

# European Option Pricing Based on Heston-Dupire

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## 1. Theory

### 1.1 Heston-Dupire Model

The Heston-Dupire immediate local volatility model is a model that combines the local volatility obtained from the Dupire formula in its unparameterized form with the Heston stochastic volatility model.

$$\begin{aligned}dS_t &= rS_t dt + L(t, S_t) \sqrt{V_t} S_t dW_{1t} \\dV_t &= a(b - V_t) dt + \sigma_v \sqrt{V_t} dW_{2t}\end{aligned}$$

### 1.2 Leverage function

#### 1.2.1 Derivation of leverage function

Let the price of the European call option be

$$C(t, K) = \frac{B_{t_0}}{B_t} E[(S_t - K)^+]$$

Differentiating the the above equation, and using Fubini's theorem, we get,

$$dC(t, K) = -\frac{r}{M_t} E[(S_t - K)^+] dt + \frac{1}{M_t} E[d(S_t - K)^+]$$

This function is not differentiable at  $x = c$  and cannot be solved directly by Ito's lemma. However, this problem can be solved by the following Tanaka-Meyer formula.

$$g(X_t) = g(X_{t_0}) + \int_{t_0}^t 1_{X_u > b} d\tilde{B}_u + \int_{t_0}^t 1_{X_u > b} dV_u + \frac{1}{2} \int_{t_0}^t g''(X_u) (d\tilde{B}_u)^2$$

To further simplify the calculation, the well-known conclusion from Feng (2010) can be used. If  $S_t, V_t$  obey the stochastic local volatility model in 1.1, then the following equation holds for the price of a European call option.

$$-\frac{\partial C(t, K)}{\partial K} = \frac{1}{B_t} E[1_{S_t > K}], \quad \frac{\partial^2 C(t, K)}{\partial K^2} = \frac{\psi_S}{B_t}$$

Then we get the leverage function

$$L^2(t, K) = \frac{\frac{\partial C(t, K)}{\partial t} + rK \frac{\partial C(t, K)}{\partial K}}{\frac{1}{2} K^2 \frac{\partial^2 C(t, K)}{\partial K^2} E[V_t | S_t = K]} = \frac{\sigma_{LV}^2(t, K)}{E[V_t | S_t = K]}$$

## 1.2.2 Calculation of leverage function

### Numerator

$$\begin{aligned} E[V_{t_i} | S_{t_i} = s_{i,j}] &\approx \frac{E[V_{t_i} \mathbb{1}_{S_{t_i} \in (b_{i,k}, b_{i,k+1}]}]}{\mathbb{Q}[S_{t_i} \in (b_{i,k}, b_{i,k+1})]} \\ &\approx \frac{\frac{1}{M} \sum_{j=1}^M v_{i,j} \mathbb{1}_{s_{i,j} \in (b_{i,k}, b_{i,k+1})}}{\mathbb{Q}[S_{t_i} \in (b_{i,k}, b_{i,k+1})]} \\ &\approx \frac{l}{M} \sum_{j \in \#_{i,k}} v_{i,j} \end{aligned}$$

### Denominator

$$\sigma_{LV}^2(T, K) = \frac{\sigma_I^2 + 2T\sigma_I \left( \frac{\partial \sigma_I}{\partial T} + rK \frac{\partial \sigma_I}{\partial K} \right)}{\left( 1 + d_1 K \sqrt{T} \frac{\partial \sigma_I}{\partial K} \right)^2 + K^2 \sigma_I T \left( \frac{\partial^2 \sigma_I}{\partial K^2} - d_1 \sqrt{T} \left( \frac{\partial \sigma_I}{\partial K} \right)^2 \right)}$$

To get  $\sigma_I$  We first use the SVI (Stochastic Volatility Inspired) function to fit in the K-direction , followed by linear interpolation in the T-direction. And then we obtain the implied volatility surface by combining the two directions.

$$\begin{aligned} \sigma_I^{SVI} &= \sqrt{\frac{\alpha_n + \beta_n \left[ \rho_n (x - m_n) + \sqrt{(x - m_n)^2 + \sigma_n} \right]}{T_n}} \\ \min J_n(\mathbf{C}) &= \frac{1}{2} \sum_{i=1}^M w_i \left( \frac{\sigma_I^{SVI}(T_n, K_i; \mathbf{C}) - \sigma_I^{\text{Market}}(T_n, K_i)}{\sigma_I^{\text{Market}}(T_n, K_i)} \right)^2 \\ \sigma_I^{SVI}(T, x) &= \frac{T_{n+1} - T}{T_{n+1} - T_n} \sigma_I^{SVI}(T_n, x) + \frac{T - T_n}{T_{n+1} - T_n} \sigma_I^{SVI}(T_{n+1}, x) \end{aligned}$$

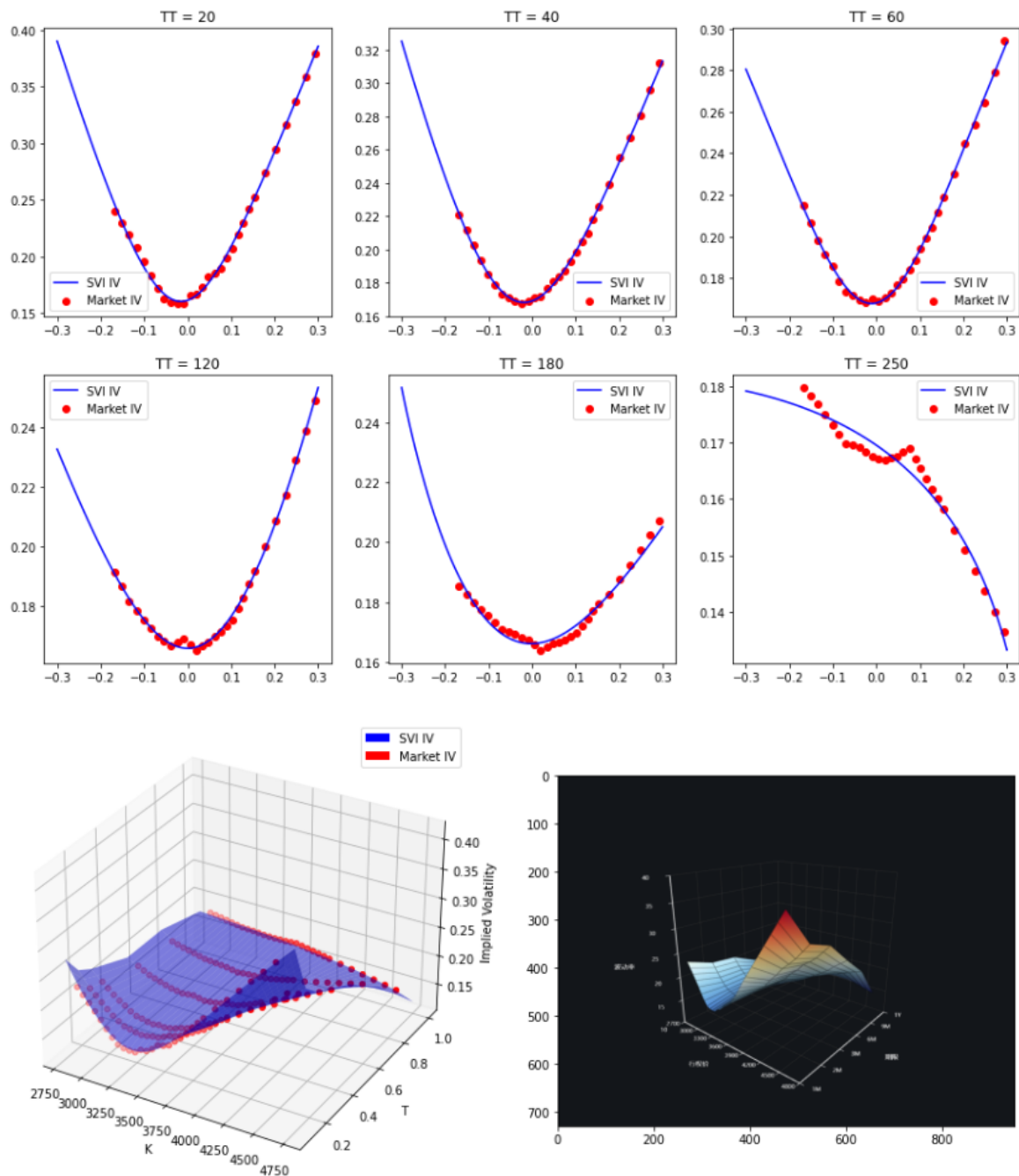
To complete the calculation of the rest of leverage function, we take the numerical derivatives by difference.

$$\begin{aligned} \sigma_I &= \sigma_I^{SVI}(T, x) = \lambda_n \sigma_I^{SVI}(T_n, x) + \lambda_{n+1} \sigma_I^{SVI}(T_{n+1}, x) \\ \frac{\partial \sigma_I}{\partial T} &= \frac{1}{T_{n+1} - T_n} [\sigma_I^{SVI}(T_{n+1}, x) - \sigma_I^{SVI}(T_n, x)] \\ \frac{\partial \sigma_I}{\partial x} &= \tau_n f_n^{(1)} + \tau_{n+1} f_{n+1}^{(1)} \\ \frac{\partial^2 \sigma_I}{\partial x^2} &= \tau_n f_n^{(2)} + \tau_{n+1} f_{n+1}^{(2)} \end{aligned}$$

## 2. Practice

### 2.1 Fitting volatility based on SVI

We select the CSI 300 put option data on December 29, 2023, and fit the implied volatility surface using the SVI method.



## 2.2 Model result

- `strike = np.arange(2900, 3500, 50)`
- `sigma, vov, mr, rho, texp, spot = 0.3, 1, 0.5, -0.9, 20, 3431.1099`

### 2.2.1 Heston model

- Lewis AL (2000) Option valuation under stochastic volatility: with Mathematica code. Finance Press

```
array([2614.7557097 , 2605.5777012 , 2596.47637983, 2587.45031286,
       2578.49811439, 2569.61844314, 2560.81000035, 2552.07152782,
       2543.40180606, 2534.79965257, 2526.2639202 , 2517.79349561])
```

- Conditional MC for Heston model based on QE discretization scheme by Andersen (2008)

```
array([2628.65334821, 2619.48966579, 2610.40231646, 2601.38987433,
       2592.45096014, 2583.58423907, 2574.78841859, 2566.06224659,
       2557.40450947, 2548.81403046, 2540.28966797, 2531.83031408])
```

- Milstein for Heston

```
[2588.3889053 2579.35324418 2570.39391599 2561.50654925 2552.6974709
 2543.95942172 2535.28801316 2526.68473049 2518.15106307 2509.68248496
 2501.27981974 2492.95075096]
```

## 2.2.2 Heston-Dupire

- Var calculation is based on QE method, and we use Euler method to obtain ST

$$dX_t = \left( r - \frac{1}{2} L^2(t, e^{X_t}) V_t \right) dt + L(t, e^{X_t}) \sqrt{V_t} dW_{1t}$$

```
Heston-Dupire: [574.08437613 558.31437139 542.96737848 528.08087414 513.67470773
499.80181511 486.51445111 473.83205918 461.79139787 450.4427644
439.77928339 429.82440945]
```

- Milstein for Heston-Dupire

$$v_{t+dt} = v_t + \kappa(\theta - v_t)dt + \sigma\sqrt{v_t}dtZ_v + \frac{1}{4}\sigma^2dt(Z_v^2 - 1)$$

$$X_{t+dt} = X_t + \left( r - \frac{1}{2} v_t L_t^2 \right) dt + \sqrt{v_t} L_t dt Z_s$$

```
[543.0012140365923, 527.5087182695403, 512.4369276030873, 497.7983599724606, 483.640501
0650162, 469.9989355729829, 456.9234524661498, 444.458064480143, 432.6333222533259, 42
1.4505778410579, 410.9189585735541, 401.04508534373315]
```