# CESIFO WORKING PAPERS

### Momentum Effects in the Cryptocurrency Market After One-Day Abnormal Returns

Guglielmo Maria Caporale, Alex Plastun



#### Impressum:

CESifo Working Papers ISSN 2364-1428 (electronic version) Publisher and distributor: Munich Society for the Promotion of Economic Research - CESifo GmbH The international platform of Ludwigs-Maximilians University's Center for Economic Studies and the ifo Institute Poschingerstr. 5, 81679 Munich, Germany Telephone +49 (0)89 2180-2740, Telefax +49 (0)89 2180-17845, email office@cesifo.de Editor: Clemens Fuest www.cesifo-group.org/wp

An electronic version of the paper may be downloaded

- · from the SSRN website: <u>www.SSRN.com</u>
- from the RePEc website: <u>www.RePEc.org</u>
- from the CESifo website: <u>www.CESifo-group.org/wp</u>

### Momentum Effects in the Cryptocurrency Market After One-Day Abnormal Returns

### Abstract

This paper examines whether there exists a momentum effect after one-day abnormal returns in the cryptocurrency market. For this purpose a number of hypotheses of interest are tested for the BitCoin, Ethereum and LiteCoin exchange rates vis-à-vis the US dollar over the period 01.01.2017-01.09.2019, specifically whether or not: H1) the intraday behaviour of hourly returns is different on overreaction days compared to normal days; H2) there is a momentum effect on overreaction days, and H3) after one-day abnormal returns. The methods used for the analysis include a number of statistical methods as well as a trading simulation approach. The results suggest that hourly returns during the day of positive/negative overreactions are significantly higher/lower than those during the average positive/negative day. Overreactions can usually be detected before the day ends by estimating specific timing parameters. Prices tend to move in the direction of the overreaction till the end of the day when it occurs, which implies the existence of a momentum effect on that day giving rise to exploitable profit opportunities. This effect (together with profit opportunities) is also observed on the following day. In two cases (BTCUSD positive overreactions and ETHUSD negative overreactions) a contrarian effect is detected instead.

JEL-Codes: G120, G170, C630.

Keywords: cryptocurrencies, anomalies, momentum effect, overreactions, abnormal returns, patterns.

Guglielmo Maria Caporale\* Department of Economics and Finance Brunel University London United Kingdom – London, UB8 3PH Guglielmo-Maria.Caporale@brunel.ac.uk Alex Plastun Sumy State University Sumy / Ukraine o.plastun@gmail.com

\*corresponding author

#### October 2019

The second-named author gratefully acknowledges financial support from the Ministry of Education and Science of Ukraine (0117U003936).

#### 1. Introduction

It is well known that the Efficient Market Hypothesis (EMH) is inconsistent with the existence of abnormal returns, i.e. of fat tails in the price distribution. However, numerous empirical studies have reported evidence of so-called market overreactions. De Bondt and Thaler (1985) developed the overreaction hypothesis to describe price patterns caused by abnormal price changes. The subsequent literature has also analysed the reasons for abnormal price changes (Griffin and Tversky, 1992; Aiyagari and Gertler, 1999; Madura and Richie, 2004; Mynhardt and Plastun, 2013); price patterns (Cutler et al., 1991; Ferri and Min, 1996); trading strategies based on overreactions (Jegadeesh, 1993; Caporale and Plastun, 2019); the influence of price overreactions on market participants (Savor, 2012), etc. According to the overreaction hypothesis there should be price reversals after abnormal price changes; various studies have analysed the US stock market (De Bondt and Thaler, 1985; Jegadeesh, 1993; Ferri and Min, 1996), other stock markets (Lobe and Rieks, 2011; Mynhardt and Plastun, 2013), the FOREX (Caporale et al., 2018), option (Poteshman, 2001) and commodity markets (Cutler et al., 1991), and most of them have provided evidence in favour of price reversals after overreactions, with a few detecting instead momentum effects (Cox and Peterson, 1994).

The cryptocurrency market represents a particularly interesting case being rather new, relatively unexplored and at the same time extremely vulnerable to overreactions, given its high volatility relative to the FOREX, stock and commodity markets etc. (Cheung et al., 2015; Aalborg et al., 2019). Recent studies have analysed its efficiency (Bartos, 2015; Urquhart, 2016), long-memory properties and persistence (Bariviera et al., 2017; ), the existence of price bubbles (Corbet et al, 2018), its competitiveness (Halaburda and Gandal, 2014), the issue of predictability (Bouri et al. 2018; Caporale et al., 2019) and the presence of anomalies (Kurihara and Fukushima, 2017; Caporale and Plastun, 2018). However, there are very few studies focusing on overreactions in the cryptocurrency market. Chevapatrakul and Mascia (2019) using the quantile autoregressive model show that days with extremely negative returns are likely to be followed by periods characterised by negative returns and weekly positive returns as Bitcoin prices continue to rise. Caporale and Plastun (2019) find evidence of price patterns after overreactions, i.e. the next-day price changes in both directions are bigger than after "normal" days; they show that a strategy based on the momentum effect (rather than counter-movements) after overreactions is profitable.

The present paper extends the analysis of Caporale and Plastun (2019) by examining whether there exists a momentum effect after one-day abnormal returns in the cryptocurrency market. For this purpose a number of hypotheses of interest are tested for the BitCoin, Ethereum and LiteCoin exchange rates vis-à-vis the US dollar over the period 01.01.2017-01.09.2019, specifically whether or not: H1) the intraday behaviour of hourly returns is different on overreaction days compared to normal days; H2) there is a momentum effect on overreaction days, and H3) after one-day abnormal returns. The methods used for the analysis include statistical tests and the cumulative abnormal returns and trading simulation approaches. The results are relevant to both academics interested in the EMH and practitioners (traders, investors, financial analysts, etc.) aiming to design profitable trading strategies using information about the existence or not of a momentum effect, the timing of the overreactions and the duration of the anomaly.

The remainder of the paper is organised as follows. Section 2 describes the methodology. Section 3 discusses the empirical results. Section 4 offers some concluding remarks.

### 2. Methodology

Our sample includes daily and hourly data for the following cryptocurrencies: BitCoin, Ethereum and LiteCoin. These are the most popular cryptocurrencies with the highest market capitalisation and longest time horizon (Table 1). We analyse their exchange rates vis-à-vis the US dollar: BTCUSD, ETHUSD and LTCUSD. The sample period is 01.01.2015-01.09.2019, and the data sources are CoinMarketCap (https://coinmarketcap.com/coins/), Gemini (https://gemini.com/) and Bitstamp (https://www.bitstamp.net). These are leading exchanges and trading platforms in the cryptocurrency market. For example, CoinMarketCap calculates prices as the volume-weighted average of all prices reported for each market. As a result BitCoin prices are the average of those from 400 markets.

#	Name	Market Cap	Price	Circulating Supply	Data start from
1	Bitcoin	\$148 657 197 170	\$8 267.84	17 980 175 BTC	28 Apr 2013
2	Ethereum	\$19 674 550 330	\$182.07	108 059 235 ETH	07 Aug 2015
3	Ripple	\$12 017 970 035	\$0.278408	43 166 787 298 XRP	04 Aug 2013
4	Bitcoin Cash	\$4 236 366 686	\$234.76	18 045 263 BCH	23 Jul 2017

 Table 1: Capitalisation of the cryptocurrency market (08.10.2019)

#	Name	Name Market Cap		Circulating Supply	Data start from
5	Litecoin	\$3 659 603 443	\$57.70	63 420 942 LTC	28 Apr 2013

Cryptocurrency Market Capitalisation. Data source: https://coinmarketcap.com/coins/

We define overreactions on the basis of the number of standard deviations to be added to the average return.

In order to avoid the distortions caused by price gaps, returns  $(R_i)$  are computed as follows:

 $R_{i} = \left(\frac{\text{Close}_{i}}{\text{Open}_{i}} - 1\right) \times 100\%, \qquad (1)$ where  $R_{i}$  - returns on the *i*-th day in %; Open\_{i}- open price on the *i*-th day; Close\_{i} - close price on the *i*-th day.

The returns calculated using (1) are divided into two data sets corresponding respectively to positive and negative overreactions with the aim of testing for possible differences in price behaviour between those two cases.

A positive overreaction is defined as follows:

$$R_i > (\overline{R}_n + k \times \delta_n) \tag{2}$$

and a negative overreaction as:

$$R_i < (\overline{R}_n - k \times \delta_n) \tag{3}$$

where k is the number of standard deviations used to identify the overreaction (k=2 for BTCUSD and k=1.5 for ETHUSD and LTCUSD, k being chosen on the basis of the sample size to generate in each case a sufficient number of overreactions);  $\overline{R}_n$  is the average size of daily returns for period n.

$$\overline{R}_n = \sum_{i=1}^n R_i / n \tag{4}$$

and  $\delta_n$  is the standard deviation of daily returns for period n

$$\delta_n = \sqrt{\frac{1}{n} \sum_{i=1}^n (R_i - \overline{R})^2}$$
(5)

Then the following hypotheses are tested:

Hypothesis 1 (H1): The intraday behaviour of hourly returns is different on overreaction days compared to normal days.

Hypothesis 2 (H2): There is a momentum effect on overreaction days.

Hypothesis 3 (H3): There is a momentum effect after one-day abnormal returns.

To test these hypotheses we use the following methods:

- Visual inspection and average analysis;
- Student's t-tests;
- A cumulative abnormal returns approach;
- A trading simulation approach.

The cumulative abnormal returns approach is based on MacKinlay (1997) and is standard for event studies. Abnormal returns are defined as follows:

$$AR_t = R_t - E(R_t) \tag{6}$$

where  $R_t$  is the return at time t and  $E(R_t)$  is corresponding average return computed over the whole sample period as follows:

$$E(R_t) = \left(\frac{1}{T}\right) \sum_{i=1}^T R_i \tag{7}$$

where *T* is the sample size.

The cumulative abnormal return denoted as  $CAR_i$  is simply the sum of the abnormal returns:

$$CAR_i = \sum_{i=1}^{24} AR_i \tag{8}$$

Parametric t-tests are also carried out for Hypothesis 1. The Null Hypothesis (H0) is that the data (hourly returns on the overreaction day and in the full sample) belong to the same population, a rejection of the null suggesting the presence of a statistical anomaly in the price behavior on the overreaction day. The test is carried out at the 95% confidence level, and the degrees of freedom are N - 1 (N being equal to N1+ N2).

The trading simulation approach replicates the actions of traders by using appropriate algorithms for trading strategies based on the observed price patterns; its aim is to establish whether the detected anomalies can be exploited to generate abnormal profits. It should be mentioned that our analysis does not incorporate transaction costs such as spreads, broker or bank fees, swaps etc., and therefore is only a proxy for actual trading. However, in the case of Internet trading such costs are typically small and ignoring them does not affect the results.

The percentage results for an individual deal are computed as follows:

$$\% result = \left(\frac{P_{close}}{P_{open}} - 1\right) \times 100\%$$
<sup>(9)</sup>

where  $P_{open}$  – opening price for the trade

 $P_{close}$  – closing price for the trade

The sum of the results from each trade is the total financial result of trading. A strategy producing positive total profits implies that there exists an exploitable market anomaly.

Another important indicator of the degree of success of the trading strategy is the percentage of successful trades:

% successful trades = 
$$\frac{100\% \times number \ of \ successful \ trades}{overall \ number \ of \ trades}$$
(10)

To establish whether or not the results obtained are statistically different from the random trading ones t-tests are carried out. These compare the means from two samples to see whether or not they come from the same population. The first sample consists of the trading results from the trading strategy, and the second one of random trading results. The null hypothesis is that the mean is the same in both samples, and the alternative that it is not. The computed values of the t-test are compared with the critical ones at the 5% significance level. Failure to reject the null implies that there are no advantages from exploiting the trading strategy being considered since the trading results do not differ from the random ones, whilst a rejection suggests that the adopted strategy can generate abnormal profits since the trading results are not random and therefore it is possible to "beat the market". As an example, the t-test results for LTCUSD in the case of Strategy 1 are shown in Table 3.

Parameter	Value
Number of the trades	38
Total profit	311.39%
Average profit per trade	8.19%
Standard deviation	7.01%
t-test	7.20
t critical value (0.95%)	1.78
Null hypothesis	rejected

Table 2: t-test for evaluating the success of the trading strategy: LTCUSD, positive overreactions, Strategy 1

#### 3. Empirical Results

We divide the results into two sets including respectively those for the day of overreactions and those for the day after in order to explore price behaviour when abnormal price movements are observed and after them. The results for BTCUSD are presented in Appendices A and B (for the day of overreactions and the day after in turn). Figures A.1 and A.2 show that returns on overreaction days differ from those on normal days; the t-test statistics confirm that these differences are statistically significant. This holds for both positive and negative overreactions (Table A.2 and Table A.3).

The cumulative abnormal returns analysis (Table A.4 and Figure A.2) provides information on the average timing of the anomaly appearance: in general positive overreactions are detected after 6pm and negative ones after 4pm, namely in both cases some time before the end of the trading day.

Concerning price behaviour on the day after overreactions, average hourly BTCUSD returns after a positive overreaction are much lower than on normal days during the first hours of the following day (Figure B.1), and these differences are statistically significant (Table B.1), which implies the existence of a contrarian effect. As for negative overreactions, on the following day prices tend to move in the direction of the overreaction (Figure B.2 and Table B.2), which represents evidence of a momentum effect.

Specific timings for trading can be determined. The contrarian strategy for BTCUSD after a positive overreaction is as follows: sell on the start of the day and close position after 4pm. The momentum strategy for the case of negative overreactions is instead to sell at the start of the day after the negative overreaction and close this position after 11am (Table B.3 and Figure B.3). A similar analysis is carried out for LTCUSD (Appendices C and D) and ETHUSD (Appendices E and F). These results are summarised in Table 3 (for positive overreactions) and Table 4 (for

negative overreactions). As can be seen, they are generally very similar to those for BTCUSD.

-										
BTCUSD	LTCUSD	ETHUSD								
Day of the overreaction										
Yes	Yes	Yes								
Yes. CAR increase till	Yes. CAR increase	Yes. CAR increase								
the end of the day	till the end of the day	till the end of the day								
18:00	13:00	12:00								
Day after the overreact	tion									
No	Yes	Yes								
Since the start of the	Since the start of the	Since the start of the								
day till 16:00*	day till 13:00	day till 21:00								
	BTCUSD Day of the overreacti Yes Yes. CAR increase till the end of the day 18:00 Day after the overreac No Since the start of the day till 16:00*	BTCUSDLTCUSDDay of the overreactionYesYesYesYesYes. CAR increase tillYes. CAR increasethe end of the daytill the end of the day18:0013:00Day after the overreactionNoYesSince the start of the day till 16:00*Since the start of the day till 13:00								

#### Table 3: Overall results for the case of positive overreactions

\* contrarian effect detected

#### Table 4: Overall results for the case of negative overreactions

Parameter/ Instrument	BTCUSD	LTCUSD	ETHUSD							
Day of the overreaction										
Are there significant differences in Yes Yes Yes										
returns (overreaction day vs usual day)?										
Any patterns in cumulative abnormal	Yes. CAR decrease	Yes. CAR decrease	Yes. CAR decrease till							
returns dynamics?	till the end of the day	till the end of the day	the end of the day							
Timing of overreaction	16:00	14:00	13:00							
	Day after the overrea	iction								
Is there momentum effect on the day	Yes	Yes	No							
after the overreaction?										
Timing parameters of contrarian	Since the start of the	Since the start of the	Since the start of the							
movements	day till 11:00	day till 10:00	day till the end of the							
			day*							

\* contrarian effect detected

Our findings can be summarised as follows:

- hourly returns during the day of positive/negative overreactions are significantly higher/lower than those during the average positive/negative day;

- overreaction can be detected before the day ends and specific timing parameters for the overreactions can be estimated;

- prices tend to move in the direction of overreactions till the end of the day, i.e. H2 cannot be rejected: there is a momentum effect on the day of the overreaction;

- the behaviour of the market after one-day abnormal returns in most cases also confirms the existence of a momentum effect. Usually it is short-term, and specific timing parameters can be estimated for the asset of interest; - in two cases (BTCUSD positive overreactions and ETHUSD negative overreactions) a contrarian effect is detected.

On the basis of these results the following profitable strategies can be developed:

Strategy 1: when it becomes clear that the current day is an overreaction day (see the timing of overreaction parameter in Tables 3 and 4) a position in the direction of overreaction should be opened. This position should then be closed at the end of the day.

Strategy 2: at the beginning of the day after the overreaction a position in the direction of the overreaction should be opened. This position should then be closed on the basis of the timing parameters for the momentum effect displayed in Tables 3 and 4. If this effect is not present, a contrarian trading strategy should be used: at the beginning of the day after the overreaction a position in the opposite direction to the overreaction should be opened.

The trading simulation results for the two strategies are presented in Tables 5 (for positive overreactions) and 6 (for negative overreactions).

	-				-			
	Number	Number of				Profit	t-test	t-test
	of	successful	Number of		Profit	% per	calculated	status
	trades.	trades.	successful	Profit.	% per	trade	value	
Instrument	units	units	trades. %	%	year			
			Strat	egy 1				
BTCUSD	49	42	86%	143.11%	28.62%	2.92%	6.62	rejected
LTCUSD	38	37	97%	311.39%	103.80%	8.19%	7.20	rejected
ETHUSD	58	45	78%	507.63%	50.76%	8.75%	11.06	rejected
			Strat	egy 2				
								not
BTCUSD*	49	29	59.2%	75.3%	15.06%	1.54%	1.75	rejected
								not
LTCUSD	38	14	37%	61.80%	20.60%	1.63%	0.76	rejected
ETHUSD	58	25	43%	130.20%	43.40%	2.24%	1.87	rejected

Table 5: Trading simulation results for the case of positive overreactions

\* A contrarian trading strategy is used

	Number	Number of				Profit	t-test	t-test			
	of	successful	Number of		Profit	% per	calculated	status			
	trades.	trades.	successful	Profit.	% per	trade	value				
Instrument	units	units	trades. %	%	year						
Strategy 1											
BTCUSD	46	42	91%	202.41%	40.48%	4.40%	8.21	rejected			
LTCUSD	39	33	85%	170.75%	56.92%	4.38%	4.31	rejected			

ETHUSD	57	43	75%	334.54%	111.51%	5.87%	7.84	rejected		
Strategy 2										
								not		
BTCUSD	46	25	54.3%	52.0%	10.4%	1.13%	1.55	rejected		
LTCUSD	39	20	51.3%	75.8%	25.3%	1.94%	1.93	rejected		
ETHUSD*	57	38	66.7%	225.6%	75.2%	3.96%	3.46	rejected		

#### \* A contrarian trading strategy is used

As can be seen, the detected anomalies can be exploited in most cases to generate abnormal profits from trading and to "beat the market".

#### 4. Conclusions

This paper explores the momentum effect in the cryptocurrency market after one-day abnormal returns. Daily and intraday data on the BTCUSD, ETHUSD and LTCUSD exchange rates over the period 01.01.2017-01.09.2019 are analysed using a number of statistical methods as well as a trading simulation approach. The results suggest that hourly returns during the day of positive/negative overreactions are significantly higher/lower than those during the average positive/negative day.

Further, overreactions can usually be detected before the day ends by estimating specific timing parameters. Prices tend to move in the direction of the overreaction till the end of the day when it occurs, which implies the existence of a momentum effect during that day giving rise to exploitable profit opportunities. This effect (together with profit opportunities) is also observed on the following day. In two cases (BTCUSD positive overreactions and ETHUSD negative overreactions) a contrarian effect is detected instead. These findings are of interest to both investors aiming to maximize their profits and academics interested in the empirical relevance of the EMH.

#### References

Aalborg, H.A., Molnár, P., and J.E. de Vries, (2019), What can explain the price, volatility and trading volume of Bitcoin? Finance Research Letters, 29, pp. 255-265.

Aiyagari, S. Rao and M. Gertler, (1999), Overreaction of Asset Prices in General Equilibrium. Review of Economic Dynamics, 2 (1), pp. 3–35.

Bariviera, A. F., (2017), The Inefficiency of Bitcoin Revisited: A Dynamic Approach. Economics Letters, Vol. 161, pp. 1-4.

Bartos, J., (2015), Does Bitcoin follow the hypothesis of efficient market?. International Journal of Economic Sciences, Vol. IV(2), pp. 10-23.

Bouri, E., Lau, C. K. M., Lucey, B., and D. Roubaud, (2018), Trading volume and the predictability of return and volatility in the cryptocurrency market. Finance Research Letters. doi:10.1016/j.frl.2018.08.015.

Caporale G. M., and A. Plastun, (2018), The day of the week effect in the cryptocurrency market Finance Research Letters Available online 19 November 2018, ISSN 1544-6123, https://doi.org/10.1016/j.frl.2018.11.012.

Caporale, G. and A. Plastun, (2019), Price overreactions in the cryptocurrency market, Journal of Economic Studies, Vol. 46 No. 5, pp. 1137-1155.

Caporale, G.M., Plastun, A. and V. Oliinyk, (2019), Bitcoin fluctuations and the frequency of price overreactions. Financial Markets and Portfolio Management, Volume 33, Issue 2, pp 109–131.

Caporale, G. M., Gil-Alana, L., and A. Plastun, (2018), Short-term Price Overreactions: Identification, Testing, Exploitation, Computational Economics, Volume 51, Issue 4, pp 913–940.

Cheung, A., E. Roca and J.-J. Su, (2015), Crypto-Currency Bubbles: An Application of the Phillips-Shi-Yu (2013) Methodology on Mt. Gox Bitcoin Prices. Applied Economics, vol. 47, 2015, pp. 2348-2358.

Chevapatrakul, T., and D. Mascia, (2019), Detecting overreaction in the Bitcoin market: A quantile autoregression approach. Finance Research Letters, Volume 30, pp. 371-377.

Corbet, S., Lucey, B., and L. Yarovaya, (2018), Datestamping the Bitcoin and Ethereum bubbles. Finance Research Letters, 26, pp. 81–88.

Cox, D. R. and D. R. Peterson, (1994), Stock Returns Following Large One-Day Declines: Evidence on Short-Term Reversals and Longer-Term Performance. Journal of Finance, 49, pp. 255-267.

Cutler, D., J. Poterba, and L. Summers, (1991), Speculative dynamics. Review of Economics Studies, 58, pp. 529–546.

De Bondt W. and R. Thaler, (1985), Does the Stock Market Overreact? Journal of Finance, 40, pp. 793-808.

Ferri, M., G. and C. Min, (1996), Evidence that the Stock Market Overreacts and Adjusts. The Journal of Portfolio Management, 22, pp. 71-76.

Griffin, D. and A. Tversky, (1992), The weighing of evidence and the determinants of confidence. Cognitive Psychology, 24, pp. 411-435.

Halaburda, H., and N. Gandal, (2014), Competition in the Cryptocurrency Market. NET Institute Working Paper No. 14-17. Available at SSRN: https://ssrn.com/abstract=2506463 or http://dx.doi.org/10.2139/ssrn.2506463.

Hong, H., and J. C. Stein, (1999), A Unified Theory of Underreaction, Momentum Trading and Overreaction in Asset Markets. Journal of Finance, 54, pp. 2143-2184.

Jegadeesh, N. and S. Titman, (1993), Returns to Buying Winners and Selling Losers: Implications for Stock Market Efficiency. The Journal of Finance, 48, pp. 65-91.

Kurihara, Y., and A. Fukushima, (2017), The Market Efficiency of Bitcoin: A Weekly Anomaly Perspective. Journal of Applied Finance & Banking, vol. 7, no. 3, pp. 57-64.

Lobe, S., and J. Rieks, (2010), Short-Term Market Overreaction on the Frankfurt Stock Exchange. Quarterly Review of Economics and Finance, Vol. 51, Issue 2, pp. 113-123.

Madura, J. and N. Richie, (2004), Overreaction of exchange traded funds during the bubble of 1998-2002. Journal of Behavioral Finance, 5 (2), pp. 91-104.

Mynhardt, R. H. and A. Plastun, (2013), The Overreaction Hypothesis: The case of Ukrainian stock market. Corporate Ownership and Control, 11, pp. 406-423.

Poteshman, A., (2001), Underreaction, overreaction and increasing misreaction to information in the options market. Journal of Finance, 56, pp. 851–876.

Savor, P., (2012), Stock Returns after Major Price Shocks: The Impact of Information. Journal of Financial Economics, 106 (3), pp. 635–659.

Urquhart, A., (2016), The Inefficiency of Bitcoin. Economics Letters, vol. 148, pp. 80-82.

#### Appendix A

#### **BTCUSD:** day of overreaction

<u> </u>		Positive	Negative	Hourly	Positive days	Negative days
Parameter	Daily data	days	days	data	(hourly data)	(hourly data)
Mean	0.0034	0.1178	-0.1086	0.0002	0.0047	-0.0045
Standard error	0.0011	0.0053	0.0033	0.0000	0.0005	0.0006
Median	0.0025	0.1051	-0.1029	0.0000	0.0029	-0.0019
Mode	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Std. Dev.	0.0409	0.0372	0.0225	0.0091	0.0162	0.0203
Variance	0.0017	0.0014	0.0005	0.0001	0.0003	0.0004
Kurtosis	4.5097	8.1529	-0.1042	27.3573	8.9389	13.6233
Skewness	0.2498	2.6496	-0.8845	-0.1641	1.1748	-0.4895
Minimum	-0.1617	0.0853	-0.1617	-0.1854	-0.0768	-0.1854
Maximum	0.2763	0.2763	-0.0789	0.1664	0.1209	0.1664
Sum	4.9027	5.7742	-4.9956	5.1913	5.5753	-5.0193
Observations	1427	49	46	34235	1176	1104

Table A.1: Descriptive statistics for BTCUSD data: daily, hourly, days with positive and negative returns

### Figure A.1: Average hourly returns on overreaction and normal days: the case of positive overreactions, BTCUSD



Figure A.2: Average hourly returns on overreaction and normal days: the case of negative overreactions, BTCUSD



### Table A.2: t-test of hourly returns on overreaction and normal days: the case of positive overreactions, BTCUSD

F = m = m = m							
	Average			Average			
	return on	Standard		return on	Standard		
	positive	deviation	Number of	usual day	deviation	Number of	
	overreaction	(OD)	observations	with positive	(UD)	observation	t
Hour	day (OD)		(OD)	returns (UD)		(UD)	criterion
0:00	0.94%	1.16%	48	0.12%	0.81%	756	4.80
1:00	0.32%	1.62%	48	0.10%	0.76%	756	0.93
2:00	0.22%	1.17%	48	0.12%	0.72%	756	0.56
<u>3:00</u>	<u>0.46%</u>	<u>1.38%</u>	<u>48</u>	0.02%	0.65%	<u>756</u>	2.24
<u>4:00</u>	<u>0.76%</u>	<u>2.19%</u>	<u>48</u>	<u>0.08%</u>	<u>0.69%</u>	<u>756</u>	2.15
<u>5:00</u>	<u>0.40%</u>	<u>1.41%</u>	<u>48</u>	<u>0.01%</u>	0.62%	<u>756</u>	<u>1.92</u>
6:00	0.17%	1.61%	48	0.08%	0.65%	756	0.37
7:00	0.11%	1.19%	48	0.11%	0.70%	756	-0.01
<u>8:00</u>	<u>0.59%</u>	<u>1.66%</u>	<u>48</u>	<u>0.06%</u>	0.72%	<u>756</u>	2.22
9:00	0.49%	1.78%	48	0.10%	0.72%	756	1.52
10:00	0.37%	1.14%	48	0.07%	0.69%	756	1.80
11:00	0.30%	2.36%	48	0.12%	0.70%	756	0.51
<u>12:00</u>	<u>1.18%</u>	<u>2.17%</u>	<u>48</u>	<u>0.10%</u>	<u>0.70%</u>	<u>756</u>	<u>3.44</u>
13:00	0.46%	1.64%	48	0.12%	0.83%	756	1.42
14:00	0.50%	2.09%	48	0.08%	0.73%	756	1.36
<u>15:00</u>	<u>0.51%</u>	<u>1.19%</u>	<u>48</u>	0.12%	0.83%	756	2.22
16:00	0.20%	1.93%	48	0.13%	0.74%	756	0.25
17:00	0.50%	1.70%	48	0.10%	0.67%	756	1.61
<u>18:00</u>	0.69%	1.20%	48	0.06%	0.58%	756	3.56

19:00	0.38%	1.55%	48	0.06%	0.69%	756	1.42
20:00	0.30%	1.80%	48	0.01%	0.70%	756	1.09
21:00	0.41%	1.25%	48	0.10%	0.79%	756	1.69
22:00	<u>0.76%</u>	<u>1.59%</u>	<u>48</u>	0.13%	<u>0.73%</u>	756	2.75
23:00	0.37%	1.14%	48	0.11%	0.73%	756	1.56

Table A.3:t-test of hourly returns on overreaction and n	ormal days: the case of
negative overreactions, BTCUSD	

		,					
	Average			Average			
	return on	Standard		return on	Standard		
	positive	deviation	Number of	usual day	deviation	Number of	
	overreaction	(OD)	observations	with positive	(UD)	observation	t
Hour	day (OD)		(OD)	returns (UD)		(UD)	criterion
0:00	-0.34%	1.83%	45	-0.18%	0.93%	570	-0.59
1:00	-0.39%	1.37%	45	-0.08%	0.82%	570	-1.50
<u>2:00</u>	<u>-0.70%</u>	<u>1.52%</u>	<u>45</u>	<u>-0.16%</u>	<u>0.87%</u>	<u>570</u>	-2.37
3:00	-0.41%	1.49%	45	-0.08%	0.74%	570	-1.48
4:00	-0.12%	1.43%	45	-0.13%	0.89%	570	0.07
5:00	0.03%	1.12%	45	-0.10%	0.87%	570	0.75
6:00	-0.38%	1.33%	45	-0.05%	0.75%	570	-1.62
<u>7:00</u>	<u>-0.42%</u>	<u>1.01%</u>	<u>45</u>	<u>0.01%</u>	<u>0.90%</u>	<u>570</u>	-2.75
<u>8:00</u>	<u>-0.59%</u>	<u>1.42%</u>	<u>45</u>	<u>-0.09%</u>	0.80%	<u>570</u>	-2.32
<u>9:00</u>	<u>-0.81%</u>	<u>1.80%</u>	<u>45</u>	-0.08%	<u>0.76%</u>	<u>570</u>	-2.70
10:00	-0.47%	1.68%	45	-0.13%	0.88%	570	-1.37
11:00	-0.25%	1.82%	45	-0.05%	0.73%	570	-0.70
12:00	-0.60%	2.16%	45	-0.08%	0.84%	570	-1.61
<u>13:00</u>	<u>-1.21%</u>	<u>2.20%</u>	<u>45</u>	-0.13%	<u>1.10%</u>	<u>570</u>	-3.24
14:00	-0.02%	2.46%	45	-0.14%	0.93%	570	0.32
15:00	-0.26%	2.04%	45	-0.08%	0.85%	570	-0.59
<u>16:00</u>	<u>-1.09%</u>	<u>2.26%</u>	<u>45</u>	<u>-0.07%</u>	<u>1.11%</u>	<u>570</u>	-3.01
<u>17:00</u>	<u>-0.95%</u>	<u>2.26%</u>	<u>45</u>	<u>-0.08%</u>	<u>0.87%</u>	<u>570</u>	-2.57
18:00	-0.38%	1.70%	45	-0.08%	0.73%	570	-1.18
19:00	-0.24%	1.45%	45	-0.06%	0.94%	570	-0.85
20:00	-0.18%	2.62%	45	-0.06%	0.88%	570	-0.29
21:00	-0.47%	2.05%	45	0.00%	1.07%	570	-1.51
22:00	-0.65%	3.79%	45	-0.02%	0.84%	570	-1.12
23:00	-0.01%	3.25%	45	-0.08%	0.93%	570	0.14

Figure A.2: Dynamics of cumulative abnormal returns: case of BTCUSD



 Table A.4: Cumulative abnormal returns: the case of positive and negative overreactions, BTCUSD

	Pos	sitive overread	tions	Ne	gative overreac	tions
	Abnormal	CAD	Overreaction	Abnormal	CAD	Overreaction
Hour	returns	CAK	cross	returns	CAK	cross
0:00	0.82%	0.82%	7.58%	-0.16%	-0.16%	-7.49%
1:00	0.22%	1.04%	7.26%	-0.31%	-0.47%	-7.10%
2:00	0.10%	1.13%	7.04%	-0.54%	-1.01%	-6.40%
3:00	0.45%	1.58%	6.58%	-0.33%	-1.35%	-5.99%
4:00	0.68%	2.26%	5.82%	0.02%	-1.33%	-5.87%
5:00	0.39%	2.65%	5.41%	0.13%	-1.20%	-5.91%
6:00	0.09%	2.74%	5.24%	-0.33%	-1.53%	-5.53%
7:00	0.00%	2.74%	5.14%	-0.43%	-1.96%	-5.11%
8:00	0.53%	3.27%	4.54%	-0.50%	-2.45%	-4.52%
9:00	0.39%	3.67%	4.05%	-0.73%	-3.18%	-3.71%
10:00	0.30%	3.96%	3.68%	-0.35%	-3.53%	-3.24%
11:00	0.17%	4.14%	3.39%	-0.19%	-3.72%	-2.99%
12:00	1.08%	5.22%	2.21%	-0.52%	-4.24%	-2.39%
13:00	0.34%	5.56%	1.75%	-1.07%	-5.31%	-1.18%
14:00	0.41%	5.97%	1.25%	0.12%	-5.20%	-1.16%
15:00	0.39%	6.35%	0.74%	-0.18%	-5.38%	-0.89%
16:00	0.07%	6.43%	0.54%	-1.02%	-6.40%	<u>0.20%</u>
17:00	0.40%	6.82%	0.04%	-0.87%	-7.27%	1.15%
18:00	0.62%	7.44%	<u>-0.65%</u>	-0.30%	-7.57%	1.53%
19:00	0.32%	7.76%	-1.03%	-0.19%	-7.76%	1.78%
20:00	0.28%	8.05%	-1.33%	-0.12%	-7.87%	1.96%
21:00	0.31%	8.36%	-1.73%	-0.47%	-8.34%	2.43%
22:00	0.63%	8.99%	-2.49%	-0.63%	-8.97%	3.08%
23:00	0.26%	9.25%	-2.86%	0.07%	-8.90%	3.08%

### Appendix B BTCUSD: day after the overreaction





### Table B.1: t-test of hourly returns on the day after the overreaction and normaldays: the case of positive overreactions, BTCUSD

Hour	Average return on day after positive overreaction (OD)	Standard deviation (OD)	Number of observations (OD)	Average return on usual day (UD)	Standard deviation (UD)	Number of observation (UD)	t criterion
0:00	<u>-0.36%</u>	<u>1.46%</u>	<u>49</u>	0.18%	<u>0.47%</u>	<u>1427</u>	-2.60
1:00	0.14%	1.44%	49	0.25%	0.85%	1427	-0.51
2:00	0.04%	1.46%	49	0.06%	0.57%	1427	-0.11
<u>3:00</u>	<u>-0.24%</u>	<u>1.16%</u>	<u>49</u>	<u>0.16%</u>	<u>0.44%</u>	<u>1427</u>	-2.43
4:00	-0.05%	1.89%	49	0.05%	0.85%	1427	-0.36
5:00	-0.32%	1.52%	49	-0.04%	0.66%	1427	-1.29
6:00	0.01%	1.37%	49	0.20%	0.53%	1427	-0.95
<u>7:00</u>	0.58%	<u>1.28%</u>	<u>49</u>	<u>0.04%</u>	<u>0.78%</u>	<u>1427</u>	<u>2.91</u>
<u>8:00</u>	<u>-0.52%</u>	<u>1.36%</u>	<u>49</u>	<u>0.11%</u>	<u>0.42%</u>	<u>1427</u>	<u>-3.28</u>
9:00	0.00%	1.36%	49	0.11%	0.80%	1427	-0.55
10:00	0.19%	1.27%	49	-0.06%	0.66%	1427	1.39
11:00	0.08%	1.53%	49	0.18%	0.51%	1427	-0.49
12:00	0.24%	2.33%	49	-0.24%	1.02%	1427	1.43
<u>13:00</u>	<u>-0.17%</u>	<u>1.36%</u>	<u>49</u>	<u>0.19%</u>	<u>0.91%</u>	<u>1427</u>	<u>-1.82</u>
14:00	0.05%	1.75%	49	0.06%	0.44%	1427	-0.04
<u>15:00</u>	<u>-0.31%</u>	<u>1.31%</u>	<u>49</u>	0.25%	<u>0.70%</u>	<u>1427</u>	-2.99
<u>16:00</u>	<u>-0.50%</u>	<u>1.70%</u>	<u>49</u>	<u>0.09%</u>	<u>0.63%</u>	<u>1427</u>	-2.42
<u>17:00</u>	0.51%	<u>1.28%</u>	<u>49</u>	<u>-0.29%</u>	<u>0.90%</u>	<u>1427</u>	4.36
18:00	0.09%	1.32%	49	-0.04%	0.48%	1427	0.68

19:00	-0.35%	1.31%	49	-0.06%	0.64%	1427	-1.56
20:00	0.11%	1.21%	49	-0.10%	1.07%	1427	1.25
21:00	0.06%	1.98%	49	0.09%	1.01%	1427	-0.09
22:00	0.10%	1.74%	49	0.05%	0.66%	1427	0.19
23:00	0.02%	1.57%	49	0.08%	0.69%	1427	-0.28

Figure	<b>B.2:</b>	Average	hourly	returns	on	the	day	after	the	overreaction	and
normal	days	: the case	of negat	tive over	reac	tions	, BT	CUSD			



Table B.2: t-test of hourly returns on the day after the overreaction and normaldays: the case of negative overreactions, BTCUSD

Hour	Average return on day after negative overreaction (OD)	Standard deviation (OD)	Number of observations (OD)	Average return on usual day (UD)	Standard deviation (UD)	Number of observation (UD)	t criterion
0:00	0.04%	1.76%	46	0.18%	0.47%	1427	-0.54
<u>1:00</u>	<u>-0.31%</u>	<u>1.83%</u>	<u>46</u>	0.25%	0.85%	<u>1427</u>	<u>-2.07</u>
2:00	-0.19%	1.91%	46	0.06%	0.57%	1427	-0.87
3:00	0.15%	1.47%	46	0.16%	0.44%	1427	-0.05
<u>4:00</u>	-0.28%	<u>1.22%</u>	<u>46</u>	<u>0.05%</u>	0.85%	<u>1427</u>	<u>-1.78</u>
5:00	-0.12%	1.89%	46	-0.04%	0.66%	1427	-0.31
6:00	0.05%	1.50%	46	0.20%	0.53%	1427	-0.65
7:00	0.24%	1.31%	46	0.04%	0.78%	1427	1.02
8:00	-0.08%	1.87%	46	0.11%	0.42%	1427	-0.70
9:00	0.06%	2.27%	46	0.11%	0.80%	1427	-0.15
10:00	-0.21%	1.67%	46	-0.06%	0.66%	1427	-0.60
11:00	-0.19%	1.59%	46	0.18%	0.51%	1427	-1.59
12:00	0.36%	2.20%	46	-0.24%	1.02%	1427	1.83

13:00	0.01%	2.66%	46	0.19%	0.91%	1427	-0.46
14:00	-0.19%	2.48%	46	0.06%	0.44%	1427	-0.69
15:00	0.20%	1.69%	46	0.25%	0.70%	1427	-0.22
16:00	0.65%	2.24%	46	0.09%	0.63%	1427	1.69
17:00	-0.34%	1.59%	46	-0.29%	0.90%	1427	-0.20
18:00	0.00%	1.38%	46	-0.04%	0.48%	1427	0.18
19:00	0.52%	1.81%	46	-0.06%	0.64%	1427	2.14
20:00	0.31%	2.02%	46	-0.10%	1.07%	1427	1.38
21:00	0.68%	1.22%	46	0.09%	1.01%	1427	3.23
22:00	<u>-0.34%</u>	<u>1.51%</u>	<u>46</u>	0.05%	0.66%	<u>1427</u>	<u>-1.75</u>
23:00	0.32%	1.01%	46	0.08%	0.69%	1427	1.58

Table   B.3	: Cumulative	abnormal	returns:	the	case	of	positive	and	negative
overreactio	ons, BTCUSD								

	Positive	e overreactions	Negativ	e overreactions
	Abnormal	Cumulative	Abnormal	Cumulative
Hour	returns	abnormal returns	returns	abnormal returns
0:00	-0.54%	-0.54%	-0.14%	0.14%
1:00	-0.11%	-0.65%	-0.56%	0.70%
2:00	-0.02%	-0.67%	-0.24%	0.94%
3:00	-0.41%	-1.08%	-0.01%	0.95%
4:00	-0.10%	-1.18%	-0.32%	1.28%
5:00	-0.28%	-1.46%	-0.09%	1.36%
6:00	-0.19%	-1.64%	-0.14%	1.51%
7:00	0.54%	-1.11%	0.20%	1.31%
8:00	-0.64%	-1.74%	-0.19%	1.50%
9:00	-0.11%	-1.85%	-0.05%	1.55%
10:00	0.25%	-1.60%	-0.15%	1.70%
<u>11:00</u>	-0.11%	-1.70%	<u>-0.37%</u>	2.07%
12:00	0.48%	-1.23%	0.60%	1.48%
13:00	-0.36%	-1.59%	-0.18%	1.66%
14:00	-0.01%	-1.60%	-0.25%	1.91%
15:00	-0.56%	-2.16%	-0.05%	1.97%
<u>16:00*</u>	<u>-0.59%</u>	-2.74%	0.56%	1.41%
17:00	0.81%	-1.94%	-0.05%	1.46%
18:00	0.13%	-1.81%	0.04%	1.42%
19:00	-0.29%	-2.10%	0.57%	0.85%
20:00	0.22%	-1.88%	0.41%	0.43%
21:00	-0.02%	-1.91%	0.59%	-0.15%
22:00	0.05%	-1.86%	-0.39%	0.24%
23:00	-0.06%	-1.92%	0.24%	0.00%

\* contrarian effect detected





### Appendix C

### LTCUSD: day of overreaction

1	8					
		Positive		Hourly	Positive days	Negative days
Parameter	Daily data	days	Negative days	data	(hourly data)	(hourly data)
Mean	0.0047	0.2057	-0.1201	0.0001	0.0072	-0.0050
Standard error	0.0021	0.0192	0.0043	0.0001	0.0009	0.0008
Median	-0.0024	0.1504	-0.1184	-0.0001	0.0027	-0.0035
Mode	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Std. Dev.	0.0698	0.1368	0.0304	0.0136	0.0265	0.0252
Variance	0.0049	0.0187	0.0009	0.0002	0.0007	0.0006
Kurtosis	27.9109	9.1535	9.5862	23.1228	10.2157	9.8648
Skewness	3.2292	2.7385	-2.3506	1.2878	1.8318	0.3360
Minimum	-0.2650	0.1095	-0.2650	-0.1773	-0.0876	-0.1773
Maximum	0.8349	0.8349	-0.0872	0.2061	0.1950	0.2002
Sum	5.1325	10.4890	-5.8864	2.2125	6.5569	-4.6369
Observations	1095	51	49	19047	912	936

### Table C.1: Descriptive statistics for LTCUSD data: daily, hourly, days with positive and negative returns

### Figure C.1: Average hourly returns on overreaction and normal days: the case of positive overreactions, LTCUSD



Figure C.2: Average hourly returns on overreaction and normal days: the case of negative overreactions, LTCUSD



### Table C.2: t-test of hourly returns on overreaction and normal days: the case of positive overreactions, LTCUSD

	Average			Average			
	return on	Standard		return on	Standard		
	positive	deviation	Number of	usual day	deviation	Number of	
	overreaction	(OD)	observations	with positive	(UD)	observation	t
Hour	day (OD)		(OD)	returns (UD)		(UD)	criterion
0:00	<u>0.67%</u>	<u>1.34%</u>	<u>37</u>	<u>0.02%</u>	<u>1.31%</u>	<u>794</u>	<u>2.88</u>
1:00	0.50%	2.63%	37	-0.04%	1.26%	794	1.24
2:00	0.88%	2.05%	<u>37</u>	<u>-0.05%</u>	<u>1.32%</u>	<u>793</u>	2.75
3:00	0.28%	1.82%	37	-0.08%	1.09%	793	1.22
4:00	0.97%	3.71%	37	0.03%	1.38%	793	1.54
5:00	0.22%	1.87%	37	-0.07%	1.21%	793	0.91
6:00	0.11%	2.03%	37	0.00%	1.21%	793	0.34
7:00	0.27%	1.61%	37	-0.04%	1.04%	793	1.16
<u>8:00</u>	<u>0.73%</u>	<u>2.19%</u>	<u>37</u>	<u>-0.06%</u>	<u>1.18%</u>	<u>793</u>	<u>2.19</u>
9:00	0.54%	2.17%	37	-0.03%	1.33%	793	1.60
10:00	0.33%	2.67%	37	0.03%	1.30%	793	0.68
<u>11:00</u>	<u>1.43%</u>	<u>2.71%</u>	<u>37</u>	<u>0.15%</u>	<u>1.30%</u>	<u>794</u>	<u>2.86</u>
<u>12:00</u>	<u>2.21%</u>	<u>3.85%</u>	<u>37</u>	<u>0.14%</u>	<u>1.54%</u>	<u>794</u>	<u>3.26</u>
<u>13:00</u>	<u>2.40%</u>	<u>4.46%</u>	<u>37</u>	<u>0.04%</u>	<u>1.84%</u>	<u>794</u>	<u>3.20</u>
14:00	0.59%	2.59%	37	-0.08%	1.55%	794	1.55
<u>15:00</u>	1.28%	2.56%	37	0.11%	<u>1.59%</u>	<u>794</u>	2.76
16:00	1.00%	3.63%	37	0.10%	1.84%	794	1.50
17:00	-0.18%	2.24%	37	-0.03%	1.26%	794	-0.40

<u>18:00</u>	0.89%	2.45%	<u>37</u>	<u>-0.01%</u>	<u>1.09%</u>	794	2.21
19:00	0.50%	1.94%	37	0.03%	1.25%	794	1.47
20:00	0.38%	2.49%	37	-0.01%	1.37%	794	0.95
21:00	0.30%	1.88%	37	0.07%	1.28%	794	0.74
22:00	0.24%	3.18%	37	0.07%	1.48%	794	0.33
23:00	0.72%	2.08%	37	0.01%	<u>1.24%</u>	<u>794</u>	2.06

Table C.3:t-test of hourly returns on overreaction and	normal	days:	the	case of
negative overreactions, LTCUSD				

Hour 0:00	Average return on negative overreaction day (OD) -0.41%	Standard deviation (OD) 2.10%	Number of observations (OD) 38	Average return on usual day with negative returns (UD) -0.17%	Standard deviation (UD) 1.36%	Number of observation (UD) 409	t criterion -0.69
1:00	-0.67%	1.67%	38	-0.23%	1.22%	409	-1.59
2:00	-0.81%	2.03%	38	-0.29%	1.33%	409	-1.55
<u>3:00</u>	<u>-0.72%</u>	<u>1.60%</u>	<u>38</u>	<u>-0.23%</u>	<u>1.11%</u>	<u>409</u>	-1.86
4:00	-0.16%	1.62%	38	-0.08%	1.12%	409	-0.29
5:00	-0.68%	2.14%	38	-0.16%	1.21%	409	-1.46
6:00	-0.42%	2.79%	38	-0.17%	1.31%	409	-0.54
7:00	-0.62%	1.51%	38	-0.19%	1.02%	409	-1.75
<u>8:00</u>	<u>-1.08%</u>	<u>1.94%</u>	<u>38</u>	<u>-0.29%</u>	<u>1.12%</u>	<u>409</u>	-2.46
9:00	-1.02%	2.68%	38	-0.25%	1.35%	409	-1.76
10:00	0.20%	1.90%	38	-0.18%	1.22%	409	1.21
11:00	-0.49%	2.44%	38	-0.08%	1.13%	409	-1.03
12:00	-0.35%	2.13%	38	-0.04%	1.31%	409	-0.87
13:00	-0.99%	3.55%	38	-0.25%	1.69%	409	-1.27
14:00	-0.96%	3.45%	38	-0.27%	1.64%	409	-1.22
15:00	0.54%	4.02%	38	-0.08%	1.64%	409	0.94
<u>16:00</u>	<u>-1.36%</u>	<u>2.39%</u>	<u>38</u>	<u>-0.15%</u>	<u>1.87%</u>	<u>409</u>	<u>-3.04</u>
17:00	-0.52%	2.84%	38	-0.12%	1.35%	409	-0.86
18:00	0.00%	2.06%	38	-0.17%	1.07%	409	0.50
19:00	-0.20%	2.20%	38	-0.09%	1.38%	409	-0.32
20:00	-0.04%	3.73%	38	-0.11%	1.54%	409	0.11
21:00	-0.41%	2.51%	38	-0.05%	1.35%	409	-0.87
22:00	-0.17%	2.20%	38	-0.03%	1.36%	409	-0.40
23:00	-0.55%	2.63%	38	-0.15%	1.27%	409	-0.92

Figure C.3: Dynamics of cumulative abnormal returns, LTCUSD



Table C.4: Cumulative abnormal returns: the case of positive and negative overreactions, LTCUSD

	Pos	Positive overreactions			Negative overreactions			
		Cumulative			Cumulative			
	Abnormal	abnormal	Overreaction	Abnormal	abnormal	Overreaction		
Hour	returns	returns	cross	returns	returns	cross		
0:00	0.65%	0.65%	10.27%	-0.24%	-0.24%	-8.19%		
1:00	0.54%	1.19%	9.77%	-0.44%	-0.68%	-7.52%		
2:00	0.93%	2.12%	8.89%	-0.52%	-1.20%	-6.71%		
3:00	0.37%	2.49%	8.61%	-0.49%	-1.70%	-5.99%		
4:00	0.94%	3.43%	7.63%	-0.08%	-1.78%	-5.83%		
5:00	0.28%	3.72%	7.42%	-0.51%	-2.29%	-5.16%		
6:00	0.11%	3.83%	7.31%	-0.25%	-2.53%	-4.74%		
7:00	0.31%	4.14%	7.04%	-0.44%	-2.97%	-4.12%		
8:00	0.79%	4.93%	6.31%	-0.78%	-3.75%	-3.04%		
9:00	0.58%	5.51%	5.77%	-0.78%	-4.53%	-2.02%		
10:00	0.30%	5.81%	5.44%	0.38%	-4.15%	-2.21%		
11:00	1.28%	7.09%	4.01%	-0.41%	-4.56%	-1.72%		
12:00	2.07%	9.16%	1.81%	-0.31%	-4.87%	-1.37%		
<u>13:00</u>	<u>2.35%</u>	<u>11.51%</u>	<u>-0.59%</u>	-0.74%	-5.61%	-0.38%		
<u>14:00</u>	0.67%	12.18%	-1.18%	<u>-0.69%</u>	<u>-6.30%</u>	<u>0.58%</u>		
15:00	1.17%	13.35%	-2.46%	0.62%	-5.69%	0.04%		
16:00	0.90%	14.25%	-3.47%	-1.21%	-6.90%	1.40%		
17:00	-0.15%	14.10%	-3.29%	-0.40%	-7.30%	1.92%		
18:00	0.89%	14.99%	-4.17%	0.17%	-7.13%	1.92%		
19:00	0.47%	15.47%	-4.68%	-0.11%	-7.24%	2.12%		
20:00	0.39%	15.86%	-5.06%	0.07%	-7.17%	2.16%		
21:00	0.23%	16.09%	-5.36%	-0.36%	-7.54%	2.56%		
22:00	0.17%	16.27%	-5.60%	-0.14%	-7.68%	2.74%		
23:00	0.71%	16.98%	-6.32%	-0.40%	-8.08%	3.29%		

### Appendix D LTCUSD: day after the overreaction





### Table D.1: t-test of hourly returns on the day after the overreaction and normal days: the case of positive overreactions, LTCUSD

Hour	Average return on day after positive overreaction (OD)	Standard deviation (OD)	Number of observations (OD)	Average return on usual day (UD)	Standard deviation (UD)	Number of observation (UD)	t criterion
0:00	0.32%	2.13%	38	0.02%	1.31%	794	0.87
1:00	0.19%	2.29%	38	-0.04%	1.26%	794	0.62
2:00	-0.18%	2.32%	38	-0.05%	1.32%	793	-0.35
3:00	-0.33%	1.75%	38	-0.08%	1.09%	793	-0.85
4:00	0.55%	3.48%	38	0.03%	1.38%	793	0.91
5:00	-0.15%	2.22%	38	-0.07%	1.21%	793	-0.24
6:00	0.32%	2.12%	38	0.00%	1.21%	793	0.93
7:00	0.11%	1.61%	38	-0.04%	1.04%	793	0.60
8:00	-0.26%	1.58%	38	-0.06%	1.18%	793	-0.77
9:00	-0.29%	2.11%	38	-0.03%	1.33%	793	-0.75
10:00	-0.37%	2.35%	38	0.03%	1.30%	793	-1.04
11:00	0.60%	2.43%	38	0.15%	1.30%	794	1.14
12:00	0.25%	2.38%	38	0.14%	1.54%	794	0.28
13:00	0.58%	3.47%	38	0.04%	1.84%	794	0.95
14:00	-0.47%	2.37%	38	-0.08%	1.55%	794	-1.00
15:00	0.18%	1.92%	38	0.11%	1.59%	794	0.22
16:00	-0.12%	2.56%	38	0.10%	1.84%	794	-0.52
17:00	0.07%	2.11%	38	-0.03%	1.26%	794	0.30
18:00	-0.23%	1.47%	38	-0.01%	1.09%	794	-0.94

19:00	-0.01%	1.17%	38	0.03%	1.25%	794	-0.22
20:00	-0.09%	1.38%	38	-0.01%	1.37%	794	-0.34
21:00	0.10%	1.92%	38	0.07%	1.28%	794	0.11
22:00	-0.29%	2.65%	38	0.07%	1.48%	794	-0.82
23:00	0.11%	1.68%	38	0.01%	1.24%	794	0.36

Figure D.2: Average hourly returns on the day after the overreaction and normal days: the case of negative overreactions, LTCUSD



Table D.2: t-test of hourly returns on the day after the overreaction and nor	mal
days: the case of negative overreactions, LTCUSD	

Hour	Average return on day after negative overreaction (OD)	Standard deviation (OD)	Number of observations (OD)	Average return on usual day (UD)	Standard deviation (UD)	Number of observation (UD)	t criterion
0:00	-0.05%	1.90%	39	0.02%	1.31%	794	-0.22
1:00	-0.36%	2.08%	39	-0.04%	1.26%	794	-0.95
2:00	0.09%	1.81%	39	-0.05%	1.32%	793	0.50
3:00	0.28%	1.70%	39	-0.08%	1.09%	793	1.31
4:00	-0.35%	1.78%	39	0.03%	1.38%	793	-1.33
5:00	0.11%	2.62%	39	-0.07%	1.21%	793	0.42
6:00	-0.35%	1.97%	39	0.00%	1.21%	793	-1.09
7:00	0.06%	1.32%	39	-0.04%	1.04%	793	0.48
8:00	-0.41%	2.03%	39	-0.06%	1.18%	793	-1.05
9:00	-0.23%	2.42%	39	-0.03%	1.33%	793	-0.51
10:00	-0.32%	2.00%	39	0.03%	1.30%	793	-1.07
11:00	0.23%	2.00%	39	0.15%	1.30%	794	0.26
12:00	1.23%	4.06%	39	0.14%	1.54%	794	1.67
13:00	0.96%	4.31%	39	0.04%	1.84%	794	1.32

14:00	-0.13%	3.36%	39	-0.08%	1.55%	794	-0.09
15:00	0.46%	2.85%	39	0.11%	1.59%	794	0.76
16:00	0.87%	3.98%	39	0.10%	1.84%	794	1.20
17:00	-0.79%	1.61%	39	-0.03%	1.26%	794	-2.91
18:00	0.05%	1.63%	39	-0.01%	1.09%	794	0.20
19:00	0.60%	1.97%	39	0.03%	1.25%	794	1.80
20:00	0.22%	1.78%	39	-0.01%	1.37%	794	0.80
21:00	0.74%	1.90%	39	0.07%	1.28%	794	2.17
22:00	-0.14%	1.90%	39	0.07%	1.48%	794	-0.67
23:00	0.23%	1.13%	39	0.01%	1.24%	794	1.16

Figure D.3: Dynamics of cumulative abnormal returns on the day after the overreaction, LTCUSD



Table D.3: Cumulative abnormal returns on the day after the overreaction: the case of positive and negative overreactions, LTCUSD

	Positiv	e overreactions	Negativ	e overreactions
	Abnormal	Cumulative	Abnormal	Cumulative
Hour	returns	abnormal returns	returns	abnormal returns
0:00	0.30%	0.30%	-0.07%	0.07%
1:00	0.23%	0.54%	-0.32%	0.39%
2:00	-0.13%	0.40%	0.15%	0.24%
3:00	-0.24%	0.16%	0.36%	-0.12%
4:00	0.52%	0.68%	-0.38%	0.26%
5:00	-0.09%	0.59%	0.18%	0.08%
6:00	0.32%	0.91%	-0.35%	0.43%
7:00	0.16%	1.07%	0.10%	0.33%
8:00	-0.20%	0.87%	-0.34%	0.67%
9:00	-0.26%	0.61%	-0.20%	0.87%

<u>10:00</u>	-0.40%	0.22%	-0.35%	<u>1.21%</u>
11:00	0.45%	0.67%	0.08%	1.13%
12:00	0.11%	0.78%	1.09%	0.04%
<u>13:00</u>	<u>0.54%</u>	<u>1.32%</u>	0.92%	-0.87%
14:00	-0.39%	0.93%	-0.05%	-0.82%
15:00	0.07%	1.00%	0.35%	-1.17%
16:00	-0.22%	0.78%	0.77%	-1.94%
17:00	0.10%	0.88%	-0.76%	-1.17%
18:00	-0.23%	0.65%	0.05%	-1.23%
19:00	-0.04%	0.61%	0.57%	-1.80%
20:00	-0.08%	0.53%	0.23%	-2.03%
21:00	0.03%	0.57%	0.67%	-2.70%
22:00	-0.36%	0.21%	-0.21%	-2.49%
23:00	0.10%	0.31%	0.22%	-2.71%

### Appendix E

#### ETHUSD: day of overreaction

		Positive	Negative		<b>Positive days</b>	Negative days
Parameter	Daily data	days	days	Hourly data	(hourly data)	(hourly data)
Mean	0.0039	0.1626	-0.1355	0.0002	0.0066	-0.0054
Standard error	0.0018	0.0060	0.0056	0.0001	0.0007	0.0010
Median	0.0000	0.1484	-0.1225	0.0000	0.0040	-0.0032
Mode	0.0000	0.0000	0.0000	0.0000	0.0000	0.0000
Std. Dev.	0.0613	0.0454	0.0426	0.0139	0.0233	0.0308
Variance	0.0038	0.0021	0.0018	0.0002	0.0005	0.0010
Kurtosis	3.3418	2.2361	1.9070	21.4268	3.9199	11.7349
Skewness	0.3911	1.5380	-1.4479	0.4050	0.4461	0.3255
Minimum	-0.2703	0.1143	-0.2703	-0.2347	-0.1138	-0.2347
Maximum	0.3029	0.3029	-0.0896	0.2371	0.1263	0.2371
Sum	4.7596	9.4321	-7.7234	5.6704	7.7909	-5.0130
Observations	1213	58	57	29100	1176	936

### Table E.1: Descriptive statistics for ETHUSD data: daily, hourly, days with positive and negative returns

### Figure E.1: Average hourly returns on overreaction and normal days: the case of positive overreactions, ETHUSD



Figure E.2: Average hourly returns on overreaction and normal days: the case of negative overreactions, ETHUSD



### Table E.2: t-test of hourly returns on overreaction and normal days: the case ofpositive overreactions, ETHUSD

	Average			Average			
	return on	Standard		return on	Standard		
	positive	deviation	Number of	usual day	deviation	Number of	
	overreaction	(OD)	observations	with positive	(UD)	observation	t
Hour	day (OD)		(OD)	returns (UD)		(UD)	criterion
0:00	0.84%	2.19%	48	0.23%	1.52%	604	1.91
1:00	0.80%	2.44%	48	0.24%	1.39%	604	1.56
2:00	0.77%	2.04%	48	0.20%	1.22%	604	1.92
3:00	0.62%	2.19%	48	0.20%	1.28%	604	1.32
4:00	0.30%	2.35%	48	0.23%	1.31%	604	0.21
5:00	0.69%	2.28%	48	0.11%	1.23%	604	1.76
6:00	-0.10%	1.89%	48	0.12%	1.07%	604	-0.80
7:00	0.65%	1.87%	48	0.19%	1.07%	604	1.67
8:00	0.46%	1.41%	48	0.14%	1.17%	604	1.56
9:00	0.13%	2.40%	48	0.10%	1.21%	604	0.10
10:00	-0.13%	2.46%	48	0.04%	1.18%	604	-0.48
11:00	0.60%	2.07%	48	0.24%	1.25%	604	1.18
12:00	1.37%	2.75%	48	0.26%	1.45%	604	2.78
13:00	0.86%	2.76%	48	0.12%	1.44%	604	1.84
14:00	0.60%	3.02%	48	0.15%	1.58%	604	1.02
15:00	0.90%	2.08%	48	0.26%	1.55%	604	2.09
16:00	0.36%	2.04%	48	0.24%	1.47%	604	0.40
17:00	0.46%	2.38%	48	0.10%	1.39%	604	1.01
18:00	0.98%	1.97%	48	0.20%	1.21%	604	2.71

19:00	0.79%	2.10%	48	0.25%	1.27%	604	1.76
20:00	0.97%	2.43%	48	0.13%	1.33%	604	2.37
21:00	1.29%	2.34%	48	0.28%	1.30%	604	2.95
22:00	0.27%	2.98%	48	0.17%	1.26%	604	0.22
23:00	1.41%	2.60%	48	0.35%	1.35%	604	2.79

Table E.3:t-testof hourly returns on ov	verreaction a	and 1	normal	days:	the	case (	of
negative overreactions, ETHUSD							

	Average			Average			
	return on	Standard		return on	Standard		
	negative	deviation	Number of	usual day	deviation	Number of	
	overreaction	(OD)	observations	with negative	(UD)	observation	t
Hour	day (OD)		(OD)	returns (UD)		(UD)	criterion
<u>0:00</u>	-1.02%	<u>2.49%</u>	<u>38</u>	<u>-0.19%</u>	<u>1.35%</u>	<u>602</u>	-2.05
1:00	-0.81%	2.87%	38	-0.23%	1.30%	602	-1.25
2:00	-0.07%	2.94%	38	-0.09%	1.40%	602	0.06
3:00	-0.73%	2.25%	38	-0.20%	1.16%	602	-1.43
4:00	-0.26%	2.16%	38	-0.12%	1.12%	602	-0.41
<u>5:00</u>	<u>-0.78%</u>	2.22%	<u>38</u>	<u>-0.13%</u>	<u>1.09%</u>	<u>602</u>	-1.79
6:00	-0.02%	2.27%	38	-0.15%	1.13%	602	0.34
<u>7:00</u>	<u>-1.59%</u>	<u>4.05%</u>	<u>38</u>	<u>-0.23%</u>	<u>1.48%</u>	<u>602</u>	-2.05
8:00	-0.37%	4.92%	38	-0.22%	1.64%	602	-0.18
9:00	-0.35%	2.77%	38	-0.24%	1.49%	602	-0.26
10:00	-0.65%	1.99%	38	-0.20%	1.17%	602	-1.38
11:00	-0.54%	2.96%	38	-0.11%	1.37%	602	-0.89
12:00	-0.59%	2.01%	38	-0.07%	1.27%	602	-1.55
13:00	-1.07%	3.94%	38	-0.15%	1.68%	603	-1.42
14:00	-0.38%	2.59%	38	-0.17%	1.66%	603	-0.50
15:00	-0.26%	4.26%	38	-0.17%	1.74%	603	-0.14
16:00	-0.55%	2.56%	38	-0.10%	1.72%	603	-1.06
<u>17:00</u>	<u>-1.00%</u>	2.65%	<u>38</u>	-0.22%	<u>1.39%</u>	<u>603</u>	<u>-1.79</u>
18:00	-0.50%	3.27%	38	-0.18%	1.38%	603	-0.61
19:00	-0.48%	2.88%	38	-0.12%	1.42%	603	-0.76
20:00	0.00%	3.84%	38	-0.12%	1.58%	603	0.19
21:00	-0.42%	2.47%	38	-0.03%	1.28%	603	-0.98
22:00	-0.85%	3.76%	38	-0.09%	1.51%	603	-1.24
23:00	0.44%	3.73%	38	-0.07%	1.54%	603	0.83

Figure E.3: Dynamics of cumulative abnormal returns, ETHUSD



Table E.4: Cumulative abnormal returns: the case of positive and negative overreactions, ETHUSD

	Pos	sitive overread	tions	Negative overreactions			
	Abnormal	CAP	Overreaction	Abnormal	CAP	Overreaction	
Hour	returns	CAK	cross	returns	CAK	cross	
0:00	0.62%	0.62%	5.68%	-0.84%	-0.84%	-7.77%	
1:00	0.56%	1.17%	4.88%	-0.59%	-1.42%	-6.96%	
2:00	0.57%	1.75%	4.11%	0.03%	-1.39%	-6.90%	
3:00	0.42%	2.17%	3.49%	-0.53%	-1.92%	-6.17%	
4:00	0.07%	2.24%	3.19%	-0.14%	-2.07%	-5.90%	
5:00	0.59%	2.83%	2.50%	-0.65%	-2.71%	-5.12%	
6:00	-0.22%	2.61%	2.60%	0.13%	-2.59%	-5.10%	
7:00	0.46%	3.06%	1.95%	-1.35%	-3.94%	-3.51%	
8:00	0.33%	3.39%	1.49%	-0.14%	-4.08%	-3.14%	
9:00	0.03%	3.42%	1.35%	-0.12%	-4.20%	-2.79%	
10:00	-0.17%	3.25%	1.49%	-0.45%	-4.65%	-2.15%	
11:00	0.36%	3.61%	0.89%	-0.43%	-5.08%	-1.60%	
<u>12:00</u>	<u>1.12%</u>	<u>4.73%</u>	<u>-0.48%</u>	-0.51%	-5.59%	-1.02%	
<u>13:00</u>	0.74%	5.47%	-1.34%	<u>-0.91%</u>	<u>-6.50%</u>	<u>0.05%</u>	
14:00	0.45%	5.92%	-1.94%	-0.21%	-6.71%	0.43%	
15:00	0.64%	6.56%	-2.85%	-0.09%	-6.81%	0.69%	
16:00	0.12%	6.68%	-3.21%	-0.45%	-7.25%	1.24%	
17:00	0.35%	7.03%	-3.67%	-0.78%	-8.03%	2.24%	
18:00	0.78%	7.81%	-4.65%	-0.33%	-8.36%	2.74%	
19:00	0.54%	8.36%	-5.44%	-0.36%	-8.72%	3.21%	
20:00	0.84%	9.20%	-6.41%	0.12%	-8.60%	3.22%	
21:00	1.01%	10.21%	-7.70%	-0.40%	-8.99%	3.64%	
22:00	0.10%	10.31%	-7.97%	-0.76%	-9.76%	4.49%	
23:00	1.06%	11.37%	-9.38%	0.50%	-9.25%	4.06%	

### Appendix F ETHUSD: day after the overreaction

## Figure F.1: Average hourly returns on the day after the overreaction and normal days: the case of positive overreactions, ETHUSD



### Table F.1: t-test of hourly returns on the day after the overreaction and normaldays: the case of positive overreactions, ETHUSD

Hour	Average return on day after positive overreaction (OD)	Standard deviation (OD)	Number of observations (OD)	Average return on usual day (UD)	Standard deviation (UD)	Number of observation (UD)	t criterion
0:00	-0.09%	2.76%	58	0.02%	1.47%	1213	-0.32
1:00	0.14%	1.94%	58	0.01%	1.36%	1212	0.52
<u>2:00</u>	<u>0.69%</u>	<u>2.54%</u>	<u>58</u>	<u>0.04%</u>	<u>1.33%</u>	1212	<u>1.92</u>
3:00	0.18%	1.93%	58	0.00%	1.24%	1212	0.69
4:00	-0.05%	1.93%	58	0.05%	1.23%	1212	-0.41
5:00	0.27%	1.85%	58	-0.01%	1.16%	1212	1.14
6:00	0.01%	1.63%	58	-0.02%	1.11%	1212	0.09
7:00	0.31%	1.77%	58	-0.02%	1.31%	1212	1.40
8:00	-0.67%	2.98%	58	-0.04%	1.44%	1212	-1.59
9:00	-0.50%	1.93%	58	-0.07%	1.36%	1212	-1.66
10:00	0.00%	1.91%	58	-0.08%	1.18%	1212	0.33
<u>11:00</u>	<u>0.50%</u>	<u>1.79%</u>	<u>58</u>	<u>0.07%</u>	<u>1.32%</u>	<u>1212</u>	<u>1.82</u>
12:00	0.37%	2.19%	58	0.09%	1.37%	1212	0.96
13:00	0.04%	1.96%	58	-0.02%	1.57%	1213	0.22
14:00	0.29%	2.88%	58	-0.01%	1.62%	1213	0.79
15:00	0.31%	2.86%	58	0.05%	1.66%	1213	0.69
16:00	-0.64%	2.76%	58	0.07%	1.61%	1213	-1.95
17:00	0.08%	2.27%	58	-0.06%	1.40%	1213	0.46
18:00	0.04%	2.21%	58	0.01%	1.31%	1213	0.10
19:00	-0.22%	2.64%	58	0.06%	1.36%	1213	-0.82

20:00	<u>0.74%</u>	2.45%	<u>58</u>	<u>0.01%</u>	<u>1.47%</u>	1213	2.24
21:00	0.62%	2.43%	58	0.13%	1.32%	1213	1.52
22:00	-0.55%	2.64%	58	0.04%	1.40%	1213	-1.69
23:00	0.19%	2.89%	58	0.14%	1.46%	1213	0.14

Figure F.2: Average hourly returns on the day after the overreaction and normal days: the case of negative overreactions, ETHUSD



Table F.2: t-test of hourly returns on the day after the overrea	ction and	normal
days: the case of negative overreactions, ETHUSD		

Hour	Average return on day after negative overreaction (OD)	Standard deviation (OD)	Number of observations (OD)	Average return on usual day (UD)	Standard deviation (UD)	Number of observation (UD)	t criterion
0:00	-0.10%	2.57%	57	0.02%	1.47%	1213	-0.36
1:00	0.00%	3.14%	57	0.01%	1.36%	1212	-0.02
2:00	0.38%	2.85%	57	0.04%	1.33%	1212	0.87
3:00	0.21%	2.22%	57	0.00%	1.24%	1212	0.69
4:00	0.29%	2.47%	57	0.05%	1.23%	1212	0.71
5:00	-0.43%	2.41%	57	-0.01%	1.16%	1212	-1.31
6:00	-0.50%	2.12%	57	-0.02%	1.11%	1212	-1.70
7:00	0.11%	3.53%	57	-0.02%	1.31%	1212	0.29
8:00	0.43%	3.68%	57	-0.04%	1.44%	1212	0.96
9:00	-0.09%	2.51%	57	-0.07%	1.36%	1212	-0.06
10:00	-0.47%	1.80%	57	-0.08%	1.18%	1212	-1.61
11:00	0.04%	1.88%	57	0.07%	1.32%	1212	-0.11
12:00	0.73%	2.70%	57	0.09%	1.37%	1212	1.80

13:00	0.23%	3.13%	57	-0.02%	1.57%	1213	0.60
14:00	0.11%	3.43%	57	-0.01%	1.62%	1213	0.27
15:00	0.48%	2.91%	57	0.05%	1.66%	1213	1.11
16:00	0.44%	2.92%	57	0.07%	1.61%	1213	0.94
17:00	-0.19%	2.17%	57	-0.06%	1.40%	1213	-0.46
<u>18:00</u>	<u>0.49%</u>	<u>1.88%</u>	<u>57</u>	<u>0.01%</u>	<u>1.31%</u>	<u>1213</u>	<u>1.91</u>
19:00	0.78%	3.12%	57	0.06%	1.36%	1213	1.72
20:00	0.40%	2.92%	57	0.01%	1.47%	1213	1.01
<u>21:00</u>	<u>0.61%</u>	<u>1.59%</u>	<u>57</u>	<u>0.13%</u>	1.32%	<u>1213</u>	2.22
22:00	0.01%	2.05%	57	0.04%	1.40%	1213	-0.09
23:00	0.39%	1.55%	57	0.14%	1.46%	1213	1.19

### Figure F.3: Dynamics of cumulative abnormal returns, ETHUSD



Table F.3: Cumulative abnormal returns: the case of positive and negative overreactions, ETHUSD

	Positive	e overreactions	Negative overreactions		
	Abnormal	Cumulative	Abnormal	Cumulative	
Hour	returns	abnormal returns	returns	abnormal returns	
0:00	-0.12%	-0.12%	-0.12%	0.12%	
1:00	0.14%	0.02%	-0.01%	0.13%	
2:00	0.64%	0.66%	0.33%	-0.20%	
3:00	0.18%	0.84%	0.21%	-0.41%	
4:00	-0.11%	0.73%	0.23%	-0.64%	
5:00	0.28%	1.01%	-0.42%	-0.22%	
6:00	0.02%	1.03%	-0.48%	0.26%	
7:00	0.33%	1.36%	0.13%	0.13%	
8:00	-0.62%	0.74%	0.47%	-0.34%	

9:00	-0.43%	0.31%	-0.02%	-0.32%
10:00	0.08%	0.40%	-0.39%	0.07%
11:00	0.43%	0.83%	-0.03%	0.10%
12:00	0.28%	1.11%	0.65%	-0.55%
13:00	0.06%	1.17%	0.25%	-0.80%
14:00	0.30%	1.47%	0.12%	-0.92%
15:00	0.26%	1.73%	0.43%	-1.35%
16:00	-0.71%	1.02%	0.36%	-1.72%
17:00	0.14%	1.16%	-0.13%	-1.59%
18:00	0.03%	1.19%	0.48%	-2.07%
19:00	-0.29%	0.90%	0.72%	-2.78%
20:00	0.73%	1.63%	0.39%	-3.17%
<u>21:00</u>	<u>0.49%</u>	2.12%	0.48%	-3.65%
22:00	-0.59%	1.53%	-0.03%	-3.62%
23:00*	0.05%	1.58%	0.25%	-3.87%

\* contrarian effect detected