

# Pricing and hedging in regime switching stochastic volatility model: Application to electricity markets.

Goutte Stéphane

CNRS - LPMA Université Paris 7

December 20, 2011

EFC11

Institute of Organization and Management  
Wroclaw University of Technology



# Outline

## 1 Introduction

- Motivation
- RS-SV Model

## 2 Local risk Minimization approach

- The contingent claim
- Some definition
- Local risk minimization approach
- Our Model

## 3 Pricing swap

- Variance swap
- Volatility swap

## 4 Numerical Application

**Benth model:** For all  $t \in [0, T]$ ,

$$F_t(T) = F_0(T) \exp \left( m_t^T + \int_0^t \sigma_L(s) dW_s + \int_0^t \sigma_S(s) e^{-\lambda(T-s)} dL_s \right)$$

Studied in Goutte and al. [2], [3] using a quadratic approach.

**Stochastic volatility:** • Complex volatility structure.

- Volatility varies over time: volatility increases when the time to maturity decreases (Samuelson hypothesis).

**Regime switching:** catches states of the world as

- "good" or "bad" economic.
- "on-peak" or "off-peak" time for electricity.

Let  $(\omega, \mathcal{F}, P)$  be a filtered probability space with filtration  $\mathbb{F} = (\mathcal{F}_t)_{t \in [0, T]}$  satisfying the usual conditions for some fixed but arbitrary time horizon  $T \in (0, \infty)$ . The Regime Switching Stochastic Volatility model (RS-SV) is defined by

$$\begin{aligned} dS_t &= \mu(t, Y_t, X_t)S_t dt + Y_t S_t dW_t^1 \\ dY_t &= a(t, Y_t, X_t)dt + b(t, Y_t, X_t)dW_t^2 \end{aligned} \quad (1)$$

where

- $W^1$  and  $W^2$  are two Brownian motion which are correlated as  $d\langle W^1, W^2 \rangle_t = \rho dt$ .
- $Y_t$  is a real stochastic process which is  $\mathcal{F}_t$ -adapted.
- $X_t$  a continuous time homogeneous Markov chain on finite space  $\mathcal{S} = \{1, 2, \dots, N\}$ .

We assume that the time invariant matrix  $Q$  denotes the generator  $(q_{ij})_{i,j=1,\dots,m}$  of  $X$ , where  $q_{ij}$  is an infinitesimal intensity of  $X$ . Then, the semi-martingale decomposition for  $X$  is given by

$$X_t = X_0 + \int_0^t QX_s ds + M_t^X$$

where  $(M_t^X)$  is an  $\mathbb{R}^N$ -valued martingale with respect to the natural filtration generated by  $X$  under  $P$ .

### Assumption 1

We will assume that we know all the trajectory of  $X$ , that is,  $\mathcal{F}_T^X$  and that the Markov chain  $X$  is independent of  $S$  and  $Y$ .

In our model, there are three source of randomness:  $W^1$ ,  $W^2$  and  $X$ . Hence we will denote by  $\mathbb{G}$ , the filtration generated by  $W^1$ ,  $W^2$  and  $X$ . So the filtration  $\mathbb{G} = (\mathcal{G}_t)_{t \in [0, T]} := \sigma(W_t^1, W_t^2, X_t, 0 \leq t \leq T)$ .

- **Hull-White:**

$$\mu(t, Y_t, X_t) = \mu Y_t, a(t, Y_t, X_t) = \frac{Y_t}{2} \left( \alpha - \frac{\beta^2}{2} \right), b(t, Y_t, X_t) = \frac{\beta}{2} Y_t$$

$$dS_t = \mu Y_t S_t dt + Y_t S_t dW_t^1$$

$$dY_t^2 = \alpha Y_t^2 dt + \beta Y_t^2 dW_t^2$$

- **Stein-Stein:**

$$\mu(t, Y_t, X_t) = \mu Y_t, a(t, Y_t, X_t) = \alpha(\omega - Y_t), b(t, Y_t, X_t) = \beta$$

$$dS_t = \mu Y_t S_t dt + Y_t S_t dW_t^1$$

$$dY_t = \alpha(\omega - Y_t) dt + \beta dW_t^2$$

- **Heston:**

$$\mu(t, Y_t, X_t) = \mu Y_t, a(t, Y_t, X_t) = \frac{4\kappa\theta - \sigma^2}{8Y_t} - \frac{\kappa}{2} Y_t, b(t, Y_t, X_t) = \frac{\sigma}{2}$$

$$dS_t = \mu Y_t S_t dt + Y_t S_t dW_t^1$$

$$dY_t^2 = \kappa(\theta - Y_t^2) dt + \sigma Y_t dW_t^2$$

$$2\kappa\theta \geq \sigma^2$$

# Outline

- 1 Introduction
  - Motivation
  - RS-SV Model
- 2 Local risk Minimization approach
  - The contingent claim
  - Some definition
  - Local risk minimization approach
  - Our Model
- 3 Pricing swap
  - Variance swap
  - Volatility swap
- 4 Numerical Application

We are interested in the hedging of an European style contingent claims with an  $\mathcal{F}_T$ -measurable square integrable random payoff  $H$  based on the dynamics given by

$$\begin{aligned}dS_t &= \mu(t, Y_t, X_t)S_t dt + Y_t S_t dW_t^1 \\dY_t &= a(t, Y_t, X_t)dt + b(t, Y_t, X_t)dW_t^2\end{aligned}$$

Hence  $H$  is a function of time,  $S$ ,  $Y$  and  $X$ :

$$\begin{aligned}H : [0, T] \times \mathbb{R} \times \mathbb{R} \times \mathcal{S} &\rightarrow \mathbb{R}^+ \\(t, S_t, Y_t, X_t) &\mapsto h(t, S_t, Y_t, X_t)\end{aligned}$$

As example of payoff, we could take classical european Call option

$$H = h(S_T) := (S_T - K)^+$$



## Definition 1

An **Hedging strategy** is a pair  $\varphi = (v, \eta)$  such that  $v = (v_t)_{t \in [0, T]}$  is a predictable process such that

$$\mathbb{E} \left[ \int_0^T v_t^2 Y_t^2 S_t^2 dt \right] + \mathbb{E} \left[ \left( \int_0^T |v_t| |\mu(t, Y_t, X_t)| dt \right)^2 \right] < \infty \quad (2)$$

and  $\eta = (\eta_t)_{t \in [0, T]}$  is an adapted process such that for all  $t \in [0, T]$ ,  $\mathbb{E} [\eta_t^2] < \infty$ .

The hedging strategy  $\varphi$  defines a portfolio where  $v_t$  denotes the number of shares of the risky asset  $S$  held by the investor at time  $t \in [0, T]$  and  $\eta_t$  denotes the amount invested at time  $t$ .

## Definition 2

Given a hedging strategy  $\varphi$ , we call for all  $t \in [0, T]$

- the **Value process**  $V(\varphi)$ , the right continuous process given by

$$V_t(\varphi) = v_t S_t + \eta_t \quad (3)$$

- the **Cost process**  $C(\varphi)$ , the process given by

$$C_t(\varphi) = V_t(\varphi) - \int_0^t v_s dS_s \quad (4)$$

The quantity  $\int_0^t v_s dS_s$  represents the hedging gains or losses up to time  $t$  following the hedging strategy  $\varphi$ . If  $C(\varphi)$  is square integrable, then the **risk process** of  $\varphi$  is defined by

$$R_t(\varphi) := \mathbb{E} \left[ (C_T(\varphi) - C_t(\varphi))^2 \mid \mathcal{G}_t \right] \quad (5)$$

The study of this minimization in a general semimartingale case is due to Schweizer [4]. Assume that  $V_T(\varphi) = H$ , the local risk minimization problem is to minimize the Risk process  $R(\varphi)$ . This requires more specific assumptions on  $S$ . We assume that  $S$  can be decomposed as

$$S_t = S_0 + M_t + A_t$$

where  $M$  is a real valued locally square integrable local  $P$ -martingale null at 0 and  $A$  is a real valued adapted continuous process of finite variation also null at 0. We recall now the Definition of **Structure Condition (SC)**. We say that  $S$  satisfies the (SC) if there exists a predictable process  $\lambda$  such that  $A$  is absolutely continuous with respect to  $\langle M \rangle$  in the sense that

$$A_t = \int_0^t \lambda_s d\langle M \rangle_s$$

and such that the so called **mean variance tradeoff process (MVT)  $K$**  satisfied

$$K_t := \int_0^t \lambda_s^2 d\langle M \rangle_s < \infty, \quad P - a.s.$$

Moreover Proposition 2.24 of Follmer and Schweizer in [1] shows that finding a locally risk minimizing strategy for a given contingent claim  $H \in L^2(P)$  is equivalent to finding a decomposition of H of the form

$$H = H_0^{lr} + \int_0^T \xi_t^{lr} dS_t + L_T^{lr} \quad (6)$$

where  $H_0^{lr}$  is a constant,  $\xi^{lr}$  is a predictable process satisfying condition (2) and  $L^{lr}$  is a square integrable P-martingale null at 0 and strongly orthogonal to M (i.e.  $L^{lr}M$  is a P-martingale). The representation (6) is usually referred to as the **Follmer-Schweizer (FS) decomposition** of H.

Once we have (6), then the desired hedging strategy  $\varphi^{lr}$  which is locally risk minimizing is then given for all  $t \in [0, T]$  by

$$v^{lr}_t = \xi_t^{lr} \quad (7)$$

and

$$\eta_t^{lr} = V_t(\varphi^{lr}) - v_t^{lr} S_t \quad (8)$$

where

$$V_t(\varphi^{lr}) = C_t(\varphi^{lr}) + \int_0^t v_s^{lr} dS_s \quad (9)$$

with

$$C_t(\varphi^{lr}) = H_0^{lr} + L_t^{lr} \quad (10)$$

As is shown in [1] and [6] there exists a measure  $\hat{P}$ , the so called **minimal equivalent local martingale measure** (minimal ELMM), such that

$$V_t(\varphi^{lr}) = \hat{\mathbb{E}}[H|\mathcal{G}_t] \quad (11)$$

where  $\hat{\mathbb{E}}$  denotes the conditional expectation under  $\hat{P}$ .

Theorem 1 of [1] allows us to construct uniquely  $\hat{P}$  such that  $\hat{P}$  exists if and only if for all  $t \in [0, T]$

$$\hat{Z}_t = \exp\left(-\int_0^t \lambda_s dM_s - \frac{1}{2} \int_0^t \lambda_s^2 d\langle S \rangle_s\right) \quad (12)$$

is a square integrable martingale under  $P$ . Then

$$\frac{\hat{P}}{P} := \hat{Z}_T \in L^2(P)$$

defines a probability measure  $\hat{P}$  equivalent to  $P$  which is in  $\mathbb{P}$  since one easily verifies that  $\hat{Z}S$  is a local  $P$ -martingale.

Let  $S$ ,  $Y$  and  $X$  given as in model (1). The local risk minimizing hedging strategy can be obtained in two step.

(I) Determine  $\hat{P}$  and the dynamic of  $(S, Y)$  under  $\hat{P}$ .

(II) Find the decomposition of  $H$  with respect to  $S$  under  $\hat{P}$ .

Then the optimal hedging strategy will defined by (7) and (8).

### Finding $\hat{P}$ :

According to previous subsection, the density process of the minimal ELMM  $\hat{P}$  with respect to  $P$  is given by

$$\hat{Z}_t = \exp \left( - \int_0^t \lambda_s dM_s - \frac{1}{2} K_t \right)$$

Hence  $S$  is continuous, we need to determine the canonical decomposition  $S_t = S_0 + M_t + \int_0^t \lambda_s d\langle M \rangle_s$  of  $S$  under  $P$ .

### Proposition 1

Assume the regime stochastic volatility model (1) then we have that

$$M_t = \int_0^t S_s Y_s dW_s^1, A_t = \int_0^t \mu(s, Y_s, X_s) S_s ds \quad , \quad \langle M \rangle_t = \int_0^t S_s^2 Y_s^2 ds$$
$$\lambda_t = \frac{dA_t}{d\langle M \rangle_t} = \frac{\mu(t, Y_t, X_t)}{S_t Y_t^2} \quad \text{and} \quad K_t = \int_0^t \left( \frac{\mu(s, Y_s, X_s)}{Y_s} \right)^2 ds$$
$$\hat{Z} = \exp \left( - \int_0^t \frac{\mu(s, Y_s, X_s)}{Y_s} dW_s^1 - \frac{1}{2} \int_0^t \left( \frac{\mu(s, Y_s, X_s)}{Y_s} \right)^2 ds \right)$$



We are now able to determine the dynamic of the model under  $\hat{P}$ .

## Proposition 2

**Assume that  $\hat{Z}$  is a true  $\mathbf{P}$ -martingale**, then the dynamic of the model (1) under  $\hat{P}$  is given for all  $t \in [0, T]$  by

$$\begin{aligned}dS_t &= Y_t S_t d\hat{W}_t^1 \\dY_t &= \hat{a}(t, Y_t, X_t)dt + b(t, Y_t, X_t)d\hat{W}_t^2 \\dY_t &= \hat{a}(t, Y_t, X_t)dt + b(t, Y_t, X_t)(\rho d\hat{W}_t^1 + \sqrt{1 - \rho^2}d\hat{W}_t^3)\end{aligned}\tag{13}$$

with

$$\hat{a}(t, Y_t, X_t) = a(t, Y_t, X_t) - \frac{\rho}{Y_t} \mu(t, Y_t, X_t) b(t, Y_t, X_t)\tag{14}$$

and  $\hat{W}^3$  is another Brownian motion independent of  $\hat{W}^1$ .

## Decomposition of H with respect to S under $\hat{P}$ :

Taking a contingent claim of the form  $H = h(S_T)$  for some given function  $h \rightarrow [0, \infty) \times \mathbb{R} \times \mathcal{S}$ . Then finding the Galtchouk-Kunita-Watanabe decomposition of H under  $\hat{P}$  reduces to solve a system of partial differential equation PDE if one exploits the Markovian structure.

Using the Markov property we can rewrite (11) in the form

$$V_t(\varphi^{lr}) = \hat{\mathbb{E}} [h(S_T) | \mathcal{F}_t] = \hat{v}(t, S_t, Y_t, X_t) \quad (15)$$

for some function  $\hat{v}(t, s, y, x)$  defined on  $[0, T] \times \mathbb{R} \times \mathbb{R} \times \mathcal{S}$ .

### Proposition 3:

We have that the conditional price of the contingent claim  $H$  is given by  $\hat{v}$  as the solution to the system of partial differential equation given for all  $i \in \mathcal{S}$  by

$$\begin{aligned} 0 &= \hat{v}_t(t, s, y, i) + \hat{a}(t, y, i)\hat{v}_y(t, s, y, i) \\ &\quad + \frac{1}{2} [s^2 y^2 \hat{v}_{ss}(t, s, y, i) + b^2(t, y, i)\hat{v}_{yy} + 2syb\rho\hat{v}_{sy}(t, s, y, i)] \\ &\quad + \sum_{j \neq i, j \in \mathcal{S}} q_{ij} (\hat{v}(t, s, y, j) - \hat{v}(t, s, y, i)) \end{aligned}$$

with terminal condition for all  $i \in \mathcal{S}$ ,

$$\hat{v}(T, s, y, i) = h(s, y, i)$$

where  $\hat{v}_t := \frac{\partial \hat{v}}{\partial t}$ ,  $\hat{v}_y := \frac{\partial \hat{v}}{\partial Y}$ ,  $\hat{v}_{ss} := \frac{\partial^2 \hat{v}}{\partial S^2}$ ,  $\hat{v}_{yy} := \frac{\partial^2 \hat{v}}{\partial Y^2}$  and  $\hat{v}_{sy} := \frac{\partial^2 \hat{v}}{\partial S \partial Y}$ .

Hence for the particular case of european call option:

$$\hat{v}(T, s, y, i) = (s_T(i) - K)^+$$

We are now able to find the decomposition of  $H$  with respect to  $S$  under  $\hat{P}$  and so the locally risk minimizing  $H$ -admissible strategy  $\varphi^{lr}$ .

### Theorem 1:

For all  $t \in [0, T]$  we have that the locally risk-minimizing hedging strategy of  $H$ ,  $\varphi^{lr} := (v^{lr}, \eta^{lr})$  is given by

$$v_t^{lr} = \hat{v}_s(u, S_u, Y_u, X_u) + \hat{v}_y(u, S_u, Y_u, X_u) \frac{\rho}{S_u Y_u} b(u, Y_u, X_u) \quad (16)$$

$$\eta_t^{lr} = V_t(\varphi^{lr}) - v_t^{lr} S_t \quad (17)$$

where  $V_t(\varphi^{lr}) = V_0(\varphi^{lr}) + \int_0^t v_s^{lr} dS_s + L_t^{lr}$  and

$$L_t^{lr} = \int_0^t \sqrt{1 - \rho^2} \hat{v}_y(u, S_u, Y_u, X_u) b(u, Y_u, X_u) d\hat{W}_u^3 \\ + \int_0^t \int_S [\hat{v}(u, S_u, Y_u, j) - \hat{v}(u, S_{u-}, Y_{u-}, X_{u-})] (\nu - \bar{\nu})(du, dj)$$

We can obtain a formulation of the conditional expected squared cost on the interval  $[t, T]$  for the locally risk-minimizing strategy  $\varphi^{lr}$  which we recall that it is denoted by  $R_t^{lr}$ .

#### Proposition 4:

We have that for all  $t \in [0, T]$  that the conditional expected squared cost on the interval  $[t, T]$  for the locally risk-minimizing strategy  $\varphi^{lr}$  is given by

$$\begin{aligned}
 R_t^{lr} &= \mathbb{E} \left[ \left( \int_t^T \sqrt{1 - \rho^2} \hat{v}_y(u) b(u) d\hat{W}_u^3 \right)^2 \middle| \mathcal{F}_t \right] \\
 &+ 2\mathbb{E} \left[ \left( \int_t^T \sqrt{1 - \rho^2} \hat{v}_y(u) b(u) d\hat{W}_u^3 \right) \left( \int_t^T \int_{\mathcal{S}} [\hat{v}(u, j) - \hat{v}(u, X_{u-})] (\nu - \bar{\nu})(du, dj) \right) \middle| \mathcal{F}_t \right] \\
 &+ \mathbb{E} \left[ \int_t^T \left[ Q\hat{v}^2(u, X_{u-}) - 2\hat{v}(u, X_{u-})Q\hat{v}(u, X_{u-}) \right] du \middle| \mathcal{F}_t \right]
 \end{aligned}$$

## Remark

To apply all the results about local risk minimizing hedging strategy , it remains to prove that  $\hat{Z}$  is a true  $P$ -martingale and square integrable under  $P$ . A well-known sufficient condition for both is boundedness of the mean variance tradeoff process  $K$  uniformly in  $t$  and  $\omega$ . This condition will be checked in our examples.

## Heston model:

$$\mu(t, Y_t, X_t) = \mu(X_t)Y_t \quad \text{with} \quad \mu = (\mu_1, \dots, \mu_N) \in \mathbb{R}^{\text{card}(S)}$$

$$a(t, Y_t, X_t) = \frac{4\kappa(X_t)\theta(X_t) - \sigma(X_t)^2}{8Y_t} - \frac{\kappa(X_t)}{2}Y_t, \quad \text{with} \quad \kappa = (\kappa_1, \dots, \kappa_N)$$

$$b(t, Y_t, X_t) = \frac{\sigma(X_t)}{2} \quad \text{and} \quad \rho = \rho_0 \in ]-1, 1[$$

The constants  $\kappa_i, \theta_i, \sigma_i$  are all nonnegative for all  $i \in S$  and for all  $i \in S$  that  $\kappa_i\theta_i \geq \frac{1}{2}\sigma_i$ . The model is then given by

$$\begin{aligned}dS_t &= \mu(X_t)S_t dt + Y_t S_t dW_t^1 \\dY_t^2 &= \kappa(X_t)(\theta(X_t) - Y_t^2)dt + \sigma(X_t)Y_t dW_t^2\end{aligned}$$

The mean variance tradeoff process is then given by

$$K_t = \int_0^t \frac{\mu(t, Y_s, X_s)^2}{Y_s^2} ds = \int_0^t \frac{\mu^2(X_s)Y_s^2}{Y_s^2} ds = \int_0^t \mu^2(X_s) ds < \infty$$

Hence the MVT process  $K$  is deterministic so bounded uniformly in  $t \in [0, T]$  and  $\omega$ . This implies that  $\hat{Z}$  is a P-martingale and so that we can apply all the result mentioned before.

# Outline

## 1 Introduction

- Motivation
- RS-SV Model

## 2 Local risk Minimization approach

- The contingent claim
- Some definition
- Local risk minimization approach
- Our Model

## 3 Pricing swap

- Variance swap
- Volatility swap

## 4 Numerical Application



## Variance swap

A variance swap is a forward contract on annualized variance, which is the square of the realized annual volatility. Let  $Y_R^2$  denote the realized annual stock variance over the life of the contract.

$$Y_R^2 = \frac{1}{T} \int_0^T Y_t^2 dt$$

Let  $K_v$  and  $N$  denote the delivery price for variance and the notional amount of the swap in dollars per annualized volatility point squared. Then, the payoff of the variance swap at expiration time  $T$  is given by  $N(Y_R^2 - K_v)$ . Intuitively, the buyer of the variance swap will receive  $N$  dollars for each point by which the realized annual variance  $Y_R^2$  has exceeded the variance delivery price  $K_v$ .

Assume that we are under the ELMM  $\hat{P}$ .

$$\begin{aligned}dS_t &= Y_t S_t d\hat{W}_t^1 \\dY_t &= \hat{a}(t, Y_t, X_t)dt + b(t, Y_t, X_t)d\hat{W}_t^2\end{aligned}$$

with  $\hat{a}(t, Y_t, X_t) = a(t, Y_t, X_t) - \frac{\rho}{Y_t}\mu(t, Y_t, X_t)b(t, Y_t, X_t)$ . In particular, given  $\mathcal{F}_T^X$ , **the conditional price of the variance swap**  $P(X)$  is given by

$$\begin{aligned}P(X) &= \hat{\mathbb{E}} \left[ \exp \left( - \int_0^T r_t dt \right) N(Y_R^2 - K_v) | \mathcal{F}_T^X \right] \\&= \exp \left( - \int_0^T r_t dt \right) N \hat{\mathbb{E}} \left[ Y_R^2 | \mathcal{F}_T^X \right] - \exp \left( - \int_0^T r_t dt \right) N K_v\end{aligned}$$

where  $N$  is the notional amount in dollars.

Itô formula gives that for all  $t \in [0, T]$

$$\frac{d\hat{\mathbb{E}}^X [Y_t^2]}{dt} = \hat{\mathbb{E}}^X [2Y_t\hat{a}(t, Y_t, X_t) + b^2(t, Y_t, X_t)] \quad (18)$$

### Assumption 2:

Assume that we know the solution of equation (18) which we denote for all  $t \in [0, T]$  by:  $y(t, Y_t, X_t)$ .

### Proposition 5:

Under Assumption 2, we have for all  $t \in [0, T]$  that the variance swap price  $P(X)$  is given by

$$P(X) = \exp\left(-\int_0^T r_t dt\right) N\left(\frac{1}{T} \int_0^T y(t, Y_t, X_t) dt - K_v\right) \quad (19)$$

## Heston Model

$\hat{a}(t, Y_t, X_t) = \frac{4\kappa(X_t)\theta(X_t) - \sigma(X_t)^2}{8Y_t} - \frac{\kappa(X_t)}{2} Y_t$  and  $b(t, Y_t, X_t) = \frac{\sigma(X_t)}{2}$ . Then (18) becomes

$$\frac{d\hat{\mathbb{E}}^X [Y_t^2]}{dt} = \kappa(X_t) \left( \theta(X_t) - \hat{\mathbb{E}}^X [Y_t^2] \right)$$

Let  $y_t := \hat{\mathbb{E}}^X [Y_t^2]$ , then we need to solve the differential equation

$$\frac{dy_t}{dt} = \kappa(X_t) (\theta(X_t) - y_t)$$

The solution of this differential equation is given for all  $t \in [0, T]$  by

$$y_t = y_0 \exp\left(-\int_0^t \kappa_s ds\right) + \frac{\int_0^t \left(\exp\left(\int_0^s \kappa_u du\right) \kappa_s \theta_s\right) ds}{\exp\left(\int_0^t \kappa_s ds\right)}$$

Hence we find

$$\hat{\mathbb{E}}^X [Y_t^2] = Y_0^2 \exp\left(-\int_0^t \kappa_s ds\right) + \frac{\int_0^t (\exp(\int_0^s \kappa_u du) \kappa_s \theta_s) ds}{\exp\left(\int_0^t \kappa_s ds\right)} \quad (20)$$

We can now obtain the variance swap price applying Proposition 5:

$$P(X) = \exp\left(-\int_0^T r_t dt\right) N\left[\frac{1}{T} \int_0^T \left(Y_0^2 \exp\left(-\int_0^t \kappa_s ds\right) + \frac{\int_0^t (\exp(\int_0^s \kappa_u du) \kappa_s \theta_s) ds}{\exp\left(\int_0^t \kappa_s ds\right)}\right) dt - K_V\right]$$

## Volatility swap

We recall that the realized annual stock variance over the life of the contract is given by

$$Y_R^2 := \frac{1}{T} \int_0^T Y_t^2 dt$$

and depend on the Markov chain  $X$ . Denote by  $I_t = \int_0^t Y_s^2 ds$  the accumulated variance where the process  $Y^2$  is solution of the stochastic differential equation (SDE) given by

$$dY_t^2 = (2Y_t \hat{a}(t, Y_t, X_t) + b^2(t, Y_t, X_t)) dt + 2Y_t b(t, Y_t, X_t) d\hat{W}_t^2 \quad (21)$$

Hence  $I_t$  is the solution of the SDE given by  $dI_t = Y_t^2 dt$ . Let define by  $E_t^T$  the expectation at time  $t \in [0, T]$  given  $X$  by

$$E_t^T = \hat{\mathbb{E}}_t \left[ \frac{1}{T} \int_0^T Y_s^2 ds | \mathcal{F}_T^X \right] = \hat{\mathbb{E}}_t^X \left[ \frac{1}{T} \int_0^T Y_s^2 ds \right] \quad (22)$$

Hence  $(E_t^T)_{t \in [0, T]}$  depends on the variance process  $Y^2$  of the underlying asset and on the Markov chain  $X$ . We call by fair variance strike price the quantity  $K_v^*$  which is such that the variance swap price  $P(X)$  vanishes:

$$\begin{aligned} P(X) &= \hat{\mathbb{E}} \left[ \exp \left( - \int_0^T r_t dt \right) N(Y_R^2 - K_v^*) | \mathcal{F}_T^X \right] \\ &= \hat{\mathbb{E}}^X \left[ \exp \left( - \int_0^T r_t dt \right) N(Y_R^2 - K_v^*) \right] = 0 \end{aligned}$$

Then we have that  $K_v^* = \hat{\mathbb{E}} [Y_R^2 | \mathcal{F}_T^X]$  and for time  $t = 0$ , we obtain that  $E_0^T = K_v^*$ . We define now the forward price process  $Z_t^T$  as

$$Z_t^T = \hat{\mathbb{E}}_t \left[ \sqrt{\frac{1}{T} \int_0^T Y_s^2 ds} | \mathcal{F}_T^X \right] = \hat{\mathbb{E}}_t^X \left[ \sqrt{\frac{1}{T} \int_0^T Y_s^2 ds} \right] \quad (23)$$

## Proposition 6:

The forward price process  $Z_t^T$  can be expressed as a function  $F(t, Y_t^2, X_t, I_t)$  and is the solution of the system of stochastic differential equation given by

$$\frac{\partial F}{\partial t} + \frac{\partial F}{\partial I} Y_t^2 + \frac{\partial F}{\partial Y^2} (2Y_t \hat{a}_t + b_t^2) dt + \frac{1}{2} \frac{\partial^2 F}{\partial (Y^2)^2} 4Y_t^2 b_t^2 dt + \langle F, QX_t \rangle = 0 \quad (2)$$

with boundary condition given by  $F(T, Y_T^2, X_T, I_T) = \sqrt{\frac{I_T}{T}}$ .



# Outline

## 1 Introduction

- Motivation
- RS-SV Model

## 2 Local risk Minimization approach

- The contingent claim
- Some definition
- Local risk minimization approach
- Our Model

## 3 Pricing swap

- Variance swap
- Volatility swap

## 4 Numerical Application

To solve system of SDE of Proposition 3, we can use numerical grid using explicit scheme. For all  $k \in \mathcal{S}$  we construct a grid of size  $[0, \bar{S}] \times [0, \bar{Y}]$ . The corresponding discretization will contains  $I + 1$  nodes in  $S$  direction and  $J + 1$  nodes in  $Y$  direction. Then all partial differentiations could be stated as following:

$$\begin{aligned} \hat{v}_y(k) &= \frac{\hat{v}_{i,j+1}^n(k) - \hat{v}_{i,j-1}^n(k)}{2\Delta Y} \\ \hat{v}_{ss}(k) &= \frac{\hat{v}_{i+1,j}^n(k) - 2\hat{v}_{i,j}^n(k) + \hat{v}_{i-1,j}^n(k)}{(\Delta S)^2} \\ \hat{v}_{yy}(k) &= \frac{\hat{v}_{i,j+1}^n(k) - 2\hat{v}_{i,j}^n(k) + \hat{v}_{i,j-1}^n(k)}{(\Delta Y)^2} \\ \hat{v}_{sy}(k) &= \frac{\hat{v}_{i+1,j+1}^n(k) + \hat{v}_{i-1,j-1}^n(k) - \hat{v}_{i-1,j+1}^n(k) - \hat{v}_{i+1,j-1}^n(k)}{4\Delta S\Delta Y} \end{aligned}$$

and

$$\hat{v}_t(k) = \frac{\hat{v}_{i,j}^{n+1} - \hat{v}_{i,j}^n}{\Delta t}$$

# Bibliography



FOLLMER, H. AND SCHWEIZER, M. (1991), *Hedging contingent claims under incomplete information*. in M.H.A. Davis and R.J. Elliott (eds.), **Applied Stochastic Analysis**, Stochastics Monographs, 5, Gordon and Breach, 389-414.



GOUTTE, S., OUDJANE, N. AND RUSSO, F. (2011), *Variance optimal hedging for continuous time additive processes and applications*. Preprint.



GOUTTE, S., OUDJANE, N. AND RUSSO, F. (2011), *Variance Optimal Hedging for discrete time processes with independent increments. Application to Electricity Markets*, To Appear in **Journal of Computational Finance**.



SCHWEIZER, M. (1991), *Option Hedging for Semimartingales*, **Stochastic Processes and their Applications** 37, 339-363



SCHWEIZER, M. (1994). *Approximating random variables by stochastic integrals*. **The Annals of Probability** Vol. 22, 1536-1575.



SCHWEIZER, M., (1995), *On the minimal martingale measure and the Follmer-Schweizer decomposition*, **Stochastic Analysis and Applications**, 13, 573-599.