms) touch or go above 701.4 at 16:00 15.05.14						

## BITCOIN WALLET



## COINKITE BITCOIN POS



# DATA CONSTRUCTION AND ANALYSIS

- The set of variables are the price of the following exchange rates: BTC/USD, BTC/EUR.
- The frame of the data spans from 19<sup>th</sup> July 2010 to 23<sup>th</sup> May 2014, which is almost 4 years of daily series data, counting 1405 observations, the computing of the data was done with Eviews 7, and the source for all data is <http://www.oanda.com/currency/historical-rates/>.
- The two series BTC/USD and BTC/EUR were transformed to a logarithmical form and they can be found under the named `l_btc_usd` respectively `l_btc_eur`, using the generate option, after which I applied the first difference operator on both logarithmical forms obtaining a daily price variation named `dl_btc_usd` respectively `dl_btc_eur`.
- The commands are:
  - `l_btc_usd=log(btc_usd)`
  - `l_btc_eur=log(btc_eur)`
  - `dl_btc_usd= l_btc_usd-l_btc_usd(-1)`
  - `dl_btc_eur= l_btc_eur-l_btc_eur(-1)`

# ABOUT STATISTICS

- **Skewness** is an indicator of asymmetry and deviation from the normal distribution. This indicator can take three different values positive, zero or negative.

- The formula for skewness is:

$$S = \frac{1}{n} \times \frac{\sum_{i=1}^n (y_i - \bar{y})^3}{\sigma^2^{3/2}} \text{ and } \sigma^2 = \frac{1}{n} \times \sum_{i=1}^n (y_i - \bar{y})^2$$

- **Kurtosis** is an indicator which describes the distribution of the series around its mean, it's a measure for the probability peakness and it's based on a scaled version of the fourth moment. This indicator can also take three values greater, equal or lower than 3.

- The formula for kurtosis is:

$$K = \frac{1}{n} \times \frac{\sum_{i=1}^n (y_i - \bar{y})^4}{\sigma^2^2}$$

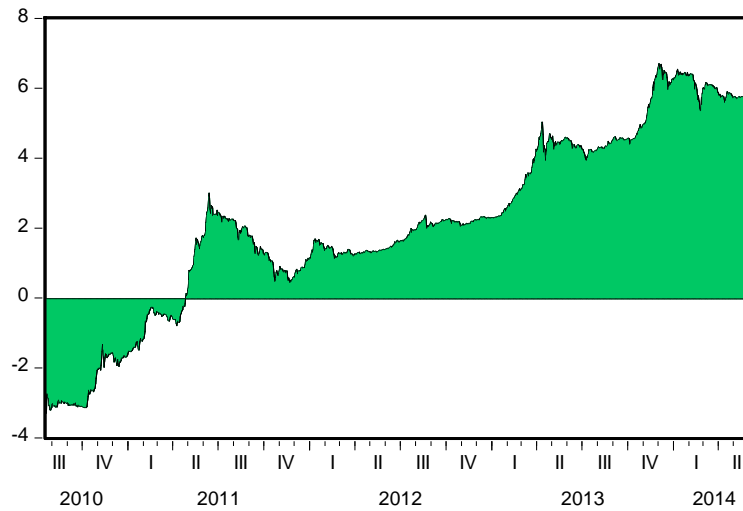
- **Jarque-Bera** (Charles Jarque and Anil Bera, 1979) is a test for goodness-of-fit, meaning that it shows whether the series have the skewness and kurtosis matching of a normal distribution.
- The formula for Jarque-Bera test is:

$$JB = \frac{n}{6} \times (S^2 + \frac{1}{4} (K - 3)^2)$$

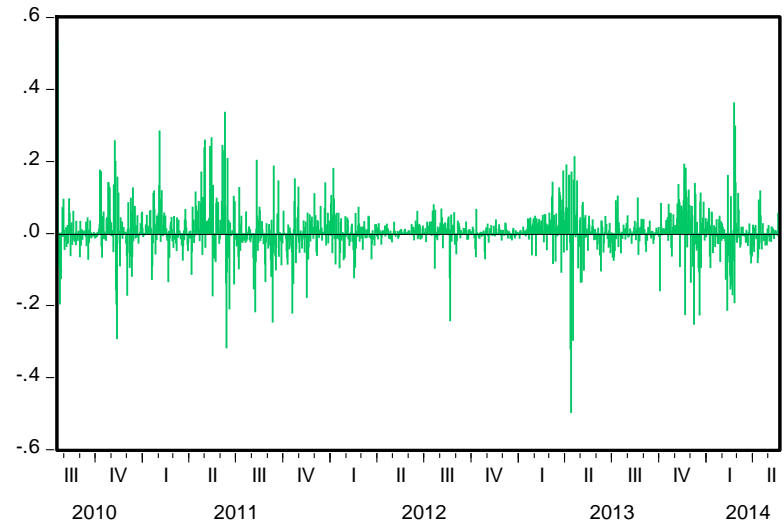


# REPRESENTATION OF L\_BTC AND DL\_BTC

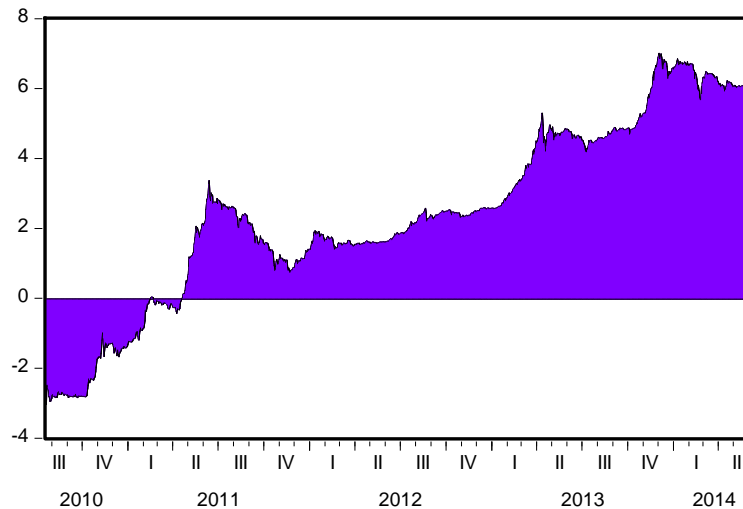
L\_BTC\_EUR



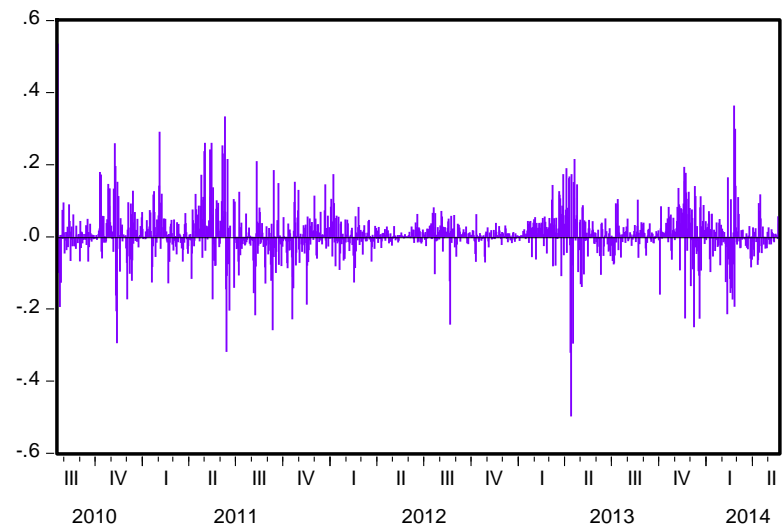
DL\_BTC\_EUR



L\_BTC\_USD



DL\_BTC\_USD



# UNIT ROOT TESTING – AUGMENTED DICKEY FULLER

- Augmented Dickey Fuller (David Dickey and Wayne Arthur Fuller, 1981) is a unit root test for large and complicated time series. The ADF t-statistic is a negative number and the more negative it is compared to the critical values, the rejection of the null hypothesis „the series has a unit root” is stronger meaning that the series is stationary.
- The test for unit root was done in level, include in test equation: intercept (constant), lag lenght was on automatic selection Schawrz Info Critirion with maximum lags:23 for l\_btc\_usd and l\_btc\_eur series.

# UNIT ROOT TESTING – KWIATKOWSKI-PHILLIPS-SCHMIDT-SHIN

- Kwiatkowski-Phillips-Schmidt-Shin (Denis Kwiatkowski, Peter Charles Bonest, Youngcheol Shin, Peter Schmidt; 1992) is used for testing the null hypothesis that a time series is stationarity. KPSS is often used to reinforce the results obtained in the ADF test. The acceptance of the null hypothesis is based on the value of  $t$  – statistic being smaller than all the critical values
- The test for unit root was done in level, include in test equation: intercept (constant), spectral estimation method: default ( Bartlett kernel), bandwidth : Newey – West Bandwidth for both series.

# ADF TEST

ADF	t-Statistic	Probability	Critical values	
l_btc_usd	-1.101536	0.7172	1%	-3.434802
			5%	-2.863393
			10%	-2.567806
l_btc_eur	-1.09288	0.7206	1%	-3.434802
			5%	-2.863393
			10%	-2.567806

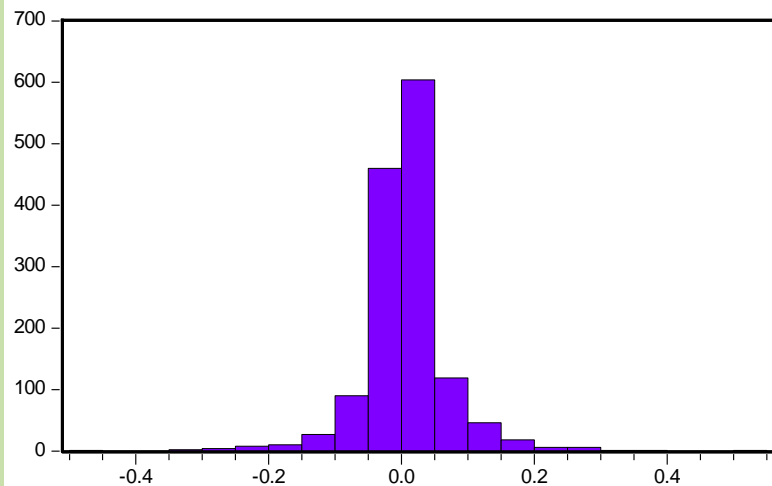
# KPSS TEST

KPSS (level)	t-Statistic	Critical values	
l_btc_usd	4.00177	1%	0.739
		5%	0.463
		10%	0.347
l_btc_eur	4.032417	1%	0.739
		5%	0.463
		10%	0.347

# UNIT ROOT TESTIN - ADF AND KPSS 1<sup>ST</sup> DIFFERENCE

Unit root test	Series	t-Statistic	Probability	Critical values	
ADF	dl_btc_usd	-30.70504	0.0000	1%	-3.434802
				5%	-2.863393
				10%	-2.567806
	dl_btc_eur	-30.68699	0.0000	1%	-3.434802
				5%	-2.863393
				10%	-2.567806
KPSS	dl_btc_usd	0.161037	-	1%	0.739000
				5%	0.463000
				10%	0.347000
	dl_btc_eur	0.153664	-	1%	0.739000
				5%	0.463000
				10%	0.347000

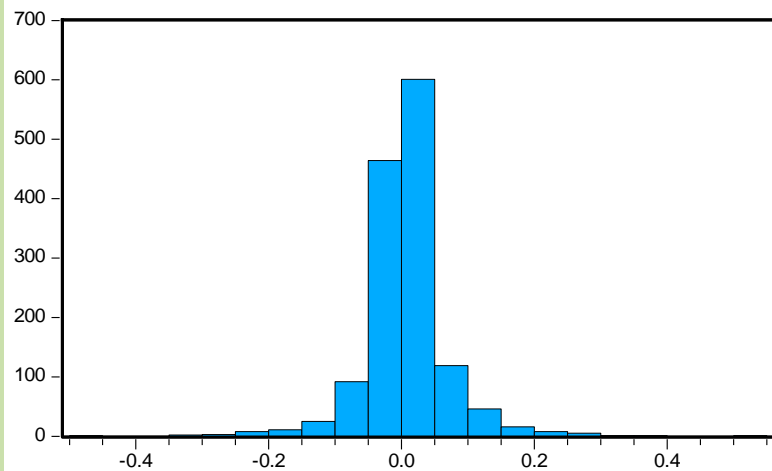
# SERIES DISTRIBUTION



Series: DL\_BTC\_USD  
Sample 7/19/2010 5/23/2014  
Observations 1404

Mean 0.006574  
Median 0.003665  
Maximum 0.535962  
Minimum -0.497501  
Std. Dev. 0.064298  
Skewness 0.123719  
Kurtosis 13.90817

Jarque-Bera 6964.388  
Probability 0.000000

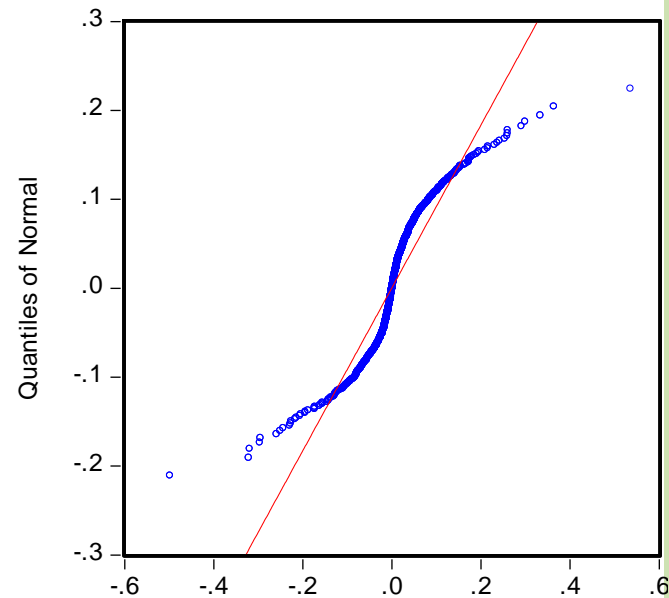


Series: DL\_BTC\_EUR  
Sample 7/19/2010 5/23/2014  
Observations 1404

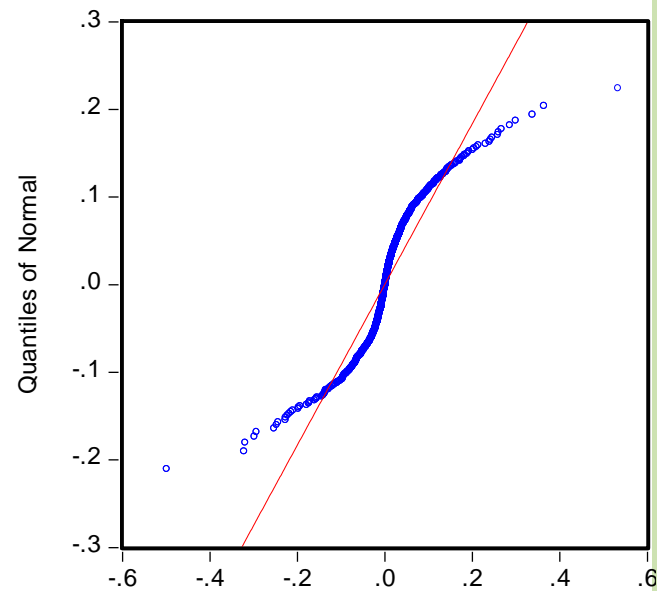
Mean 0.006534  
Median 0.003788  
Maximum 0.533542  
Minimum -0.497720  
Std. Dev. 0.064210  
Skewness 0.130415  
Kurtosis 13.87479

Jarque-Bera 6922.248  
Probability 0.000000

# KERNEL DENSITY GRAPH



Quantiles of DL\_BTC\_USD



Quantiles of DL\_BTC\_EUR

# ARMA - CORRELOGRAM

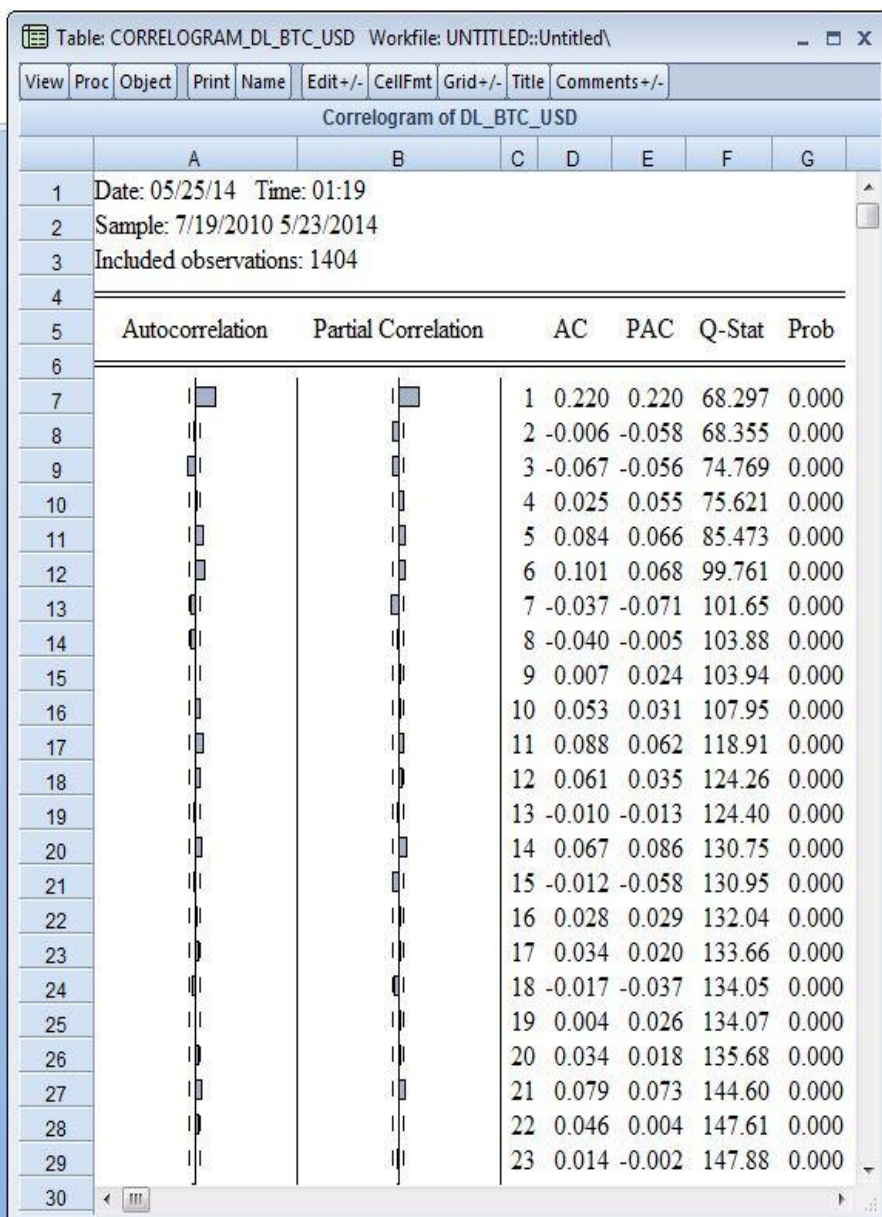
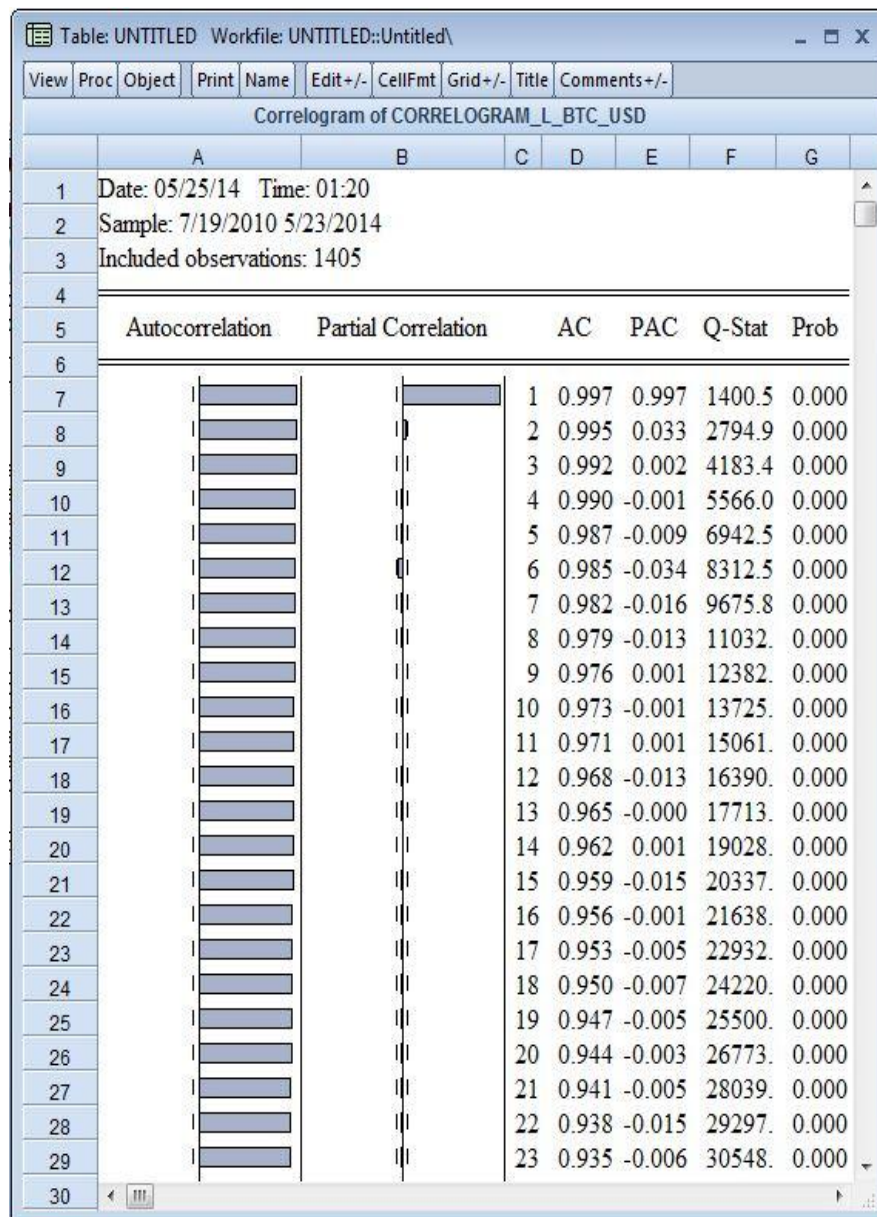




Table: CORRELOGRAM_L_BTC_EUR Workfile: UNTITLED::Untitled\									
View	Proc	Object	Print	Name	Edit+/-	CellFmt	Grid+/-	Title	Comments+/-
Correlogram of L_BTC_EUR									
	A	B	C	D	E	F	G		
1	Date: 05/25/14 Time: 01:19								
2	Sample: 7/19/2010 5/23/2014								
3	Included observations: 1405								
	Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob			
7			1	0.997	0.997	1400.5	0.000		
8			2	0.995	0.033	2795.1	0.000		
9			3	0.992	0.002	4183.8	0.000		
10			4	0.990	-0.000	5566.7	0.000		
11			5	0.987	-0.009	6943.5	0.000		
12			6	0.985	-0.034	8313.9	0.000		
13			7	0.982	-0.016	9677.7	0.000		
14			8	0.979	-0.013	11035.	0.000		
15			9	0.976	0.001	12385.	0.000		
16			10	0.974	-0.002	13729.	0.000		
17			11	0.971	0.001	15065.	0.000		
18			12	0.968	-0.014	16395.	0.000		
19			13	0.965	-0.000	17719.	0.000		
20			14	0.962	0.001	19035.	0.000		
21			15	0.960	-0.014	20344.	0.000		
22			16	0.957	-0.001	21647.	0.000		
23			17	0.954	-0.006	22942.	0.000		
24			18	0.951	-0.006	24230.	0.000		
25			19	0.948	-0.005	25511.	0.000		
26			20	0.945	-0.003	26786.	0.000		
27			21	0.942	-0.005	28053.	0.000		
28			22	0.939	-0.015	29312.	0.000		
29			23	0.936	-0.007	30565.	0.000		

Table: CORRELOGRAM_DL_BTC_EUR Workfile: UNTITLED::Untitled\									
View	Proc	Object	Print	Name	Edit+/-	CellFmt	Grid+/-	Title	Comments+/-
Correlogram of DL_BTC_EUR									
	A	B	C	D	E	F	G		
1	Date: 05/25/14 Time: 01:19								
2	Sample: 7/19/2010 5/23/2014								
3	Included observations: 1404								
	Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob			
7			1	0.221	0.221	68.540	0.000		
8			2	-0.003	-0.054	68.552	0.000		
9			3	-0.066	-0.056	74.729	0.000		
10			4	0.022	0.053	75.427	0.000		
11			5	0.082	0.066	84.889	0.000		
12			6	0.103	0.071	99.983	0.000		
13			7	-0.035	-0.072	101.74	0.000		
14			8	-0.038	-0.004	103.82	0.000		
15			9	0.008	0.026	103.91	0.000		
16			10	0.053	0.030	107.84	0.000		
17			11	0.090	0.065	119.42	0.000		
18			12	0.062	0.033	124.90	0.000		
19			13	-0.009	-0.012	125.01	0.000		
20			14	0.064	0.082	130.75	0.000		
21			15	-0.015	-0.060	131.08	0.000		
22			16	0.028	0.030	132.16	0.000		
23			17	0.032	0.017	133.65	0.000		
24			18	-0.017	-0.037	134.08	0.000		
25			19	0.001	0.023	134.08	0.000		
26			20	0.032	0.017	135.50	0.000		
27			21	0.079	0.074	144.45	0.000		
28			22	0.049	0.006	147.94	0.000		
29			23	0.015	-0.003	148.26	0.000		



# ARMA – AUTOREGRESSIVE MOVING AVERAGE

- Autoregressive Moving Average model, ARMA for short, was first described by Peter Whittle in 1951. ARMA models are used to describe stationary time series and it represents time series that are generated by passing a white noise test (recursive filter) – AR(p), and also a non-recursive linear filter – MA(q), consecutively.

- 

- The AR(p) model is describes by:

$$X_t = c + \sum_{i=1}^p \rho_i X_{t-i} + \varepsilon_t$$

- Where:

- $\varepsilon_t$  – white noise process, with a zero mean and a constant variance if the series is normal distributed;
- c – constant;
- $\rho_i$  – coefficients.

- MA(q) model represent the moving average of older q with white noise errors and is described by:

$$X_t = \varepsilon_t + \sum_{i=1}^q \theta_i \varepsilon_{t-i}$$

- Where  $\theta_i$  are the MA coefficients.

- Combining the two previous model we get ARMA(p,q) model and is described as follows:

$$X_t = \varepsilon_t + \sum_{i=1}^p \rho_i X_{t-i} + \sum_{i=1}^q \theta_i \varepsilon_{t-i}$$

# ARMA ESTIMATION – MODEL SELECTION

- AIC (1973-1974) is a measure of the relative quality, of a statistical model for a given set of data, provides a mean for model selection; deals with a tradeoff between the goodness of fit of the model. Can't tell anything about a tests null hypothesis.
- SC (Gideon E Schwarz 1978), a criterion for model selection among a finite set of models. It is based on the likelihood function and it's closely related to AIC.
- The basic information criteria are given by the following formulas:

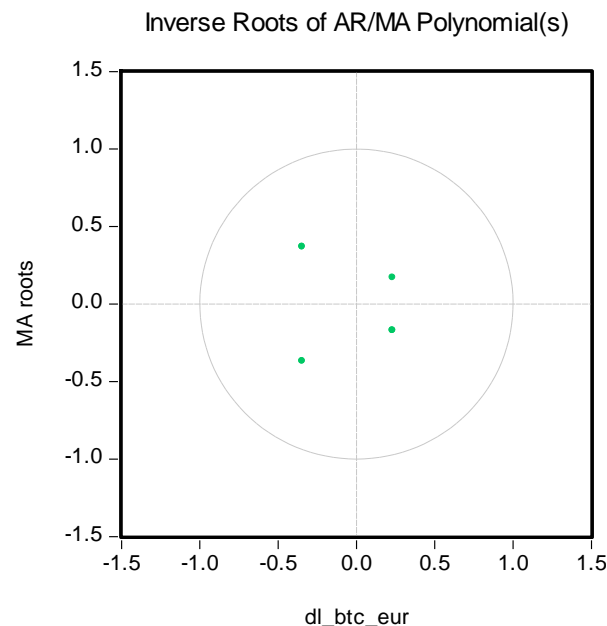
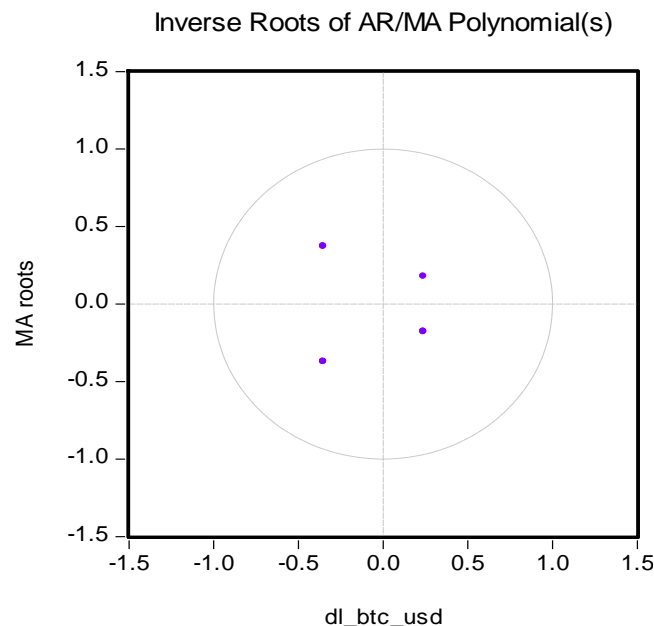
$$AIC = -2\frac{l}{T} + 2\frac{k}{T}$$

$$SC = -2\frac{l}{T} + \frac{k \log T}{T}$$

- Where  $l$  – the value of the likelihood function with  $k$  parameters estimated using  $T$  observation and the criterions are based on  $(-2)$  times the average log likelihood function, adjusted by a penalty function.

# MOVING AVERAGE ESTIMATION

- To estimate the MA(4) I generated a new equation for each exchange rate that looks like:
- $dl\_btc\_usd = c + dl\_btc\_usd(-1) + ma(1) + ma(2) + ma(3) + ma(4)$
- $dl\_btc\_eur = c + dl\_btc\_eur(-1) + ma(1) + ma(2) + ma(3) + ma(4)$
- and it's estimated with LS- Least Sequence Squared (NLD and ARMA) method
- Let's take a look at the at the characteristic polynomial roots, graph and tabel representations.



### Inverse Roots of AR/MA Polynomial(s)

Specification: DL\_BTC\_USD C DL\_BTC\_USD(-1)

MA(1) MA(2) MA(3) MA(4)

Date: 05/25/14 Time: 02:09

Sample: 7/19/2010 5/23/2014

Included observations: 1403

MA Root(s)	Modulus	Cycle
-0.351885 ± 0.372220i	0.512221	2.698825
0.237943 ± 0.178296i	0.297332	9.770665

No root lies outside the unit circle.

ARMA model is invertible.

### Inverse Roots of AR/MA Polynomial(s)

Specification: DL\_BTC\_EUR C DL\_BTC\_EUR(-1)

MA(1) MA(2) MA(3) MA(4)

Date: 05/25/14 Time: 02:22

Sample: 7/19/2010 5/23/2014

Included observations: 1403

MA Root(s)	Modulus	Cycle
-0.344400 ± 0.368567i	0.504433	2.705575
0.231383 ± 0.170535i	0.287437	9.892523

No root lies outside the unit circle.

ARMA model is invertible.

Table: MA4_CORREL_DL_BTC_USD Workfile: UNTITLED::Untitled\									
View	Proc	Object	Print	Name	Edit+/-	CellFmt	Grid+/-	Title	Comments+/-
Correlogram of Residuals									
	A		B		C	D	E	F	G
1	Date: 05/25/14 Time: 02:10								
2	Sample: 7/21/2010 5/23/2014								
3	Included observations: 1403								
4	Q-statistic probabilities adjusted for 4 ARMA term(s)								
5									
6	Autocorrelation		Partial Correlation		AC	PAC	Q-Stat	Prob	
7									
8					1	0.001	0.001	0.0004	
9					2	0.001	0.001	0.0024	
10					3	-0.009	-0.009	0.1143	
11					4	0.010	0.010	0.2533	
12					5	0.057	0.057	4.8406	0.028
13					6	0.118	0.119	24.619	0.000
14					7	-0.059	-0.059	29.458	0.000
15					8	-0.025	-0.026	30.368	0.000
16					9	0.008	0.010	30.465	0.000
17					10	0.029	0.023	31.656	0.000
18					11	0.071	0.059	38.704	0.000
19					12	0.055	0.049	42.915	0.000
20					13	-0.038	-0.023	44.995	0.000
21					14	0.095	0.099	57.805	0.000
22					15	-0.039	-0.048	59.926	0.000
23					16	0.023	0.008	60.664	0.000
24					17	0.047	0.034	63.868	0.000
25					18	-0.031	-0.035	65.269	0.000
26					19	0.006	0.014	65.324	0.000
27					20	0.025	0.006	66.180	0.000
28					21	0.068	0.083	72.821	0.000
29					22	0.023	0.012	73.589	0.000
30									

Table: MA4_CORREL_DL_BTC_EUR Workfile: UNTITLED::Untitled\									
View	Proc	Object	Print	Name	Edit+/-	CellFmt	Grid+/-	Title	Comments+/-
Correlogram of Residuals									
	A		B		C	D	E	F	G
1	Date: 05/25/14 Time: 02:22								
2	Sample: 7/21/2010 5/23/2014								
3	Included observations: 1403								
4	Q-statistic probabilities adjusted for 4 ARMA term(s)								
5									
6	Autocorrelation		Partial Correlation		AC	PAC	Q-Stat	Prob	
7									
8					1	0.000	0.000	0.0003	
9					2	0.001	0.001	0.0023	
10					3	-0.009	-0.009	0.1105	
11					4	0.010	0.010	0.2456	
12					5	0.055	0.055	4.5135	0.034
13					6	0.122	0.122	25.419	0.000
14					7	-0.058	-0.058	30.165	0.000
15					8	-0.025	-0.025	31.024	0.000
16					9	0.010	0.012	31.168	0.000
17					10	0.028	0.023	32.303	0.000
18					11	0.073	0.062	39.823	0.000
19					12	0.054	0.047	43.979	0.000
20					13	-0.036	-0.021	45.854	0.000
21					14	0.091	0.095	57.712	0.000
22					15	-0.041	-0.050	60.048	0.000
23					16	0.023	0.009	60.813	0.000
24					17	0.046	0.032	63.767	0.000
25					18	-0.031	-0.035	65.124	0.000
26					19	0.004	0.012	65.144	0.000
27					20	0.023	0.004	65.881	0.000
28					21	0.068	0.083	72.469	0.000
29					22	0.026	0.014	73.458	0.000
30									

# AUTOREGRESSIVE ESTIMATION

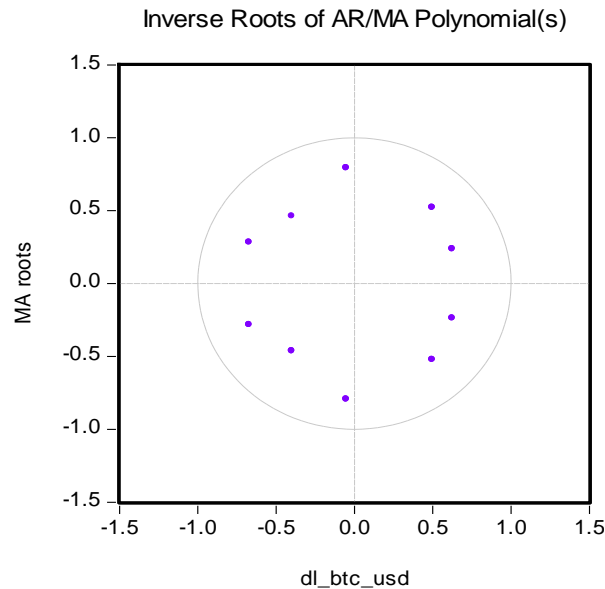
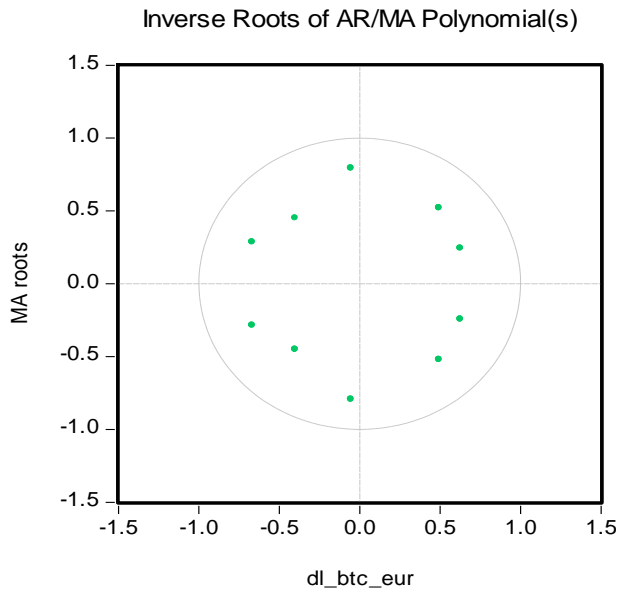
- To estimate the AR(1) I generated a new equation for each exchange rate, that looks like:
- $dl\_btc\_usd \sim c + \phi \cdot dl\_btc\_usd(-1)$
- $dl\_btc\_eur \sim c + \phi \cdot dl\_btc\_eur(-1)$
- and it's estimated with LS- Least Squared (NLD and ARMA) method.
- AR coefficient is smaller than 1 which means that the equation is stable.

Dependent Variable: DL_BTC_USD				
Method: Least Squares				
Date: 05/25/14 Time: 07:35				
Sample (adjusted): 7/21/2010 5/23/2014				
Included observations: 1403 after adjustments				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.004753	0.001641	2.896699	0.0038
DL_BTC_USD(-1)	0.220356	0.025391	8.678376	0.0000
R-squared	0.051015	Mean dependent var		0.006197
Adjusted R-squared	0.050338	S.D. dependent var		0.062747
S.E. of regression	0.061147	Akaike info criterion		-2.749631
Sum squared resid	5.238361	Schwarz criterion		-2.742152
Log likelihood	1930.866	Hannan-Quinn criter.		-2.746836
F-statistic	75.31422	Durbin-Watson stat		1.945339
Prob(F-statistic)	0.000000			

Dependent Variable: DL_BTC_EUR				
Method: Least Squares				
Date: 05/25/14 Time: 07:41				
Sample (adjusted): 7/21/2010 5/23/2014				
Included observations: 1403 after adjustments				
Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	0.004722	0.001639	2.881223	0.0040
DL_BTC_EUR(-1)	0.220752	0.025393	8.693270	0.0000
R-squared	0.051181	Mean dependent var		0.006159
Adjusted R-squared	0.050504	S.D. dependent var		0.062671
S.E. of regression	0.061068	Akaike info criterion		-2.752228
Sum squared resid	5.224774	Schwarz criterion		-2.744750
Log likelihood	1932.688	Hannan-Quinn criter.		-2.749433
F-statistic	75.57294	Durbin-Watson stat		1.947826
Prob(F-statistic)	0.000000			

# ARMA ESTIMATION

- To estimate ARMA, I generated a new equation for each exchange rate that looks like:
- $dl\_btc\_usd \sim c + dl\_btc\_usd(-1) + ma(5) + ma(6) + ma(7) + ma(8) + ma(9) + ma(10)$
- $dl\_btc\_eur \sim c + dl\_btc\_eur(-1) + ma(5) + ma(6) + ma(7) + ma(8) + ma(9) + ma(10)$
- and the MA value starting at 5 is justified by the serial correlation that exists starting with the fifth lag. The equation is estimated with LS- Least Squares Squared (NLS and ARMA) method.
- The inverted roots are in the unit circle for both series and the modulus of the characteristic polynomial roots is smaller than 1, meaning that the equation is stable.





## Inverse Roots of AR/MA Polynomial(s)

Specification: DL\_BTC\_USD C DL\_BTC\_USD(-1)

MA(5) MA(6) MA(7) MA(8) MA(9) MA(10)

Date: 05/25/14 Time: 08:30

Sample: 7/19/2010 5/23/2014

Included observations: 1403

MA Root(s)	Modulus	Cycle
-0.051429 ± 0.792755i	0.794421	3.841566
-0.672177 ± 0.282881i	0.729277	2.290418
0.497047 ± 0.522779i	0.721356	7.751041
0.625330 ± 0.237736i	0.668997	17.29469
-0.398771 ± 0.461824i	0.610164	2.752091

No root lies outside the unit circle.

ARMA model is invertible.

## Inverse Roots of AR/MA Polynomial(s)

Specification: DL\_BTC\_EUR C DL\_BTC\_EUR(-1)

MA(5) MA(6) MA(7) MA(8) MA(9) MA(10)

Date: 05/25/14 Time: 08:45

Sample: 7/19/2010 5/23/2014

Included observations: 1403

MA Root(s)	Modulus	Cycle
-0.052625 ± 0.793801i	0.795544	3.838244
-0.666866 ± 0.285577i	0.725440	2.295661
0.493270 ± 0.520966i	0.717440	7.731258
0.627357 ± 0.243019i	0.672782	17.00131
-0.401136 ± 0.450714i	0.603368	2.734126

No root lies outside the unit circle.

ARMA model is invertible.

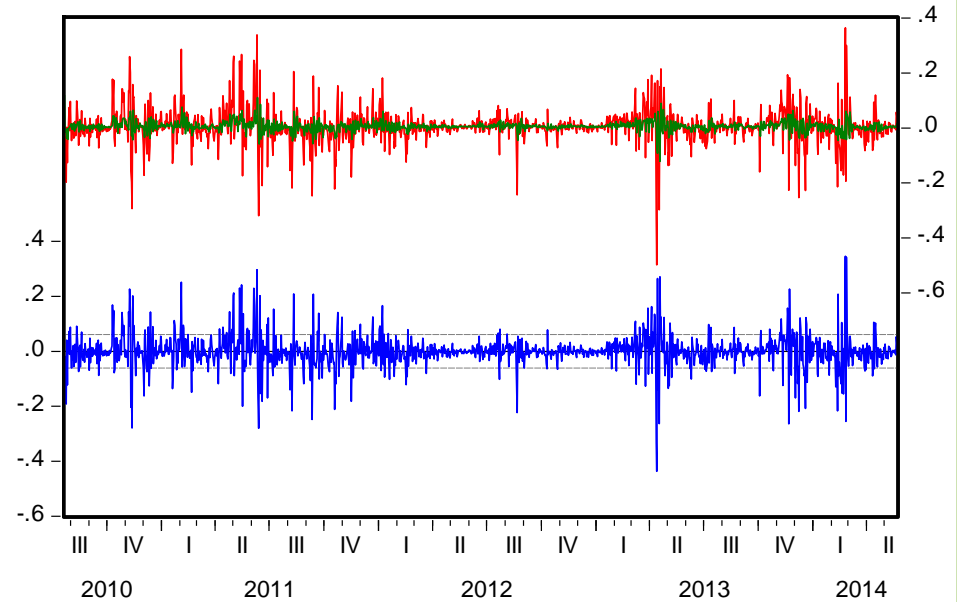
# CORRELOGRAM ARMA(1, 10)

Table: ARMA110_CORRE_DL_BTC_USD Workfile: UNTITLED::Untitled\							
View	Proc	Object	Print	Name	Edit+/-	CellFmt	Grid+/-
Correlogram of Residuals							
	A	B	C	D	E	F	G
1	Date: 05/25/14 Time: 08:30						
2	Sample: 7/21/2010 5/23/2014						
3	Included observations: 1403						
4	Q-statistic probabilities adjusted for 6 ARMA term(s)						
	Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob	
8			1 0.022	0.022	0.6873		
9			2 -0.028	-0.029	1.7946		
10			3 -0.064	-0.063	7.5976		
11			4 0.042	0.044	10.092		
12			5 -0.002	-0.007	10.096		
13			6 -0.007	-0.008	10.161		
14			7 -0.000	0.006	10.161	0.001	
15			8 -0.014	-0.017	10.436	0.005	
16			9 -0.000	0.000	10.436	0.015	
17			10 0.005	0.005	10.469	0.033	
18			11 0.058	0.056	15.275	0.009	
19			12 0.059	0.059	20.225	0.003	
20			13 -0.032	-0.031	21.655	0.003	
21			14 0.087	0.099	32.284	0.000	
22			15 -0.045	-0.049	35.126	0.000	
23			16 0.016	0.015	35.490	0.000	
24			17 0.027	0.041	36.522	0.000	
25			18 -0.029	-0.046	37.752	0.000	
26			19 0.007	0.021	37.829	0.000	
27			20 0.015	0.018	38.159	0.000	
28			21 0.074	0.065	45.922	0.000	
29			22 0.012	0.014	46.116	0.000	
30			23 0.000	0.000	46.216	0.000	

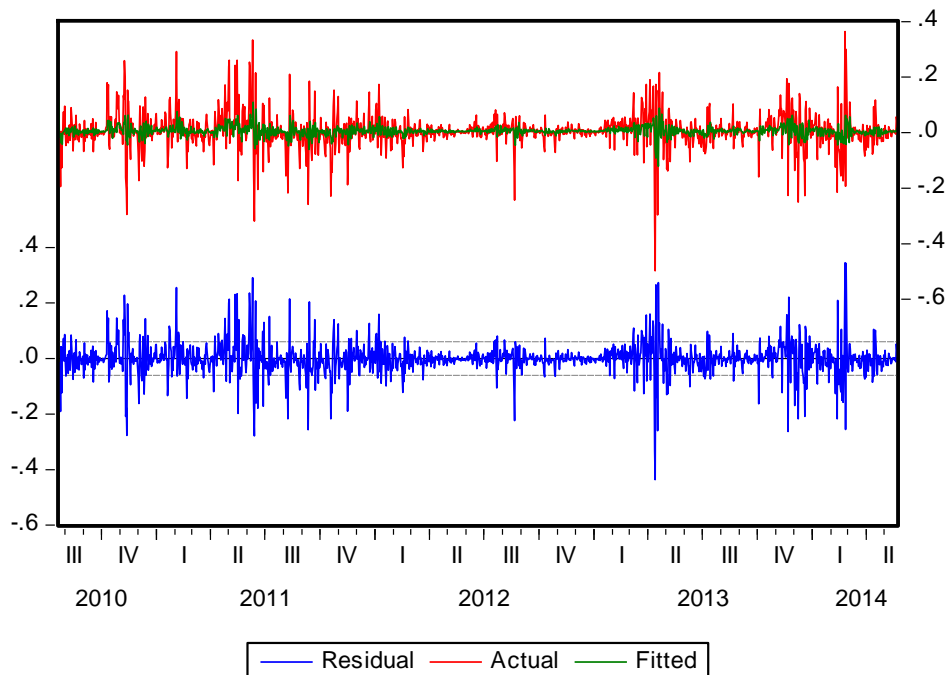
Table: ARMA110_CORRE_DL_BTC_EUR Workfile: UNTITLED::Untitled\							
View	Proc	Object	Print	Name	Edit+/-	CellFmt	Grid+/-
Correlogram of Residuals							
	A	B	C	D	E	F	G
1	Date: 05/25/14 Time: 08:45						
2	Sample: 7/21/2010 5/23/2014						
3	Included observations: 1403						
4	Q-statistic probabilities adjusted for 6 ARMA term(s)						
	Autocorrelation	Partial Correlation	AC	PAC	Q-Stat	Prob	
8			1 0.021	0.021	0.6252		
9			2 -0.025	-0.026	1.5097		
10			3 -0.064	-0.063	7.2077		
11			4 0.040	0.042	9.4155		
12			5 -0.002	-0.007	9.4238		
13			6 -0.007	-0.009	9.4891		
14			7 -0.000	0.005	9.4891	0.002	
15			8 -0.014	-0.017	9.7737	0.008	
16			9 0.001	0.001	9.7743	0.021	
17			10 0.005	0.005	9.8034	0.044	
18			11 0.061	0.059	15.078	0.010	
19			12 0.059	0.058	20.016	0.003	
20			13 -0.030	-0.030	21.305	0.003	
21			14 0.084	0.096	31.299	0.000	
22			15 -0.047	-0.051	34.445	0.000	
23			16 0.016	0.015	34.819	0.000	
24			17 0.025	0.039	35.718	0.000	
25			18 -0.029	-0.045	36.914	0.000	
26			19 0.005	0.018	36.948	0.000	
27			20 0.013	0.016	37.206	0.001	
28			21 0.073	0.065	44.864	0.000	
29			22 0.014	0.016	45.158	0.000	
30			23 0.000	0.000	45.216	0.000	

# RESIDUAL, ACTUAL, FILTERED DATA ARMA(1, 10)

dl\_btc\_eur



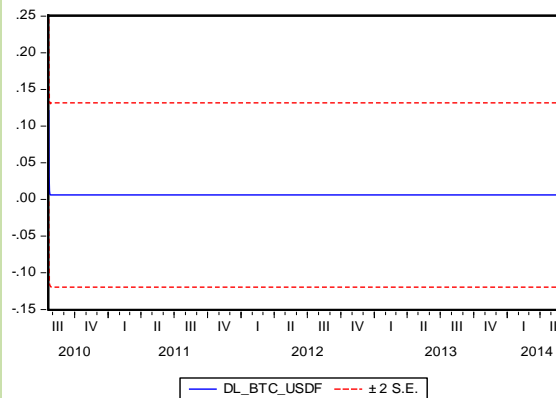
dl\_btc\_usd



# ARMA MODELS - COMPARISON

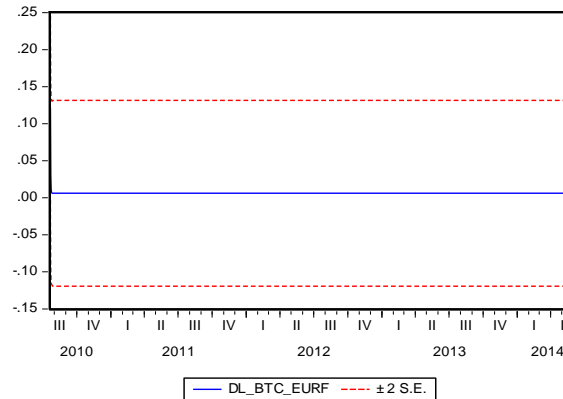
	MA(4)		AR(1)		ARMA(1,10)	
	dl_btc_usd	dl_btc_eur	dl_btc_usd	dl_btc_eur	dl_btc_usd	dl_btc_eur
Adjusted R-squared	0.058673	0.058137	0.050338	0.0504	<b>0.066713</b>	<b>0.06716</b>
Akaike info criterion	-2.755604	-2.757457	-2.749631	-2.752228	<b>-2.762763</b>	<b>-2.76581</b>
Schwarz criterion	-2.733168	-2.735021	<b>-2.742152</b>	<b>-2.74475</b>	-2.732848	-2.735766

# FORECAST ARMA(1,10)



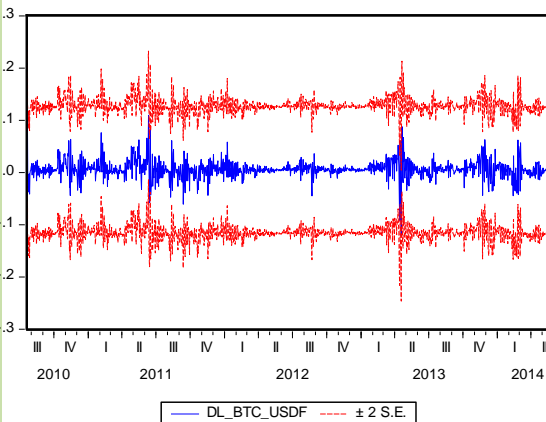
Forecast: DL\_BTC\_USDF  
Actual: DL\_BTC\_USD  
Forecast sample: 7/19/2010 5/23/2014  
Adjusted sample: 7/21/2010 5/23/2014  
Included observations: 1403

Root Mean Squared Error	0.062848
Mean Absolute Error	0.037851
Mean Abs. Percent Error	195.9082
Theil Inequality Coefficient	0.897754
Bias Proportion	0.000000
Variance Proportion	0.898179
Covariance Proportion	0.101821



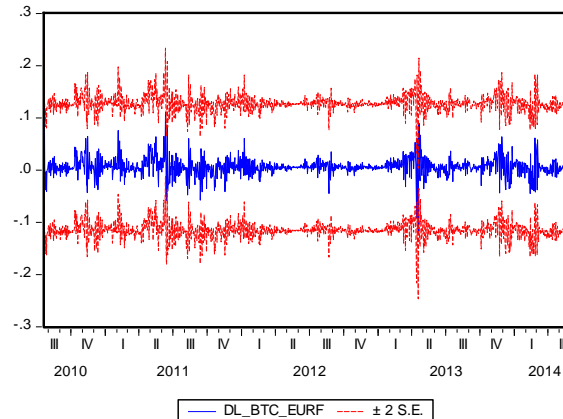
Forecast: DL\_BTC\_EURF  
Actual: DL\_BTC\_EUR  
Forecast sample: 7/19/2010 5/23/2014  
Adjusted sample: 7/21/2010 5/23/2014  
Included observations: 1403

Root Mean Squared Error	0.062766
Mean Absolute Error	0.037945
Mean Abs. Percent Error	190.1320
Theil Inequality Coefficient	0.898045
Bias Proportion	0.000000
Variance Proportion	0.898338
Covariance Proportion	0.101662



Forecast: DL\_BTC\_USDF  
Actual: DL\_BTC\_USD  
Forecast sample: 7/19/2010 5/23/2014  
Adjusted sample: 7/21/2010 5/23/2014  
Included observations: 1403

Root Mean Squared Error	0.060445
Mean Absolute Error	0.036774
Mean Abs. Percent Error	241.6502
Theil Inequality Coefficient	0.747047
Bias Proportion	0.000000
Variance Proportion	0.577921
Covariance Proportion	0.422079



Forecast: DL\_BTC\_EURF  
Actual: DL\_BTC\_EUR  
Forecast sample: 7/19/2010 5/23/2014  
Adjusted sample: 7/21/2010 5/23/2014  
Included observations: 1403

Root Mean Squared Error	0.060357
Mean Absolute Error	0.036856
Mean Abs. Percent Error	222.6150
Theil Inequality Coefficient	0.746456
Bias Proportion	0.000000
Variance Proportion	0.576666
Covariance Proportion	0.423334

# VOLATILITY ANALYSIS

## ARCH

- ARCH (autoregressive conditional heteroskedasticity, introduced by Robert Fry Engle, 1982) models are used for modelling financial time series and assumes that the variance of the current error term or innovation (the difference between the observed value of a variable at time  $i$  and the optimal forecast of that value based on information available prior to time  $i$ ) to be function of the actual size of the previous time periods error terms, often the variance is related to the squares of previous innovations.
- The error terms  $\varepsilon_t$  is composed of two stochastic terms  $z_t$  (white noise) and standard deviation  $\sigma_t$ .

$$\varepsilon_t = z_t \sigma_t$$

- The representation of ARCH(p) variance  $\sigma_t^2$  is as followed:

$$\sigma_t^2 = \alpha_0 + \alpha_1 \varepsilon_{t-1}^2 + \dots + \alpha_q \varepsilon_{t-q}^2 = \alpha_0 + \sum_{i=1}^q \alpha_i \varepsilon_{t-i}^2$$

- Where:  $\alpha_0 > 0$  and  $\alpha_i > 0, i > 0$ .
- The ARCH testing is done with the estimation method: ARCH and the orders are as follows: ARCH – 1, GARCH – 0 and Threshold – 0, error distribution – normal.

# GARCH

- GARCH (generalized autoregressive conditional heteroskedasticity) is a generalized ARCH model introduced by Tim Petter Bollerslev in 1986.
- The representation of GARCH (p,q) variance is as follows:

$$\sigma_t^2 = \omega + \sum_{j=1}^q \beta_j \sigma_{t-j}^2 + \sum_{i=1}^p \alpha_i \epsilon_{t-i}^2$$

- Where  $\omega$  a constant.
- The GARCH testing is done with the estimation method: ARCH and the orders are as follows: ARCH – 1, GARCH – 1 and Threshold - 0, error distribution – normal.

# EGARCH

- EGARCH model (exponential generalized autoregressive conditional heteroskedasticity) was introduced by Nelson in 1991 and is another form of GARCH.
- The representation of EGARCH variance is as follows:
- $\log (\sigma_t^2) =$   
$$\omega + \sum_{j=1}^q \beta_j \log(\sigma_{t-j}^2) + \sum_{i=1}^p \alpha_i \left| \frac{\varepsilon_{t-1}}{\sigma_{t-1}} \right| + \sum_{k=1}^r \gamma_k \frac{\varepsilon_{t-k}}{\sigma_{t-k}}$$
- The GARCH testing for is done with the estimation method: EGARCH and the orders are as follows: ARCH – 1, GARCH – 1 and Asymmetric order – 0, error distribution – normal.



# TARCH

- TARCH (Threshold autoregressive conditional heteroskedasticity) was introduced independently by Zakoian (1994) and Glosten.
- The generalized representation of TARCH variance is as followed:
- $$\sigma_t^2 = \omega + \sum_{j=1}^q \beta_j \sigma_{t-j}^2 + \sum_{i=1}^q \alpha_i \varepsilon_{t-i}^2 + \sum_{k=1}^r \delta_k \varepsilon_{t-k}^2 \bar{I}_{t-k}$$
- Where:  $\bar{I}_{t-k} = 1$  if  $\varepsilon_t < 0$  and zero otherwise.
- The GARCH testing for is done with the estimation method: EGARCH and the orders are as follows: ARCH – 1, GARCH – 1 and threshold – 1, error distribution – normal.

# VOLATILITY MODELS - COMPARISON

	ARCH 1		GARCH 1_1		EGARCH 1_1		TARCH 1_1_1	
	dl_btc_usd	dl_btc_eur	dl_btc_usd	dl_btc_eur	dl_btc_usd	dl_btc_eur	dl_btc_usd	dl_btc_eur
Log likelihood	2120.676	2121.752	2309.084	2306.274	<b>2316.511</b>	<b>2313.593</b>	2309.173	2306.443
Akaike info criterion	-3.018057	-3.01959	-3.28505	-3.281017	<b>-3.2956</b>	<b>-3.290019</b>	-3.283722	-3.279834
Schwarz criterion	-3.010583	-3.012115	-3.273808	-3.269805	<b>-3.284389</b>	<b>-3.27507</b>	-3.268773	-3.264885

# ARCH LM-TEST

Heteroskedasticity Test: ARCH		dl_btc_usd		
F-statistic	0.201248	Prob. F(1,1401)		0.6538
Obs*R-squared	0.201506	Prob. Chi-Square(1)		0.6535

Heteroskedasticity Test: ARCH		dl_btc_eur		
F-statistic	0.285457	Prob. F(1,1401)		0.5932
Obs*R-squared	0.285806	Prob. Chi-Square(1)		0.5929

# CONCLUSION

- Comparison of any classic or gold coins can not be achieved in practice without involving state organizations, and this contradicts the fundamental principle of virtual currency, therefore not be under state control and be managed by specific rules (banking and financial system).
- The increase in the use of bitcoins and other similar virtual currencies might lead to a decrease in the use of real money, reducing the cash needed to conduct transactions.
- From the conducted research, besides that bitcoin can be currency financial assets, property, also appears the probability that bitcoin can be used as a real weapon to destroy and control the financial system.
- ARMA (1,10) is the best model to estimate the forecast, based on the values of adjuster R-squared, Akaike info criterion and Schwarz criterion.
- Although I chose the best ARMA model, the forecast of bitcoin is very complicated and the result are not conclusive, because the series are very volatile and it's very difficult to estimate future outcome. The future of bitcoin relying only on it's developers and users.
- The volatility analysis has determined that the best model is EGARCH (1,1), and both series have an assymetrical distribution to the left, leptokurtic and normal distribution.
- Attempts of econometric application on the current series have proven that can be applied to a point, beyond which analysis parameters can not be common to Bitcoin and other currencies.
- In my opinion Bitcoin power should not be neglected, but rather should support users of this currency with respect to the risks and benefits of transactions made.

# PROPOSAL

- The creation on an educational system by the developers, for users.
- Collect all the official statements of state organisation, financial institutions, states etc and centralize all the oppinions about bitcoin and see the frequency with which they appear.
- I commit to post this paper, to be public to all persons interested in this topic, to discuss and clarify what is Bitcoin ultimately.
- To try and understand the general trend of bitcoin and see what and how influences it.

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THANK YOU VERY  
MUCH FOR YOUR  
ATTENTION !

Q & A

