

Empirical Finance: Methods & Applications

Volatility Trading

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Week 6

Introduction

The trade-off between **risk** and **return** is a key relationship in finance

- We want to quantify the amount of risk an investor faces over a given investment period,
- The return standard deviation (or simply **volatility**), is commonly as a measure of **risk**,
- Volatility is a key **input** for portfolio allocation, derivatives pricing, and risk management.

No matter how we model it, **volatility is a backward-looking measure of risk**

- Do we have a forward-looking measure of risk?
- Can market participants trade volatility as an asset class?
- Why would they trade volatility?

Volatility Trading

As a **speculator** you can bet on the future path of volatility

- If you expect high volatility after a forthcoming election, you can take a position in order to profit if actual volatility increases.

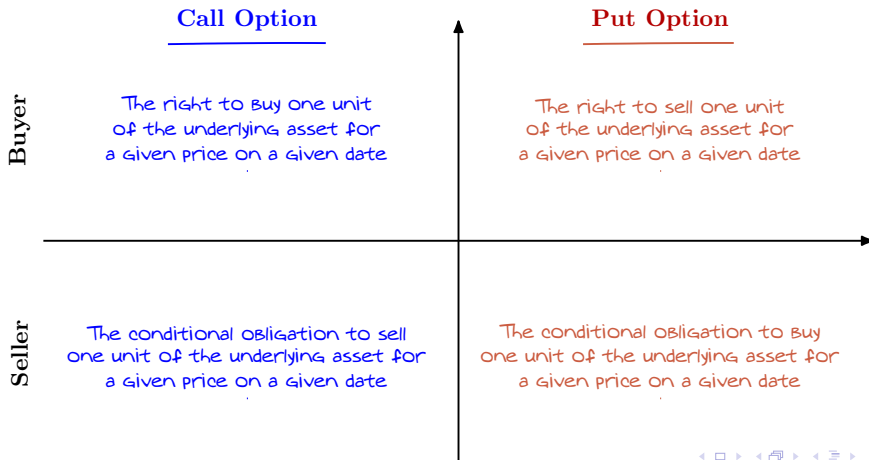
As an **hedger** you can offset your portfolio against **market uncertainty**:

- During a financial turmoil volatility tends to increase and you may need to rebalance your portfolio quite often incurring in large transaction costs,
- You can hedge against such an event by trading a volatility derivative.

Volatility trading took off

- After the meltdown of LTCM, when volatility reached unprecedented levels (Gatheral, 2006),
- The first volatility contract was traded in 1993 by UBS (Carr and Lee, 2009).

European Options



Pricing European Options

Consider the BlackScholesMerton model

Value at time t of an option with strike price K and maturity $T-t$ on one unit of the underlying asset

$$V_t(K, T) = \phi B_{t,T} [F_{t,T} \mathcal{N}(\phi d_+) - K \mathcal{N}(\phi d_-)]$$

$\phi = +1$ for a call option

$\phi = -1$ for a put option

Pricing European Options

$\mathcal{N}(\cdot)$ denotes the CDF of a standard normal with arguments defined as

Forward rate at time t
with delivery date T on one
unit of the underlying asset

Strike price at time t with
delivery date T on one unit
of the underlying asset

$$d_{\pm} = \frac{\ln(F_{t,T}/K)}{\sigma_{t,T}\sqrt{T-t}} \pm \frac{1}{2}\sigma_{t,T}\sqrt{T-t}$$

Implied volatility between
times t and T (or
maturity $T-t$) per annum

Option's maturity as
 $T-t = \frac{\text{calendar days}}{365}$

Pricing European Options

The forward price with maturity $T - t$ is

$$F_{t,T} = P_t e^{(r_{t,T} - q_{t,T})(T-t)}$$

Underlying asset price at
time t for the underlying

Continuously compounded risk-
less between t and T per annum

Continuously compounded dividend
yield between t and T per annum

Pricing European Options

The present value of expected dividend yield between times t and T is

$$DY_{t,T} = \frac{\sum_{\tau \in (t,T)} \mathbb{E}_t[D_\tau] e^{-r_{t,\tau}(\tau-t)}}{P_t}$$

Expected dividends between t and T

Discount rates based on riskless rates from t to τ

The continuously compounded dividend yield between times t and T is

$$q_{t,T} = -\frac{1}{T-t} \ln(1 - DY_{t,T})$$

To make $q_{t,T}$ in annual terms

The option is on the final price and dividends are 'stolen' before maturity.

Pricing European Options

The bond price with maturity $T - t$ is

$$B_{t,T} = e^{-r_{t,T}(T-t)}$$

Continuously compounded risk-less between t and T per annum

Pricing European Options

By setting $\phi = 1$, we have a call option

$$V_t(K, T) = C_t(K, T) \rightsquigarrow \text{Call at time } t \text{ with strike } K \text{ and maturity } T-t$$

By setting $\phi = -1$, we have a put option

$$V_t(K, T) = P_t(K, T) \rightsquigarrow \text{Put at time } t \text{ with strike } K \text{ and maturity } T-t$$

Volatility Trading

Can we use plain-vanilla call (or put) to trade volatility? **No**

$$C_{t,T}(K) = f(P_t, \sigma_{t,T}, t)$$

depends on the underlying
asset price, volatility,
and time to maturity

Using a second-order Taylor approximation, we can show that

Change in Option Value $\leftarrow dC_{t,T} \approx \frac{\partial C_{t,T}}{\partial P_t} dP_t + \frac{\partial C_{t,T}}{\partial \sigma_{t,T}} d\sigma_{t,T} + \frac{\partial C_{t,T}}{\partial t} dt + \frac{1}{2} \frac{\partial^2 C_{t,T}}{\partial P_t^2} (dP_t)^2 + \dots$

Directional risk (or Delta Δ)

Volatility risk (or Vega ν)

Time decay (or Theta Θ)

Convexity risk (or Gamma Γ)

A vanilla option is a joint bet on (P, σ, t) , not a pure vol instrument

Volatility Trading

How does the option respond to price changes?

Option sensitivity to
the underlying asset price

$$\Delta_{t,T} = \frac{\partial C_{t,T}}{\partial P_t} > 0$$

As P increases, the probability
of exercise rises, making the
call more valuable (opposite for put).

How does the option respond to volatility changes?

Option sensitivity to
implied volatility

$$v_{t,T} = \frac{\partial C_{t,T}}{\partial \sigma_{t,T}} > 0$$

Higher σ raises the option value
by increasing the chance of a
"big win" (same for put)

Volatility Trading

What is the cost of time?

Option sensitivity to time decay $\leftarrow \Theta_{t,T} = \frac{\partial C_{t,T}}{\partial t} < 0 \rightarrow$ As time passes ($T-t$ falls), the option loses value each day, all else equal (same for puts).

How does Δ change with price?

Delta's sensitivity to price changes $\leftarrow \Gamma_{t,T} = \frac{\partial^2 C_{t,T}}{\partial P_t^2} > 0 \rightarrow$ Measures how fast Δ changes as P moves. High Γ means the option price can accelerate quickly (same for put).

Volatility Trading

Can we use a delta-hedged strategy to trade volatility? **No**

$$\Pi_{t,T} = C_{t,T} - \Delta \times P_t \longrightarrow \text{Buy a call + sell } \Delta \text{ units of the underlying asset}$$

We neutralize Δ at the inception but ...

Change in Portfolio Value $\longleftarrow d\Pi_{t,T} \approx \frac{\partial \Pi_{t,T}}{\partial \sigma_{t,T}} d\sigma_{t,T} + \frac{\partial \Pi_{t,T}}{\partial t} dt + \frac{1}{2} \frac{\partial^2 \Pi_{t,T}}{\partial P_t^2} (dP_t)^2 + \dots$

ν
Profit if implied volatility rises

\ominus
The daily cost of holding the trade

Γ
The benefit of large price swings

Volatility Trading

A **delta-hedged** portfolio must be frequently adjusted to be riskless

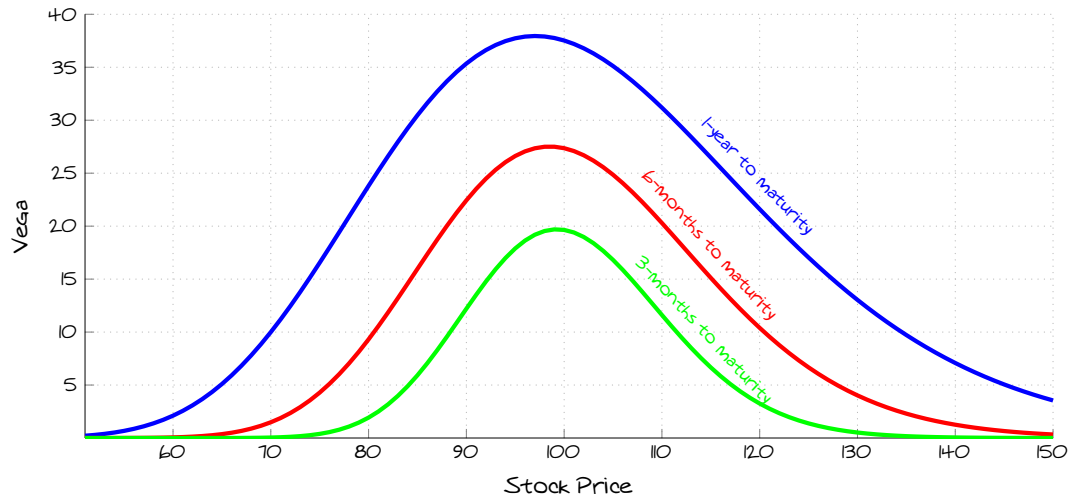
- Γ risk implies that $\Delta_{t,T}$ changes as prices move, requiring continuous rebalancing.
- Θ risk implies that even with perfect hedging, the portfolio loses value as time passes.

Also, the volatility of **delta-hedged** portfolio is not constant over time

$$v_{t,T} = \frac{\partial C_{t,T}(K)}{\partial \sigma_{t,T}} = P_t \sqrt{T-t} N'(d_+) e^{-q_{t,T}(T-t)} \longrightarrow \text{Maximum value for ATM options and declines for ITM and OTM options.}$$

A delta-hedged portfolio provides an imperfect bet on volatility.

Option's Vega



Data source: Author's calculation using $K = 100$, $r = 5\%$ and $\sigma = 20\%$.

Volatility Trading

Can we use a straddle to trade volatility? **Yes, But imperfectly**

$$\Pi_{t,T} = C_{t,T}(K) + P_{t,T}(K) \longrightarrow \text{Buy a call and a put with the same } K, T$$

We neutralize Δ at the inception but ...

Change in Portfolio Value $\longleftarrow d\Pi_{t,T} \approx \underbrace{\frac{\partial \Pi_{t,T}}{\partial \sigma_{t,T}} d\sigma_{t,T}}_{\substack{\nu \\ \text{Profit if implied} \\ \text{volatility rises}}} + \underbrace{\frac{\partial \Pi_{t,T}}{\partial t} dt}_{\substack{\ominus \\ \text{The daily cost of} \\ \text{holding the trade}}} + \underbrace{\frac{1}{2} \frac{\partial^2 \Pi_{t,T}}{\partial P_t^2} (dP_t)^2 + \dots}_{\substack{\Gamma \\ \text{The benefit of large} \\ \text{price swings}}}$

A straddle provides a more direct, but still imperfect, bet on volatility

Other hidden terms in the "... " for straddles

What is the cost volatilitydirection interaction?

Delta's sensitivity to implied volatility \leftarrow $\boxed{\text{Vanna}_{t,T}} = \frac{\partial^2 C_{t,T}}{\partial P_t \partial \sigma_{t,T}} = \frac{\partial \Delta_{t,T}}{\partial \sigma_{t,T}} \rightarrow$ Large volatility moves can cause call and put deltas to adjust at different speeds, "leaking" directional exposure into a straddle.

What is the cost of convexity in volatility?

Vega's sensitivity to implied volatility \leftarrow $\boxed{\text{Volga}_{t,T}} = \frac{\partial^2 C_{t,T}}{\partial \sigma_{t,T}^2} = \frac{\partial v_{t,T}}{\partial \sigma_{t,T}} \rightarrow$ A straddle is long two options, so Volga is high. If σ rises, Vega increases and you are "more long" volatility. If σ falls, Vega shrinks, making premium recovery much harder.

Volatility Trading

The **volatility swap** is the cleanest and simplest way to trade volatility

- A **forward contract** on the future **realized volatility** of a given asset,
- It requires a dynamic replication strategy.

The **variance swap** is not a pure bet but is related volatility

- A **forward contract** on the future **realized variance** of a given asset,
- It requires a static replication strategy.

Players talk about volatility but variance swaps are easier to replicate than volatility swaps

- Variance swaps are more popular in equity (traded on exchanges),
- Volatility swaps are more popular in currency (traded over-the-counter).

Variance Swap

Variance Swap

What is a **variance swap**? A forward contract on future realized variance

- A derivative instrument that offers a direct exposure to the return variance of an asset,
- You enter a contract on day t but the payoff is calculate on the maturity day T .

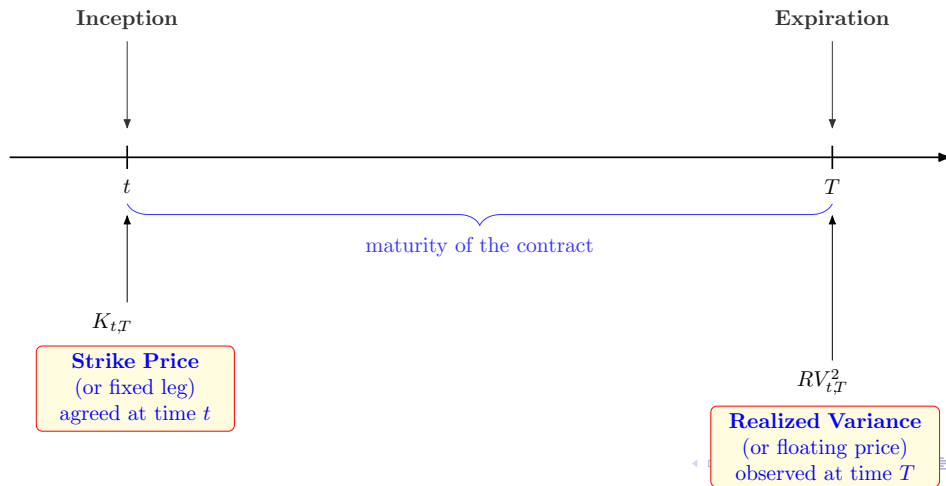
What is the **payoff** of a variance swap at maturity?

Buyer's payoff $\leftarrow X_{t,T} = \left(RV_{t,T}^2 - K_{t,T} \right) \times N_{var} \rightarrow$

She pays the strike
and receives the
realized variance

- $RV_{t,T}^2$ is the annualized realized variance known ex-post at time T
- $K_{t,T}$ is the annualized variance strike known ex-ante at time t
- N_{var} is the variance notional amount.

Variance Swap



Variance Swap

Take the S&P500 index as an example

- The market expects a 20% volatility over the next year,
- You expect volatility to be higher and wish to gain exposure,
- You decide to buy a variance swap with variance notional $N_{var} = \$2,500$.

At maturity, the realized volatility is 15% and your payoff is

$$X_{t,T} = (15^2 - 20^2) \times \$2,500 = -\$437,500$$

The Buyer will make a payment to the seller

Variance Swap

The notional of a variance swap can be expressed either as **variance notional** or **vega notional**

- The variance notional is the true notional for the calculation of the payoff,
- But variance is not intuitive and market participants think in terms of volatility,
- It is thus common to express the notional amount in terms of vega notional.

The **vega notional** is expressed as

$$N_{vega} = N_{var} \times 2K_{t,T}$$

Variance Swap

Take the S&P500 index as an example

- The market expects a 20% volatility over the next year,
- You expect volatility to be higher and wish to gain exposure,
- You decide to buy a variance swap with vega notional $N_{vega} = \$100,000$.

At maturity, the realized volatility is 15% and your payoff is

$$X_{t,T} = (15^2 - 20^2) \times \frac{\$100,000}{2 \times 20} = -\$437,500$$

The variance notional is implied from the vega notional

Variance Swap

The risk-neutral valuation (present value of the future expected payoff) is

$$B_{t,T}E_t^Q(X_{t,T}) = B_{t,T}E_t^Q(RV_{t,T}^2 - K_{t,T}) = 0$$

We thus have the **fair price** of a variance swap

$$K_{t,T} = E_t^Q[RV_{t,T}^2]$$

The strike price of variance swap is the risk-neutral expectation of its future realized variance

Realized Variance

The annualized **realized variance** measured from day t to T

$$RV_{t,T}^2 = \frac{365}{T-t} \sum_{t=1}^T r_t^2$$

- where r_t is the calendar asset return ($r_t = 0$ for holidays/weekends),
- it is also common to use business days with **252** to annualize.

Let x be a random variable

$$\text{Var}(x) = E(x^2) - [E(x)]^2 \simeq E(x^2)$$

as $E(x) \approx 0$ for daily observations.

Pricing Variance Swap

Pricing Variance Swaps

We can price variance swaps using [Britten-Jones and Neuberger \(2000\)](#)

- They derive the so-called model-free implied variance,
- Volatility can be either deterministic or stochastic,

Assumptions

- The stock price has a continuous sample path (no jumps) and positive at all times,
- There is no assumption regarding the underlying volatility process.

Pricing Variance Swaps

The dynamics of the stock price under risk neutrality follows a geometric Brownian motion

$$\frac{dS_t}{S_t} = (r - q) dt + \sigma_t dW_t$$

where σ_t is the volatility which is

$\sigma_t \longrightarrow$ an arbitrary function of other parameters

- $S_t \longrightarrow$ stock price at time t ,
- $dW_t \longrightarrow$ standard Brownian increments,
- $r \longrightarrow$ riskless rate rate,
- $q \longrightarrow$ dividend yield.

Pricing Variance Swaps

By taking the Ito's lemma, we obtain the standard Brownian motion for **log price**

$$d \ln S_t = \left(\mu - \frac{1}{2} \sigma_t^2 \right) dt + \sigma_t dW_t.$$

Take the difference between the geometric and standard Brownian motion

$$\frac{dS_t}{S_t} - d \ln S_t = \frac{1}{2} \sigma_t^2 dt.$$

and rearrange as

$$\sigma_t^2 dt = 2 \left[\frac{dS_t}{S_t} - d \ln S_t \right]. \quad (1)$$

Pricing Variance Swaps

The risk-neutral integrated variance between t to T is defined

$$V_{t,T}^2 = \frac{1}{T-t} \int_t^T \sigma_t^2 dt \quad (2)$$

Combine Equations (1) and (2), and obtain

$$V_{t,T}^2 = \frac{2}{T-t} \left[\int_t^T \frac{dS_\tau}{S_\tau} - \int_t^T d \ln S_\tau \right]$$

This is the sum of
daily log returns

$\ln(S_T/S_t)$

Pricing Variance Swaps

This identity dictates the replication strategy

$$V_{t,T}^2 = \frac{2}{T-t} \left[\int_t^T \frac{dS_\tau}{S_\tau} - \ln \frac{S_T}{S_t} \right]$$

- The **first term** denotes gains/losses on a continuously rebalancing strategy where you hold $1/S_\tau$ units of the underlying asset at time τ and your position is always worth \$1,
- The **second term** is a static position into a contract that pays the log of the stock price at T normalised by its current price at time t . This is the *log contract* of Neuberger (1994).

Pricing Variance Swaps

The swap rate K_t is the risk-neutral expectation of the integrated variance

$$\begin{aligned} K_t &= E_t^Q \left[V_{t,T}^2 \right] \\ &= \frac{2}{T-t} E_t^Q \left[\int_t^T \frac{dS_\tau}{S_\tau} - \ln \frac{S_T}{S_t} \right] \\ &= \frac{2}{T-t} \left\{ E_t^Q \left[\int_t^T \frac{dS_\tau}{S_\tau} \right] - E_t^Q \left[\ln \frac{S_T}{S_t} \right] \right\} \end{aligned}$$

Pricing Variance Swaps

The first risk-neutral expectation is

$$\begin{aligned} E_t^Q \left[\int_t^T \frac{dS_\tau}{S_\tau} \right] &= E_t^Q \left[\int_t^T (r - q) d\tau + \int_t^T \sigma_\tau^2 dW_\tau \right] \\ &= (r - q) (T - t) \end{aligned}$$

the expected terminal value of the dynamically rebalanced strategy.

Pricing Variance Swaps

The **second risk-neutral expectation** is

$$E_t^Q \left[\ln \frac{S_T}{S_t} \right] = ?$$

the expected payoff of a **log contract** (Neuberger, 1994).

How to replicate such a payoff?

- Bakshi and Madan (2000) show that a payoff function with bounded expectations can be spanned by a continuum of out-of-the-money call and put options.
- We will use this result and replicate the payoff of a log contract using a static position in plain-vanilla call & put options.

Bakshi and Madan (2000)

Bakshi and Madan (2000)

Consider a payoff function with bounded expectations

$g(S_T) \longrightarrow$ This payoff depends on the terminal value S_T

and define the following quantities

$g'(\cdot) = \frac{\partial g(\cdot)}{\partial S_T} \longrightarrow$ first-order derivative

$g''(\cdot) = \frac{\partial^2 g(\cdot)}{\partial S_T^2} \longrightarrow$ second-order derivative

Bakshi and Madan (2000)

The payoff function $g(S_T)$ can be spanned as follows

$$g(S_T) = g(\bar{S}) + g'(\bar{S})(S_T - \bar{S}) \longrightarrow \text{tangent approximation}$$

$$+ \int_0^{\bar{S}} g''(K) (K - S_T)^+ dK \longrightarrow \text{tangent correction}$$

$$+ \int_{\bar{S}}^{\infty} g''(K) (S_T - K)^+ dK \longrightarrow \text{tangent correction}$$

where

$$\bar{S} \geq 0 \longrightarrow \text{scalar for evaluation}$$

Bakshi and Madan (2000)

Apply the risk-neutral expectation and obtain the arbitrage-free price as

$$B_{t,T}E_t^Q [g(S_T)] = B_{t,T}E_t^Q [g(\bar{S})] - B_{t,T}E_t^Q [g'(\bar{S})\bar{S}]$$

$$+ g'(\bar{S})B_{t,T}E_t^Q [S_T]$$

$$+ \int_0^{\bar{S}} g''(K)B_{t,T}E_t^Q [(K - S_T)^+] dK$$

$$+ \int_{\bar{S}}^{\infty} g''(K)B_{t,T}E_t^Q [(S_T - K)^+] dK$$

Bakshi and Madan (2000)

Apply the risk-neutral expectation and obtain the arbitrage-free price as

$$\begin{aligned} B_{t,T} E_t^Q [g(S_T)] &= B_{t,T} E_t^Q [g(\bar{S})] - B_{t,T} E_t^Q [g'(\bar{S}) \bar{S}] \\ &\quad + g'(\bar{S}) B_{t,T} E_t^Q [S_T] \\ &\quad + \int_0^{\bar{S}} g''(K) B_{t,T} E_t^Q [(K - S_T)^+] dK \\ &\quad + \int_{\bar{S}}^{\infty} g''(K) B_{t,T} E_t^Q [(S_T - K)^+] dK \end{aligned}$$

Bond Price $e^{-r(T-t)}$

constant terms

Current Asset Price S_t

Put Option $P_{t,T}(K)$

Call Option $C_{t,T}(K)$

Bakshi and Madan (2000)

Apply the risk-neutral expectation and obtain the arbitrage-free price as

$$B_{t,T}E_t^Q[g(S_T)] = B_{t,T}g(\bar{S}) - B_{t,T}g'(\bar{S})\bar{S}$$

$$+ g'(\bar{S})S_t$$

$$+ \int_0^{\bar{S}} g''(K)P_{t,T}(K)dK$$

$$+ \int_{\bar{S}}^{\infty} g''(K)C_{t,T}(K)dK$$

Bakshi and Madan (2000)

Similar to Taylor expansion, any payoff can be replicated by trading

$$B_{t,T} E_t^Q [g(S_T)] = B_{t,T} [g(\bar{S}) - g'(\bar{S})\bar{S}] \longrightarrow \text{Long position in a \$1 zero coupon bond with weight } g(\bar{S}) - g'(\bar{S})\bar{S}$$

$$+ g'(\bar{S})S_t \longrightarrow \text{Long position in the asset } S_t \text{ with weight } g'(\bar{S})$$

$$+ \int_0^{\bar{S}} g''(K) P_{t,T}(K) dK \longrightarrow \text{Long position in a portfolio of put options with weights } g''(K)$$

$$+ \int_{\bar{S}}^{\infty} g''(K) C_{t,T}(K) dK \longrightarrow \text{Long position in a portfolio of call options with weights } g''(K)$$

Replication of a Log Contract

Replication of a Log Contract

Take the forward price $F_{t,T}$ and rewrite

$$\ln \frac{S_T}{S_t} = \ln \frac{F_{t,T}}{S_t} + \ln \frac{S_T}{F_{t,T}}$$

independent of the final price S_T
and requires no replication

depends on the final price S_T
and must be replicated

and use Bakshi and Madan (2000) to replicate the second term.

Replication of a Log Contract

The payoff function is defined as

$$g(S_T) = \ln \frac{S_T}{F_{t,T}}$$

The derivatives are given by

$$g'(S_T) = \frac{1}{S_T}$$

$$g''(S_T) = -\frac{1}{S_T^2}$$

Replication of a Log Contract

Use the replicating strategy of Bakshi and Madan (2000) while setting $\bar{S} = F_{t,T}$ as

$$g(S_T) = g(F_{t,T}) + g'(F_{t,T})(S_T - F_{t,T}) \\ + \int_0^{F_{t,T}} g''(K)(K - S_T)^+ dK + \int_{F_{t,T}}^{\infty} g''(K)(S_T - K)^+ dK$$

Replication of a Log Contract

Use the replicating strategy of Bakshi and Madan (2000) while setting $\bar{S} = F_{t,T}$ as

$$\ln \frac{S_T}{F_{t,T}} \longleftarrow g(S_T) = g(F_{t,T}) + g'(F_{t,T})(S_T - F_{t,T}) + \int_0^{F_{t,T}} g''(K)(K - S_T)^+ dK + \int_{F_{t,T}}^{\infty} g''(K)(S_T - K)^+ dK$$

Diagram illustrating the replication of a Log Contract using the Bakshi and Madan (2000) strategy. The equation shows the payoff $\ln \frac{S_T}{F_{t,T}}$ (green) is replicated by a portfolio of a call spread and a butterfly spread. The components are:

- $g(F_{t,T})$ (pink) is linked to $\ln \frac{F_{t,T}}{F_{t,T}}$ (pink).
- $g'(F_{t,T})(S_T - F_{t,T})$ (pink) is linked to $\frac{1}{F_{t,T}}$ (pink).
- $\int_0^{F_{t,T}} g''(K)(K - S_T)^+ dK$ (pink) is linked to $-\frac{1}{K^2}$ (pink).
- $\int_{F_{t,T}}^{\infty} g''(K)(S_T - K)^+ dK$ (pink) is linked to $-\frac{1}{K^2}$ (pink).

Replication of a Log Contract

Rearrange and obtain

$$\ln \frac{S_T}{F_{t,T}} = \frac{S_T - F_{t,T}}{F_{t,T}} - \left(\int_0^{F_{t,T}} \frac{1}{K^2} (K - S_T)^+ dK + \int_{F_{t,T}}^{\infty} \frac{1}{K^2} (S_T - K)^+ dK \right)$$

Take the risk-neutral valuation and obtain

$$\begin{aligned} B_{t,T} E_t^Q \left[\ln \frac{S_T}{F_{t,T}} \right] &= B_{t,T} E_t^Q \left[\frac{S_T - F_{t,T}}{F_{t,T}} \right] \\ &\quad - \left(\int_0^{F_{t,T}} \frac{1}{K^2} B_{t,T} E_t^Q [(K - S_T)^+] dK + \int_{F_{t,T}}^{\infty} \frac{1}{K^2} B_{t,T} E_t^Q [(S_T - K)^+] dK \right) \end{aligned}$$

Replication of a Log Contract

Recall

$$P_{t,T}(K) = B_{t,T} E_t^Q [(K - S_T)^+]$$

$$C_{t,T}(K) = B_{t,T} E_t^Q [(S_T - K)^+]$$

$$F_{t,T}(K) = E_t^Q [S_T]$$

and simplify as

$$E_t^Q \left[\ln \frac{S_T}{F_{t,T}} \right] = -\frac{1}{B_{t,T}} \left(\int_0^{F_{t,T}} \frac{1}{K^2} P_{t,T}(K) dK + \int_{F_{t,T}}^{\infty} \frac{1}{K^2} C_{t,T}(K) dK \right)$$

Pricing Variance Swap

Pricing Variance Swaps

We now have the ingredients to solve the risk-neutral expectation

$$\begin{aligned} K_t &= E_t^Q \left[V_{t,T}^2 \right] \\ &= \frac{2}{T-t} E_t^Q \left[\int_t^T \frac{dS_\tau}{S_\tau} - \ln \frac{S_T}{S_t} \right] \\ &= \frac{2}{T-t} \left\{ E_t^Q \left[\int_t^T \frac{dS_\tau}{S_\tau} \right] - E_t^Q \left[\ln \frac{S_T}{S_t} \right] \right\} \end{aligned}$$

Pricing Variance Swaps

We now have the ingredients to solve the risk-neutral expectation

$$\begin{aligned} K_t &= \frac{2}{T-t} \left\{ (r-q)(T-t) - \ln \frac{F_{t,T}}{S_t} - E_t^Q \left[\ln \frac{S_T}{F_{t,T}} \right] \right\} \\ &= \frac{2}{T-t} \left\{ (r-q)(T-t) - \ln \frac{F_{t,T}}{S_t} + \frac{1}{B_{t,T}} \left(\int_0^{F_{t,T}} \frac{1}{K^2} P_{t,T}(K) dK + \int_{F_{t,T}}^{\infty} \frac{1}{K^2} C_{t,T}(K) dK \right) \right\} \end{aligned}$$

Pricing Variance Swaps

Recall that

$$F_{t,T} = S_t e^{(r-q)(T-t)}$$

thus having

$$\begin{aligned} K_t &= \frac{2}{T-t} \left\{ (r-q)(T-t) - \ln \frac{S_t e^{(r-q)(T-t)}}{S_t} + \frac{1}{B_{t,T}} \left(\int_0^{F_{t,T}} \frac{1}{K^2} P_{t,T}(K) dK + \int_{F_{t,T}}^{\infty} \frac{1}{K^2} C_{t,T}(K) dK \right) \right\} \\ &= \frac{2}{T-t} \left\{ (r-q)(T-t) - (r-q)(T-t) + \frac{1}{B_{t,T}} \left(\int_0^{F_{t,T}} \frac{1}{K^2} P_{t,T}(K) dK + \int_{F_{t,T}}^{\infty} \frac{1}{K^2} C_{t,T}(K) dK \right) \right\} \end{aligned}$$

Pricing Variance Swaps

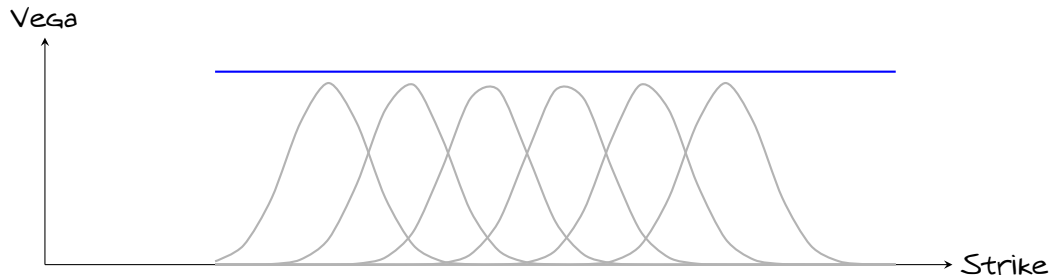
Rearrange as

$$K_t = \frac{2B_{t,T}^{-1}}{T-t} \left(\int_0^{F_{t,T}} \frac{1}{K^2} P_{t,T}(K) dK + \int_{F_{t,T}}^{\infty} \frac{1}{K^2} C_{t,T}(K) dK \right)$$

The strike price of a variance swap can be replicated using a portfolio of out-of-the money call and put options, each weighted by the inverse of their squared strike price.

Pricing Variance Swaps

Stacking nearby vegas:
Flatten aggregate vega



A continuum of nearby strikes delivers approximately constant volatility exposure.

Volatility Swap

Volatility Swap

What is a **volatility swap**? *A forward contract on future realized volatility*

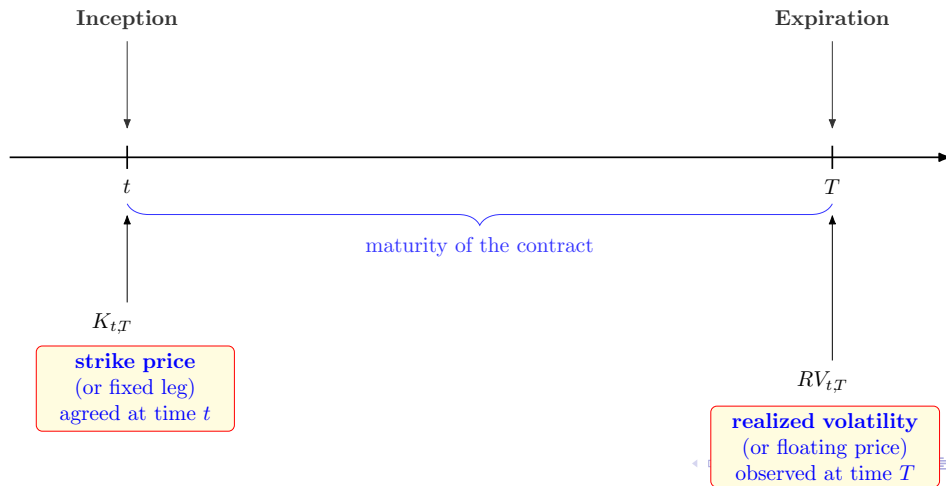
- A derivative instrument that offers a direct exposure to the return volatility of an asset,
- You enter a contract on day t but the payoff is calculate on the maturity day T .

What is the **payoff** of a volatility swap at maturity?

Buyer's payoff $\longleftarrow X_{t,T} = (RV_{t,T} - K_{t,T}) \times N_{vol} \longrightarrow$ *She pays the strike and receives the realized volatility*

- $RV_{t,T}$ is the annualized realized volatility known ex-post at time T
- $K_{t,T}$ is the annualized volatility strike known ex-ante at time t
- N_{vol} is the volatility (or vega) notional amount.

Volatility Swap



Volatility Swap

Take the EURUSD exchange rate as an example

- The market expects a 10% volatility over the next year,
- You expect volatility to be higher and wish to gain exposure,
- You decide to buy a volatility swap with $N_{vol} = \$100,000$.

At maturity, the realized volatility is 15% and your payoff is

$$X_{t,T} = (15 - 10) \times \$100,000 = \$500,000$$

The Buyer will receive a payment from the seller

Volatility Swap

The risk-neutral valuation (present value of the future expected payoff) is

$$B_{t,T}E_t^Q(X_{t,T}) = B_{t,T}E_t^Q(RV_{t,T} - K_{t,T}) = 0$$

We thus have the **fair price** of a volatility swap

$$K_{t,T} = E_t^Q[RV_{t,T}]$$

The strike price of volatility swap is the risk-neutral expectation of its future realized volatility

Realized Volatility

The annualized **realized variance** measured from day t to T

$$RV_{t,T}^2 = \frac{365}{T-t} \sum_{t=1}^T r_t^2$$

- where r_t is the calendar asset return ($r_t = 0$ for holidays/weekends),
- it is also common to use business days with **252** to annualize.

The annualized **realized volatility** is simply

$$RV_{t,T} = \sqrt{RV_{t,T}^2}$$

Variance Swap vs. Volatility Swap

Variance Swap vs. Volatility Swap

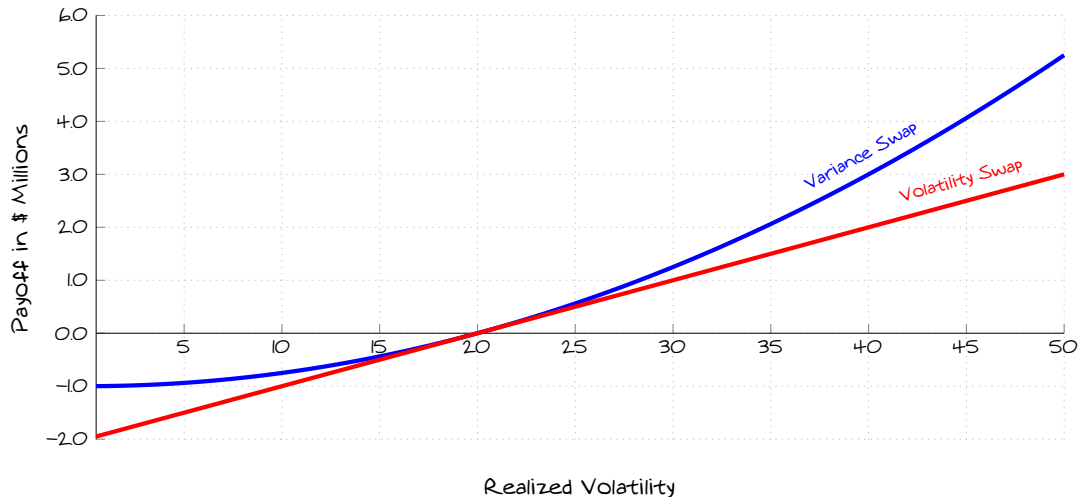
The payoff of a variance swap is linear with variance but convex with volatility

- A long variance swap always gains more from an increase in volatility than it loses for a corresponding decrease in volatility,
- This difference between the magnitude of gains and losses increases with the change in volatility, which is known as the **convexity of the variance swap**.

What is effect of convexity?

- A loss from a long variance swap is limited,
- But the loss from a short variance swap is not unlimited!

Variance Swap vs. Volatility Swap



Data source: Author's calculation using $N_{vega} = 100,000$, and $K_{vol} = 20\%$.

Pricing Volatility Swap

Pricing Volatility Swaps

Can we infer the **model-free implied volatility** from the model-free implied variance?

What we need $\leftarrow E_t^Q \left(\sqrt{RV_{t,T}^2} \right) \leq \sqrt{E_t^Q \left(RV_{t,T}^2 \right)}$ \rightarrow What we have from Britte-Jones and Neuberger (2000)

We take a shortcut and calculate

$$E_t^Q (RV_{t,T}) \simeq \sqrt{E_t^Q (RV_{t,T}^2)}$$

the **model-free implied volatility** as the square root of the model-free implied variance.

Pricing Volatility Swaps

Convexity bias arise from the pricing approximation

$$\text{Convexity} = \sqrt{E_t^Q (RV_{t,T}^2)} - E_t^Q (RV_{t,T})$$

$\sqrt{K_{var}}$ K_{vol}

Brockhaus and Long (2000) derive the convexity using a second order Taylor expansion as

$$K_{vol} = \sqrt{K_{var}^*} - \frac{Var[RV_{t,T}^2]}{8\sqrt{(E[RV_{t,T}^2])^3}}.$$

Pricing Volatility Swaps

Recall the second-order Taylor expansion around x_0 of random variable x

$$f(x) \approx f(x_0) + f'(x_0)(x - x_0) + \frac{1}{2}f''(x_0)(x - x_0)^2$$

By applying this expansion to \sqrt{x} , we have

$$\begin{aligned}\sqrt{x} &\approx \sqrt{x_0} + \frac{(x - x_0)}{2\sqrt{x_0}} - \frac{1}{8} \frac{(x - x_0)^2}{\sqrt{x_0^3}} \\ &\approx \frac{(x + x_0)}{2\sqrt{x_0}} - \frac{(x - x_0)^2}{8\sqrt{x_0^3}}\end{aligned}$$

Pricing Volatility Swaps

By taking the expectation on both sides and rearranging, we obtain

$$\begin{aligned} E[\sqrt{x}] &\approx \frac{E[x] + x_0}{2\sqrt{x_0}} - \frac{E[x^2] - 2x_0E[x] + x_0^2}{8\sqrt{x_0^3}} \\ &\approx \frac{E[x] + x_0}{2\sqrt{x_0}} - \frac{E[x^2] - E[x]^2 + E[x]^2 - 2x_0E[x] + x_0^2}{8\sqrt{x_0^3}} \\ &\approx \frac{E[x] + x_0}{2\sqrt{x_0}} - \frac{V[x] + (E[x] - x_0)^2}{8\sqrt{x_0^3}}. \end{aligned}$$

Pricing Volatility Swaps

If we set

$$x_0 = E[x]$$

We can further simplify the expansion as follows

$$E[\sqrt{x}] \approx \sqrt{E[x]} - \frac{V[x]}{8\sqrt{(E[x])^3}}$$

where the approximate value of the convexity bias is

$$\frac{V[x]}{8\sqrt{E[x]^3}}$$

An Application

Volatility Risk Premium

The **volatility risk premium** is the ex-ante payoff to a volatility swap

$$VRP_t = E_t(RV_{t,T}) - K_{t,T}$$

- E_t denotes the real-world expectation at time t ,
- $E_t(RV_{t,T}) = RV_{t-T,t}$ as in Bollerslev, Tauchen and Zhou (2009).

The **empirical evidence** suggests

- The volatility risk premium is generally negative,
- A volatility swap viewed as an insurance contract to hedge unexpected changes in volatility,
- The protection buyer pays a premium to the insurance seller.

Currency Strategy Based on Volatility Risk Premia

At the end of each month t

- Sort currencies into 5 baskets using the 1-year volatility risk premia,
- **First Portfolio (P_C)**: Currencies with cheap volatility insurance (or high VRP),
- **Last Portfolio (P_E)**: Currencies with expensive volatility insurance (or low VRP).

At the end of next month $t + 1$

- Currencies are equally-weighted within each portfolios,
- Long the **first portfolio** and short **the last portfolio**, i.e., $P_C - P_E$
- This is the **VRP strategy**.

Constructing Other Currency Portfolios

Carry Portfolios

- *CAR* is a portfolio which is long high interest rate and short low interest rate currencies

Momentum Portfolios

- *MOM* is a portfolio which is long winner and short loser currencies (1-month FX return).

Value Portfolios

- *VAL* is a portfolio which is long undervalued ($RER < 1$) and short overvalued ($RER > 1$) currencies according to PPP.

Risk Reversal Portfolios

- *RR* is a portfolio which is long low risk reversal (negative skew) and short high risk reversal (positive skew) currencies.

Dataset

The 'Developed & Emerging Countries' sample (20):

- Australia, Brazil, Canada, Czech Republic, Denmark, Euro Area, Hungary, Japan, Mexico, New Zealand, Norway, Poland, Singapore, South Africa, South Korea, Sweden, Switzerland, Taiwan, Turkey, United Kingdom.

The 'Developed Countries' sample (10):

- highlighted in blue. I will only present this evidence.

Over-the-Counter Currency Option Data

- 1-year implied volatility data from JP Morgan from Jan98 to Dec13,
- five deltas: ATM, 10 Δ Calls & Puts, 25 Δ Calls & Puts.

Currency Strategies: Summary Statistics

Panel A: Excess Returns					
	<i>CAR</i>	<i>MOM</i>	<i>VAL</i>	<i>RR</i>	<i>VRP</i>
<i>Mean</i>	4.10	0.92	3.66	5.10*	4.95**
<i>Sdev</i>	10.73	9.81	8.97	11.49	8.15
<i>Skew</i>	−0.71	0.26	−0.16	−0.47	−0.03
<i>Kurt</i>	5.25	3.75	3.71	5.41	3.97
<i>SR</i>	0.38	0.09	0.41	0.44	0.61
<i>MDD</i>	−37.8	−22.8	−15.1	−35.1	−17.0
<i>Freq_L</i>	0.10	0.48	0.09	0.08	0.29
<i>Freq_S</i>	0.07	0.44	0.07	0.22	0.35
Panel B: FX Returns					
<i>Mean</i>	−0.81	0.94	2.15	1.84	5.45***
<i>Sdev</i>	10.76	9.87	9.02	11.58	8.12
<i>Skew</i>	−0.72	0.33	−0.24	−0.50	−0.03
<i>Kurt</i>	5.43	3.94	3.76	5.63	4.04
<i>SR</i>	−0.08	0.09	0.24	0.16	0.67
<i>MDD</i>	−43.3	−23.2	−22.5	−40.3	−14.5
<i>Freq_L</i>	0.10	0.48	0.09	0.08	0.29
<i>Freq_S</i>	0.07	0.44	0.07	0.22	0.35

Excess Return Decomposition

Excess Return

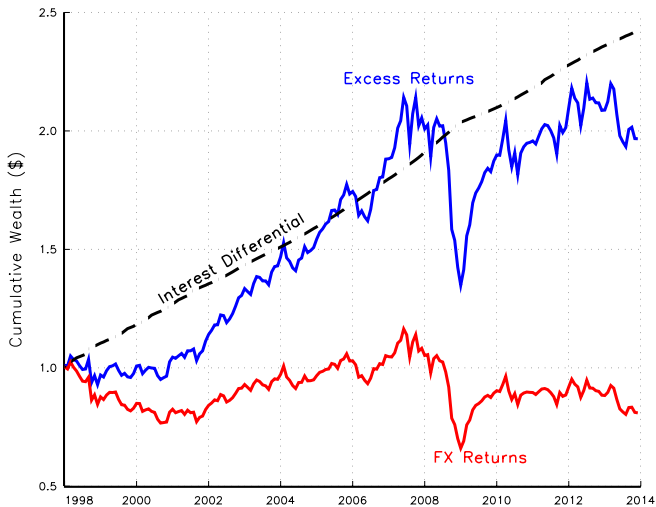
- Buy the foreign currency at time t in forward market and offset the position at time $t + 1$ in the spot market

$$RX_{t+1} = \frac{S_{t+1} - F_t}{S_t}$$

- Perform the simple decomposition

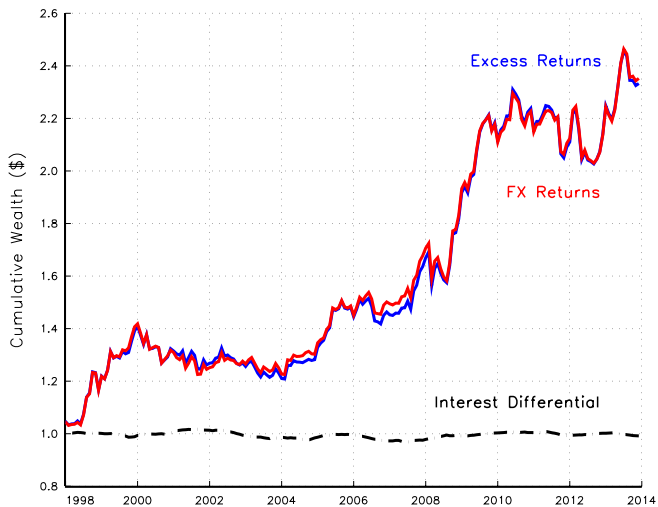
$$\begin{aligned} RX_{t+1} &= \frac{S_{t+1} - S_t}{S_t} - \frac{F_t - S_t}{S_t} \\ &\approx \frac{S_{t+1} - S_t}{S_t} + i_t^* - i_t \end{aligned}$$

Carry Strategy



Carry returns depend mostly on the interest rate differential

VRP Strategy

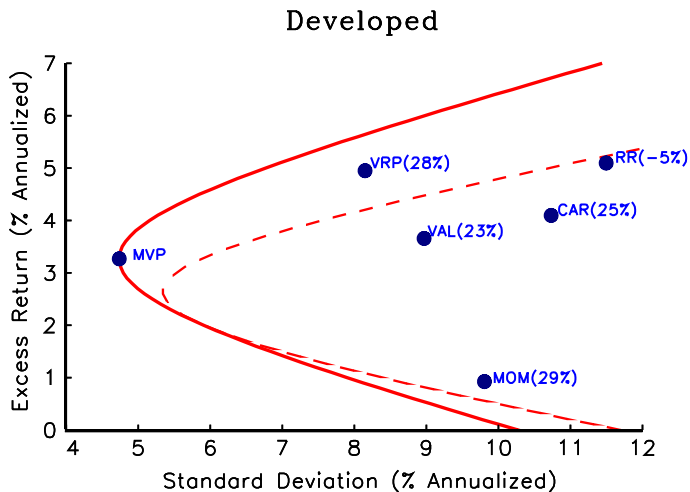


VRP returns depend mostly on FX returns.

Currency Strategy: Correlations

	<i>CAR</i>	<i>MOM</i>	<i>VAL</i>	<i>RR</i>
<i>CAR</i>	1.00			
<i>MOM</i>	-0.20	1.00		
<i>VAL</i>	0.30	-0.20	1.00	
<i>RR</i>	0.76	-0.23	0.46	1.00
<i>VRP</i>	-0.08	0.11	0.19	0.10

Global Minimum Variance Portfolio



VRP adds diversification benefits to a global currency strategy

What is the Story?

VRP has strong predictive power for the cross-section of currency returns

- Economically valuable and statistically significant,
- VRP information relates to future spot exchange rate returns, not to interest differentials.

VRP returns essentially uncorrelated with traditional currency strategies

- Potential diversification gains from adding *VRP* strategy to those normally followed by currency managers,
- Evidence consistent with *VRP* returns driven by speculator-hedger interactions in the presence of time-varying capital constraints.