

# Modeling and Pricing of Variance Swaps for Local Stochastic Volatilities with Delay and Jumps\*

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## Outline of Presentation

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2. Motivation: Why Delay and Jumps?
3. Modeling of Local SV with Delay and Jumps (LSVDJ)
4. Pricing of Variance Swaps for LSVDJ
5. Delay as a Measure of Risk
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## Introduction: Stock Price

$$dS(t) = \mu S(t)dt + \sigma(t, S_t)S(t)dW(t), \quad t > 0,$$

where  $\mu \in R$  is the mean rate of return, the volatility term  $\sigma > 0$  is a bounded function and  $W(t)$  is a Brownian motion on a probability space  $(\Omega, \mathcal{F}, P)$  with a filtration  $\mathcal{F}_t$ . We also let  $r > 0$  be the risk-free rate of return of the market. We denote  $S_t = S(t - \tau)$ ,  $t > 0$  and the initial data of  $S(t)$  is defined by  $S(t) = \varphi(t)$ , where  $\varphi(t)$  is a deterministic function with  $t \in [-\tau, 0]$ ,  $\tau > 0$ .

## Introduction: Stochastic Volatility with Delay and Jumps

$$\begin{aligned} \frac{d\sigma^2(t, S_t)}{dt} &= \gamma V + \frac{\alpha}{\tau} \left[ \int_{t-\tau}^t \sigma(u, S_u) dW(u) + \int_{t-\tau}^t \sigma(u, S_u) d\tilde{N}(u) \right]^2 \\ &- (\alpha + \gamma) \sigma^2(t, S_t) \end{aligned}$$

where  $N(t)$  is a Poisson process independent of  $W(t)$  with intensity  $\lambda > 0$  and  $\tilde{N}(t) := N(t) - \lambda t$ .

Here,  $V > 0$  is a mean-reverting level (or long-term equilibrium of  $\sigma^2(t, S_t)$ ),  $\alpha, \gamma > 0$ , and  $\alpha + \gamma < 1$ .

## Introduction: Stochastic Volatility with Delay and Jumps

$$\begin{aligned} \frac{d\sigma^2(t, S_t)}{dt} &= \gamma V + \frac{\alpha}{\tau} \left[ \int_{t-\tau}^t \sigma(u, S_u) dW(u) + \int_{t-\tau}^t \sigma(u, S_u) d\tilde{N}(u) \right]^2 \\ &\quad - (\alpha + \gamma) \sigma^2(t, S_t) \end{aligned}$$

We suppose that  $\gamma > \alpha\lambda$ .

We note that if  $\gamma > \alpha\lambda$ , then the process  $\sigma^2(t, S_t)$  remains positive a.s. It is similar to the CIR process, which remains positive if all the coefficients are positive and satisfy some inequality. We also note that in many applications this condition  $\gamma > \alpha\lambda$  is satisfied (see our Numerical Examples).

## **Introduction: Stochastic Volatility with Delay and Jumps (cntd)**

Our model of stochastic volatility exhibits jumps and also past-dependence: the behavior of a stock price right after a given time  $t$  not only depends on the situation at  $t$ , but also on the whole past (history) of the process  $S(t)$  up to time  $t$ .

## **Introduction: Stochastic Volatility with Delay and Jumps (cntd)**

This draws some similarities with fractional Brownian motion models (see Mandelbrot (1997)) due to a long-range dependence property.

Another advantage of this model is mean-reversion.

This model is also a continuous-time version of GARCH(1,1) model (see Bollerslev (1986)) with jumps.

## Motivation: Why Delay?

Some statistical studies of stock prices indicate the dependence on past returns:

- Sheinkman and LeBaron (1989),
- Akgiray (1989)
- Kind, Liptser and Runggaldier (1991)

## **Motivation: Why Delay? (cntd)**

- Hobson and Rogers (1998)
- Chang and Yoree (1999)
- Mohammed, Arriojas and Pap (2001)

## Motivation: Why Delay? (cntd)

Our work is also based on the GARCH(1,1) model (see Bollerslev (1986))

$$\sigma_n^2 = \gamma V + \alpha \ln^2(S_{n-1}/S_{n-2}) + (1 - \alpha - \gamma)\sigma_{n-1}^2$$

or, more general,

$$\sigma_n^2 = \gamma V + \frac{\alpha}{l} \ln^2(S_{n-1}/S_{n-1-l}) + (1 - \alpha - \gamma)\sigma_{n-1}^2$$

and the work of Duan (1995) where he showed that it is possible to use the GARCH model as the basis for an internally consistent option pricing model.

## Motivation: Why Delay? (cntd)

If we write down the last equation in differential form we can get the continuous-time GARCH with expectation of log-returns of zero:

$$\frac{d\sigma^2(t)}{dt} = \gamma V + \frac{\alpha}{\tau} \ln^2\left(\frac{S(t)}{S(t-\tau)}\right) - (\alpha + \gamma)\sigma^2(t)$$

If we incorporate non-zero expectation of log-return (using Itô Lemma for  $\ln \frac{S(t)}{S(t-\tau)}$ ) then we arrive to our continuous-time GARCH model for stochastic volatility with delay:

$$\frac{d\sigma^2(t, S_t)}{dt} = \gamma V + \frac{\alpha}{\tau} \left[ \int_{t-\tau}^t \sigma(s, S_s) dW(s) \right]^2 - (\alpha + \gamma)\sigma^2(t, S_t).$$

## **Motivation: Why Jumps?**

The key risk factors considered in option pricing models, besides the diffusive price risk of the underlying asset, are stochastic volatility and jumps, both in the asset price and its volatility. Models that include some or all of these factors were developed by Merton (1976), Heston (1993), Bates (1996), Bakshi et al. (1997) and Duffie et al. (2000).

## **Motivation: Why Jumps? (cntd)**

The jumps in stock market volatility are found to be so active that this discredits many recently proposed stochastic volatility models without jumps (see Todorov and Tauchen (2008), Bollerslev *et al* (2008)).

## **Motivation: Why Jumps?**

There is currently fairly compelling evidence for jumps in the level of financial prices. The most convincing evidence comes from recent nonparametric work using high-frequency data as in Barndorff-Nielsen and Shephard (2007) and Ait-Sahalia and Jacod (2008) among others.

## Motivation: Why Jumps? (contd)

Paper by Todorov and Tauchen (2008) conducts a non-parametric analysis of the market volatility dynamics using high-frequency data on the VIX index compiled by the CBOE and the *S&P500* index. Some attempts have been made to incorporate jumps in stochastic volatility to price variance and volatility swaps (see Howison *et al.* (2004)).

## **Motivation: Why Jumps? (cntd)**

Eraker *et al.* (2000) use returns data to investigate the performance of models with jumps in volatility using the class of jump-in-volatility models proposed by Duffie *et al.* (2000). The results in Eraker *et al.* (2000) show that the jump-in-volatility models provide a significant better fit to the returns data.

## The Model

Stock Price:

$$dS(t) = \mu S(t)dt + \sigma(t, S_t)S(t)dW(t), \quad t > 0,$$

SV with Delay and Jumps:

$$\begin{aligned} \frac{d\sigma^2(t, S_t)}{dt} &= \gamma V + \frac{\alpha}{\tau} \left[ \int_{t-\tau}^t \sigma(u, S_u) dW(u) + \int_{t-\tau}^t \sigma(u, S_u) d\tilde{N}(u) \right]^2 \\ &- (\alpha + \gamma) \sigma^2(t, S_t) \end{aligned}$$

## Conditions

C1)  $\sigma(t, S_t)$  satisfies local Lipschitz and growth conditions;

C2)  $\int_0^T E\sigma^2(t, S_t)dt < +\infty$ ;

C3)  $\int_0^T \left(\frac{r-\mu}{\sigma(t, S_t)}\right)^2 dt < +\infty$  a.s.

Condition C1) guarantees the existence and uniqueness of a solution of equations for  $S(t)$  and  $\sigma^2(t)$  in Section 2 (see Mohammed (1998)). Condition C2) guarantees the existence of Itô integral in equation for  $\sigma^2(t)$  and C3) guarantees the existence of risk-neutral measure  $P^*$  (see below).

## Risk-Neutral World

1) There is a probability measure  $\mathbb{P}^*$  equivalent to  $\mathbb{P}$  such that

$$\frac{d\mathbb{P}^*}{d\mathbb{P}} = \exp \left\{ - \int_0^T \theta(s) dW(s) - \frac{1}{2} \int_0^T \theta^2(s) ds \right\}$$

is its Radon-Nikodym density, where

$$\theta(t) = \frac{\mu - r}{\sigma(t, S_t)}.$$

## Risk-Neutral World (cntd)

2) The discounted asset price  $D(t)$  is a positive local martingale with respect to  $\mathbb{P}^*$ , and

$$W^*(t) = \int_0^t \theta(s) ds + W(t)$$

is a standard Brownian motion with respect to  $\mathbb{P}^*$ .

## Risk-Neutral World (cntd)

$$dS(t) = rS(t)dt + \sigma(t, S_t)S(t)dW^*(t)$$

and the asset volatility is defined then as follows:

$$\frac{d\sigma^2(t, S_t)}{dt} = \gamma V + \frac{\alpha}{\tau} \left[ \int_{t-\tau}^t \sigma(s, S_s) dW^*(s) + \int_{t-\tau}^t \sigma(u, S_u) d\tilde{N}(u) - (\mu - r)\tau \right]^2 - (\alpha + \gamma)\sigma^2(t, S_t).$$

## Variance Swaps

Variance swaps are forward contracts on future realized stock variance, the square of the future volatility. The easy way to trade variance is to use variance swaps, sometimes called realized variance forward contracts (see Carr and Madan (1998)). Although options market participants talk of volatility, it is variance, or volatility squared, that has more fundamental significance (see Demeterfi, K., Derman, E., Kamal, M., and Zou, J. (1999)).

## Variance Swaps (cntd)

A *variance swap* is a forward contract on annualized variance, the square of the realized volatility. Its payoff at expiration is equal to

$$N(\sigma_R^2(S) - K_{var}),$$

where  $\sigma_R^2(S)$  is the realized stock variance(quoted in annual terms) over the life of the contract,

## Variance Swaps (cntd)

$$\sigma_R^2(S) := \frac{1}{T} \int_0^T \sigma^2(s) ds,$$

$K_{var}$  is the delivery price for variance, and  $N$  is the notional amount of the swap in dollars per annualized volatility point squared.

## Variance Swaps (cntd)

The holder of variance swap at expiration receives  $N$  dollars for every point by which the stock's realized variance  $\sigma_R^2(S)$  has exceeded the variance delivery price  $K_{var}$ . We note that usually  $N = \alpha I$ , where  $\alpha$  is a converting parameter such as 1 per volatility-square, and  $I$  is a long-short index (+1 for long and -1 for short).

## Variance Swaps (cntd)

Valuing a variance forward contract or swap is no different from valuing any other derivative security. The value of a forward contract  $P$  on future realized variance with strike price  $K_{var}$  is the expected present value of the future payoff in the risk-neutral world:

$$\mathcal{P}^* = E^* \{ e^{-rT} (\sigma_R^2(S) - K_{var}) \},$$

where  $r$  is the risk-free discount rate corresponding to the expiration date  $T$ , and  $E^*$  denotes the expectation under the risk-neutral measure  $P^*$ .

## Variance Swaps (cntd)

In tis way, a *variance swap for stochastic volatility with delay* is a forward contract on annualized variance  $\sigma_R^2(t, S_t)$ . Its payoff at expiration equals to

$$N(\sigma_R^2(S) - K_{var}),$$

where  $\sigma_R^2(S)$  is the realized stock variance(quoted in annual terms) over the life of the contract,

$$\sigma_R^2(S) := \frac{1}{T} \int_0^T \sigma^2(u, S(u - \tau)) du, \quad \tau > 0.$$

## Pricing of Variance Swaps (cntd)

$$\frac{d\sigma^2(t, S_t)}{dt} = \gamma V + \frac{\alpha}{\tau} \left[ \int_{t-\tau}^t \sigma(s, S_s) dW^*(s) + \int_{t-\tau}^t \sigma(u, S_u) d\tilde{N}(u) - (\mu - r)\tau \right]^2 - (\alpha + \gamma)\sigma^2(t, S_t).$$

Let us take the expectations under risk-neutral measure  $\mathbb{P}^*$  on the both sides of the equation above.

## Pricing of Variance Swaps (cntd)

Denoting  $v(t) = \mathbb{E}^*[\sigma^2(t, S_t)]$ , we obtain the following deterministic delay differential equation:

$$\frac{dv(t)}{dt} = \gamma V + \alpha\tau(\mu - r)^2 + \frac{\alpha(1 + \lambda)}{\tau} \int_{t-\tau}^t v(s)ds - (\alpha + \gamma)v(t).$$

## Pricing of Variance Swaps (cntd)

Notice that the above equation has a stationary solution

$$v(t) \equiv X = \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}.$$

## Pricing of Variance Swaps (cntd)

Hence, the expectation of the realized variance, or say the fair delivery price  $K_{var}$  of a variance swap for stochastic volatility with delay in stationary regime under risk-neutral measure  $\mathbb{P}^*$  equals to

$$\begin{aligned} K_{var} &= \mathbb{E}^*[\sigma^2(t, S_t)] = \frac{1}{T} \int_0^T v(t) dt \\ &= \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}. \end{aligned}$$

## Pricing of Variance Swaps (cntd)

The price  $P$  of a variance swap at time  $t$  given delivery price  $K$  in this case should be:

$$P = e^{-r(T-t)} \left[ \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} - K \right].$$

## Pricing of Variance Swaps (cntd)

$$\frac{dv(t)}{dt} = \gamma V + \alpha \tau (\mu - r)^2 + \frac{\alpha(1 + \lambda)}{\tau} \int_{t-\tau}^t v(s) ds - (\alpha + \gamma)v(t).$$

In general case, there is no way to write a solution of this equation in explicit form for arbitrarily given initial data.

## Pricing of Variance Swaps (cntd)

But we can write an approximate solution for  $v(t)$  when  $t$  has large values:

$$v(t) \approx X + Ce^{-\gamma t} = \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} + Ce^{(\alpha \lambda - \gamma)t}.$$

where

$$C = v(0) - X = \sigma_0^2 - \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}.$$

## Pricing of Variance Swaps (cntd)

We note, that the characteristic equation for the above equation in this case has the following look

$$\rho^2 + \rho(\gamma - \alpha\lambda) = 0$$

and the solution of the equation is

$$\rho = (\alpha\lambda - \gamma).$$

## Pricing of Variance Swaps (cntd)

Hence, the expectation of the realized variance, or say the fair delivery price  $K_{var}$  of *variance swap for stochastic volatility with delay in general case under risk-neutral measure*  $\mathbb{P}^*$  equals to

$$\begin{aligned} K_{var} &= \mathbb{E}^*[\sigma^2(t, S_t)] = \frac{1}{T} \int_0^T v(t) dt \\ &\approx \frac{1}{T} \int_0^T [V + \alpha\tau(\mu - r)^2/\gamma + (\sigma_0^2 - V - \alpha\tau(\mu - r)^2/\gamma)e^{(\alpha\lambda - \gamma)t}] dt \\ &= \frac{\gamma V + \alpha\tau(\mu - r)^2}{\gamma - \alpha\lambda} + (\sigma_0^2 - \frac{\gamma V + \alpha\tau(\mu - r)^2}{\gamma - \alpha\lambda}) \frac{e^{(\alpha\lambda - \gamma)T} - 1}{T(\alpha\lambda - \gamma)}. \end{aligned}$$

## Pricing of Variance Swaps (cntd)

The price  $P$  of a variance swap at time  $t$  given delivery price  $K$  in this case should be:

$$P \approx e^{-r(T-t)} \left[ \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} + \left( \sigma_0^2 - \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} \right) \frac{e^{(\alpha \lambda - \gamma)T} - 1}{T(\alpha \lambda - \gamma)} - K \right].$$

## Delay as a Measure of Risk

By previous results we have:

$$\begin{aligned}\mathbb{E}^*[\sigma^2(t, S_t)] &= \frac{1}{T} \int_0^T v(t) dt \\ &\approx \frac{1}{T} \int_0^T [V + \alpha\tau(\mu - r)^2/\gamma + (\sigma_0^2 - V - \alpha\tau(\mu - r)^2/\gamma) \\ &\quad \times e^{(\alpha\lambda - \gamma)t}] dt \\ &= \frac{\gamma V + \alpha\tau(\mu - r)^2}{\gamma - \alpha\lambda} + (\sigma_0^2 - \frac{\gamma V + \alpha\tau(\mu - r)^2}{\gamma - \alpha\lambda}) \frac{e^{(\alpha\lambda - \gamma)T} - 1}{T(\alpha\lambda - \gamma)}.\end{aligned}$$

## Delay as a Measure of Risk (contd)

$$\mathbb{E}^*[\sigma^2(t, S_t)] = \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} + \left( \sigma_0^2 - \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} \right) \frac{e^{(\alpha \lambda - \gamma)T} - 1}{T(\alpha \lambda - \gamma)}.$$

This expression contains all the information about our model, since it contains all the initial parameters.

We note that  $\sigma_0^2 = \sigma^2(0, \varphi(-\tau))$ .

So the sign of the second term in the above expression depends on the relationship between  $\sigma_0^2$  and  $\frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}$ .

## Delay as a Measure of Risk (contd)

$$\mathbb{E}^*[\sigma^2(t, S_t)] = \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} + \left( \sigma_0^2 - \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} \right) \frac{e^{(\alpha \lambda - \gamma)T} - 1}{T(\alpha \lambda - \gamma)}.$$

If  $\sigma_0^2 > \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}$ , the second term is positive and  $\mathbb{E}^*[\sigma^2(t, S_t)]$  stays above  $\frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}$ , which means the risk is high.

## Delay as a Measure of Risk (contd)

$$\mathbb{E}^*[\sigma^2(t, S_t)] = \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} + \left( \sigma_0^2 - \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} \right) \frac{e^{(\alpha \lambda - \gamma)T} - 1}{T(\alpha \lambda - \gamma)}.$$

If

$$\sigma_0^2 < \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda},$$

the second term is negative and  $\mathbb{E}^*[\sigma^2(t, S_t)]$  stays below  $\frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}$ , which means the risk is low.

## Delay as a Measure of Risk (contd)

Therefore,

$$\sigma_0^2 = \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}$$

defines the measure of risk in the stochastic volatility model with delay and jumps.

## Delay as a Measure of Risk (contd)

If

$$\sigma_0^2 = \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda}$$

is satisfied, then the value of risk is the following constant

$$\frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} = \sigma_0^2.$$

## Delay as a Measure of Risk (contd)

To reduce and control the risk we need to take into account the following relationship with respect to the delay  $\tau$  :

$$\sigma_0^2 < \frac{\gamma V + \alpha \tau (\mu - r)^2}{\gamma - \alpha \lambda} \quad \text{or} \quad \tau > \frac{\sigma_0^2 (\gamma - \alpha \lambda) - \gamma V}{\alpha (\mu - r)^2}.$$

It follows that if delay is larger (or longer) of this value  $\frac{\sigma_0^2 (\gamma - \alpha \lambda) - \gamma V}{\alpha (\mu - r)^2}$ , then the risk is lower.

## Delay as a Measure of Risk (contd)

Also, there is a way to control the delay but paying a higher risk. If we bound the delay by the following relation

$$\tau < \frac{\sigma_0^2(\gamma - \alpha\lambda) - \gamma V}{\alpha(\mu - r)^2},$$

then

$$\sigma_0^2 > \frac{\gamma V + \alpha\tau(\mu - r)^2}{\gamma - \alpha\lambda}.$$

It follows that if delay is smaller (shorter) of the value  $\frac{\sigma_0^2(\gamma - \alpha\lambda) - \gamma V}{\alpha(\mu - r)^2}$ , then the risk is higher.

## Numerical Example 1: *S&P60* Canada Index

<b>Statistics on Log Returns <i>S&amp;P60</i> Canada Index</b>	
<b>Series:</b>	Log Returns <i>S&amp;P60</i> Canada Index
<b>Sample:</b>	1 1300
<b>Observations:</b>	1300
Mean	0.000235
Median	0.000593
Maximum	0.051983
Minimum	-0.101108
Std. Dev.	0.013567
Skewness	-0.665741
Kurtosis	7.787327

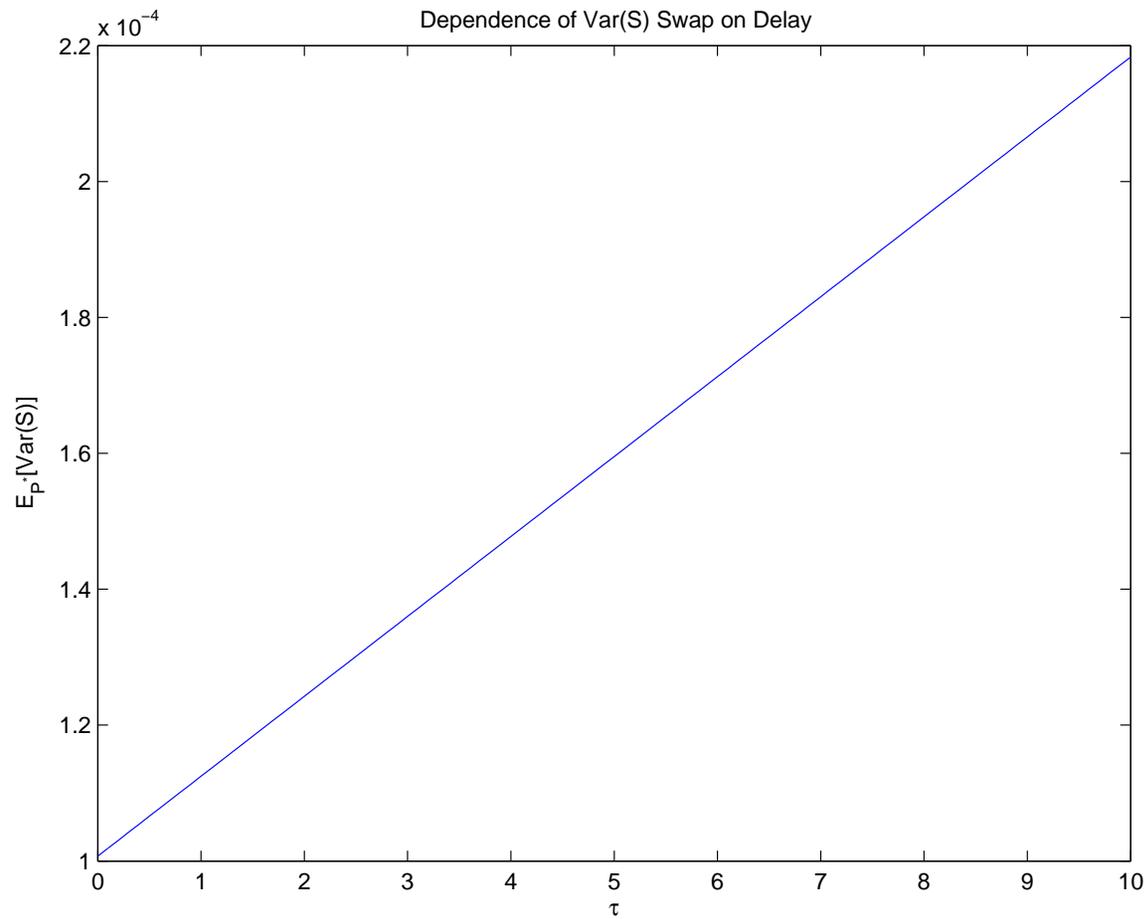


Fig. 1. Dependence of Variance Swap with Delay on Delay (*S&P60* Canada Index).

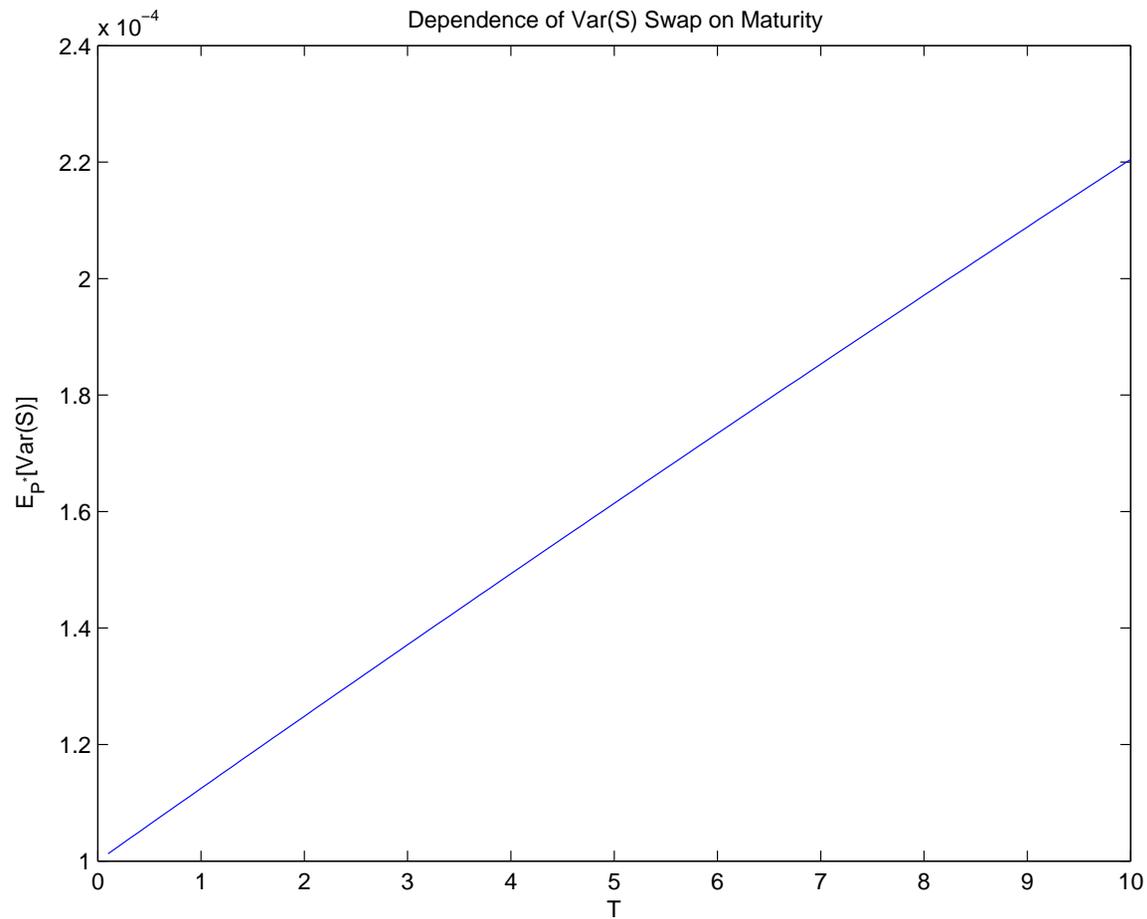


Fig. 2. Dependence of Variance Swap with Delay on Maturity (*S&P60* Canada Index).

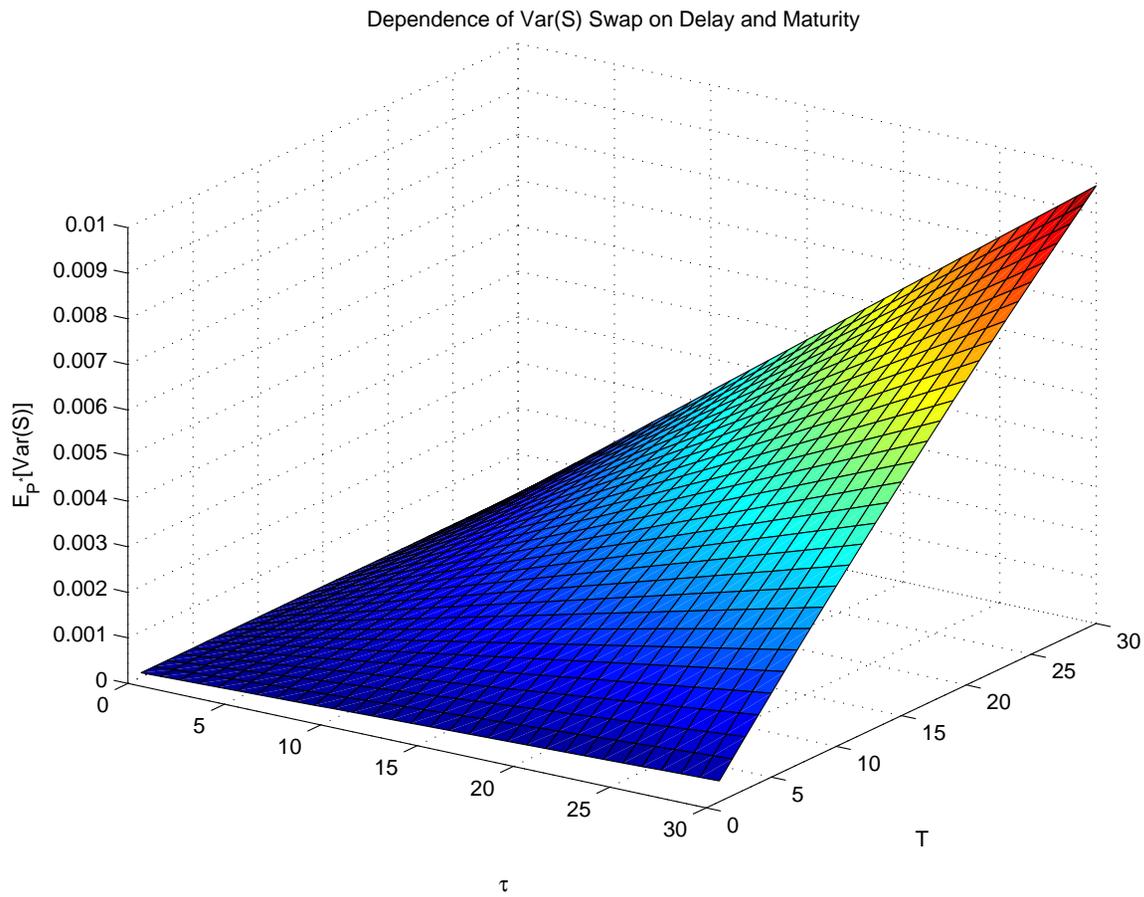


Fig. 3. Variance Swap with Delay for *S&P60* Canada Index.

## Numerical Example 2: *S&P500* Index

<b>Statistics on Log Returns <i>S&amp;P500</i> Index</b>	
<b>Series:</b>	Log returns <i>S&amp;P500</i> Index
<b>Sample:</b>	1 1006
<b>Observations:</b>	1006
Mean	0.000263014
Median	8.84424E-05
Maximum	0.034025839
Minimum	-0.045371484
Std. Dev.	0.00796645
Variance	6.34643E-05
Skewness	-0.178481359
Kurtosis	3.296144083

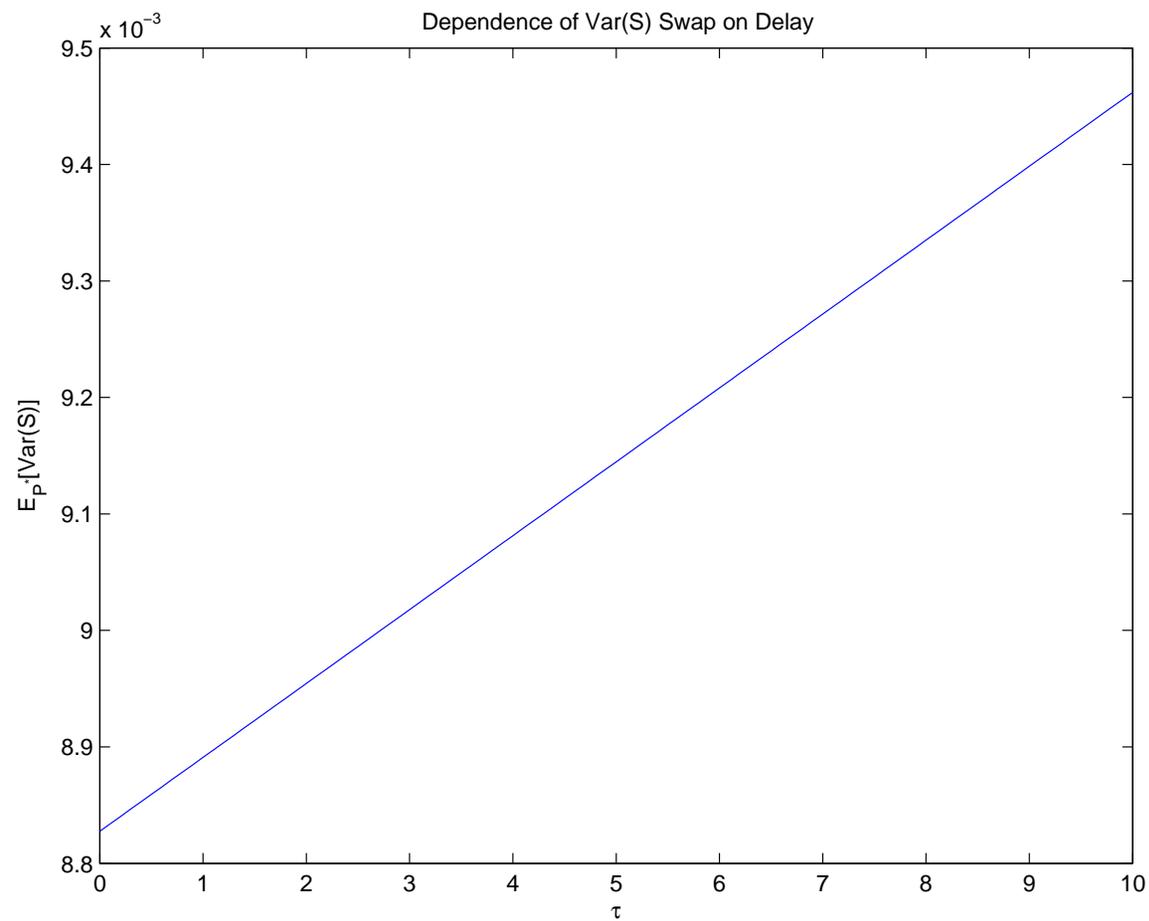


Fig. 4. Dependence of Variance Swap with Delay on Delay (*S&P500* Index).

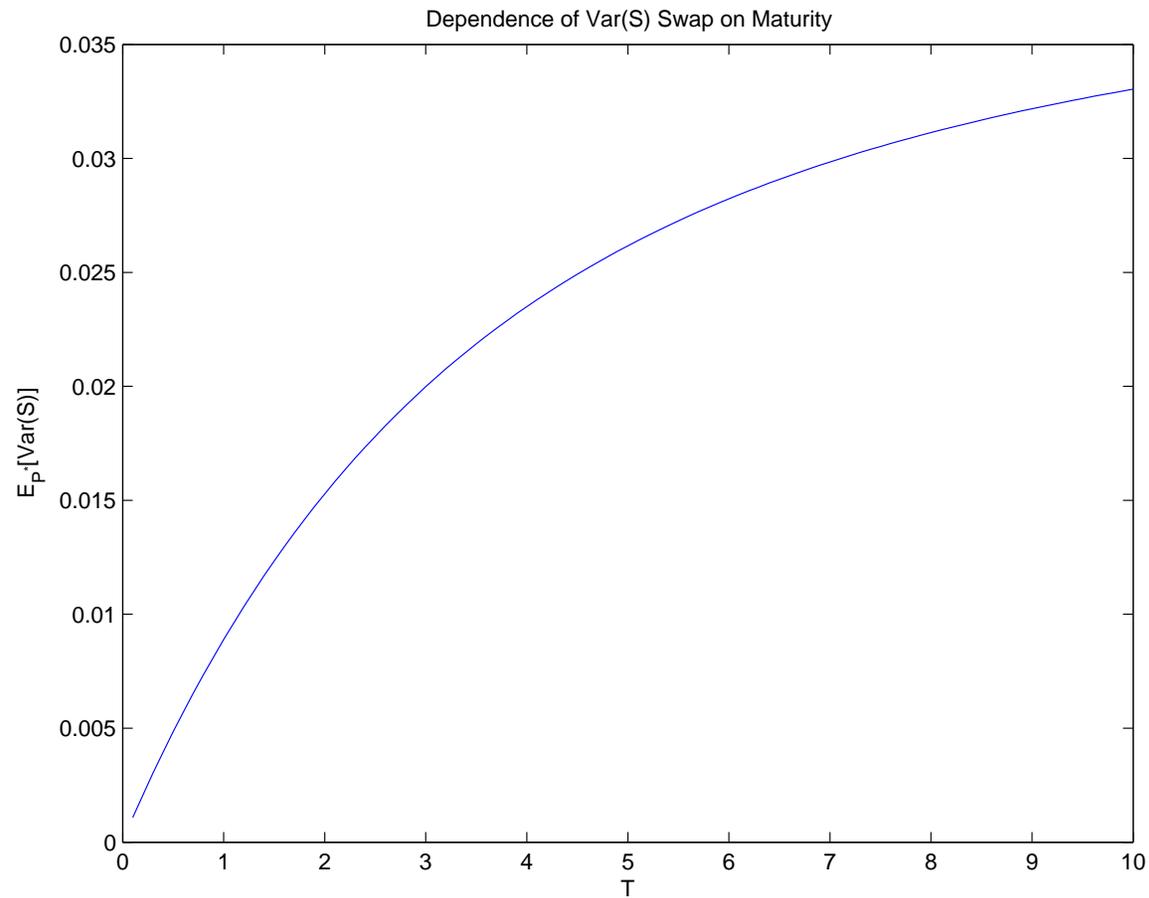


Fig. 5. Dependence of Variance Swap with Delay on Maturity (*S&P500* Index).

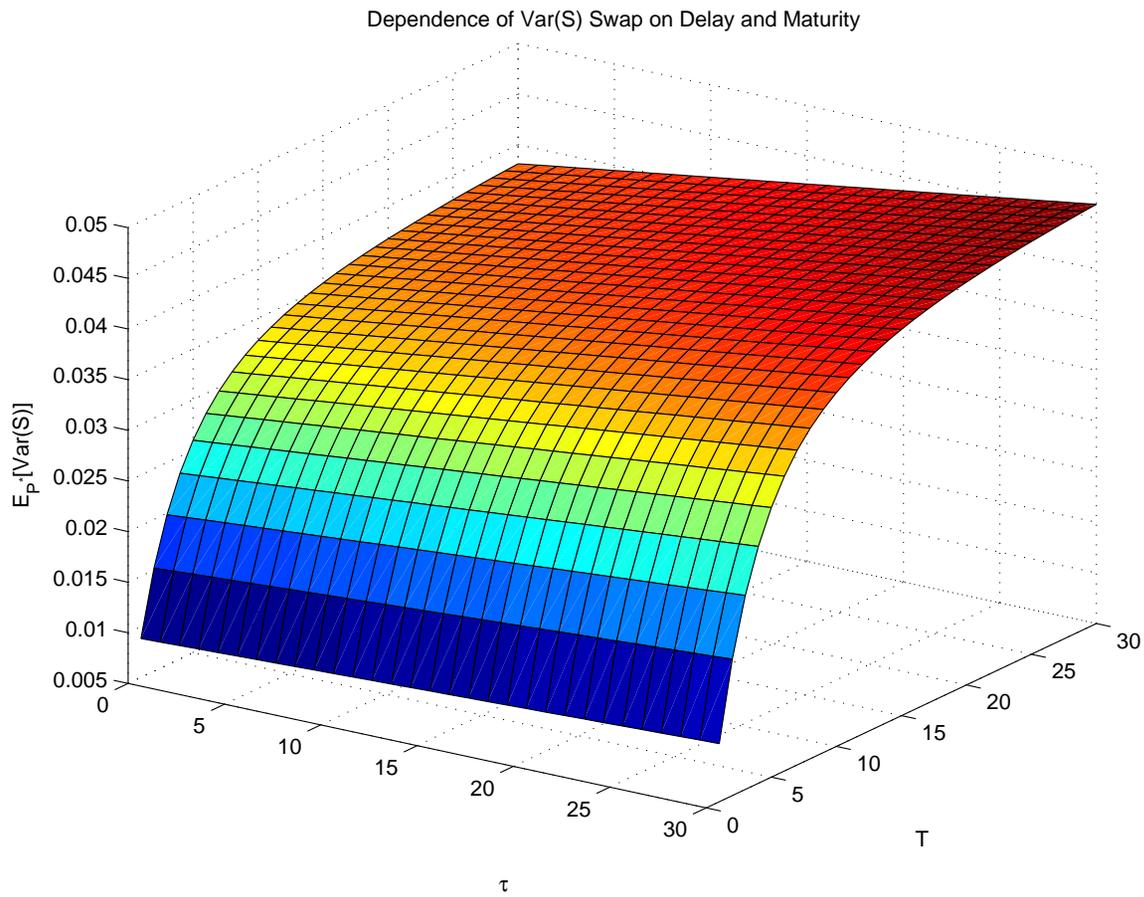


Fig. 6. Variance Swap with Delay for  
*S&P500* Index.

## Conclusion

1. The Model of Stock Price with Local SV with Delay and Jumps (LSVDJ)
2. Motivation: Why Delay and Jumps?
3. Pricing of Variance Swaps for LSVDJ
4. Delay as a Measure of Risk
5. Numerical Examples: *S&P60* Canada and *S&P500* Indexes

**The End**

Thank you for your time and attention!



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