

Dupire's Local Dependent Volatility Model

Idea: Suppose we have option prices. Can we build a model which exactly fits those option prices.

$$dS = \sum_t dW_t + b_t dt$$

- Option prices are calculated under \mathbb{Q} , so we look for a model under \mathbb{Q}

$$dS = \sum_t dW_t^{\mathbb{Q}} + rS_t dt$$

- Try first for a (time and space inhomogeneous) model

$$dS_t = S_t \sigma(S_t, t) dW_t^{\mathbb{Q}} + rS_t dt.$$

Suppose the \mathbb{Q} and write $dS_t = S_t \sigma(S_t, t) dW + rS_t dt$; \mathbb{E} not $\mathbb{P}^{\mathbb{Q}}$.

We assume we have a double continuum of option prices

$$\{C(K, T)\}_{T \geq 0, K \geq 0}$$

- We suppose they are smooth so that we can find $\frac{\partial C}{\partial T}$ or $\frac{\partial C}{\partial K}$ etc
- Change of viewpoint: before for fixed (K, T) we found a PDE for C as function of (t, S_t) . Now we fix $t=0$ and S_0 and consider C as a function of option characteristics.

$$C(K, T) = \mathbb{E}[e^{-rT}(S_T - K)^+] = e^{-rT} \mathbb{E}(S_T - K)^+$$

Differentiating wrt K $\frac{\partial}{\partial K}(S_T - K)^+ = -1_{(S_T > K)}$ $\frac{\partial^+}{\partial K^-}(S_T - K)^+ = -\delta_K(S_T)$

$$\left. \begin{aligned} \frac{\partial C}{\partial K} &= -\mathbb{E}[e^{-rT} \cdot 1_{(S_T > K)}] = -e^{-rT} \mathbb{E}[1_{(S_T > K)}] \\ \frac{\partial^+ C}{\partial K^-} &= +e^{-rT} \mathbb{E}\left[\delta(S_T + \Delta K)\right] = e^{-rT} p(T, K) \\ &= e^{-rT} \int_{-\infty}^K p_T(s) \delta_K(s) ds = e^{-rT} p_T(K). \end{aligned} \right\}$$

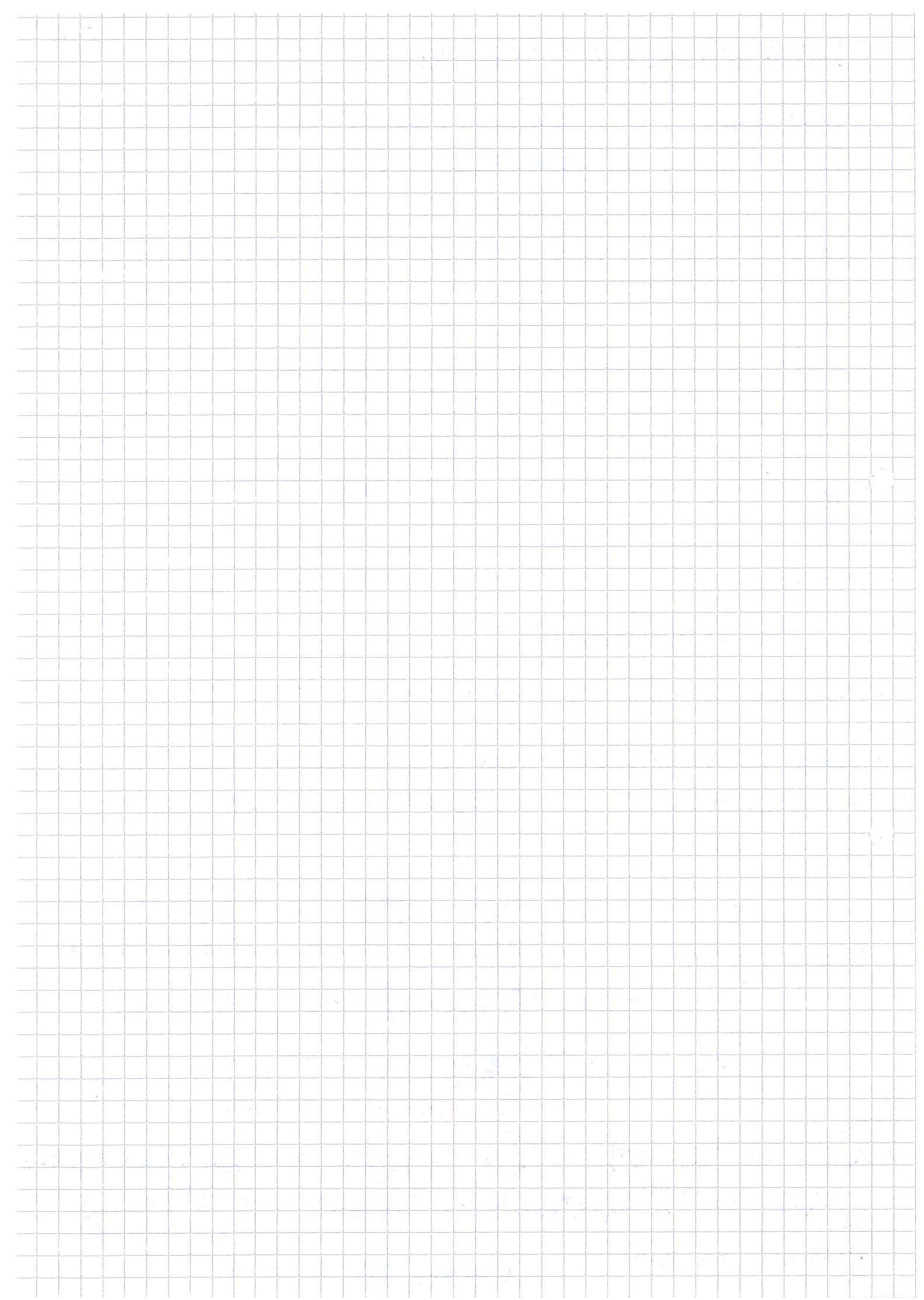
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$$C(K, T) = \mathbb{E}[e^{-rT}(S_T - K)^+] = e^{-rT} \int_0^\infty (s - K)^+ p_T(s) ds \quad \text{where } p_T(s) \text{ is the density of } S_T$$

$$\frac{\partial C(K, T)}{\partial K} = -e^{-rT} \int_0^\infty 1_{(S_T > K)} p_T(s) ds = -e^{-rT} \int_K^\infty p_T(s) ds = -e^{-rT} \mathbb{E}^{\mathbb{Q}}[1_{(S_T > K)}] = -e^{-rT} \mathbb{Q}(S_T > K)$$

$$\frac{\partial^+ C(K, T)}{\partial K^-} = +e^{-rT} p_T(K)$$

All assuming $p_T(s)$ has a density



For any regular f and $\Delta > 0$

$$\begin{aligned} e^{-r(T+\Delta)} f(S_{T+\Delta}) &= e^{-rT} f(S_T) + \int_T^{T+\Delta} e^{-ru} f'(S_u) dS_u + \int_T^{T+\Delta} \frac{1}{2} e^{-ru} f''(S_u) d[u]_u \\ &\quad - r \int_T^{T+\Delta} e^{-ru} f(S_u) du \\ &= e^{-rT} f(S_T) + \int_T^{T+\Delta} e^{-ru} r S_u f'(S_u) du \\ &\quad + \int_T^{T+\Delta} e^{-ru} \sigma(S_u, u) S_u dW_u + \int_T^{T+\Delta} \frac{1}{2} d[u]_u e^{-ru} f''(S_u) S_u \sigma(S_u, u) \\ &\quad - r \int_T^{T+\Delta} e^{-ru} f(S_u) du \end{aligned}$$

$$\frac{\mathbb{E}[e^{-r(T+\Delta)} f(S_{T+\Delta}) - e^{-rT} f(S_T)]}{\Delta}$$

$$\begin{aligned} &= \frac{1}{\Delta} \mathbb{E}\left[\int_T^{T+\Delta} r e^{-ru} S_u f'(S_u) du\right] \\ &\quad + \frac{1}{\Delta} \mathbb{E}\left[\int_T^{T+\Delta} e^{-ru} \frac{S_u^2}{2} \sigma(S_u, u)^2 f''(S_u) du\right] \\ &\quad - \frac{1}{\Delta} \mathbb{E}\left[\int_T^{T+\Delta} e^{-ru} f(S_u) du\right] \end{aligned}$$

Take $f(s) = (s - K)^+$

$$\begin{aligned} \frac{\partial C(K, T)}{\partial T} &= r e^{-rT} \mathbb{E}[S_T 1_{(S_T > K)}] + \frac{1}{2} e^{-rT} K^2 \sigma(K, T)^2 \rho(T, K) \\ &\quad - r e^{-rT} \mathbb{E}[(S_T - K)^+] \end{aligned}$$

$$= e^{-rT} r K \mathbb{E}[1_{(S_T > K)}] + e^{-rT} \frac{K^2}{2} \sigma(K, T)^2 \rho(T, K) - rC$$

$$C' = -rK C' + \frac{1}{2} K^2 \sigma(K, T)^2 C'' - rC$$

$$\boxed{\sigma(K, T)^2 = \frac{C' + rKC' + rC}{\frac{1}{2} K^2 C''}} = \frac{2(C' + rKC')}{K^2 C''}$$

$C \downarrow K$, $C' + rKC' > 0$, C convex in K .

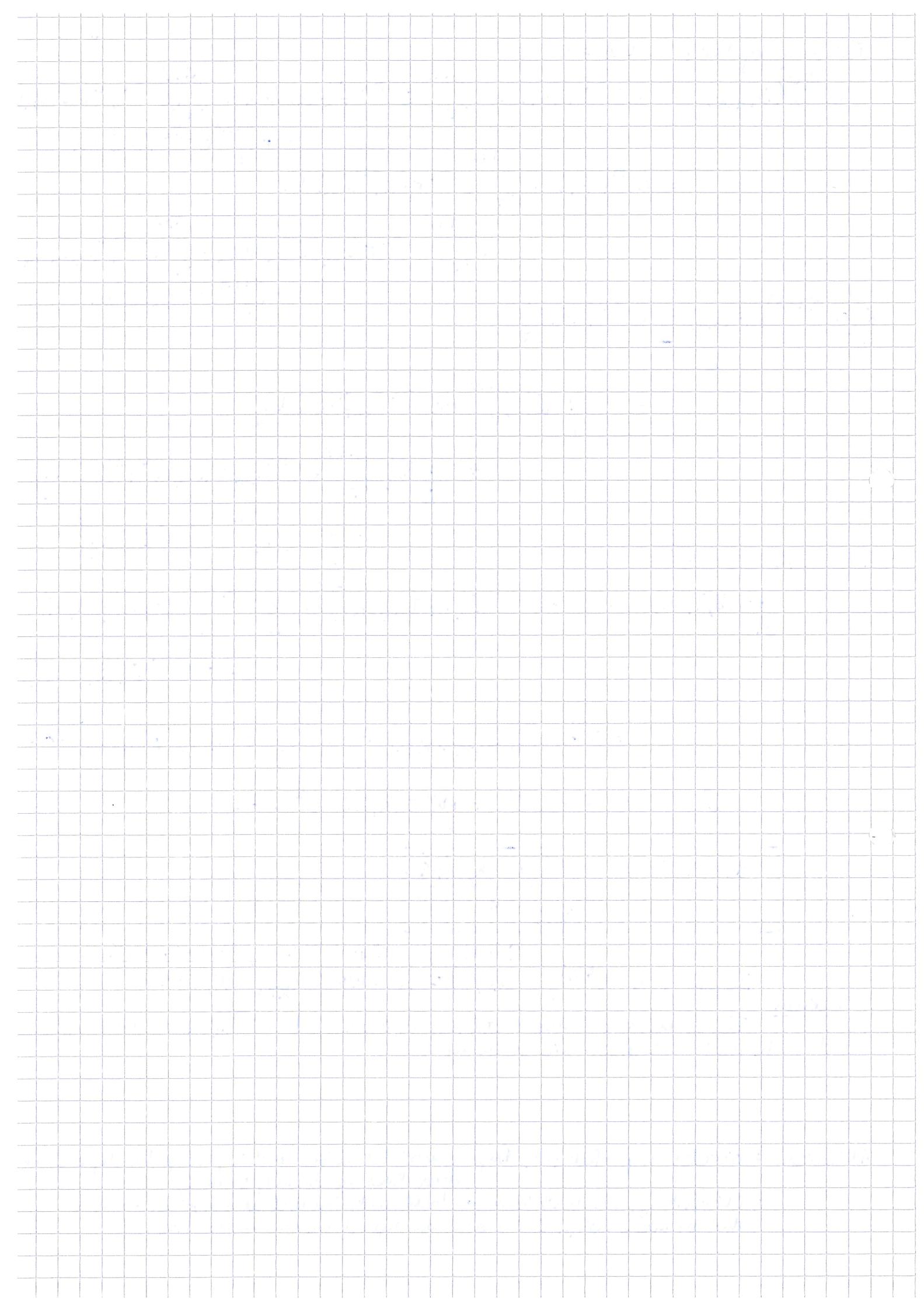
Suppose $C \in C^{2,1}$, $C(0, T) = S_0$ and $\lim_{K \uparrow \infty} C(K, T) = 0$

Theorem

There exists a unique diffusion such that $e^{-rt} S_t$ is a martingale and such that call prices are given by C .

Advantages: Provides a perfect fit. Dupire model can be used to price path-dependent derivatives eg. traps, barriers,

: Overfitting, we don't have a double continuum of option prices, so must start by interpolating. Our answer will depend on how we do this.



Diffusion Models of Volatility

The idea is to model the volatility as a process in its own right (typically as a diffusion,

Allows for historic variation in volatility levels, and future uncertainty over volatility

Can incorporate leverage, either explicitly, or through correlation between price level and volatility.

Model (S, Y) is a bi-variate diffusion under \mathbb{P}

$$(*) \begin{cases} dS = S \left\{ \sigma(Y) dW + \mu(Y) dt \right\} \\ dY = \alpha(Y) dS + \beta(Y) dt \end{cases} \quad \begin{array}{l} S_0 = s \\ Y_0 = y \end{array}$$

$$d[\theta, w] = \rho(Y) dt$$

More generally we could allow σ and μ to depend on S .

- We have written down the model under \mathbb{P} . Sometimes want to write it down under \mathbb{Q} .
- Making μ and σ depend on S is not necessary for leverage. Can we use correlation instead.

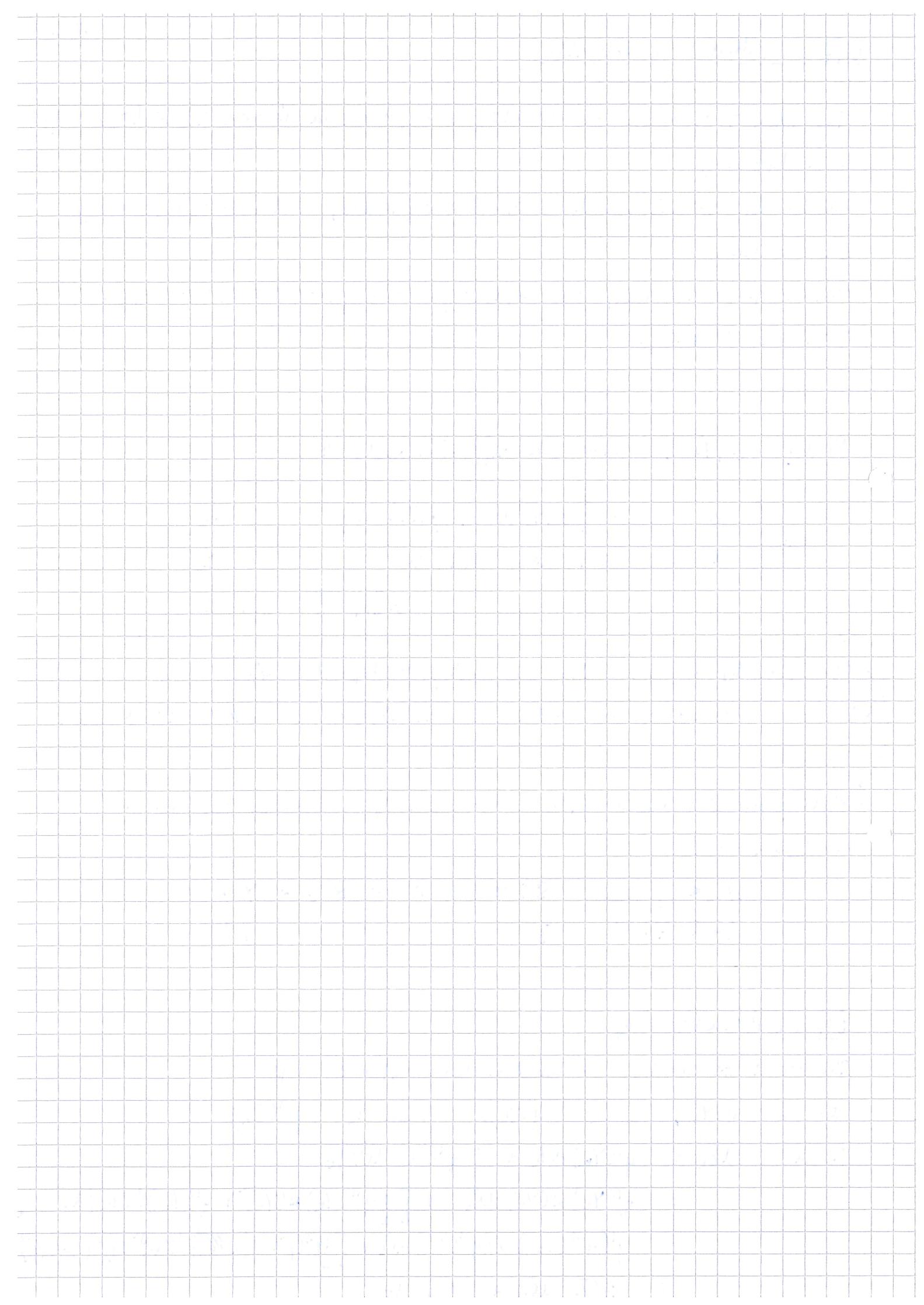
If $\rho < 0$, then $S \uparrow \overset{\text{N}}{\leftrightarrow} W \uparrow \sim B \downarrow \sim Y \downarrow$
 (and $\sigma(\cdot) > 0$) $S \downarrow \qquad \qquad \qquad Y \uparrow$

Different Parameterizations

The model (*) can be parameterized in many different ways

Let $Z = y(Y)$, i.e. $Z_t = y(Y_t)$. Then $Y_t = y^{-1}(Z_t)$

$$\begin{aligned} dZ_t &= y'(Y_t) dY_t + \frac{1}{2} y''(Y_t) d[Y]_t \\ &= y'(Y_t) \alpha(Y_t) dB_t + \left[\frac{1}{2} y''(Y_t) \alpha(Y_t)^2 + y'(Y_t) \beta(Y_t) \right] dt \\ &= y'(y^{-1}(Z_t)) \alpha(y^{-1}(Z_t)) dB_t + \left[\frac{1}{2} y''(y^{-1}(Z_t)) \alpha(y^{-1}(Z_t))^2 + y'(y^{-1}) \cdot \beta(y^{-1}) \right] dt \end{aligned}$$



There are two natural choices: $\Sigma = \sigma(Y)$

$$\text{or } V = \mathbb{E}[\sigma(Y)^2]$$

In the former case

$$dS_t = S_t [\Sigma_t dW_t + \mu_\Sigma(\Sigma_t) dt]$$

$$d\Sigma_t = a_\Sigma(\Sigma_t) d\delta_t + b_\Sigma(\Sigma_t) dt$$

$$d[\delta, \omega] = \rho(\Sigma_t) dt$$

In the latter case

$$dS_t = S_t [\sqrt{V_t} dW_t + \mu_V(V_t) dt]$$

$$dV_t = a_V(V_t) d\delta_t + b_V(V_t) dt$$

Choices & Model

We choose models to be tractable, so we can calculate option prices/expect have realistic properties (eg $\lambda \leq 0$) provide a good fit to data.

Some We want If we model via V , we should require $V \geq 0$.

If we model using Σ we have the issue that we see $[S]_t$ and therefore Σ^L , but may not be able to observe Σ .

Question: Is $\mathbb{F}_t^S = \sigma((S_u); 0 \leq u \leq t)$ equal to

$$\mathbb{F}_t^{S, \omega} = \mathbb{F}_t^{S, \Sigma} = \sigma((S_u, \Sigma_u); 0 \leq u \leq t)?$$

Some models

i) Σ is exponential brownian motion

If $\rho = 0$ Hull-White

If $\rho \neq 0$ Wiggins

ii) $\Sigma = e^u$ where u is an OU process.

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$$du = \delta dB + (\beta - \lambda u) dt$$

β is non-centrality

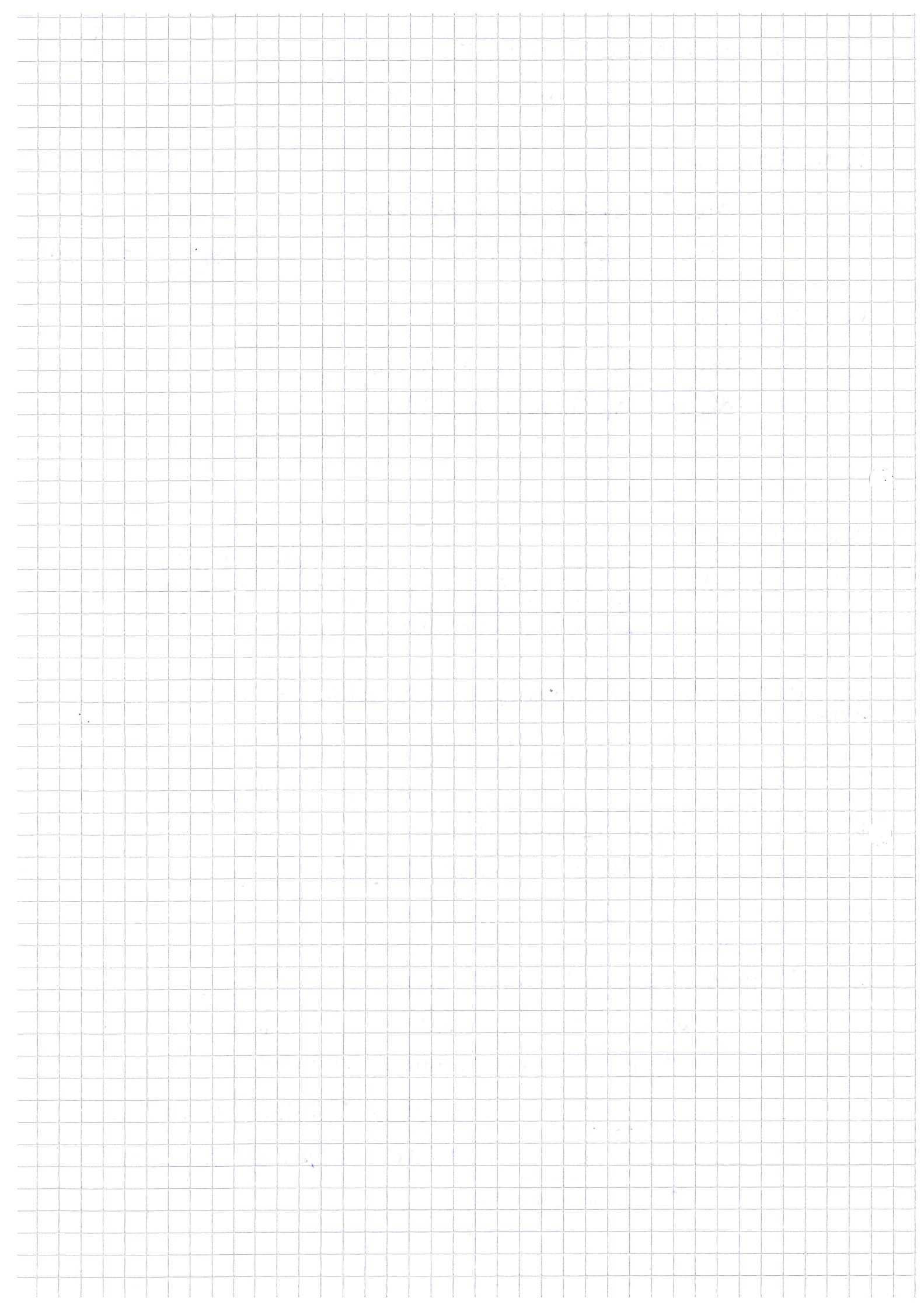
$$\Sigma = e^u$$

$$d\Sigma = e^u du + \frac{1}{2} e^{2u} d[u]$$

$$= \delta \Sigma dB + \Sigma (\beta - \lambda u \Sigma) dt + \frac{\delta^2}{2} \Sigma dt$$

$$= \delta \Sigma dB + \Sigma \left(\beta + \frac{\delta^2}{2} - \lambda u \Sigma \right) dt$$

Σ is a mean reverting process.



3) Σ is an OU process.

Note: we can instantaneously have $\Sigma = 0$, and Σ is R-valued.

From observing S we can see $V = \Sigma^L$, but we do not know if $\Sigma = +\sqrt{V}$ or $\Sigma = -\sqrt{V}$

$$d\Sigma = \gamma dB + (\beta - \lambda \Sigma) dt$$

Exercise
 $dU_t^{(i)} = \theta dB_t^{(i)} - \frac{\lambda}{2} U_t^{(i)} dt$

4) Square-root Model (Heston)

$$dV_t = a\sqrt{V_t} dB_t + (b - cV_t) dt$$

$$\frac{d\Sigma_t}{\Sigma_t} = \sqrt{V_t} d\overset{W}{B}_t + \mu(V_t) dt$$

$$\text{Then } \Sigma_t = \sqrt{V_t}$$

$$\begin{aligned} d\Sigma_t &= \frac{1}{2\sqrt{V_t}} dV_t - \frac{1}{8} \frac{(dV_t)^2}{V_t^{3/2}} \\ &= \frac{1}{2\Sigma_t} (a\Sigma_t dB_t + (b - c\Sigma_t^L) dt) - \frac{1}{8} \frac{\Sigma_t^L}{\Sigma_t^3} dt \\ &= \frac{a}{2} dB_t + \left(\frac{b}{2} - \frac{a^2}{8} \right) \frac{1}{\Sigma_t} dt - \frac{c\Sigma_t}{2} dt \end{aligned}$$

5) β_2 Model

$$dV_t = a\sqrt{V_t} dB_t + (bV - cV^2) dt$$

$$d\Sigma_t = \frac{a}{2} \Sigma_t^L dB_t + \left(\frac{b}{2} \Sigma_t - \frac{c}{2} \Sigma_t^3 \right) dt - \frac{a}{8} \Sigma_t^3 dt$$

$$= \frac{a}{2} \Sigma_t^L dB_t + \left(\frac{b}{2} \Sigma_t - \left(\frac{c}{2} + \frac{a^2}{8} \right) \Sigma_t^3 \right) dt$$

6) Heston-Nandi Model.

Square-root model, but value $\rho = -1$.

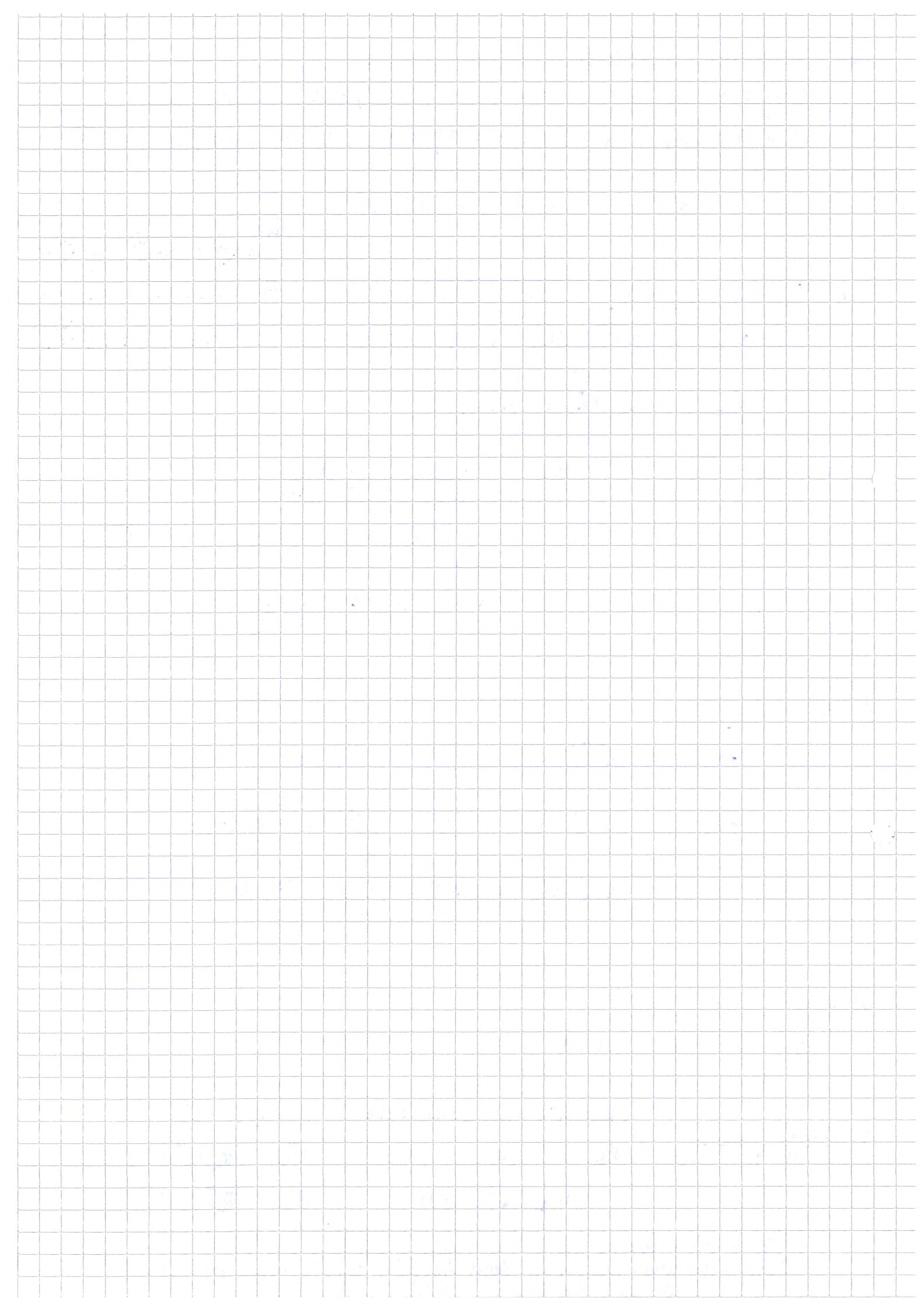
(Hagan, Kumar, Lesniewski & Woodward) (2002)

7) SABR model
 $\ln F_t = S_t e^{-rt} d\tilde{S}_t \approx$ (Forward price) $dF_t = \sum_i F_t^{\beta_i} dW_t$

$$d\{B_t, W_t\} = \rho dt \quad d\Sigma_t = \sum_i \Sigma_t^i dB_t$$

Stochastic - alpha-beta-rho model.

Designed for small maturities. Allegedly easy to relate parameters to the shape of the implied volatility smile.



Option pricing in Stochastic Volatility Models

Suppose $dS = S\sqrt{V}dW + M^{(S,V)}dt$

Definition

\mathbb{Q} is an equivalent martingale measure if \mathbb{Q} is equivalent to \mathbb{P} and $(e^{-rt}S_t)$ is a \mathbb{Q} -martingale. (For an EMM - $(e^{-rt}S_t)$ is a \mathbb{Q} -local martingale)

A natural approach is to understand price under an EMM

i) Suppose the model is written down under \mathbb{Q}

$$dS = S\sqrt{V}dW + rSdt \quad dV = a(V)dS + b(V)dt$$

$$= a(V)\{p(V)dW + \sqrt{1-p(V)^2}dW^H\} + b(V)dt$$

W and B are \mathbb{Q} -Brownian motions

$$C_t := \mathbb{E}^{\mathbb{Q}}[e^{-r(T-t)}H(S_T) | \mathcal{F}_t]$$

$$C_t = C(S_t, V_t, t) = e^{rt}\mathbb{E}[H(S_T) | \mathcal{F}_t]$$

$$\therefore (e^{-rt}C(S_t, V_t, t))_{t \geq 0} =: M_t \text{ is a martingale}$$

$$\text{eg } H(S_T) = (S_T - K)_+$$

$K > 0$ and $T > 0$ are fixed constants.

Since M is a martingale we can derive an ODE for C :

$$\begin{aligned} dM_t &= d[e^{-rt}C_t] = -re^{-rt}C(S_t, V_t, t)dt + e^{-rt}\left\{C_t dS + C_v dV + \frac{C_{ss}}{2}d[S]_t + C_{sv}d[S, V]_t + C_{vv}d[V]_t\right\} \\ &= e^{-rt}C_s S\sqrt{V}dW + e^{-rt}C_v a(V)\{p(V)dW + \sqrt{1-p(V)^2}dW^H\} \\ &\quad + e^{-rt}\left[-rC + rSC_s + b(V)C_v + \frac{8}{2}V C_{ss} + S\sqrt{V}p(V)C_{sv}a(V) + \frac{1}{2}a(V)^2C_{vv}\right] dt \end{aligned}$$

C solves

$$-rC + rSC_s + b(V)C_v + \frac{s^2V}{2}C_{ss} + S\sqrt{V}a(V)p(V)C_{sv} + \frac{1}{2}a(V)^2C_{vv} = 0$$

subject to

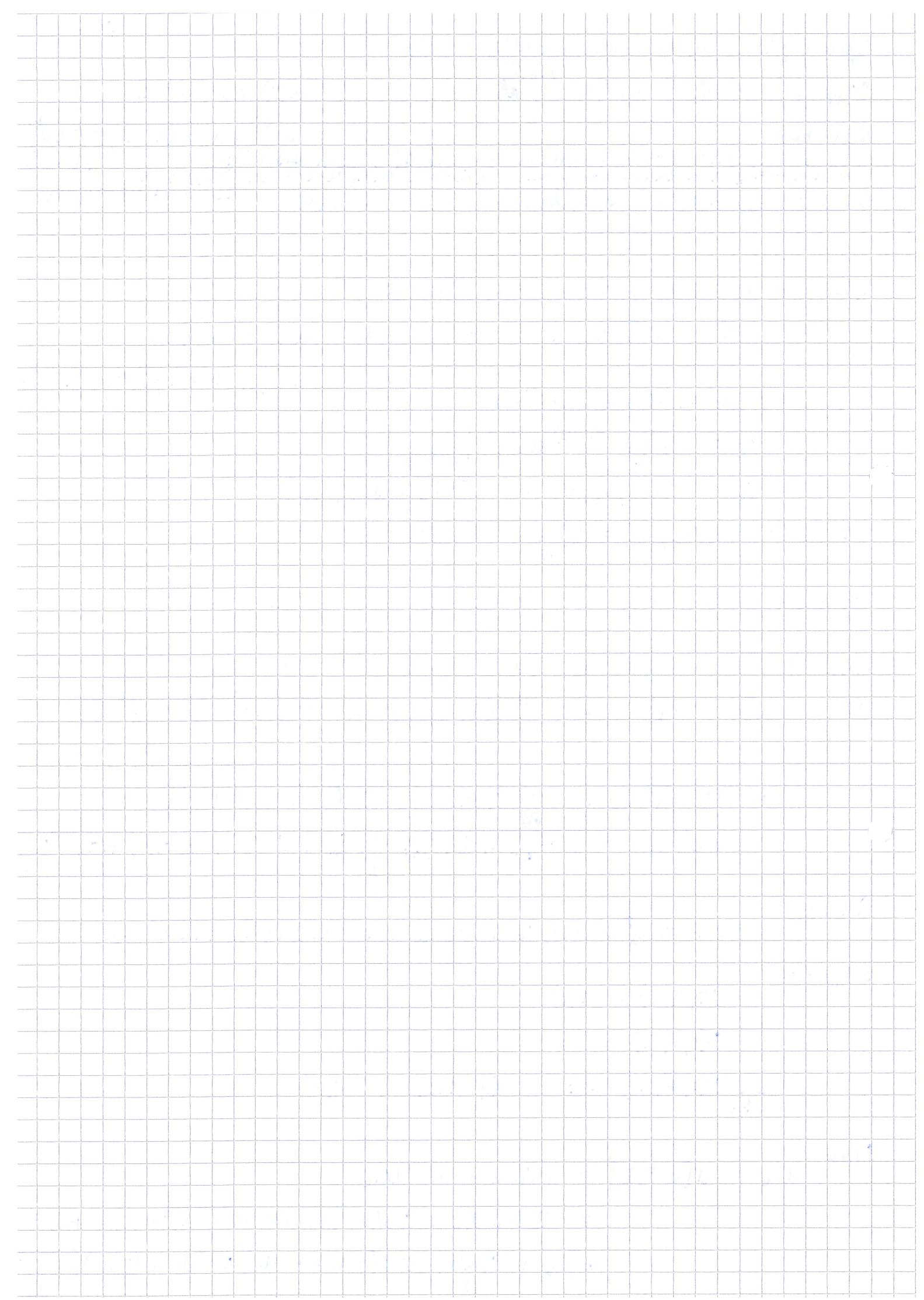
$$C(s, v, T) = H(s)$$

$$C(0, v, T) = 0$$

$$\lim_{v \downarrow 0} C(s, v, T) = s \quad (\text{if } H(s) = (s - K)_+)$$

Note

$$\begin{aligned} dC_t &= d[e^{rt}M_t] = rC_t dt + e^{rt}dM_t \\ &= C_s S\sqrt{V}dW + C_v a(V)\{p(V)dW + \sqrt{1-p(V)^2}dW^H\} + rC_t dt \\ &= \{C_s + C_v a(V)p(V)\}((dS - rSdt) + rC_t dt + C_v a(V)\sqrt{1-p(V)^2}dW^H) \end{aligned}$$



$$\text{Set } \mathbb{H} = C_0 + \frac{\text{Cva}(V)}{S\sqrt{V}} \rho(V)$$

$$dC_t = \mathbb{H} dS_t + r(C - \mathbb{H} S_t) dt + \alpha(V) C \sqrt{1 - \rho(V)^2} dW_t^+$$

If we hold \mathbb{H} shares

$$H(S_T) = C(S_T, V_T, T) = C_0 + \int_0^T \mathbb{H}_u dS_u + \int_0^T r(C - \mathbb{H}_u S_u) du + \int_0^T \alpha(V_u) C_u \sqrt{1 - \rho(V_u)^2} dW_u^+$$

$$C_0 + \underbrace{\int_0^T \mathbb{H}_u dS_u + \int_0^T r(C_u - \mathbb{H}_u S_u) du}_{\text{Terminal wealth ; gains from trade}} = H(S_T) - \underbrace{\int_0^T \alpha(V_u) C_u \sqrt{1 - \rho(V_u)^2} dW_u^+}_{\text{Option payout} + \text{hedging error}}$$

A) $\int_0^T \alpha(V_u) C_u \sqrt{1 - \rho(V_u)^2} dW_u^+$ is unhedgeable.

Ignore it. Hope that it cancels out if we have a large portfolio of options sometimes win, sometimes lose.

B) Suppose there was a second option which traded on the market, with price $P = P(S_t, V_t, t)$.

We could write

$$dS_t = \Phi dS_t + r(P - \Phi S_t) dt + \alpha(V) \sqrt{1 - \rho(V)^2} dW_t^+ P_V$$

Consider holding a portfolio of $\mathbb{H}_t - \frac{C_t}{P_V}$ shares

C_V/P_V units of the second option

$$C - \left(\mathbb{H} - \frac{C_V}{P_V} \right) S_t - \frac{C_V P}{P_V} \text{ cash}$$

Change in value of this portfolio is

$$\left(\mathbb{H}_t - \frac{C_V}{P_V} \right) dS_t + C_V dP_V + r \left(C - \left(\mathbb{H} - \frac{C_V}{P_V} \right) S - \frac{C_V P}{P_V} \right) dt$$

$$= \mathbb{H} dS_t + r(C - \mathbb{H} S) dt + \frac{C_V}{P_V} \left[dP_V - \Phi dS_t + \Phi S r dt - r P dt \right] = dC_t$$

