# Nonparametric estimation of multivariate volatility: A frame duality approach

Wale Dare

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#### 1 Model

Let  $\{p_t\}_{0 \le t \le T}$  be log prices with dynamics given by

$$dp_t = \mu_t dt + \sigma_t dW_t, \qquad t \in [0, T], \tag{1.1}$$

where  $\{W_t\}$  is a standard Brownian motion with respect to the filtered probability space  $(\Omega, \mathcal{F}, \{\mathcal{F}\}_t, P)$  satisfying the usual conditions;  $\{\mu_t\}$  is locally bounded and cadlag, whereas  $\{\sigma_t\}$  is predictable and locally bounded; both  $\{\mu_t\}$  and  $\{\sigma_t\}$  satisfy the Lipschitz and growth conditions required for the existence of a strong solution. The observation horizon T is fixed and finite.

Suppose n log prices  $p_{t_i} := p_i$ ,  $i = 1, 2, \dots, n$ , are observed discretely at equidistant intervals  $\Delta_n := T/n$ . Given this data we wish to obtain an estimate of the spot volatility,  $\sigma$ , within the observation interval [0, T]. If all paths of  $\sigma$  are such that

$$\int_0^T \sigma_t^2 \mathrm{d}t < \infty,$$

then the spot volatility is a random element (function) in  $L^2(0,T)$ , the set of square integrable functions on [0,T]. Without loss of generality we set T equal to 1.

Now  $L^2(0,1)$  is a separable Hilbert space so that it admits a frame representation. That is, there is a sequence,  $\{\phi_k\}_{k=1}^{\infty}$ , of elements in  $L^2(0,1)$  such that for all  $f \in L^2(0,1)$ ,  $f = \sum_{k=1}^{\infty} c_k \phi_k$ , where  $\{c_k\}_{k=1}^{\infty}$  is a sequence in  $\ell^2(\mathbb{N})$ , the set of square summable sequences. Frames generalize the notion of orthogonal basis: both share the representation property, but frames need not have elements that are mutually orthogonal. As a result, the representation in terms of the elements of a frame need not be unique. This is by no means a disadvantage; the redundancies in the frame yield computational stability and parsimony in the representation.

#### 2 Frames and Riesz bases

We start by giving a general definition of frames. The specialization to the space of interest  $L^2(0,1)$  is immediate.

**2.1 Definition** Let  $\mathcal{H}$  be a separable Hilbert space with inner product  $\langle \cdot, \cdot \rangle$  linear in the first argument. A sequence  $\{\phi_k\}_{k\in\mathbb{N}}$ , with  $\phi_k \in \mathcal{H}$  for all k, is a frame if there exists positive constants c and C such that

$$c||f||^2 \le \sum_{k \in \mathbb{N}} |\langle f, \phi_k \rangle|^2 \le C||f||^2, \tag{2.1}$$

for all  $f \in \mathcal{H}$ .

The constants c and C are the frame bounds. If  $\{\phi_k\}$  is a frame then we may associate with it a bounded operator,  $A: \ell^2(\mathbb{N}) \to \mathcal{H}$ , known as the synthesis operator and given by  $A \{c_k\} := \sum_{k \in \mathbb{N}} c_k \phi_k$ . Its adjoint,  $A^*: \mathcal{H} \to \ell^2(\mathbb{N})$ , is known as the analysis operator and is given by  $A^*f := \{\langle f, \phi_k \rangle\}$ . By composing the analysis operator with the synthesis operator, we obtain the frame operator,  $F: \mathcal{H} \to \mathcal{H}$ , given by

$$Ff := AA^*f = \sum_{k \in \mathbb{N}} \langle f, \phi_k \rangle \phi_k.$$

The frame operator F is bounded, invertible, and self-adjoint<sup>1</sup>. This yields the representation result

$$f = FF^{-1}f = \sum_{k \in \mathbb{N}} \langle f, F^{-1}\phi_k \rangle \phi_k. \tag{2.2}$$

The sequence  $\{F^{-1}\phi_k\}_{k\in\mathbb{N}}$  is also a frame, and it is called the *canonical dual* of  $\{\phi_k\}_{k\in\mathbb{N}}$ . A frame will generally have other duals besides the canonical dual. Frames are quite general objects. What is needed is some control over the type of redundancies allowed in a frame. Without such a restriction results about the rate of convergence of the frame expansion would be impossible to come by. A Riesz basis provides just the type of control needed. Informally, a Riesz basis is a frame whose elements are all essential.

**2.2 Definition** A sequence  $\{\phi_k\}_{k\in\mathbb{N}}$ , with  $\phi_k \in \mathcal{H}$  for all k, is a Riesz basis if there exists an orthonormal basis  $\{\xi_k\}_{k\in\mathbb{N}}$  of  $\mathcal{H}$  and a bounded invertible operator  $T: \mathcal{H} \to \mathcal{H}$  such that  $\phi_k = T\xi_k$ , for all k.

<sup>&</sup>lt;sup>1</sup>See Christensen (2001) and the references therein.

A frame is Riesz basis if it is *complete*; i.e. whenever  $\langle f, \phi_k \rangle = 0$  for all k then f = 0; and there are positive cosntants c, C such that

$$c\sum_{k=1}^{N}|c_k|^2 \le \left\|\sum_{k=1}^{N}c_k\phi_k\right\|^2 \le C\sum_{k=1}^{N}|c_k|^2,\tag{2.3}$$

for all finite sequences  $\{c_k\}_{1\leq k\leq N}$ . This is equivalent to the condition

$$c \le \sum_{k \in \mathbb{Z}} |\hat{\phi}(\omega + 2\pi k)| \le C, \qquad \forall \omega \in [0, 2\pi],$$
 (2.4)

where  $\hat{\phi}$  is the Fourier transform of  $\phi$ . To proceed in our analysis, we specialize further the type of Riesz basis to those that may be generated by a single element (function),  $\phi \in L^2(0,1)$ . The entire Riesz basis is then generated by translating  $\phi$  across the closed unit interval. By appropriately scaling the  $\phi$  we end up with different levels of granularity in the representation. That is, we have in mind a collection<sup>2</sup>  $\{\phi_{h,k}\}_{k,h\in\mathbb{Z}}$ , where  $\phi_{h,k} := \phi(x/h - k)$ . We denote the function space generated by this basis as follows:

$$V_h(\phi) := \left\{ \sum_{k \in \mathbb{Z}} c_{h,k} \phi_{h,k} : \{ c_{h,k} \} \in \ell^2(\mathbb{N}) \right\}$$
 (2.5)

### 3 Volatility estimation by duality

To obtain estimates for  $\sigma^2 \in L^2(0,1)$ , we appeal to the duality theorem of Riesz basis. That is, there exists a Riesz basis  $\{\psi_{h,k}\}_{k\in\mathbb{Z}}$  such that the projection of  $\sigma^2$  onto  $V_h(\phi)$  is given by

$$\sigma_{h,k}^2(t) = \sum_{k \in \mathbb{Z}} c_{h,k} \,\phi_{h,k}(t), \text{ where}$$
(3.1)

$$c_{h,k} = \langle \sigma^2, \psi_{h,k} \rangle \tag{3.2}$$

Now given n observations of the log price process p, we propose the following estimator of the volatility

$$\hat{\sigma}_n^2(t) = \sum_{k=-K}^K \hat{c}_{h,k} \,\phi_{h,k}(t) \tag{3.3}$$

$$\hat{c}_{h,k} = \sum_{i=2}^{n} \psi_{h,k}((i-1)/n)(p_{i\Delta_n} - p_{(i-1)\Delta_n})$$
(3.4)

<sup>&</sup>lt;sup>2</sup>See Unser & Daubechies (1997) for further elaboration on these ideas.

## References

Christensen, Ole (2001) "Frames, Riesz bases, and discrete Gabor/wavelet expansions",  $Bull.\ Amer.\ Math.\ Soc,\ pp.\ 273–291.$ 

Unser, Michael and Daubechies, Ingrid (1997) "On the approximation power of convolution-based least squares versus interpolation", *IEEE Transactions on Signal Processing*, Vol. 45, No. 7, pp. 1697–1711.