



# Bitcoin and News Around the World in Twenty-Six Languages

Lucia Alessi, Eric Ghysels, Marco Petracco, Zhe Wang

[marco.Petracco@ec.Europa.eu](mailto:marco.Petracco@ec.Europa.eu)

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# Problem definition

- Cryptocurrencies are famous for their high volatility
- Is volatility driven by fundamentals or by other factors
- Price discovery “noisy” for cryptos wrt other asset classes
- → **What is the role of sentiment in news for cryptos pricing**

## The data

- Time span: April 16, 2014 to August 31, 2020
- High frequency data on BTC quotes from Kaiko
- Sentiment scoring of news extracted from a large variety of sources and languages, from EC's Europe Media Monitor

## The data: Price data (1)

- 6 exchanges: three in Asia (OkCoin in China, Bitfinex in China (Hong Kong), and Quoine in Japan), one in Europe (Bitstamp in Luxembourg), and two in US (Coinbase and Kraken).
- We choose these exchanges because they have long trading history to match with our EMM sample period, and high enough trading volume to alleviate liquidity issues.
- **Returns**
- **Netbuy**

# The data: Europe Media Monitor (EMM) (1)

- EMM newsbrief is an automatic system that collects and analyses news media.
- Monitors a selected list of sources for news items, in up to 72 languages.
- Generates structured (meta) data (entities, events, categories ...)
- Includes three methods for sentiment, emotion, tonality
- “JRC tonality” algorithm: +/- 1 for slightly positive/negative terms, +/- 4 for strongly positive/negative terms. Normalized by word count.

## The data: Europe Media Monitor (EMM) (2)

- Articles belonging to “Bitcoin” category (inclusion based on appearance of keywords and criteria determined by subject matter experts).
- List of sources include most major online newspapers, and a set of specialised news outlets
- 403,112 articles in 26 languages (184,354 of which in English).
- **Time-stamp** (retrieval), **language**, “JRC tonality”
- Aggregate in 1-hour intervals, for English and all other languages  
Retain all intervals with articles both in English and other languages
- About 30 thousand 1-hr time intervals (67% of available)



# Methods:

- Divide every day in 24 1-hour time intervals  $t$ . ( $t = 0 : 23$ )
- Calculate tonality over period  $(t-1\text{hr}:t)$
- Calculate returns and netbuy of periods  $(t:t+k)$ , where  $k = 30', 60', 1\text{d}, 7\text{d}$

$t - 60'$        $t$        $t + 30'$



$t + 60'$



$t + 1\text{d}$



$t + 7\text{d}$



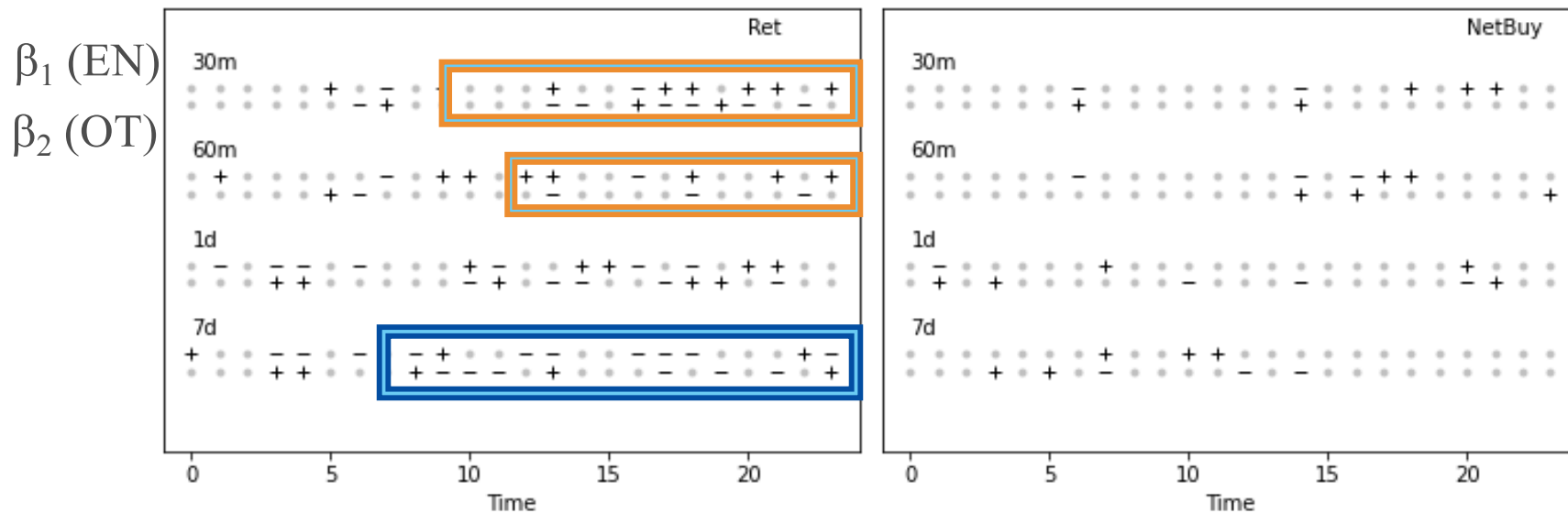
## Methods:

$$\text{Sign}(\text{Ret}_{t+k}) = \text{Logistic}(\alpha + \beta_1 \text{TonalityEN}_t + \beta_2 \text{TonalityOT}_t + \gamma \text{Controls}_t + \epsilon_{t+1})$$

$$\text{NetBuy}_{t+k} = \alpha + \beta_1 \text{TonalityEN}_t + \beta_2 \text{TonalityOT}_t + \gamma \text{Controls}_t + \epsilon_{t+1}$$

- $\text{Ret}(t+k)$  and  $\text{NetBuy}(t+k)$  with  $k = 30$  mins, 60 mins, 1 day, and 7 days, starting from the end of the time interval  $t$ . 24, 1-hour time intervals.
- $\text{TonalityEN}(t)$  is the average tonality of all English news over time period  $t$ .  $\text{TonalityOT}(t)$  is the average of tonality of news published in other languages over the same time interval.
- Controls include a macroeconomic sentiment index constructed by the Federal Reserve Bank of San Francisco.
- For each time block, dummies for “excess difference” between EN and OT tonality.
- Also estimated with fixed effects by Exchange, and with different slopes by exchange
- Mainly interested in the **SIGN** of significant coefficients for news sentiment in EN and OT

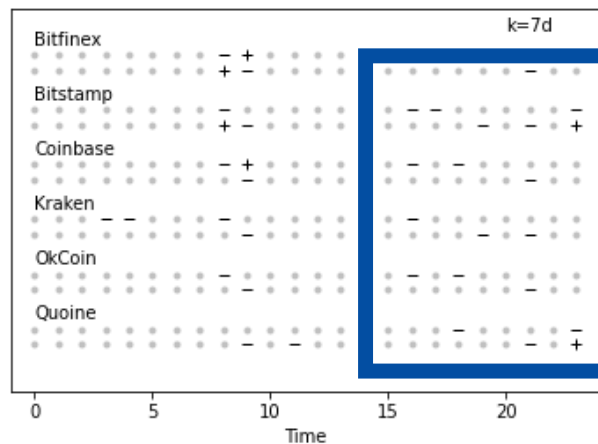
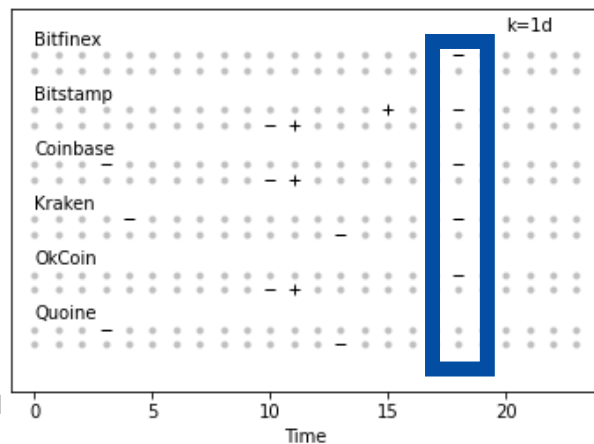
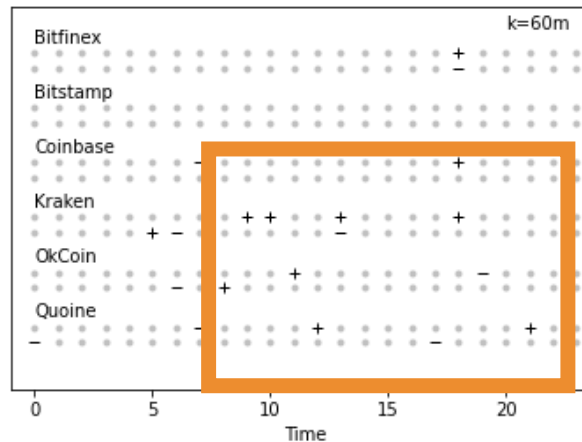
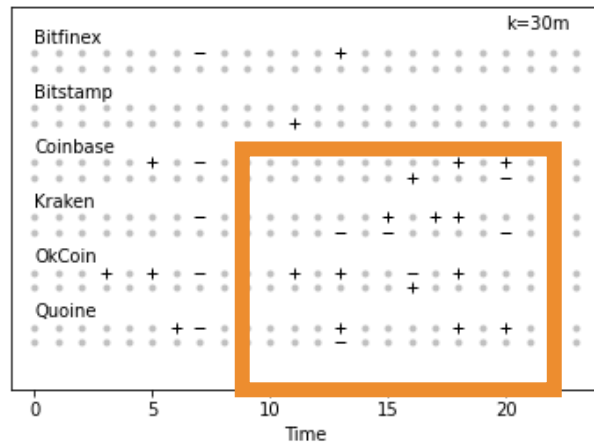
## Results: returns and netbuy (pooled and FE)



- 24 time slots; +, significant, positive; -, significant, negative; top line (EN), bottom line (OT)
- On average, returns positively correlated with news tonality at short time intervals, and negatively correlated at the longest time interval

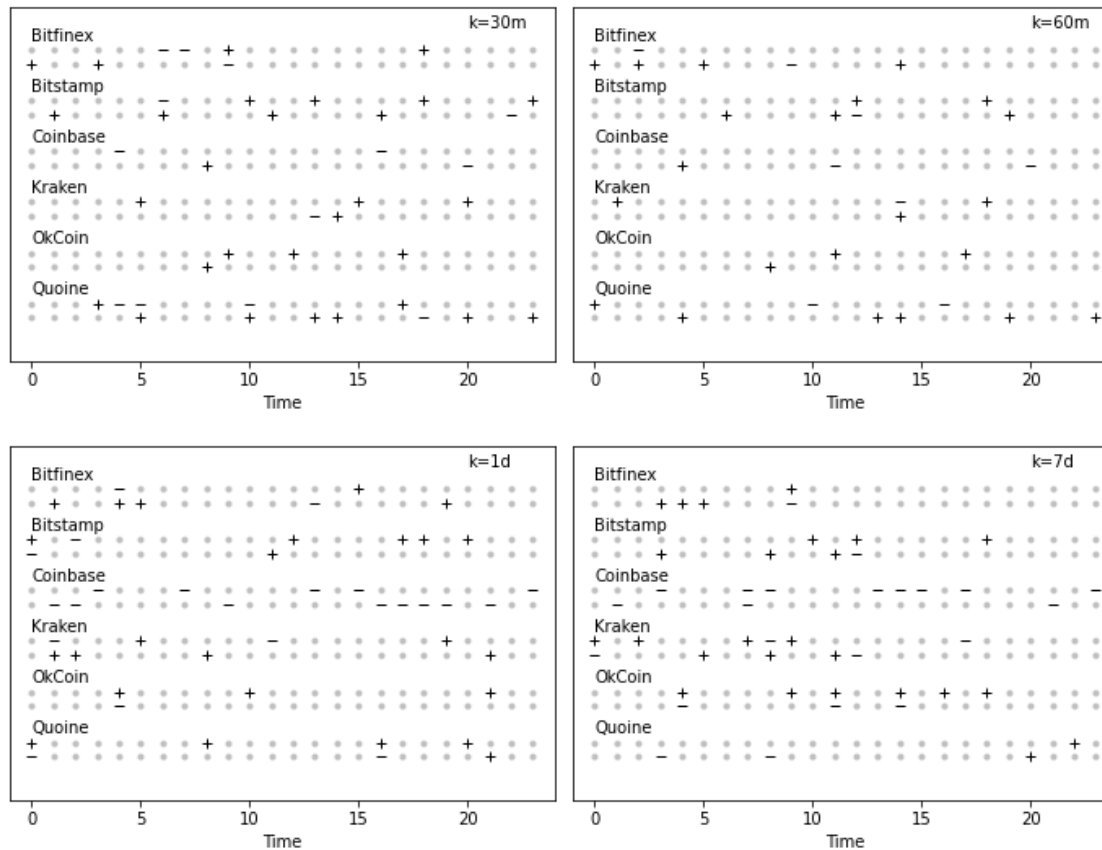
- 11 • Netbuy volume more muted relation.

# Results: returns (by exchange)



- On average, returns positively correlated with news tonality at short time intervals, and negatively correlated at the longest time interval
- For bitfinex and Bitstamp short term relationship disappears

# Results: netbuy (by exchange)



- More significant coefficients when exchange specific intercepts are allowed
- Some exchanges seem to have a prevalence of one sign ... artefact or ...

## Conclusions:

- Positive correlation between tonality of news articles in English and return on BTC over the next 30' and 60'
- agree with previous research pointing to a role for sentiment in determining returns in crypto-currencies, and with the fact that news sentiment could affect short term and intraday returns
- evidence of a reversal of these effects over longer time horizons, especially 7 days . This is suggestive of the fact that these dynamics could be tied to extensive noise trading on BTC markets.
- further supported by the analysis of results at individual exchange level, where the effects seems to disappear on Bitfinex, which is reportedly a more efficient BTC exchange.
- NetBuy dynamics more muted at aggregate level, but different exchanges seem to respond more to news.

## Further research:

- Relationships seems to be inverted for non-English language news. This result could point to a different use of news by noise traders in different countries, or to the existence of dynamics in news tonalities across different languages and will need to be investigated further.
- Netbuy dynamics might be compatible with interest and noise driven dynamics, and/or with different roles for different exchanges, or BTC trading for other purposes
- Subject of further investigation to understand if more extreme cases of market “one sidedness” could be the object of prediction.
- Introduce social media sentiment (e.g. Reddit, Twitter) ...

# Thank you



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# The data: Europe Media Monitor (EMM) (3)

**Table 1: Summary Statistics of Tonality**

Hourly Block	TonalityEN	TonalityOT	Hourly Block	TonalityEN	TonalityOT
0	-2.638	-0.119	12	-1.186	0.357
1	-3.408	0.364	13	-1.182	0.385
2	-3.259	0.022	14	-1.592	0.337
3	-3.226	0.218	15	-1.945	0.251
4	-2.596	0.232	16	-1.661	0.257
5	-2.023	0.511	17	-2.152	0.294
6	-2.246	0.495	18	-2.285	0.026
7	-2.274	0.432	19	-2.134	-0.104
8	-1.839	0.445	20	-2.563	0.275
9	-1.626	0.458	21	-2.451	-0.061
10	-1.797	0.476	22	-2.868	0.033
11	-1.860	0.866	23	-3.029	0.139
All				-2.164	0.292

Notes: Entries to the table are the average tonality of all hourly time blocks.

# The data: Price data (2)

**Table 2: Summary Statistics of Return and NetBuy**

	Ret							
	30 mins		60 mins		1 day		7 days	
	N	Mean	N	Mean	N	Mean	N	Mean
Bitfinex	30148	2.68E-05	30146	7.84E-05	30098	1.75E-03	29930	0.016
Bitstamp	30103	2.29E-05	30100	8.15E-05	30036	1.85E-03	29802	0.017
Coinbase	28479	9.80E-06	28463	6.81E-05	28360	2.22E-03	28237	0.018
Kraken	29439	3.12E-05	29369	5.56E-05	28989	2.02E-03	28795	0.017
OkCoin	26106	7.20E-05	26091	6.93E-05	25943	1.52E-03	25621	0.014
Quoine	27257	-8.95E-05	27167	6.27E-05	26681	2.39E-03	26519	0.019

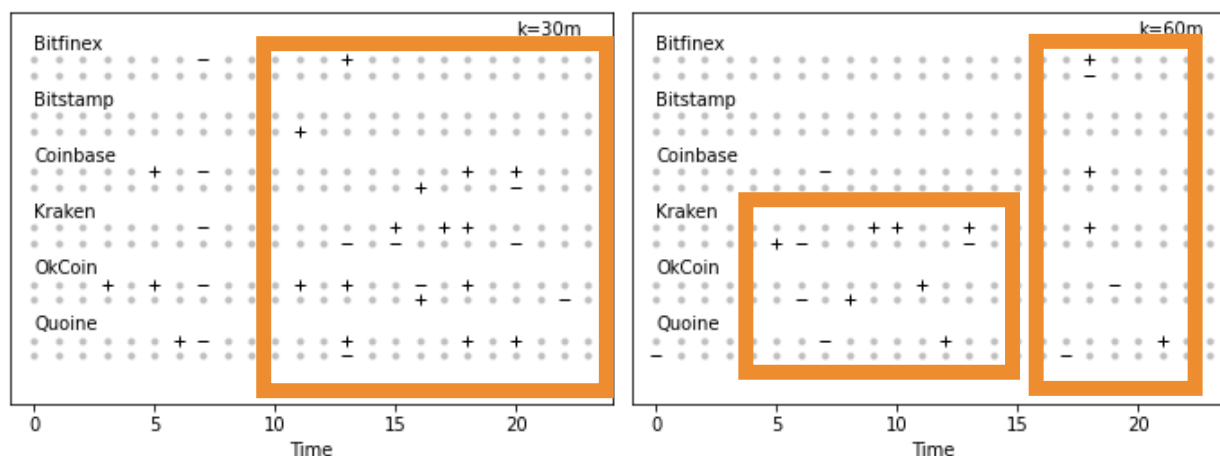
  

	NetBuy							
	30 mins		60 mins		1 day		7 days	
	N	Mean	N	Mean	N	Mean	N	Mean
Bitfinex	29934	-0.004	30061	-0.009	30150	-0.020	30140	-0.023
Bitstamp	25893	0.074	25897	0.073	25976	0.054	26090	0.048
Coinbase	28363	0.084	28388	0.078	28638	0.068	28604	0.063
Kraken	27407	0.009	28041	0.003	30143	-0.011	30153	-0.017
OkCoin	25120	-0.055	25615	-0.056	26240	-0.068	26439	-0.061
Quoine	24189	-0.028	25379	-0.034	28347	-0.019	28977	-0.019

Notes: Entries are the statistics of  $Ret_{t+k}$  and  $NetBuy_{t+k}$  for  $k=30$  mins, 60 mins, 1 day and 7 days.

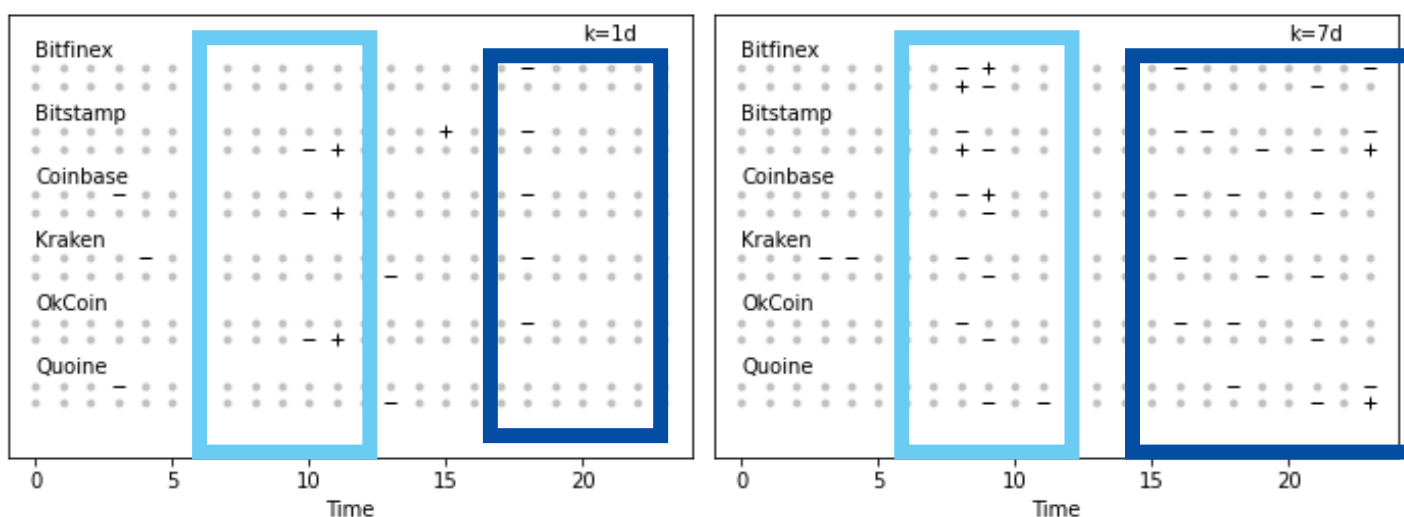
# Results: returns (by exchange – 30' and 60')

**Table 5: Tonality Predicts Homogeneous Future Returns at Different Exchanges.**



- On average, returns positively correlated with news tonality at short time intervals, and negatively correlated at the longest time interval
- Netbuy volume more muted relation.

## Results: returns (by exchange – 1d and 7d)



- On average, returns positively correlated with news tonality at short time intervals, and negatively correlated at the longest time interval
- Netbuy volume more muted relation.