

Modification of Stop-loss Start-gain strategy. Distribution of hedger losses

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1. Settings

Strip (deadband)

$$H \triangleq \{(y, t) : y \in [K, K(1 + d)], t \in [0, T]\}. \quad (1)$$

Asset price at the moment t :

$$S(t) \triangleq S_0 - \beta t + \sigma W(t). \quad (2)$$

Minimize mean losses

$$\bar{L}(d, T) \triangleq \mathbf{M}[L(d, T)] \rightarrow \min_d. \quad (3)$$

Maximize probability function

$$P_\varphi(d, T) \triangleq \mathcal{P}\{L(d, T) \leq \varphi\} \rightarrow \max_d. \quad (4)$$

Minimize quantile function

$$\varphi_\alpha(d, T) \triangleq \min\{\varphi : P_\varphi(d, T) \geq \alpha\} \rightarrow \min_d. \quad (5)$$

2. Strip crossing count

Let's denote:

η_T^- — the number of downward crossings of the strip H by $S(t)$ in a time T ;

η_T^+ — the number of upward crossings of the strip H by $S(t)$ in a time T ;

$\rho^- = -K$ — hedger's losses on one downward crossig, if option was not exercized;

$\rho^+ = K(1 + d)(1 + \theta)$ — hedger's losses on one upward crossig, if option was not exercized;

Values η_T^- и η_T^+ are random and dependent. For each possible realization of η_T^- value η_T^+ can be equal to η_T^- or $\eta_T^- + 1$.

3. Exercise of option

We denote the probability of exercising of the option by optionholder on one crossing as

$$p(d) = 1 - e^{-\lambda Kd}, \lambda > 0.$$

Let $\{\nu_i\}_{i=1}^{\infty}$ be the series of i.i.d Bernoulli random values, which describes the possible exercise of the option at the i -th crossing:

$$\mathcal{P}\{\nu_i = 1\} = p(d), \text{ for each } i = 1, 2, \dots.$$

The asset price at the moment of early exercise is random and equal to $K + \zeta$, wher ζ is a random value, with support $[0, Kd]$.

The cumulative probability distribution function of ζ let's define as

$$F_{\zeta}(s) = \begin{cases} 1, & \text{если } s > Kd; \\ \frac{1 - e^{-\lambda s}}{1 - e^{-\lambda Kd}}, & \text{если } 0 \leq s \leq Kd; \\ 0, & \text{если } s < 0. \end{cases}$$

4. Cost of hedging

Hedger's losses on i -th crossing:

$$l_i \triangleq \begin{cases} \nu_i((K + \zeta)(1 + \theta) - K) + (1 - \nu_i)\rho^+, & i = 1; \\ \left(\prod_{j=1}^{i-1} (1 - \nu_j) \right) (\nu_i((K + \zeta)(1 + \theta) - K) + (1 - \nu_i)\rho^+), & i = 2m + 1; \\ \left(\prod_{j=1}^{i-1} (1 - \nu_j) \right) \rho^-, & i = 2m. \end{cases}$$

Possible losses when no crossings happened:

$$l_0 \triangleq \nu_0((K + \zeta)(1 + \theta) - K).$$

Total losses of the hedger are equal to

$$L(d, T) \triangleq \begin{cases} \sum_{i=0}^{2m} l_i, & \text{if } \eta^+ + \eta^- = 2m; \\ \sum_{i=0}^{2m+1} l_i - \left(\prod_{j=1}^{2m+1} (1 - \nu_j) \right) K, & \text{if } \eta^+ + \eta^- = 2m + 1, \end{cases} \quad (6)$$

где $m = 0, 1, 2, \dots$.

5. Distribution of strip crossing count

The distributions of random values η_T^- и η_T^+ are defined as follows:

Theorem 1

The number of downward crossings is less than or equal to m with probability

$$\mathcal{P}\{\eta_T^- \geq m\} = \exp\left\{\frac{2m\beta Kd}{\sigma}\right\} \left[1 - \Phi\left(\frac{\beta}{\sigma}\sqrt{T} + \frac{2mKd}{\sigma\sqrt{T}}\right) + \exp\left\{-\frac{2m\beta Kd}{\sigma}\right\} \Phi\left(\frac{\beta}{\sigma}\sqrt{T} - \frac{2mKd}{\sigma\sqrt{T}}\right)\right].$$

The number of upward crossings is less than or equal to m with probability

$$\mathcal{P}\{\eta_T^+ \geq m\} = \exp\left\{-\frac{2m\beta Kd}{\sigma}\right\} \left[1 - \Phi\left(-\frac{\beta}{\sigma}\sqrt{T} + \frac{(2m-1)Kd}{\sigma\sqrt{T}}\right) + \exp\left\{\frac{(2m-1)\beta Kd}{\sigma}\right\} \Phi\left(-\frac{\beta}{\sigma}\sqrt{T} - \frac{(2m-1)Kd}{\sigma\sqrt{T}}\right)\right].$$

6. Hedger mean losses

Let's write the mean losses of the hedger's as total losses expectation:

$$\bar{L}(d, T) = \sum_{i=0}^{\infty} \mathbf{M} [L(d, T) | \eta^+ + \eta^- = i] \mathcal{P}\{\eta^+ + \eta^- = i\}.$$

Considering, that option could remain "in-money" and not exercised before T , we obtain the value of mean losses

$$\begin{aligned} \bar{L}(d, T) = & \sum_{k=0}^{\infty} \sum_{j=0}^k \mathbf{M}[l_j] \mathcal{P}\{\eta^+ + \eta^- = k\} - \\ & - \sum_{k=0}^{\infty} (1 - p(d))^{2k+1} \mathcal{K} \mathcal{P}\{\eta^+ + \eta^- = 2k + 1\}. \quad (7) \end{aligned}$$

7. Conditional distribution of losses

Let's define the conditional distribution function of the losses

$$P_\varphi(d, T, k) \triangleq \mathcal{P}\{L(d, T) \leq \varphi | \eta^+ + \eta^- = k\}, \text{ for each } \varphi > 0.$$

term “conditional” means known number of strip crossings. Using total probability formula, we obtain unconditional distribution function

$$P_\varphi(d, T) = \sum_{k=0}^{\infty} P_\varphi(d, T, k) \cdot \mathcal{P}\{\eta^+ + \eta^- = k\}.$$

Let us define conditional and unconditional quantile functions as

$$\varphi_\alpha(d, T, k) \triangleq \min\{\varphi : P_\varphi(d, T, k) \geq \alpha\}, \quad (8)$$

$$\varphi_\alpha(d, T) \triangleq \min\{\varphi : P_\varphi(d, T) \geq \alpha\}. \quad (9)$$

8. Conditional distribution of losses

Let's denote

$$k^*(\varphi) = \left\lceil \frac{\varphi}{\rho^+ - \rho^-} \right\rceil, \varphi^* = \frac{\varphi - 2k^*(\varphi)(\rho^+ + \rho^-) + K}{1 + \theta}$$

(i) If $i < 2k^*(\varphi)$: $P_\varphi(d, T, k) = 1$.

(ii) If $i = 2k^*(\varphi)$:

$$P_\varphi(d, T, k) = 1 - (1 - p(d))^{2k^*(\varphi)} + (1 - p(d))^{2k^*(\varphi)} \frac{\mathcal{P}\{K \leq S(T) \leq \varphi^*\}}{\mathcal{P}\{S(T) \leq K(1 + d)\}}.$$

(iii) If $i = 2k^*(\varphi) + 1$:

$$P_\varphi(d, T, k) = (1 - p(d))^{2k^*(\varphi)} p(d) \mathcal{P}\{\zeta \leq \varphi^* - K\} + \\ + 1 - (1 - p(d))^{2k^*(\varphi)} + (1 - p(d))^{2k^*(\varphi)+1} \frac{\mathcal{P}\{K \leq S(T) \leq \varphi^*\}}{\mathcal{P}\{K \leq S(T)\}}.$$

(iv) if $i > 2k^*(\varphi) + 1$:

$$P_\varphi(d, T, k) = 1 - (1 - p(d))^{2k^*(\varphi)} + (1 - p(d))^{2k^*(\varphi)} p(d) \mathcal{P}\{\zeta \leq \varphi^* - K\}.$$

9. Conditional quantile of losses

It's easily seen, that values $\varphi_\alpha(d, T, k)$ are increasing with k and

$$\begin{aligned} P_{\varphi_\alpha(d, T, 0)}(d, T) &= \sum_{k=0}^{\infty} P_{\varphi_\alpha(d, T, 0)}(d, T, k) \cdot \mathcal{P}\{\eta^+ + \eta^- = k\} \leq \\ &\leq \sum_{k=0}^{\infty} \alpha \cdot \mathcal{P}\{\eta^+ + \eta^- = k\} = \alpha. \end{aligned}$$

Therefore, exists $m > 0$ that:

$$\begin{aligned} P_{\varphi_\alpha(d, T, m)}(d, T) &\geq \alpha \\ P_{\varphi_\alpha(d, T, m-1)}(d, T) &\leq \alpha. \end{aligned}$$

Values $\varphi_\alpha(d, T, m-1)$ and $\varphi_\alpha(d, T, m)$ can be used as upper and lower estimates for unknown unconditional quantile $\varphi_\alpha(d, T)$.

10. Adjustment

Let's assume, that we know the exact value or upper estimate for Lipschitz constant L for function $P_\varphi(d, T)$ on interval $[\varphi_\alpha(d, T, m-1); \varphi_\alpha(d, T, m)]$, now, we can easily construct upper and lower envelopes for function $P_\varphi(d, T)$

$$F^+(\varphi, d, T) = \min\{P_{\varphi_\alpha(d, T, m)}(d, T), P_{\varphi_\alpha(d, T, m-1)}(d, T) + L(\varphi - \varphi_\alpha(d, T, m-1))\},$$

$$F^-(\varphi, d, T) = \min\{P_{\varphi_\alpha(d, T, m-1)}(d, T), P_{\varphi_\alpha(d, T, m)}(d, T) + L(\varphi - \varphi_\alpha(d, T, m))\}.$$

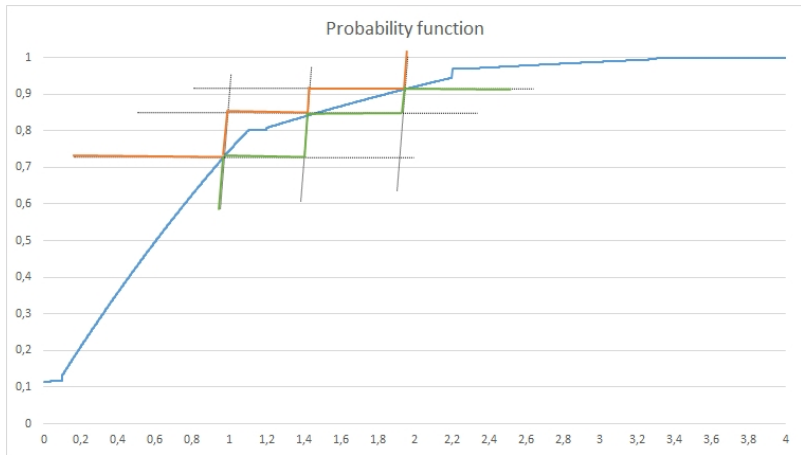
We obtain amended estimates of unconditional quantile:

$$\varphi_\alpha^+(d, T) = \frac{\alpha - P_{\varphi_\alpha(d, T, m)}(d, T)}{L} + \varphi_\alpha(d, T, m),$$

$$\varphi_\alpha^-(d, T) = \frac{\alpha - P_{\varphi_\alpha(d, T, m+1)}(d, T)}{L} + \varphi_\alpha(d, T, m+1).$$

This approach can be used together with dichotomy method for consequent adjustment of estimates of $\varphi_\alpha^+(d, T)$.

11. Adjustment



12. Estimation of Lipschitz constant

Considering that for calculation of $P_\varphi(d, T)$ one needs to calculate the sum of series, estimation of Lipschitz constant L can be very time-consuming. Let's assume that we can calculate or estimate constants L_k for conditional distribution functions $P_\varphi(d, T, k)$, then

$$\begin{aligned} |P_{x_1}(d, T) - P_{x_2}(d, T)| &= \left| \sum_{k=0}^{\infty} P_{x_1}(d, T, k) p_k - \sum_{k=0}^{\infty} P_{x_2}(d, T, k) p_k \right| = \left| \sum_{k=0}^{\infty} (P_{x_1}(d, T, k) - P_{x_2}(d, T, k)) p_k \right| \\ &\leq \sum_{k=0}^{\infty} |P_{x_1}(d, T, k) - P_{x_2}(d, T, k)| p_k \leq \sum_{k=0}^{\infty} L_k |x_1 - x_2| p_k = |x_1 - x_2| \sum_{k=0}^{\infty} L_k p_k, \end{aligned}$$

where $p_k \triangleq \mathcal{P}\{\eta^+ + \eta^- = k\}$. So, we obtain upper estimate for the Lipschitz constant L :

$$L \leq \sum_{k=0}^{\infty} L_k p_k. \quad (10)$$