



# SVCJ Model

## (Stochastic Volatility with Correlated Jumps)

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# SV Model

Expected log return

$$\frac{dS_t}{S_t} = \mu dt + \sqrt{V_t} dW_t^s$$

Brownian motion

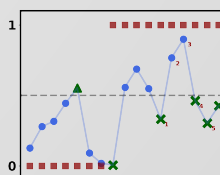
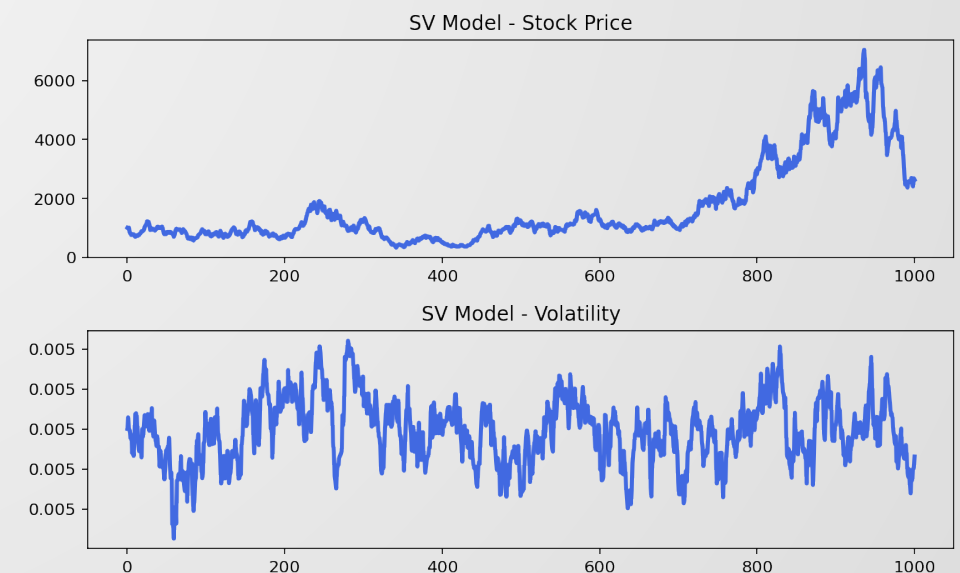
$$dW \sim N(0, 1)$$

$$dV_t = \kappa (\theta - V_t) dt + \sigma_v \sqrt{V_t} dW_t^v$$

$$\text{Cov} (dW_t^s, dW_t^v) = \rho dt$$

CIR model

Parameter:  $(\mu, \rho, \kappa, \theta, V_0, \sigma_v)$



# SVJ Model

$$\frac{dS_t}{S_t} = \mu dt + \sqrt{V_t} dW_t^s + Z_t^y dN_t$$

Jump size

Jump process

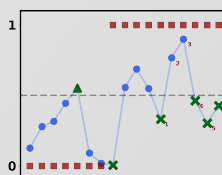
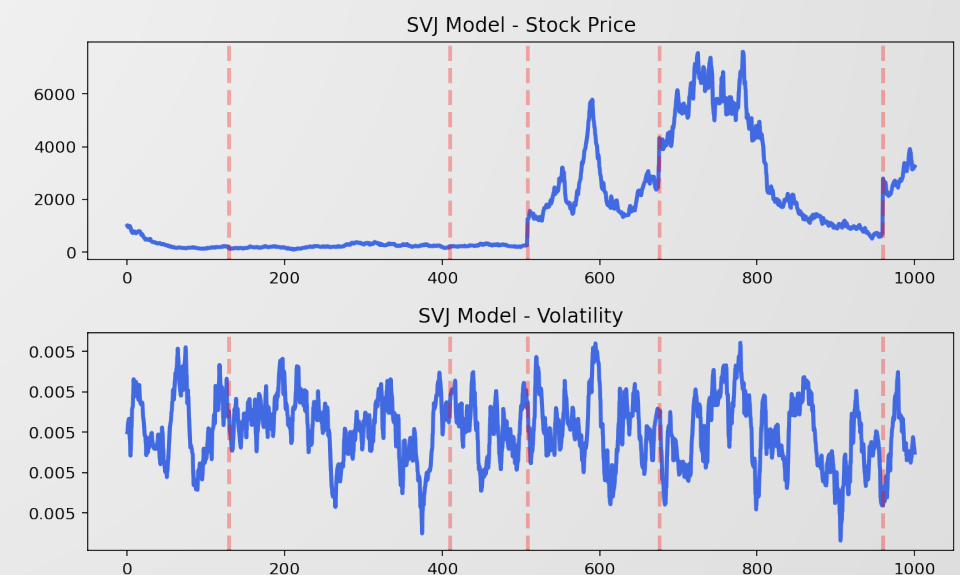
$$dV_t = \kappa (\theta - V_t) dt + \sigma_v \sqrt{V_t} dW_t^v$$

$$\text{Cov} (dW_t^s, dW_t^v) = \rho dt$$

$$P (dN_t = 1) = \lambda dt$$

$$Z_t^y \sim N (\mu_y, \sigma_y^2)$$

Parameter:  $(\mu, \rho, \kappa, \theta, V_0, \sigma_v, \lambda, \mu_y, \sigma_y)$



# SVCJ Model

$$\frac{dS_t}{S_t} = \mu dt + \sqrt{V_t} dW_t^s + Z_t^y dN_t$$

Jump size

$$dV_t = \kappa (\theta - V_t) dt + \sigma_v \sqrt{V_t} dW_t^v + Z_t^v dN_t$$

$$\text{Cov} (dW_t^s, dW_t^v) = \rho dt$$

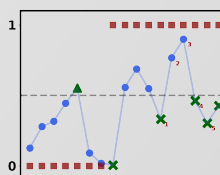
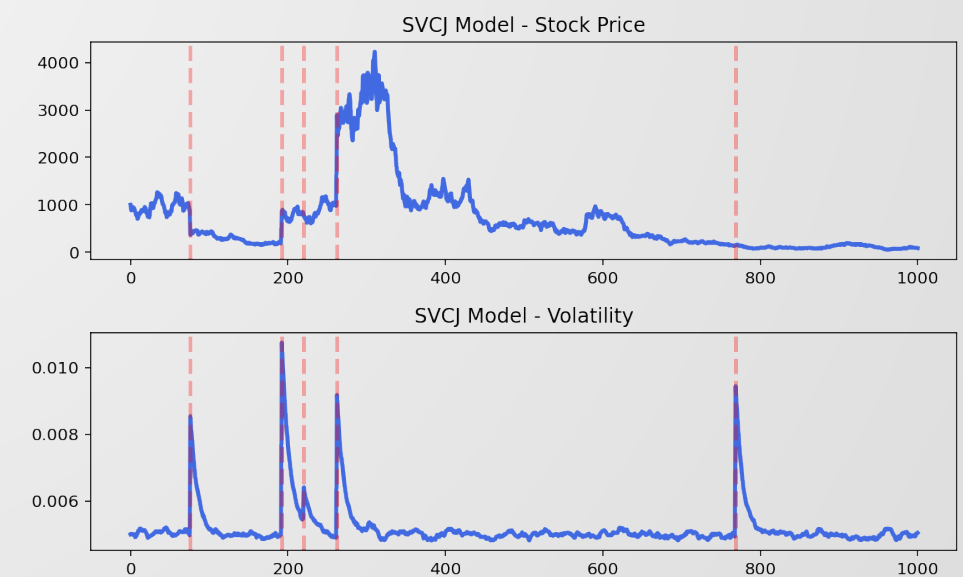
$$P (dN_t = 1) = \lambda dt$$

$$Z_t^y \mid Z_t^v \sim N \left( \mu_y + \rho_j Z_t^v, \sigma_y^2 \right)$$

$$Z_t^v \sim \text{Exp} (\mu_v)$$

Parameter:  $(\mu, \rho, \kappa, \theta, V_0, \sigma_v, \lambda, \mu_y, \sigma_y, \rho_j, \mu_v)$

Jump process



# Euler-Maruyama discretization

□ Define  $Y_{t+1} = \log \left( \frac{S_{t+1}}{S_t} \right), t = 1, \dots, T$

$$\implies S_t = S_0 \exp (Y_1 + \dots + Y_t), t = 1, \dots, T$$

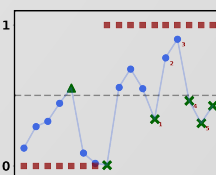
□ Set  $\alpha = \kappa\theta, \beta = 1 - \kappa$

□ Then,

$$Y_t = \mu + \sqrt{V_{t-1}}\varepsilon_t^y + Z_t^y J_t,$$

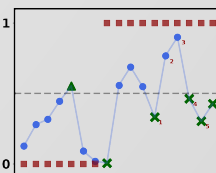
$$V_t = \alpha + \beta V_{t-1} + \sigma_v \sqrt{V_{t-1}}\varepsilon_t^v + Z_t^v J_t$$

□ where  $\varepsilon_t^t, \varepsilon_t^v \sim N(0, 1), P(J_t = 1) = \lambda$



## Parameter of SV, SVJ, SVCJ Models

- SV :  $\Theta = (\mu, \rho, \alpha, \beta, V_0, \sigma_v), \quad |\Theta| = 6$
- SVJ :  $\Theta = (\mu, \rho, \alpha, \beta, V_0, \sigma_v, \lambda, \mu_y, \sigma_y), \quad |\Theta| = 9$
- SVCJ :  $\Theta = (\mu, \rho, \alpha, \beta, V_0, \sigma_v, \lambda, \mu_y, \sigma_y, \rho_j, \mu_v), \quad |\Theta| = 11$



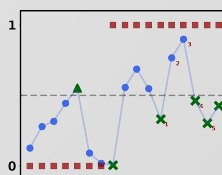
# Estimate the parameters of SVCJ

- Use  $p$  to denote the market price;  $m$  to denote model price
- Market price:  $p(\omega, \tau, K)$
- Model price:  $m(\Theta; \omega, \tau, K)$
- Define the loss function as

$$\mathcal{L}(\Theta) = \sum_i \left\{ p(\omega_i, T_i, K_i) - m(\Theta; \omega_i, T_i, K_i) \right\}^2$$

$$\approx \sum_i \left\{ p(\omega_i, T_i, K_i) - \widehat{m}(\Theta; \omega_i, T_i, K_i) \right\}^2$$

Approximated by Monte Carlo Simulation

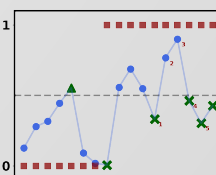


## Estimate the parameters of SVCJ

- The parameters are calibrated by minimizing the loss function

$$\Theta^* = \arg \min_{\Theta \in \mathcal{A}} \mathcal{L}(\Theta)$$

- $\mathcal{A}$  is defined to guarantee the positiveness of the CIR models.
  - ▶ Feller condition:  $2\kappa\theta > \sigma_v^2$ , or,  $2\alpha > \sigma_v^2$ .
  - ▶ Non-negativity Conditions:  
 $\kappa > 0, \theta > 0, V_0 > 0$ , or,  $\alpha > 0, \beta < 1, V_0 > 0$ .
  - ▶ In the SVCJ model, there should be another restriction  $\mu_v \geq 0$





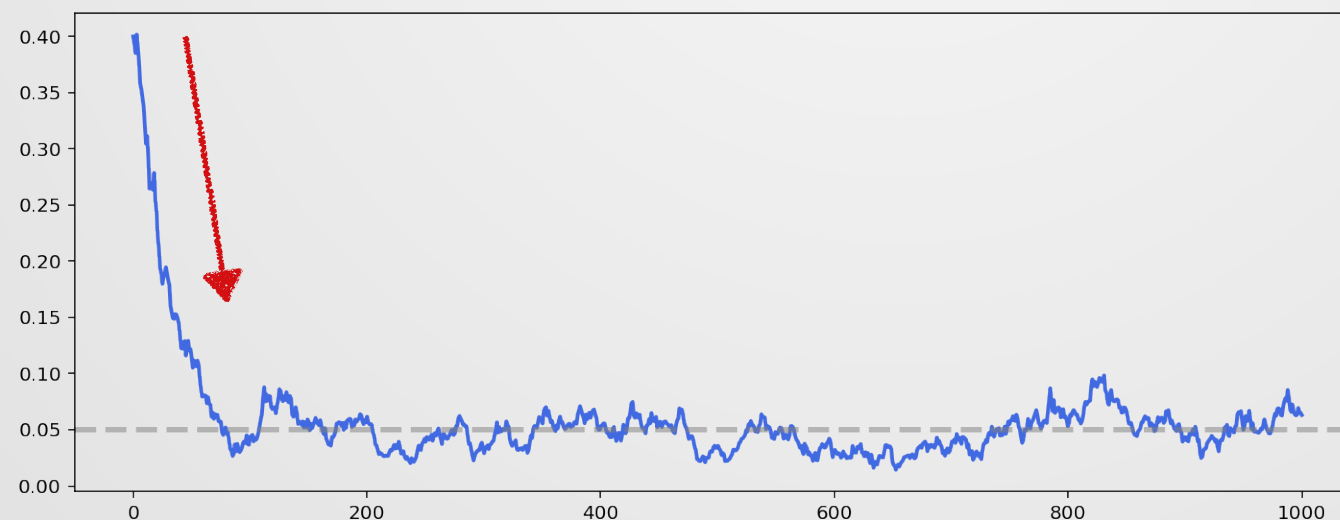
# CIR (Cox-Ingersoll-Ross) Model

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- Originally describes the evolution of interest rates
- Also suitable for describing non-negative variables such as volatility
- One-factor model (driven by standard Wiener's process)

$$dV_t = \kappa (\theta - V_t) dt + \sigma_v \sqrt{V_t} dW_t^v$$

- Applying the Feller condition  $2\kappa\theta > \sigma_v^2$  to make sure it won't reach 0



$$\begin{aligned}\kappa &= 0.2 \\ \theta &= 0.05 \\ \sigma_v &= 0.02 \\ V_0 &= 0.4 \\ T &= 1000\end{aligned}$$

