



Smile Curve and Local Volatility

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Mathematical Finance

- Undergraduate
 - Department of Mathematics
 - (Department of Mathematical Science)
 - Mathematics Major
 - Statistics Major
- Graduate School
 - Fundamental Science and Technology
 - Integrative Design Engineering
 - Science for Open and Environmental Systems

Probability
Theory for Life
Science

Statistics Major

3rd year
Spring

Statistical
Science

Introduction
to Probability
Theory

3rd year
Autumn

Data Analysis

Seminar on
Statistical Science

Mathematical
Statistics I

Data
Sampling

Probability
Theory I

Risk Mathematics

4th year

Time Series
Analysis

Mathematical
Planning

Mathematical
Statistics II

Non-linear
Models

Advances
in Statistical
Mathematics I

Advances
in Statistical
Mathematics II

Probability
Theory II

Actuarial
Mathematics

Graduate School

Data Literacy

by Ritei Shibata

Data Sciences

by Invited professors

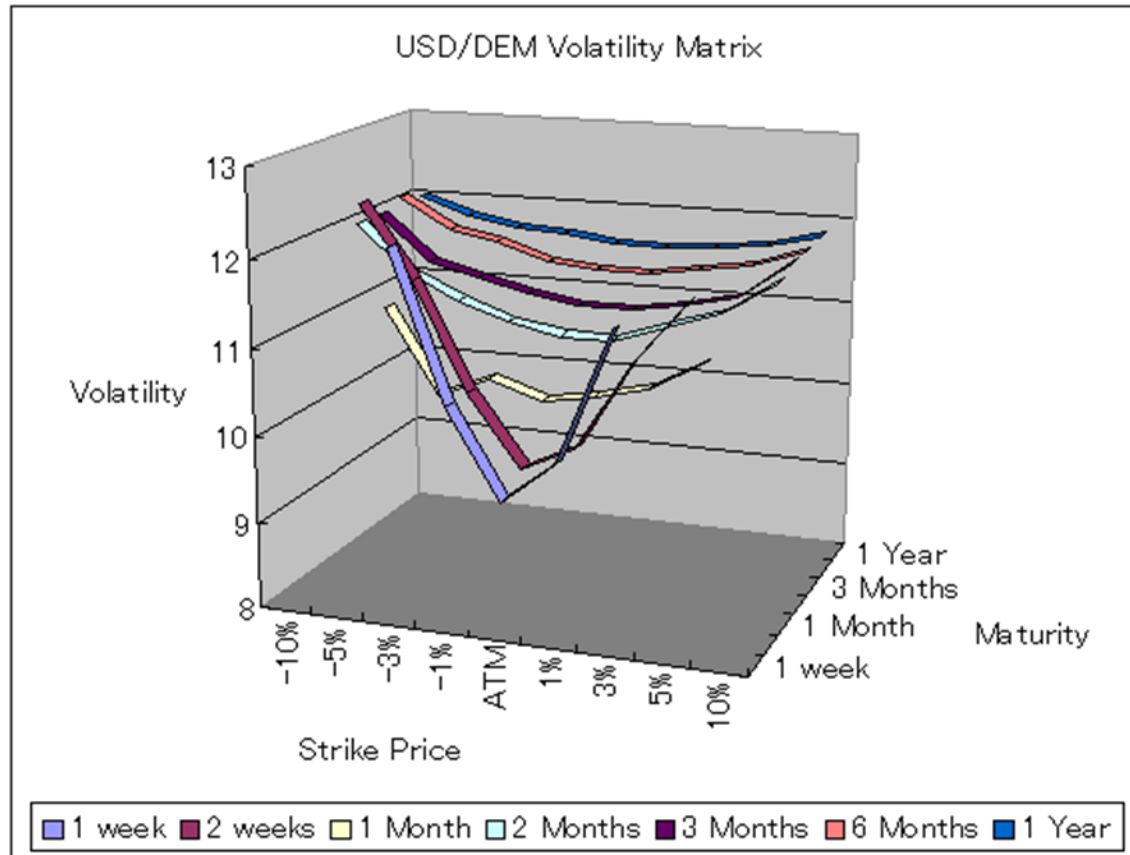
Mathematical Finance

by Ritei Shibata and Makoto Maejima

Advances in Mathematical
Finance

by Nobuhiro Nakamura
(Hitotsubashi University)

Financial Smiles



A smile is a curve that sets everything straight



Icon	Meaning
: -)	classic smile with nose
: - (	classic frown with nose (Unicode: ☹)
:)	classic smile without nose
: (	classic frown without nose

Option Price

$$c(t, x) = e^{-rt} \int_{\log(x)}^{\infty} (e^y - x) f_t(y) dy : \text{Call Option Price}$$

t : *Time of Execution (Maturity)*

x : *Execution Price*

f_t : *Distribution of $\log(S_t / S_0)$*

S_t : *Underlying Asset Price*

Black-Scholes

Assumption

$$dS_t = r S_t dt + \sigma S_t dB_t \quad (\text{log normal process})$$

$$c(t, x) = e^{-rt} \int_{\log(x)}^{\infty} (e^y - x) f_t(y) dy$$



$$= S_0 \Phi(-\lambda - \sigma\sqrt{t}) - x e^{-rt} \Phi(-\lambda),$$

$$\lambda = \frac{\log(x / S_0) - (r - \sigma^2 / 2)t}{\sigma\sqrt{t}}$$

Implied Volatility

$c(t, x) = S_0 \Phi(-\lambda - \sigma\sqrt{t}) - x e^{-rt} \Phi(-\lambda) : \text{Black Scholes Formula}$

$$\lambda = \frac{\log(x / S_0) - (r - \sigma^2 / 2)t}{\sigma\sqrt{t}}$$

$c(t, x) : \text{Option Price at } t = 0$

 $\sigma_{ip}(t, x) : \text{Implied Volatility}$

Smile or not Smile

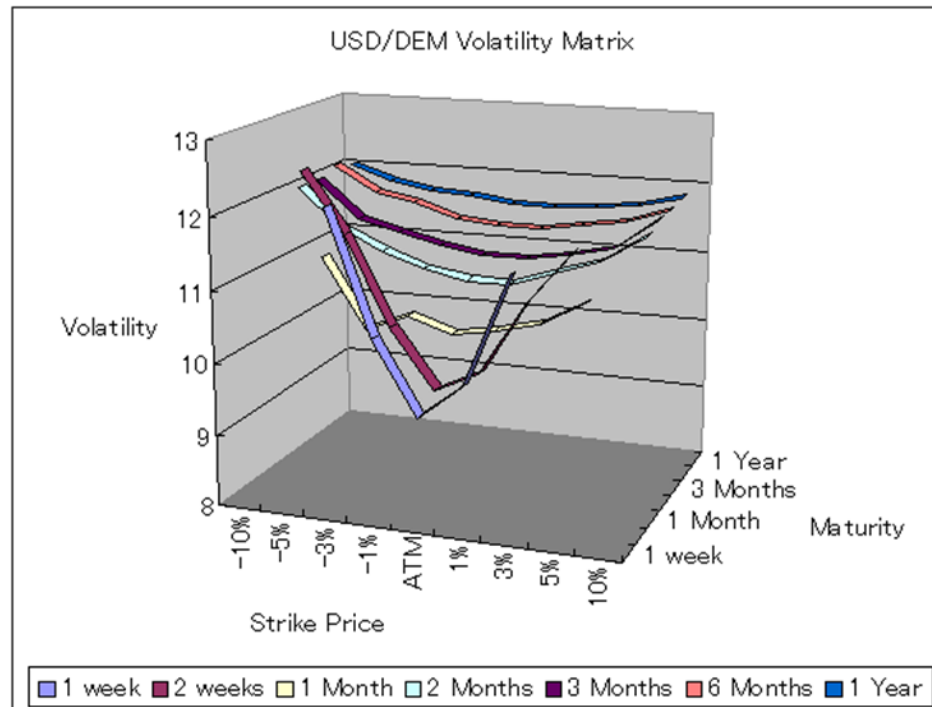
Assumption

$$dS_t = r S_t dt + \sigma S_t dB_t \text{ (log normal process)}$$



$$\sigma_{ip}(t, x) \equiv \sigma$$

Reality



What is wrong

Assumption

$$dS_t = r S_t dt + \sigma S_t dB_t \text{ (log normal process)}$$

$$\longleftrightarrow \log(S_t / S_0) \sim N((r - \sigma^2 / 2)t, \sigma^2 t)$$

$$\longleftrightarrow f_t : \text{Normal density}$$

The distribution?

$$c(t, x) = e^{-rt} \int_{\log(x)}^{\infty} (e^y - x) f_t(y) dy$$

$$\Rightarrow \frac{\partial}{\partial x} c(t, x) = -e^{-rt} (1 - F_t(\log x)) \text{ for } x > 0$$

$$\Rightarrow \frac{\partial^2}{\partial x^2} c(t, x) = -e^{-rt} \frac{1}{x} f_t(\log x) \text{ for } x > 0$$

Estimation of the distribution

$$\frac{\partial^2}{\partial x^2} c(t, x) = -e^{-rt} \frac{1}{x} f_t(\log x) \text{ for } x > 0$$

$$c(t, x) = S_0 \Phi(-\lambda - \sigma_{ip}(t, x)\sqrt{t}) - x e^{-rt} \Phi(-\lambda),$$

$$\lambda = \frac{\log(x/S_0) - (r - \sigma_{ip}(t, x)^2/2)t}{\sigma_{ip}(t, x)\sqrt{t}}$$



$$\sigma_{ip}(t, x) \Rightarrow f_t$$

Some Attempts

- Shimko(1993), RISK 6, 33-37
 - quadratic model for implied volatility
- Brunner and Hafner (2003), J. Compt. Finance
 - mixture of log normals for the distribution

$$\frac{\partial^2}{\partial x^2} c(t, x) = e^{-rt} \frac{1}{x} f_t(\log x) \text{ for } x > 0$$

$$c(t, x) = S_0 \Phi(-\lambda - \sigma_{ip}(t, x)\sqrt{t}) - x e^{-rt} \Phi(-\lambda)$$



Assumption

$$dS_t = r S_t dt + \sigma S_t dB_t \text{ (log normal process)}$$

Local Volatility Model

$$dS_t = r S_t dt + \sigma(t, S_t) S_t dB_t$$

$$\tilde{c}(t, x) = e^{rt} c(t, x)$$



$$\frac{\partial}{\partial t} \tilde{c}(t, x) = \frac{x}{2} \left(\sigma^2(t, x) f_t(\log x) \right)$$



$$\frac{\partial^2}{\partial x^2} \tilde{c}(t, x) = \frac{1}{x} f_t(\log x) \text{ for } x > 0$$

$$\sigma^2(t, x) = \frac{2 \frac{\partial}{\partial t} \tilde{c}(t, x)}{x^2 \frac{\partial^2}{\partial x^2} \tilde{c}(t, x)} ; \text{ Dupire(1994)}$$

Estimation of local volatility

Assumption

$$dS_t = r S_t dt + \sigma(t, S_t) S_t dB_t$$

$$\tilde{c}(t, x) = e^{rt} c(t, x)$$

$$\frac{\partial}{\partial t} \tilde{c}(t, x) = \frac{x}{2} \left(\sigma^2(t, x) f_t(\log x) \right)$$

Nicolas Rousseanu (2007)

$$\text{Scaling Assumption : } \frac{\log(S_t / S_0) - \mu(t)}{s(t)} \stackrel{d}{=} \frac{\log(S_1 / S_0) - \mu(1)}{s(1)}, t > 0$$

Gram – Charlier Expansion of f_t

$$\frac{\partial}{\partial t} \tilde{c}(t, x), \mu(t), s(t), \text{ Cummulants of } \frac{\log(S_t / S_0) - \mu(t)}{s(t)} \Rightarrow \sigma^2(t, x)$$

$\sigma_{ip}(t, x) : \textit{Smile Surface}$

$\longleftrightarrow \frac{\partial}{\partial t} \tilde{c}(t, x)$

$\longleftrightarrow \sigma^2(t, x) : \textit{Local Volatility}$

$$\frac{\partial}{\partial t} \tilde{c}(t, x) = \frac{x}{2} \left(\sigma^2(t, x) f_t(\log x) \right) > 0 ?$$

$$\frac{\partial^2}{\partial x^2} \tilde{c}(t, x) = \frac{1}{x} f_t(\log x) > 0 ?$$

Our results

$$(A) \quad \frac{\partial}{\partial t} \log \sigma_{ip}(t, x) > -\frac{1}{2t} \quad \Leftrightarrow \quad \frac{\partial}{\partial t} \tilde{c}(t, x) > 0$$

$$(B) \quad 1 + x \left(\frac{\partial}{\partial x} \log \sigma_{ip}(t, x) \right) \left(\frac{t}{2} \sigma_{ip}^2(t, x) - \log(x / S_0) \right) > 0 \quad \Rightarrow \quad \frac{\partial^2}{\partial x^2} \tilde{c}(t, x) > 0$$

$$(A), (B) \Rightarrow \quad \sigma^2(t, x) = \sigma_{ip}^2(t, x) \frac{1 + 2t \frac{\partial}{\partial t} \log \sigma_{ip}(t, x)}{1 + x \left(\frac{\partial}{\partial x} \log \sigma_{ip}(t, x) \right) \left(\frac{t}{2} \sigma_{ip}^2(t, x) - \log(x / S_0) \right)}$$

An example

$$\sigma_{ip}(t, x) = at^{\xi}e^{bx}$$

$$\xi > -\frac{1}{2} \Leftrightarrow (A)$$

$$a^2 > \frac{1}{bt^{2\xi+1}S_0} \left(\frac{1}{bS_0} + 1 \right) \Rightarrow (B)$$

$$\sigma^2(t, x) = \frac{(at^{\xi}e^{bx})^2(2\xi+1)}{1 + \frac{bxt^{2\xi+1}}{2}(ae^{bx})^2 - bx \log \frac{x}{S_0}}$$

$$f_t(\log x) = \frac{\phi(\lambda) \left\{ 1 + bx \left(\frac{t}{2} (at^{\xi}e^{bx})^2 - \log \frac{x}{S_0} \right) \right\}}{at^{\xi} \sqrt{t} e^{bx}}$$

A smile is a curve which invites us to
a fascinating world.

Von Voyage!