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

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August 13th, 2020

Project Motivation and Context

Linear Regression

Testing Stability

Estimating Parameter Paths

Plotting the Paths

Questions / Comments

Project Motivation and Context



Figure 1: Minute-by-minute candlechart of the USD/CAD asset on the 19th of June between 15h00 and 16h00 GMT+2. An example of of the sudden price change that can occur during one of the news releases. Green/Red candles represent an increase/decrease in price. The gray indicator is amount of trading activity taking place (a rough guideline as it only includes transactions from that particular brokerage).

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

Country	News.Event	Pair.used	GMT.Time	Frequency	Observations	Dates
Single News						
Canada	Consumer Price Index	USD/CAD	13:30	Monthly	103	Jan 2008 to Dec 2019
Canada	Core Retail Sales	USD/CAD	12:30	Monthly	94	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						
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

Note: The Canadian Consumer Price Index and Core Retail Sales coincided in Date and Time 41 times and these observations were removed.

Data

Left: News figure data. Right: Currency pair price

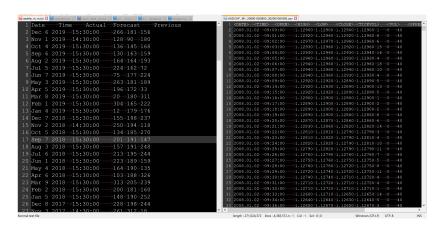


Figure 2: Data observations of our study

Linear Regression

Variables

5 Minute Price Change

$$R_t = \beta_0 + \beta_1 S_t + \varepsilon_t \tag{1}$$

"Suprise" Component of macroeconomic news

$$S_t = \frac{A_t - E_t}{\sigma_d} \tag{2}$$



Results - OLS

News.Event	M5.Coefficient	std.error	HAC.std.error
Single News			
UK CPI	12.681***	1.729	2.601
CA CPI	-8.6286***	1.958	2.502
CA CRS	-10.756***	1.624	2.249
US CPI	3.968**	1.207	1.23
NZ CPI	24.109***	2.910	4.727
AU CPI	22.293***	4.145	4.226
AU RET	9.647***	1.215	2.656
UK RET	16.574***	1.968	2.572
Grouped News			•
US AHE	10.295***	2.577	2.665
US NFP	17.938***	2.572	4.143
US UR°	1.975	2.579	2.198
CA EMC	-25.588***	3.940	4.211
CA UR	1.053	3.940	4.262
AU EMC	01 50571***	1 040	2.780
	21.59571***	1.842	
AU UR	-11.93102***	1.842	1.828

Note: The result of OLS estimation, referred to as the 'time invariant' or 'stable' case is presented in this table. The standard errors of the estimator as well as the Newey-West corrected standard errors are also included. The *,**,*** are for 10%, 5% and 1% significance levels respectively.

Testing Stability

Three Different Test

- ▶ qLL (Elliott and Müller 2006)
- Cusum (Brown, Durbin, and Evans 1975)
- Cusum-Squared (Brown, Durbin, and Evans 1975)

qLL Test

Likelihood Function under the Null Hypothesis: Stable Assumption with OLS.

$$L_{H0} = -\frac{1}{2\sigma^2} \sum_{t=1}^{I} (\Delta R_t)^2$$
 (3)

Likelihood Function under the Alternative Hypothesis: Unstable Assumption Moving Average order 1 $\Delta R_t \sim \eta_t + \psi_\eta \eta_{t-1}$

$$L_{HA} = -\frac{1}{2\sigma^2} \sum_{t=1}^{I} \eta^2 \tag{4}$$

Obtain the Test Statistic

$$\frac{\sigma_{\epsilon}^2}{\sigma_n^2} \sum_{t=1}^T \eta^2 - \sum_{t=1}^T (\Delta R_t)^2 \tag{5}$$

Steps to obtain qLL statistic (Elliott and Müller 2006)

- 1. Compute the OLS residuals $\hat{\varepsilon}_t$ by regressing R_t on S_t, Z_t ;
- 2. Construct a consistent estimator \hat{V}_X of the k*k long-run covariance matrix of $S_t \varepsilon_t$. When ε_t can be assumed uncorrelated, a natural choice is the heteroscedasticity robust estimator $\hat{V}_X = T^{-1} \sum_{t=1}^T X_t X_t' \varepsilon_t^2$
- 3. Compute $\hat{U}_t = \hat{V}_X^{-1/2} X_t \hat{\varepsilon}_t$ and denote the k elements of \hat{U}_t by $\hat{U}_{t,i}$, i=1,...,k.
- 4. For each series $\hat{U}_{t,i}$, compute a new series, $\hat{w}_{t,i}$ via $w_{t,i} = \bar{r}\hat{w}_{t-1,i} + \Delta \hat{U}_{t,i}$, and $\hat{w}_{1,i} = \hat{U}_{1,i}$, where $\bar{r} = 1 10/T$.
- 5. Compute the squared residuals from OLS regressions of $\hat{w}_{t,i}$ on \bar{r}^t individually, and sum all of those over i = 1, ..., k.
- 6. Multiply this sum of sum of squared residuals by \bar{r} , and subtract $\sum_{i=1}^{k} \sum_{t=1}^{T} (\hat{U}_{t,i})^2$

qLL Statistic

- Monotone transformation of the LRT test.
- $\triangleright \beta_t \beta_0$ following a Gaussian Random Walk assumption
- ► Leads to: First Differences follow Gaussain Moving Average

Table 2: 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

Note:

Extract of the critical values of the qLL Statistic. k represents the number of potential unstable coefficients (number of parameters in the model) whereas 1%, 5% and 10% are the significance levels where a lower value is stronger evidence that instability is present.

Cusum Test

- Use of the Recursive Least Squares Algorithm (to obtain the errors!)
- OLS but adding additional observations sequentially.

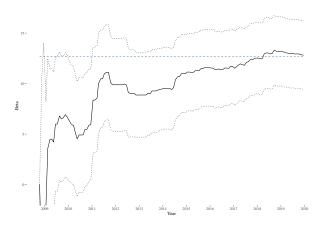


Figure 4: RLS applied to the UK CPI

RLS Algorithm

1.
$$\hat{\beta}_t = \hat{\beta}_{t-1} + g_t(R_t - S_t^T \hat{\beta}_{t-1})$$

2. $g_t = P_{t-1}^* S_t(\hat{\sigma}^2 + S_t^T P_{t-1}^* S_t)^{-1}$

3.
$$P_t^* = P_{t-1}^* - g_t S_t^T P_{t-1}^*$$

(Young 2011)

$$u_t = R_t - E(\hat{R}_t | R_{t-1})$$

Normalized:

$$u_{n,t} = \frac{u_t}{(1 + S_t^T P_t S_t)^{0.5}}$$

Compounded:

$$W_t = \frac{1}{\hat{\sigma}_{cs}} \sum_{i=k+1}^t u_{n,i}$$

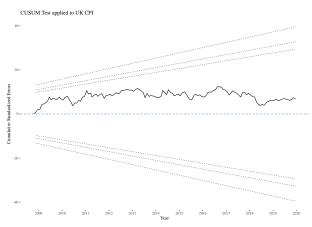
(8)

(6)

(7)

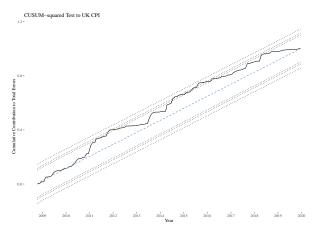
Cusum Plot

Use the confidence bands suggested by Brown, Durbin, and Evans (1975) starting at time k: $\pm a(T-k)^{0.5}$ and ending at time T: $\pm 3a(T-k)^{0.5}$.



Cusum-Squared

$$V_t = \frac{\sum_{i=k+1}^t u_{n,i}^2}{\sum_{i=k+1}^T u_{n,i}^2} \tag{9}$$



Under the null, the cumulative sum of squares follow a beta distribution and the mean is (k - h)/(N - h). As significance levels: ^{18/45}

Test Results

Table 3: Instability Test Results

News.Event	qLL	CUSUM	CUSUM.sq
Single News			
UK CPI	-12.964***	n.s	**
CA CPI	-13.015***	n.s	***
CA CRS	-9.713**	n.s	**
US CPI	-21.582***	n.s	***
NZ CPI	-7.267*	n.s	***
AU CPI	-9.022**	n.s	***
AU RET	-17.623***	**	***
UK RET	-4.503	n.s	***
Grouped News			
US Batch			
Test 1 All News	-28.14***	*	***
Test 2 AHE&NFP	-24.015***		
Test 3 NFP&UR	-20.326***		
Test 4 AHE&UR	-7.078		
Test 5 NFP	-17.643***		
CA Batch			
Test 1 All News	-9.30	n.s	***
AUD Batch			
Test 1 All News	-22.912***	n.s	***
Test 2 EMC	-6.85		
Test 3 UR	-5.66		

Note: Results of the three instability tests performed for each piece of macroeconomic news.

Estimating Parameter Paths

Two Different Methods

- ► WAR Minimizing (Elliott and Müller 2006)
- ► Stochastic Time Varying Parameter (Young 2011)

WAR Minimizing

- ▶ Start from a likelihood function in its general form.
- From a "stable" case to unknown "unstable" model where:

Likelihood: $\sum_{t=1}^{T} \ell_t(\beta)$ to: $\sum_{t=1}^{T} \ell_t(\beta + \delta_t)$

▶ 2nd order Taylor Approximation in its classical form:

$$f(x) = f(a) + f'(x - a) + \frac{f''}{2}(x - a)^2$$
 (10)

▶ Taylor Expansion of ℓ_t around $\hat{\beta}_{OLS}$ the unstable model (Univariate):

$$\ell_t(\beta + \delta_t) = \ell_t(\hat{\beta}) + s_t(\hat{\beta})(\beta + \delta_t - \hat{\beta}) - \frac{1}{2}(h_t(\hat{\beta})(\beta + \delta_t - \hat{\beta})^2$$
(11)

lt is possible to find $s_t(\hat{\beta})$ and $h_t(\hat{\beta})$!

WAR Minimizing (continued) (1)

An average \hat{H} is used instead of individual hessians at each observation.

$$\ell_t(\beta + \delta_t) = \ell_t(\hat{\beta}) + s_t(\hat{\beta})(\beta + \delta_t - \hat{\beta}) - \frac{1}{2}(\hat{H}(\hat{\beta})(\beta + \delta_t - \hat{\beta})$$
(12)

▶ Rearranging and subtracting $\frac{1}{2}s_t(\hat{\beta})\hat{H}^{-1}$ from each side.

$$\ell_t(\beta + \delta_t) - \ell_t(\hat{\beta}) - \frac{1}{2} s_t(\beta) \hat{H}^{-1} \approx -\frac{1}{2} (s_t(\hat{\beta}) - \hat{H}(\beta + \delta_t - \hat{\beta}) \hat{H}^{-1}$$
(13)

Comparing to the log-likelihood of an arbitrary Gaussian Random Variable X_n (standard nomenclature)

$$-\frac{N}{2}\log(2\pi) - N\log(\sigma^2) - \frac{1}{2\sigma^2}\sum_{n=0}^{N}(X_n - \mu)^2$$
 (14)

WAR Minimizing (continued) (2)

Arbitrary Gaussian Random Variable X_i: (standard nomenclature)

$$X_i \sim \mathcal{N}(\mu, \sigma^2)$$

$$X_i = \mu + \epsilon_i \quad \epsilon_i \sim \mathcal{N}(0, \sigma^2)$$
(15)

Scenario at hand:

$$s_t(\hat{\beta}) + \hat{H}\hat{\beta} \sim \mathcal{N}(\hat{H}(\beta + \delta_t), \hat{H})$$

$$s_t(\hat{\beta}) + \hat{H}\hat{\beta} = \hat{H}(\beta + \delta_t) + v_t \quad v_t \sim \mathcal{N}(0, \hat{H})$$
(16)

WAR Minimizing (continued) (3)

$$\beta_{OLS} = \beta_t + T^{-1/2} \hat{H}^{-1} v_0 \tag{17}$$

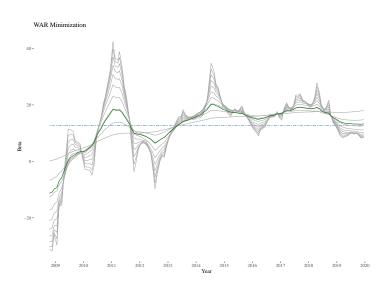
$$s_t(\beta) = \hat{H}\delta_t + v_t, t = 1, ..., T$$
 (18)

- ▶ Obtain 11 Different Random Walk weighting functions (paths).
- They all use the model above but have a different variance.
- $ightharpoonup s_t(\hat{\beta})$ and \hat{H} as:

$$s_t = \frac{\partial I}{\partial \beta_1} = \sigma^{-2} (R_t - \hat{\beta}_0 - \hat{\beta}_1 S_t) S_t$$
 (19)

$$\hat{H} = \frac{1}{T} \sum_{t=1}^{I} h_t(\hat{\beta}) = -\frac{\partial I^2}{\partial^2 \beta_1^T} = \sigma^{-2} \sum_{t=1}^{I} S_t^2$$
 (20)

WAR Plot



Steps to obtain WAR minimization path

- 1. For $t=1,\ldots,T$, let a_t and b_t be the first p elements of $\hat{H}^{-1}s_t(\hat{\theta})$ and $\hat{H}\hat{V}^{-1}s_t(\hat{\theta})$ respectively.
- 2. For $c_i \in C = 0, 5, 10, ..., 50, i = 1, ..., 11$ compute (a) $r_i = 1 c_i/T$,
- $z_{i,1} = x_1$ and $z_{i,t} = r_i z_{i,t-1} + x_t x_{t-1}, t = 2, ..., T$;
- (b) the residuals $\{\tilde{z}_{i,t}\}_{t=1}^T$ of a linear regression of $\{z_{i,t}\}_{t=1}^T$ on $\{r_i^{t-1}I_p\}_{t=1}^T$
- (c) $\bar{z}_{i,T} = \tilde{z}_{i,T}$, and $\bar{z}_{i,t} = r_i \bar{z}_{i,t+1} + \tilde{z}_{i,t} \tilde{z}_{i,t+1}$, t = 1, ..., T 1;
- (d) $\{\hat{\beta}_{i,t}\}_{t=1}^T = \{\hat{\theta}_{T} + a_t r_i \bar{z}_{i,t}\}_{t=1}^T;$
- (e) $qLL(c_i) = \sum_{t=1}^{T} (r_i) \bar{z}_{i,t} a_t)' \tilde{b}_t$ and

$$\tilde{w}_i = \sqrt{T(1 - r_i^2)r_i^{T-1}/(1 - r_i^{2T})}e^{-\frac{1}{2}qLL(c_i)}$$
 (set $\tilde{w}_0 = 1$) 3. Compute $w_i = \tilde{w}_i/\sum_{i=1}^{11} \tilde{w}_i$.

- 4. The parameter path estimator is given by $\{\hat{\beta}_t\}_{t=1}^T = \{\sum_{i=1}^{11} w_i \hat{\beta}_{i,t}\}_{t=1}^T$.
- 5. The statistic qLL(10) tests the null hypothesis of stability of β and rejects for small values.

(Müller and Petalas 2010)

STVP Algorithm

► RLS (used for the tests earlier)

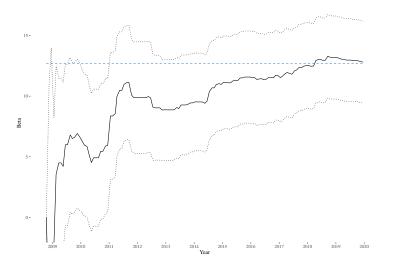


Figure 5: SRLS applied to the UK CPI

STVP Algorithm (continued)

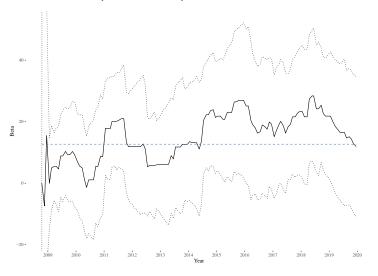


Figure 6: STVP applied to the UK CPI

Standard Recursive Time Variable Parameter Algorithm (STVP)

- Additional assumption: β_t is following a Gaussian Random Walk $\mathcal{N}(0, q_a)$ so that: $\beta_t = \beta_{t-1} + \eta_{t-1}$
- Diagonals of Qa are 25% of the OLS Coefficient
- A = D = I. (with I as the identity matrix)

Prediction (Prior)

$$1. \hat{\beta}_t | \hat{\beta}_{t-1} = A\beta_{t-1}$$

2.
$$P_t^*|P_{t-1}^* = AP_{t-1}^*A^T + DQ_aD^T$$

Correction (Posterior, same as the RLS seen earlier)

3.
$$\hat{\beta}_t = \hat{\beta}_t | \hat{\beta}_{t-1} + g_t (R_t - S_t^T (\hat{\beta}_t | \hat{\beta}_{t-1}))$$

4.
$$g_t = (P_t^*|P_{t-1}^*)S_t(\hat{\sigma}^2 + S_t^T(P_t^*|P_{t-1}^*)S_t)^{-1}$$

5.
$$P_t^* = P_t^* | P_{t-1}^* - g_t S_t^T (P_t^* | P_{t-1}^*)$$

(Young 2011)

Plotting the Paths

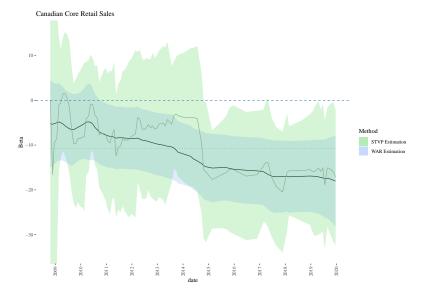


Figure 7: Note: 1 Std Dev is +0.605% greater than expected. Currency pair: USD/CAD.

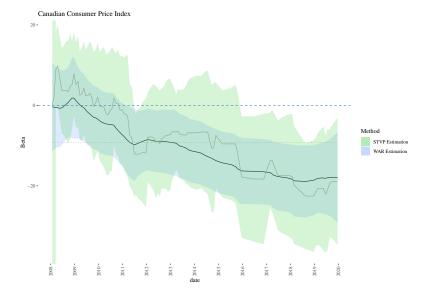


Figure 8: Note: 1 Std Dev is +0.229% greater than expected. Currency pair: USD/CAD.

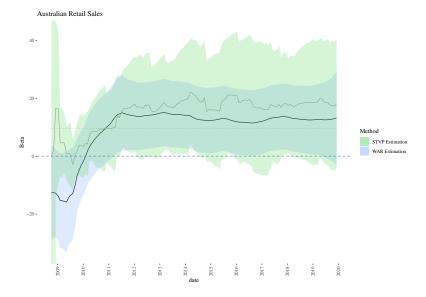


Figure 9: Note: 1 Std Dev is +0.569% greater than expected. Currency pair: AUD/USD.

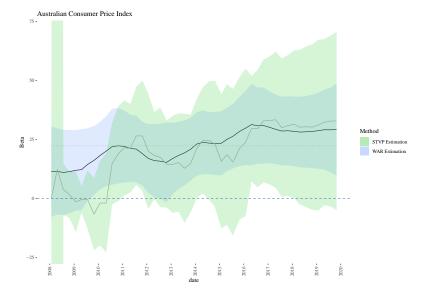


Figure 10: Note: 1 Std Dev is +0.229% greater than expected. Currency pair: AUD/USD.

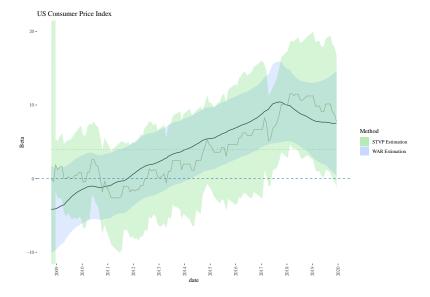


Figure 11: Note: 1 Std Dev is +0.117% greater than expected. Currency pair: USD/CHF.

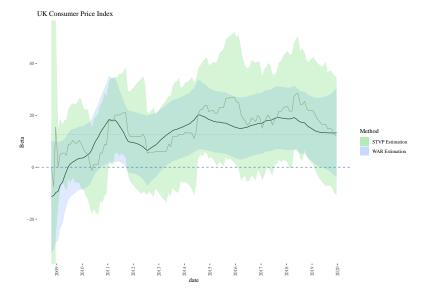


Figure 12: Note: 1 Std Dev is +0.173% greater than expected. Currency pair: GBP/USD.

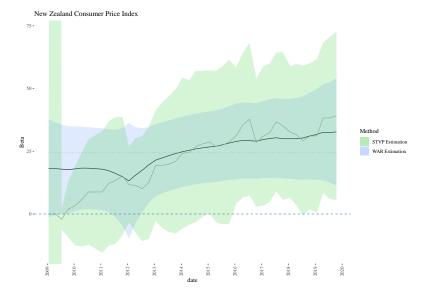


Figure 13: Note: 1 Std Dev is +0.206% greater than expected. Currency pair: NZD/USD.

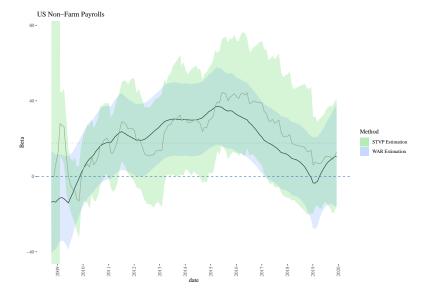


Figure 14: Note: 1 Std Dev is +66.5 thousand more people than expected. Currency pair: USDCHF.

Prediction

Calculating the Prediction Intervals

$$E(\hat{R}_{t+1}|\mathcal{F}_t, S_{t+1}) = \hat{\beta}_t S_{t+1}$$
 (21)

$$Var(\hat{R}_{t+1}|\mathcal{F}_t, S_{t+1}) = S_{t+1}^2 \sigma_{\beta}^2 + \sigma_{\epsilon}^2$$
(22)



Figure 15: GBP/USD on the 15th of January between 14h00 and 15h00 GMT+2. UK CPI released with a figure of 1.3% versus the expected 1.5% (equivalently a -1.1579697 Std. Dev shock). Point-Estimates and 95% prediction bounds are colored in blue and green for the WAR and STVP estimations respectively. The time-invariant OLS case is also added in black for reference.



Figure 16: AUD/USD on the 10th of January between 02h00 and 03h00 GMT+2. Retail Sales announced at 0.9% versus the expected 0.4% that was expected (equivalently a +0.8801813 Std. Dev shock).

Point-Estimates and 95% prediction bounds are colored in blue and green for the WAR and STVP estimations respectively. The time-invariant OLS estimate is also added in black for reference.

Questions / Comments

References

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