

# Summary of Xiaoyan Zhang's Master's Thesis

Michael Kovaliov

22-2-2021

## Introduction

The structure of the Rough Fractional Stochastic Volatility (RFSV) model is:

$$\frac{dS(t)}{S(t)} = \mu(t)dt + \sigma(t)dZ(t)$$

Where  $\mu(t)$  is the drift term,  $Z(t)$  is a standard Brownian motion,  $X(t)$  is a Ornstein-Uhlenbeck process driven by fractional Brownian motion  $W^H(t)$  given by:

$$dX(t) = -\alpha(X(t) - m)dt + \nu dW^H(t)$$

where  $m \in \mathbb{R}$  and  $\nu, \alpha$  are positive constants, with  $Z$  and  $W^H$  correlated in general.

## Summary of chapter 1

In chapter 1 there were 3 main parts

The first talked about some generic facts about volatility and modeling assets, both theoretical and practical aspects. One practical aspect which is sometimes neglected in theory is leverage - the more debt a firm has w.r.t. its earning, its volatility is going to be higher. In practice volatility also clusters and has a long memory.

The next part talked about how to actually find the volatility of a stock price. One method was using the derived from the Black & Scholes framework. This is inconsistent with practice because in practice the volatility is not constant. Another idea was time-dependent volatility model, which is still built upon B&S. It considers the dynamic of a stock price with time-dependent (yet deterministic) volatility:

$$\frac{dS(t)}{S(t)} = \mu dt + \sigma(t)dW(t)$$

Merton(1973) proposed the pricing formula using dependent volatility. If we use the annualized variance:

$$\sigma_a = \sqrt{\frac{1}{T-t} \int_t^T \sigma^2(\tau) d\tau}$$

And view  $\sigma_a$  as a constant dependent upon  $t$  and  $T$ , you can plug it into B&S and receive a formula which when solved gives you time-dependent volatility. Local volatility and mixture models were also mentioned in this part.

The last part talks about the following model:  $\frac{dS(t)}{S(t)} = \mu_t dt + \sigma(t)dW(t)_1$

The stock price is driven by the wiener process  $W_1$  and  $\mu_t$  is the drift term. However the volatility  $\sigma(t)$  now is also stochastic. Typically the volatility will be governed by a SDE which is driven by another Wiener process  $W_2$  that is correlated with  $W_1$ . The author then described popular SDEs that are used with this model. For example:

- Scott(1987)

$$dS(t) = \mu_1 S(t)dt + \sigma(t)S(t)^\kappa dW_1(t),$$

$$d\sigma(t) = \mu_2 \sigma(t)dt + \delta \sigma^\beta(t) dW_2(t)$$

and that is the end of chapter 1.

## **Summary of chapter 2**