## Time Variation of Regression Coefficients related to Macroeconomic News affecting Currency Prices

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### Contents

| L | Introduction   | 1 |
|---|--|---|
| 2 | Data   | 2 |
| 3 | Construction of the model                            | 2 |
| 1 | Testing for instability of the news impact parameter | 3 |
| 5 | Parameter Path Estimations                           | 7 |
| 3 | Conclusion   | 7 |
| 7 | Appendix7.1 Parameter Paths                          | 7 |
|   |  |   |

#### abstract here when I'm done with the writing and have all results

### 1 Introduction

There exists a number of macroeconomic figures that are released on a predetermined schedule for certain countries. These include for example the Non-Farm Employment change that is released on the first friday of every month informing economists and investors alike of the status of employment in the United States. Classic economic theory helps us understand that an increase of interest rates is warranted when economies are performing well and prices are generally increasing. Those who decide to increase national interest rates, the central banks, typically refer to measures of inflation in order to make their decisions. Because of this, investors and traders alike pay close attention to news releases (such as inflation, and also the Non-Farm-Payrolls in the case of the United States) and react according to the results. These news releases are not made public until specific times on specific days and since investors and traders react to the same news the moment it is released, the result is often a violent reaction of price in one direction or another. The common discourse is that the direction and the magnitude of the change of price depends on the difference between the expectation of the market (combined expectation of worldwide investors) and the result of the news release.

In this paper, we decide to use currency pairs to measure the price shocks. As certain news pertaining to a particular country affects its respective currency more than other ones, it makes sense to observe the currency most relevant to the news announcement. As currency prices are typically measured in pairs, the second chosen currency will be another major currency that is known for its high liquidity (EUR, USD or CHF) and does not have any other news announcements at the same time<sup>1</sup>. As an example, we would use the

 $<sup>^{1}</sup>$ When simultaneous news cannot be avoided, a sequence of stability tests will be applied to ensure time variation is identified on a specific news release

| Table 1: Summary | of the news | figures | considered | in | the study |
|------------------|-------------|---------|------------|----|-----------|
|                  |             |         |            |    |           |
|                  |             |         |            |    |           |

| Country        | News.Event                     | Pair.used | TimeGMT. | Frequency | Observations | Dates                |
|----------------|--------------------------------|-----------|----------|-----------|--------------|----------------------|
| Single News    |                                |           |          |           | NA           |                      |
| Canada         | Consumer Price Index           | USD/CAD   | 13:30    | Monthly   | 135          | Jan 2008 to Dec 2019 |
| Canada         | Core Retail Sales              | USD/CAD   | 12:30    | Monthly   | 144          | Oct 2008 to Dec 2019 |
| United States  | Consumer Price Index           | USD/CHF   | 13:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| New Zealand    | Consumer Price Index           | NZD/USD   | 21:45    | Quarterly | 43           | Jan 2009 to Oct 2019 |
| Australia      | Consumer Price Index           | AUD/USD   | 00:30    | Quarterly | 48           | Jan 2008 to Oct 2019 |
| Australia      | Retail Sales                   | AUD/USD   | 00:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| United Kingdom | Consumer Price Index           | GBP/USD   | 09:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| United Kingdom | Retail Sales                   | GBP/USD   | 09:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| Grouped News   |                                | ·         |          |           | NA           |                      |
| United States  | Average Hourly Earnings Change | USD/CHF   | 13:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| United States  | NonFarm Employment Change      | USD/CHF   | 13:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| United States  | Unemployment Rate              | USD/CHF   | 13:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| Canada         | Employment Change              | USD/CAD   | 13:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| Canada         | Unemployment Rate              | USD/CAD   | 13:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| Australia      | Employment Change              | AUD/USD   | 00:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |
| Australia      | Unemployment Rate              | AUD/USD   | 00:30    | Monthly   | 135          | Oct 2008 to Dec 2019 |

USD/CAD currency pair to measure the effect of the Canadian Consumer Price Index (CPI).

The paper of (Board of Governors of the Federal Reserve System et al. 2003) reveals that over the time period between 1987 and 2002 there has been little time-variation in the reaction to news. Some more recently published literature of (Ben Omrane et al. 2019) involving an analysis on euro-dollar contracts has determined that unlike the decade(s) encompassing the "Great Moderation" where there was lower relative volatility in the financial markets, the time period between 2004 and 2014 is characterized by evolving reactions to macroeconomic news.

This paper aims to

### 2 Data

The minute-by-minute OHLC Data of 7 currency pairs were collected from the Metatrader5 platform. This represents over 4 million data points for each pair. Only a small fraction of this data is actually used since we consider only the 5 time frame from when each piece of monthly or quarterly news is released until 5 minutSes afterwards.

### 3 Construction of the model

Being consistent with previous literature on the subject, the first step involves estimating the impact that each piece of news has on its respective currency assuming 1.) That the news effects are constant over time. 2.) The surprise element  $S_t$  of the regression is evaluated as:

$$S_t = \frac{A_t - E_t}{\sigma_d} \tag{1}$$

 $A_t$  is the actual result of the news at time t,  $E_t$  is the expected result aggregated from experts and  $\sigma_d$  is the empirical standard deviation of this difference over the entire sample. We use the expectation numbers from the ForexFactory website<sup>2</sup> Thereafter, we use this surprise element in a first simple OLS model.

$$R_t = \beta_0 + \beta_1 S_t + \varepsilon_t \tag{2}$$

<sup>&</sup>lt;sup>2</sup>The expectations of most online sources such as "Investing.com" or "DailyFX" are very similar. The aggregation methods are not disclosed to the public.

| Table 2: | Results of | of regressions | tests - | HAC standard errors |
|----------|------------|----------------|---------|---------------------|
|          |            |                |         |                     |

| News.Event | M5.Coefficient | std.error | HAC.std.error | qLL        |
|------------|----------------|-----------|---------------|------------|
| UK CPI     | 12.681***      | 1.729     | 2.601         | -12.964*** |
| CA CPI     | -12.896***     | 2.254     | 2.673         | -18.387*** |
| CA CRS     | -13.773***     | 2.112     | 2.871         | -14.445*** |
| US CPI     | 3.968**        | 1.207     | 1.230         | -21.582*** |
| NZ CPI     | 24.109***      | 2.910     | 4.727         | -7.267*    |
| AU CPI     | 22.293***      | 4.145     | 4.226         | -9.022**   |
| AU RET     | 9.647***       | 1.215     | 2.656         | -17.623*** |
| UK RET     | 16.574***      | 1.968     | 2.572         | -10.387**  |

Moreover, because we are working with a dataset where subsequent observations are suspected to be related to one another, one could expect that the errors of the basic model above be autocorrelated. Specifically, adjacent  $R_t$  would be more similar to one another than reactions that are separated in time to a greater extent. In this case, the inference on the  $\beta_1$  would be flawed. Previous researchers have used what is called Heteroskedasticity and Autocorrelation-Consistent (HAC) estimators for the variance of the OLS estimator  $\beta_1$ . Using the Newey-West estimator for this variance from (Newey and West 1987), we use modified standard errors of the  $\beta_1$  in our results. If we are wrong in our assumption in some of the news instances, and there is no underlying autocorrelation of the  $R_t$  observations, our estimation of  $\beta_1$  is less efficient than the original estimator in those cases. Nonetheless, it remains consistent and ensures we avoid type 1 error of rejecting a true null hypothesis suggesting  $\beta_1 = 0$ .

Table 2 shows the result of the different  $\beta_1$  coefficients for separate news reports. The construction of the truncation parameter in the Newey-West estimator is such that our monthly news reports consider 2 autocorrelation coefficients whereas the quarterly ones only contain 1. This is due to the difference in the number of observations in our data. A higher estimated autocorrelation between the errors of the regression will result in a stronger correction of the variance of  $\beta_1$ . In all of the news in Table 2, the higher standard error does not affect the significance of the term. While it is not formal evidence, we suspect that the news for which the HAC standard error is vastly different than its unedited counterpart contains many unknown regressors that come from any of the findings that are summarized in (Goldberg and Grisse 2013) (mentioned in the introduction).

$$R_t = \beta_0 + \beta_{1,t} S_t + \varepsilon_t \tag{3}$$

## 4 Testing for instability of the news impact parameter

There exists many ways to test whether  $\beta_t$  is time dependent or not. We choose the methodology of the authors of (Elliott and Müller 2006) and briefly replicate their method. The advantage of their test is that it identifies instability no matter whether it comes in the form of a single break, many breaks, or a continuous change (all of which are feasible in our context).

The quasi-Local-Level (qLL) test enables one to test for many different types of persistent processes of the  $\beta_t$ . It is explained that many of these breaking processes can have a "temporary memory" (strictly speaking are strongly mixing) but will be well approximated by a Wiener process.<sup>3</sup>. This is extremely practical in our scenario as there are many possibilities for the possible variation of the  $\beta_t$ . The Null Hypothesis implies there is a stable parameter as in a familiar OLS regression. We obtain the likelihood under the Null assuming that the  $R_t$  observations are independently and identically distributed (and therefore so are their first differences):

<sup>&</sup>lt;sup>3</sup>Theorem 7.30 of (White 2001) can be applied since certain assumptions are made about the process

Table 3: Asymptotic critical values of the qLL statistic

| k   | 1      | 2      | 3      | 4      | 5      |
|-----|--------|--------|--------|--------|--------|
| 1%  | -11.05 | -17.57 | -23.42 | -29.18 | -35.09 |
| 5%  | -8.36  | -14.32 | -19.84 | -25.28 | -30.60 |
| 10% | -7.14  | -12.80 | -18.07 | -23.37 | -28.55 |

$$L_{H0}(\beta_0, \beta_1, \sigma^2) = \log \prod_{t=1}^{T} p(\Delta R_t | S_t; \beta_0, \beta_1, \sigma^2)$$
(4)

$$= -\frac{T}{2}log(2\pi) - Tlog(\sigma) - \frac{1}{2\sigma^2} \sum_{t=1}^{T} (\Delta R_t - (\Delta \beta_0 + \Delta \beta_1 S_t))^2$$
 (5)

Only the last term of (5) is kept as the first constants will cancel out.  $\Delta \beta_0 + \Delta \beta_1 S_t$  becomes 0 as the terms do not change with time.

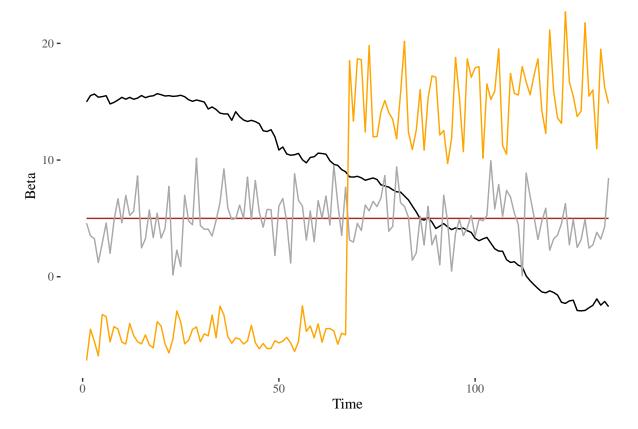
$$L_{H0} = \frac{1}{2\sigma^2} \sum_{t=1}^{T} (\Delta R_t)^2 \tag{6}$$

This contrasts with the alternative where instability is implied. We assume  $\beta_t - \beta_0$  is approximated by the Gaussian random walk and  $\Delta R_t$  is therefore a Gaussian moving average of order 1 MA(1) with the specification:  $\Delta R_t \sim \eta_t + \psi_\eta \eta_{t-1}$ ,  $\eta_t \sim iidN(0, \sigma_\eta^2)$ , constant  $\psi_\eta < 1$ . Using the same *iid* assumption we obtain the likelihood of this alternative process:

$$L_{HA} = \frac{1}{2\sigma^2} \sum_{t=1}^{T} (\Delta \eta)^2 \tag{7}$$

The qLL statistic is obtained by subtracting  $L_{HA}$  from  $L_{H0}$ . The qLL test is therefore a monotone transformation of the Likelihood Ratio Test  $(LR_T)$ , so while it does not follow a chi-square distribution exactly, it does follow a certain related distribution that has its percentiles defined by (Elliott and Müller 2006) and reported in their table, reproduced here as Table 3. The general extension to the  $LR_T$  can be made where we can reject the model related to the Null Hypothesis (the stable model) if the critical value is sufficiently negative.

In order to see this in practice and ensure it works in our case, we create some hypothetical paths and observe the result of the tests.



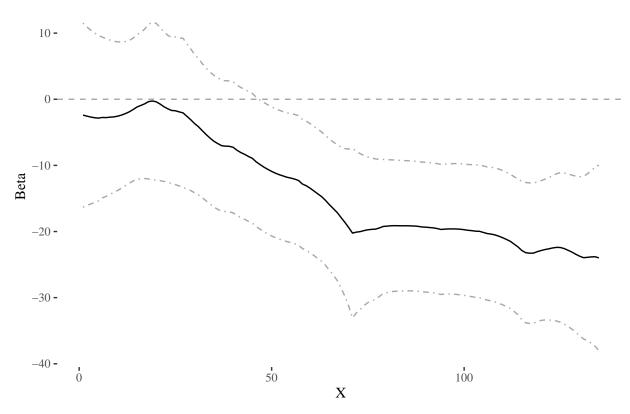
## • Assumptions of the test - appropriate in our case???

| News.Events                       | Pair    | TimeGMT. | Frequency | M5.Coefficient | std.error | obs | Dates                | qLL          |
|-----------------------------------|---------|----------|-----------|----------------|-----------|-----|----------------------|--------------|
| US Batch                          | USD/JPY | 13:30    | Monthly   |                | NA        | 135 | Oct 2008 to Dec 2019 |              |
| 1: Average Hourly Earnings Change |         |          |           | 10.295***      | 2.577     | NA  |                      |              |
| 2: US NonFarm Employment Change   |         |          |           | 17.938***      | 2.572     | NA  |                      |              |
| 3: Unemployment Rate°             |         |          |           | 1.975          | 2.579     | NA  |                      |              |
| Test 1 All News                   |         |          |           |                | NA        | NA  |                      | -28.14001*** |
| Test 2 News 1-2                   |         |          |           |                | NA        | NA  |                      | -24.01494*** |
| Test 3 News 2-3                   |         |          |           |                | NA        | NA  |                      | -20.32601*** |
| Test 4 News 1-3                   |         |          |           |                | NA        | NA  |                      | -7.07818     |
| Test 5 News 2                     |         |          |           |                | NA        | NA  |                      | -17.64316*** |
| CAD Batch                         | USD/CAD | 13:30    | Monthly   |                | NA        | 135 | Oct 2008 to Dec 2019 |              |
| 1: Employment Change              |         |          |           | -25.588***     | 3.940     | NA  |                      |              |
| 2: Unemployment Rate              |         |          |           | 1.053          | 3.940     | NA  |                      |              |
| Test 1 All News                   |         |          |           |                | NA        | NA  |                      | -9.299685    |
| AUD Batch                         | AUD/USD | 00:30    | Monthly   |                | NA        | 135 | Oct 2008 to Dec 2019 |              |
| 1: Employment Change              |         |          |           | 21.596***      | 1.842     | NA  |                      |              |
| 2: Unemployment Rate              |         |          |           | -11.931***     | 1.842     | NA  |                      |              |
| Test 1 All News                   |         |          |           |                | NA        | NA  |                      | -22.91171*** |
| Test 2 News 1                     |         |          |           |                | NA        | NA  |                      | -6.849966    |
| Test 3 News 2                     |         |          |           |                | NA        | NA  |                      | -5.655749    |

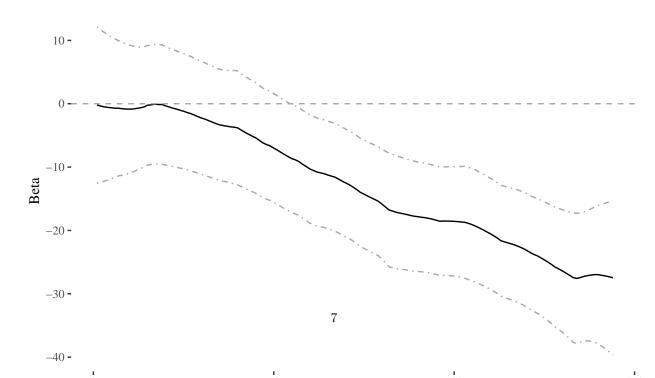
## 5 Parameter Path Estimations

- 6 Conclusion
- 7 Appendix
- 7.1 Parameter Paths

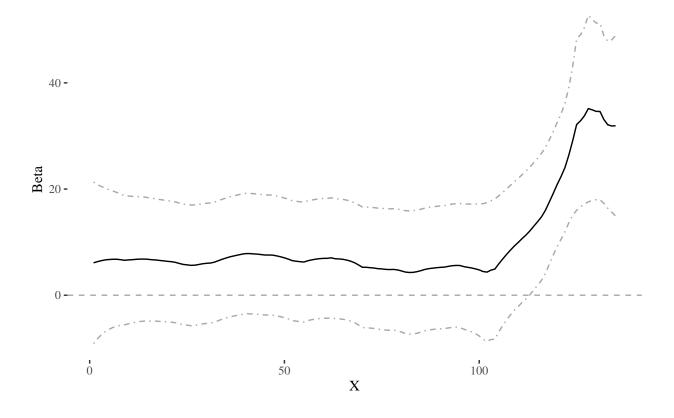
Canadian Core Retail Sales



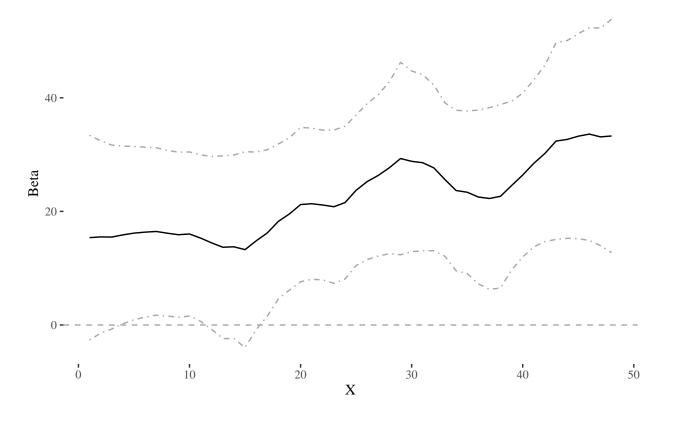
## Canadian Consumer Price Index



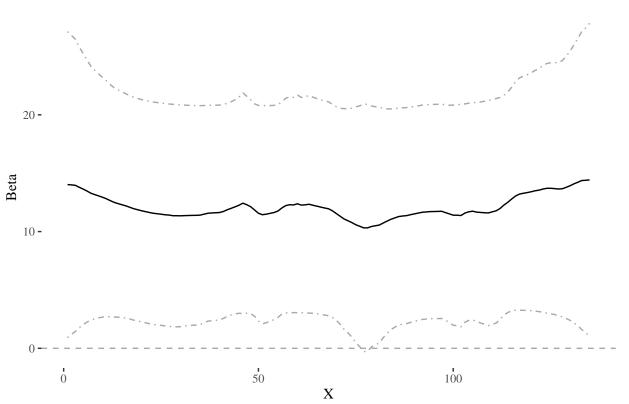
## Australian Retail Sales



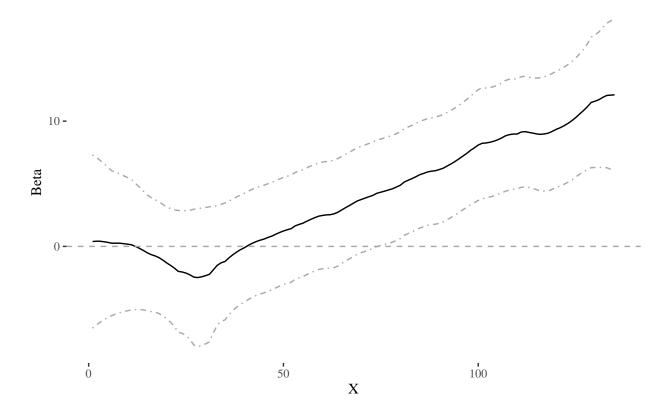
## Australian Consumer Price Index



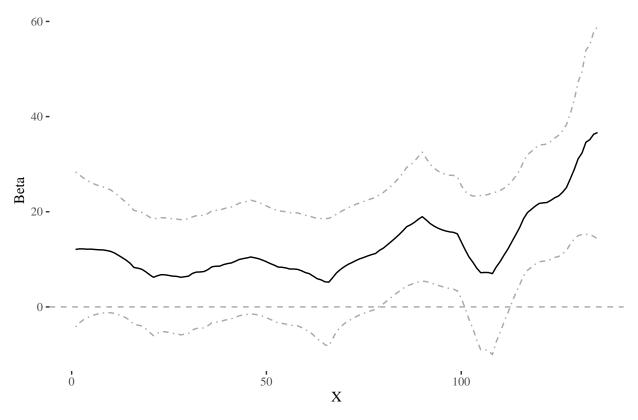
# US NonFarm Employment Change



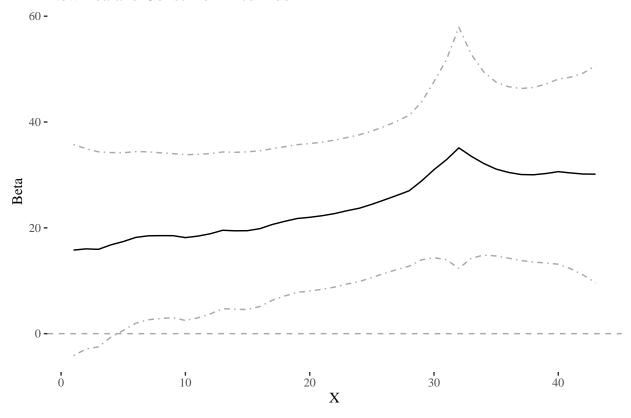
## US Consumer Price Index



## UK Consumer Price Index



### New Zealand Consumer Price Index



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