

Variance swaps

Section 5, *Stochastic Volatility Modelling*, L. Bergomi

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Understanding Variance Swaps: Synthesis, Replication, and Implications

Payoff :

$$\frac{1}{T-t} \sum_{i=0}^{N-1} \ln^2 \left(\frac{S_{i+1}}{S_i} \right) - \hat{\sigma}_{\text{VS},T}^2(t)$$

with:

- t : initial time
- T : contract's maturity
- S_i asset closed price observed at time t_i

$\hat{\sigma}_{\text{VS},T}^2(t)$ is chosen such that the value of the swap at the initial time is zero
(realized volatility estimator)

Synthesis of Variance Swaps

Definition: Variance swaps → financial derivatives to trade volatility directly, no directional risk.

Payoff = (realized - implied)*volatility over a specified period.

Synthesis of VS with European payoffs using Carr-Madan formula on log contracts.

Replicate the payoff of a variance swap by constructing a portfolio of European options with different strikes & weights.

Replication -> exposure to volatility and managing risk more effectively exposure to volatility.

$$\hat{\sigma}_{\text{VS},T}^2 = \frac{e^{rT}}{T} \int_0^\infty \frac{2}{K^2} (P_{\text{market}}^{KT} - P_{\hat{\sigma}=0}^{KT}) dK \qquad P_{\hat{\sigma}=0}^{KT} = e^{-rT} \left(S e^{(r-q)T} - K \right)^+$$

Discrete Forward Variance

Strategy considered at time t :

- Purchase of $(T_2 - t)$ VS of maturity T_2
- Sell of $(T_1 - t) e^{-r(T_2 - T_1)}$ VS of maturity T_1

Payoff in T_2 :

$$\begin{aligned} & \sum_{T_1}^{T_2} \ln^2 \left(\frac{S_{i+1}}{S_i} \right) - ((T_2 - t) \hat{\sigma}_{\text{VS}, T_2}^2(t) - (T_1 - t) \hat{\sigma}_{\text{VS}, T_1}^2(t)) \\ &= \sum_{T_1}^{T_2} \ln^2 \left(\frac{S_{i+1}}{S_i} \right) - (T_2 - T_1) \hat{\sigma}_{\text{VS}, T_1 T_2}^2(t) \end{aligned}$$

Continuous and Discrete Forward Variance

Discrete Forward Variance :

$$\hat{\sigma}_{\text{VS}, T_1 T_2}^2(t) = \frac{(T_2 - t) \hat{\sigma}_{\text{VS}, T_2}^2(t) - (T_1 - t) \hat{\sigma}_{\text{VS}, T_1}^2(t)}{T_2 - T_1}$$

→ Always positive by construction (well defined)

Continuous Forward Variance :

$$\frac{d}{dT} ((T - t) \hat{\sigma}_{\text{VS}, T}^2(t))$$

More interesting strategy

Two steps strategy :

- Same strategy at time t
- Opposite strategy at time t'

Payoff in T_2 :

$$(T_2 - T_1) \left(\hat{\sigma}_{VS, T_1 T_2}^2 (t') - \hat{\sigma}_{VS, T_1 T_2}^2 (t) \right)$$

→ Initial stake of 0 (implies P&L with no dependence in r)

→ No More dependence on the realized variance of S

→ Linear P&L in the variation of $\hat{\sigma}_{VS, T_1 T_2}^2$

Relationship between Variance Swap and log-contract

Reminder : payoff of variance swap : $\frac{1}{T-t} \sum_{i=0}^{N-1} \ln^2 \left(\frac{S_{i+1}}{S_i} \right) - \hat{\sigma}_{\text{VS},T}^2(t)$

First observation:

$$\begin{aligned} e^{-rT} \sum_{i=0}^{N-1} \ln^2 \left(\frac{S_{i+1}}{S_i} \right) &= e^{-rT} \sum_{i=0}^{N-1} \ln^2 \left(1 + \frac{S_{i+1} - S_i}{S_i} \right) \simeq e^{-rT} \sum_{i=0}^{N-1} \left(\frac{S_{i+1} - S_i}{S_i} \right)^2 = e^{-rT} \sum_{i=0}^{N-1} \left(\frac{\delta S_i}{S_i} \right)^2 \\ &= \sum_{i=0}^{N-1} e^{-rt_i} e^{-r(T-t_i)} \left(\frac{\delta S_i}{S_i} \right)^2 \end{aligned}$$

Reminder: P&L theta gamma :

$$P\&L = - \sum_i e^{-rt_i} \frac{S_i^2}{2} \frac{d^2 P_{\hat{\sigma}}}{dS^2}(t_i, S_i) (r_i^2 - \hat{\sigma}^2 \delta t) \quad \text{avec} \quad r_i = \left(\frac{\delta S_i}{S_i} \right)^2$$

Relationship between Variance Swap and log-contract

→ Same formula with $\frac{1}{2}S^2 \frac{d^2 P_{\hat{\sigma}=0}}{dS^2} = e^{-r(T-t)}$

→ Condition satisfied by the log-contract $Q^T(t, S) = -2e^{-r(T-t)} \left(\ln S + (r - q)(T - t) - \frac{\hat{\sigma}^2}{2}(T - t) \right)$

with $\hat{\sigma} = 0$:

$$Q_{\hat{\sigma}=0}^T(t, S) = -2e^{-r(T-t)} (\ln S + (r - q)(T - t))$$

because $\frac{dQ_{\hat{\sigma}=0}^T}{dS} = -e^{-r(T-t)} \frac{2}{S}$ and $\frac{1}{2}S^2 \frac{d^2 Q_{\hat{\sigma}=0}^T}{dS^2} = e^{-r(T-t)}$

Let's go a step further...

→ Buying $Q_{\hat{\sigma}=0}^T$ generates a mark-to-market of the form $-(Q_{\text{market}}^T - Q_{\hat{\sigma}=0}^T)$

therefore :

$$\hat{\sigma}_{\text{VS},T}^2 = \frac{e^{rT}}{T} (Q_{\text{market}}^T - Q_{\hat{\sigma}=0}^T)$$

Note the $\hat{\sigma}_T$ implicit volatility of the log-contract :

$$Q_{\text{market}}^T = -2e^{-r(T-t)} (\ln S + (r - q)(T - t) - \frac{\hat{\sigma}_T^2}{2}(T - t))$$

This gives: with the previous formula $\hat{\sigma}_{\text{VS},T} = \hat{\sigma}_T$

Let's go a step further...

Now, as the log-contract is replicated by :

$$-2 \ln S = -2 \ln S_0 - \frac{2}{S_0} (S - S_0) + \int_0^{S_0} \frac{2}{K^2} (K - S)^+ dK + \int_{S_0}^{\infty} \frac{2}{K^2} (S - K)^+ dK$$

(comes from section 3 of the book where Bergomi shown that

$$f(S) = f(K_0) + \left. \frac{df}{dK} \right|_{K_0} (S - K_0) + \int_0^{K_0} \frac{d^2 f}{dK^2} (K - S)^+ dK + \int_{K_0}^{\infty} \frac{d^2 f}{dK^2} (S - K)^+ dK \quad \text{with } f(S) = -2 \ln S$$

→ Payoff replicated (order 2) with delta-hedge of this portfolio (with zero implied vol)

Impact of Large Returns

→ We have done a Taylor expansion of order 2 in $\frac{\delta S}{S}$

→ Remember that we have : $\frac{1}{2} S^2 \frac{d^2 P_{\bar{\sigma}=0}}{dS^2} = e^{-r(T-t)}$

→ Suppose that S follows : $dS_t = (r - q)S_t dt + \bar{\sigma}_t S_t dW_t$

→ But one knows that, in a local volatility model : $P_{\bar{\sigma}}(t=0) = P_{\sigma}(0, S_0) + E_{\bar{\sigma}} \left[\int_0^T \frac{1}{2} e^{-rt} S_t^2 \frac{d^2 P_{\sigma}}{dS^2} (\bar{\sigma}_t^2 - \sigma(t, S_t)^2) dt \right]$

ROBUSTESSE DE LA FORMULE DE BLACK-SCHOLES

Ref : El Karoui, Jeanblanc, Shreve '98

Théorème. Supposons qu'un trader couvre un call avec un modèle calibré au marché à $t = 0$. Alors son P&L en T est

$$P\&L_T = V_T - (S_T - K)_+ = e^{rT} \int_0^T e^{-rt} \frac{1}{2} S_t^2 \partial_x^2 u^{\text{Model}}(t, S_t) ((\sigma^{\text{Model}})^2(t, S_t) - \sigma_t^2) dt$$

avec σ_t la vol. instantanée de S et u^{Model} la fonction de pricing utilisée.

- Moyenne trajectorielle entre la différence des variances calibrées et réalisées pondérée par les gammas
- Intérêt d'un marché de variance swaps pour couvrir le risque de modèle

Remarque (Propagation de convexité). Dans un modèle à volatilité locale, u hérite de la convexité en x du payoff $\Rightarrow \partial_x^2 u(t, S_t) \geq 0$

Impact of Large Returns

By setting $\sigma(t, S) \equiv 0$ we have :

$$P_{\bar{\sigma}}^T = P_{\hat{\sigma}=0}^T + e^{-rT} E_{\bar{\sigma}} \left[\int_0^T \bar{\sigma}_t^2 dt \right]$$

Let's go back to our variance swap... We have :

$$d \ln S_t = (r - q - \frac{1}{2} \bar{\sigma}_t^2) dt + \bar{\sigma}_t dW_t$$

So $\lim_{dt \rightarrow 0} \frac{1}{dt} (d \ln S_t)^2 = \bar{\sigma}_t^2$ and $\lim_{\Delta t \rightarrow 0} \sum_{i=0}^{N-1} \ln^2 \left(\frac{S_{i+1}}{S_i} \right) = \int_0^T \bar{\sigma}_t^2 dt$

Impact of Large Returns

As a consequence, we have :

$$\hat{\sigma}_{\text{VS},T}^2 = \frac{1}{T} E \left[\lim_{\Delta t \rightarrow 0} \sum_{i=0}^{N-1} \ln^2 \left(\frac{S_{i+1}}{S_i} \right) \right] = E \left[\frac{1}{T} \int_0^T \bar{\sigma}_t^2 dt \right]$$

Assuming that our model is calibrated to the market smile $P_{\bar{\sigma}}^T = P_{\text{Market}}^T$, we get :

$$\hat{\sigma}_{\text{VS},T}^2 = E_{\bar{\sigma}} \left[\frac{1}{T} \int_0^T \bar{\sigma}_t^2 dt \right] = \frac{e^{rT}}{T} (P_{\text{Market}}^T - P_{\hat{\sigma}=0}^T)$$

→ All diffusive models price variance swaps identically

Impact of Large Returns

Now remember that :

$$P_{\bar{\sigma}}(t=0) = P_{\sigma}(0, S_0) + E_{\bar{\sigma}} \left[\int_0^T \frac{1}{2} e^{-rt} S_t^2 \frac{d^2 P_{\sigma}}{dS^2} (\bar{\sigma}_t^2 - \sigma(t, S_t)^2) dt \right]$$

By setting $\sigma(t, S) \equiv 0$ we get :

$$P_{\text{Market}}^T = P_{\hat{\sigma}}^T = P_{\hat{\sigma}=0}^T + e^{-rT} T \hat{\sigma}_T^2$$

With the previous expression $\hat{\sigma}_{\text{VS},T}^2 = E_{\bar{\sigma}} \left[\frac{1}{T} \int_0^T \bar{\sigma}_t^2 dt \right] = \frac{e^{rT}}{T} (P_{\text{Market}}^T - P_{\hat{\sigma}=0}^T)$ we have :

$$\hat{\sigma}_{\text{VS},T} = \hat{\sigma}_T$$

Impact of daily-return skewness

→ Short variance swap & long delta-hedged log-contrat with zero interest rate and repo (for simplicity)

→ Remember that : $Q^T(t, S) = -2e^{-r(T-t)} \left(\ln S + (r - q)(T - t) - \frac{\hat{\sigma}^2}{2} (T - t) \right)$

so we have : $Q^T(t, S) = -2\ln S + \hat{\sigma}^2(T - t)$

Then, we get : $Q^T(t_{i+1}, S_{i+1}) - Q^T(t_i, S_i) = 2(e^{r_i} - 1) - 2r_i - \hat{\sigma}_T^2 \Delta t$ $r_i = \ln\left(\frac{S_{i+1}}{S_i}\right)$

Impact of daily-return skewness

→ Total P&L (short VS / long delta-hedged log-contract) :

$$\begin{aligned} P\&L &= (Q^T(t_{i+1}, S_{i+1}) - Q^T(t_i, S_i)) - \frac{dQ^T}{dS}(t_i, S_i)(S_{i+1} - S_i) \\ &\quad - (r_i^2 - \hat{\sigma}_{VS,T}^2 \Delta t) \\ &= (2(e^{r_i} - 1) - 2r_i - \hat{\sigma}_T^2 \Delta t) - (r_i^2 - \hat{\sigma}_{VS,T}^2 \Delta t) \\ &= (2(e^{r_i} - 1) - 2r_i - r_i^2) - (\hat{\sigma}_T^2 - \hat{\sigma}_{VS,T}^2) \Delta t \end{aligned}$$

with $r_i = \ln\left(\frac{S_{i+1}}{S_i}\right)$

Impact of daily-return skewness

Then $P\&L \simeq \frac{r_i^3}{3} - (\hat{\sigma}_T^2 - \hat{\sigma}_{\text{vs},T}^2)\Delta t$ (Taylor expansion)

By imposing $\mathbb{E}(P\&L) = 0$ we obtain :

$$\hat{\sigma}_{\text{vs},T}^2 - \hat{\sigma}_T^2 \simeq -\frac{\langle r^3 \rangle}{3\Delta t} \simeq -\frac{s_{\Delta t}}{3}\hat{\sigma}_T^3\sqrt{\Delta t}$$

with $s_{\Delta t} = \langle r^3 \rangle / \langle r^2 \rangle^{\frac{3}{2}}$.

and $\langle r^2 \rangle = \hat{\sigma}_T^2\Delta t$

Impact of daily-return skewness

Writing: $\sigma_{\hat{V}S,T}^2 - \hat{\sigma}_T^2 \simeq 2\hat{\sigma}_T(\sigma_{\hat{V}S,T} - \hat{\sigma}_T)$

We obtain:

$$\frac{\hat{\sigma}_{VS,T}}{\hat{\sigma}_T} - 1 \simeq -\frac{s\Delta t}{6}\hat{\sigma}_T\sqrt{\Delta t}$$

$$\hat{\sigma}_{VS,T} \simeq \hat{\sigma}_T \left(1 - \frac{s\Delta t}{6}\hat{\sigma}_T\sqrt{\Delta t}\right)$$

→ The difference between $\hat{\sigma}_T$ and $\hat{\sigma}_{VS,T}$ depends on the non-Gaussian character of the daily log-returns (skewness)

Impact of strike discreteness

- Difference between VS payoff & P&L from delta-hedging a log contract.
- Neither the log nor the delta-hedged log contract is traded, and cannot be perfectly synthesized from vanilla options due to discrete strikes.
- Practical Replication: Replication of a 1-year VS using daily closing quotes of the S&P 500. Log contract \sim vanilla options with specified strike intervals.
- Coarser discretization implies less accurate replication of VS payoff
- Impact on Volatility Adjustment: Coarser discretization implies larger adjustment in volatility.

Conclusion

- Variance Swaps are replicated by delta-hedging a log contract
- The difference between $\hat{\sigma}_T$ and $\hat{\sigma}_{VS,T}$ is a measure of the implied skewness of daily returns
- Variance Swaps are highly liquid, unlike options that are very out-of-the-money: market makers extrapolate implied volatility with VS, which reinforces $\hat{\sigma}_T = \hat{\sigma}_{VS,T}$
- In reality, strikes are not continuous so there is a discretization (and a mismatch)

Conclusion

Market Liquidity Impact: Despite the absence of actively traded log contracts, variance swaps (VS) are more liquid than far out-of-the-money vanilla options in many cases, influencing their pricing dynamics.

Model Dependence: The choice of model for pricing VS, whether jump-diffusion or other, significantly affects the implied volatilities and the spread between $\hat{\sigma}_{VS,T}$ and $\hat{\sigma}_T$.

Practical Adjustments: Equations such as (5.42) provide practical adjustments to dissociate $\hat{\sigma}_{VS,T}$ from $\hat{\sigma}_T$ allowing for accurate pricing of VS even without direct market trading of log contracts.

Risk Management: Automatic adjustment mechanisms in pricing libraries ensure consistency and facilitate risk monitoring, especially concerning interest rate volatility and dividend models for longer-dated VSs.

Continuous Evaluation: Ongoing evaluation and adjustment of market practices are necessary to adapt to changing market conditions and improve the accuracy of VS pricing and risk management strategies.