

SABR

Antoine Jacquier

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AUTHOR:	Antoine Jacquier
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1 SABR Expansions and arbitrage

```
In [9]: import numpy as np
import matplotlib.pyplot as plt
```

The SABR model is the most popular model on Fixed Income desks. We investigate here some of its pitfalls. The forward process $(F_t)_{t \geq 0}$ is the unique strong solution to the following stochastic differential equation:

$$\begin{aligned} dF_t &= \sigma_t F_t^\beta \left(\rho dW_t^1 + \sqrt{1 - \rho^2} dW_t^2 \right), \\ d\sigma_t &= \nu \sigma_t dW_t^2, \end{aligned}$$

with $F_0 = 1$, $\sigma_0 = \alpha$, $\rho \in [-1, 1]$ and $\beta \leq 1$. We shall denote by $I_t(x)$, the implied volatility corresponding to European options with maturity t and log strike $x = K/F_0$.

2 Computing the density

We compute the density function corresponding to a given smile, at some fixed time horizon $t > 0$.

```
In [15]: def density(smile, t, x, eps):
    # Computes the density at a point  $x = \log(K/S)$  given a smile
    # function
    w = smile(x) * t
    wp = smile(x + eps) * smile(x + eps) * t
    wm = smile(x - eps) * smile(x - eps) * t
    w1 = (wp - wm) / (2. * eps)
    w2 = (wp - 2.0 * w + wm) / (eps * eps)
    g = (1. - 0.5 * x * w1 / w) * (1. - 0.5 * x * w1 / w) - \
        0.25 * w1 * w1 * (1. / w + 0.25) + 0.5 * w2
    sqw = np.sqrt(w)
    return g * np.exp(-x - 0.5 * (-x / sqw - 0.5 * sqw) * (-x / sqw - 0.5 * sqw)) / np.sqrt(2. * np

def density2(smile, smileP, smileM, t, x, eps):
    # Given three vectors of smile (evaluated at  $x$ ,  $e+eps$ ,
    #  $x-eps$ ), computes the density at  $x = \log(K/S)$ 
    w = smile * smile * t
    wp = smileP * smileP * t
    wm = smileM * smileM * t
    w1 = (wp - wm) / (2. * eps)
    w2 = (wp - 2. * w + wm) / (eps * eps)
    g = (1. - 0.5 * x * w1 / w) * (1. - 0.5 * x * w1 / w) - \
        0.25 * w1 * w1 * (1. / w + 0.25) + 0.5 * w2
    sqw = np.sqrt(w)
    return g * np.exp(-0.5 * (-x / sqw - 0.5 * sqw) * (-x / sqw - 0.5 * sqw)) / np.sqrt(2. * np

def plotDensity(smile, smileP, smileM, t, eps, logStrikes, col):
    # Plots the density using the function density2 above
    mydensity = []
    for n in range(len(logStrikes)):
        mydensity.append(
            density2(smile[n], smileP[n], smileM[n], t, logStrikes[n], eps))
    plt.plot(logStrikes, mydensity, col, linewidth=2)
```

3 Obloj's Expansion

Jan Obloj proved the following expansion for the implied volatility in the SABR model for a maturity t (where $\tilde{x} = -x = \log(F_0/K)$):

$$I_t(\tilde{x}) = I^0(\tilde{x}) (1 + I_H^1(\tilde{x})t),$$

where

$$I_H^1(\tilde{x}) = \frac{(\beta - 1)^2}{24} \frac{y_0^2}{(fK)^{1-\beta}} + \frac{1}{4} \frac{\rho \nu y_0 \beta}{(fK)^{(1-\beta)/2}} + \frac{2 - 3\rho^2}{24} \nu^2.$$

At the money ($x = 0$), the zero-order term reads $I^0(0) = y_0 K^{\beta-1}$. Otherwise, when $\tilde{x} \neq 0$, for $\nu = 0$ the zero-order term is $I^0(\tilde{x}) = \frac{\tilde{x} y_0 (1-\beta)}{x_0^{1-\beta} - K^{1-\beta}}$. Otherwise, for $\nu > 0$, when $\beta < 1$, two zero-order terms exist:

- Hagan et al's:

$$I_H^0(\tilde{x}) = \frac{\nu \tilde{x} \zeta}{z} \left(\log \left(\frac{\sqrt{1 - 2\rho\zeta + \zeta^2} + \zeta - \rho}{1 - \rho} \right) \right)^{-1},$$

with

$$z = \frac{\nu}{y_0} \frac{x_0^{1-\beta} - K^{1-\beta}}{1 - \beta}, \quad \text{and} \quad \zeta = \frac{\nu}{y_0} \frac{x_0 - K}{(x_0 - K)^{\beta/2}};$$

- Berestycki-Busca-Florent's:

$$I_B^0(\tilde{x}) = \nu \tilde{x} / \log \left(\frac{\sqrt{1 - 2\rho z + z^2} + z - \rho}{1 - \rho} \right),$$

with z as above.

Hence, for $\beta < 1$ the two different expansions are $I_B(\tilde{x}) = I_B^0 (1 + I_H^1(\tilde{x})t)$, and $I_H(\tilde{x}) = I_H^0 (1 + I_H^1(\tilde{x})t)$. We implemented the BBF one one below.

For $\beta = 1$ and $\nu > 0$, the zero-order term reads

$$I^0(\tilde{x}) = \nu \tilde{x} / \log \left(\frac{\sqrt{1 - 2\rho z + z^2} + z - \rho}{1 - \rho} \right),$$

with $z = \nu x_0 / y_0$.

In [16]: *# Note that we normalise the stock price to F_0 = 1*

```
def sabr_vol_HKLW(y0, nu, beta, rho, x, t):
    # returns the implied volatility (square root) using Hagan et al.'s expansion
    # and the mapping x -> -x
    K = np.exp(x)
    b1 = 1. - beta
    if abs(x) < 1E-7:
        H0 = y0 * pow(K, b1)
    elif nu < 0.00000011:
```



```

    logStrikeMin, logStrikeMax, num=nbStrikes)
oblojformula, hklwformula = [], []

for k in logStrikes:
    expansion = sabr_vol_Obloj(y0, nu, beta, rho, k, t)
    expansion2 = sabr_vol_HKLW(y0, nu, beta, rho, k, t)
    oblojformula.append(expansion)
    hklwformula.append(expansion2)

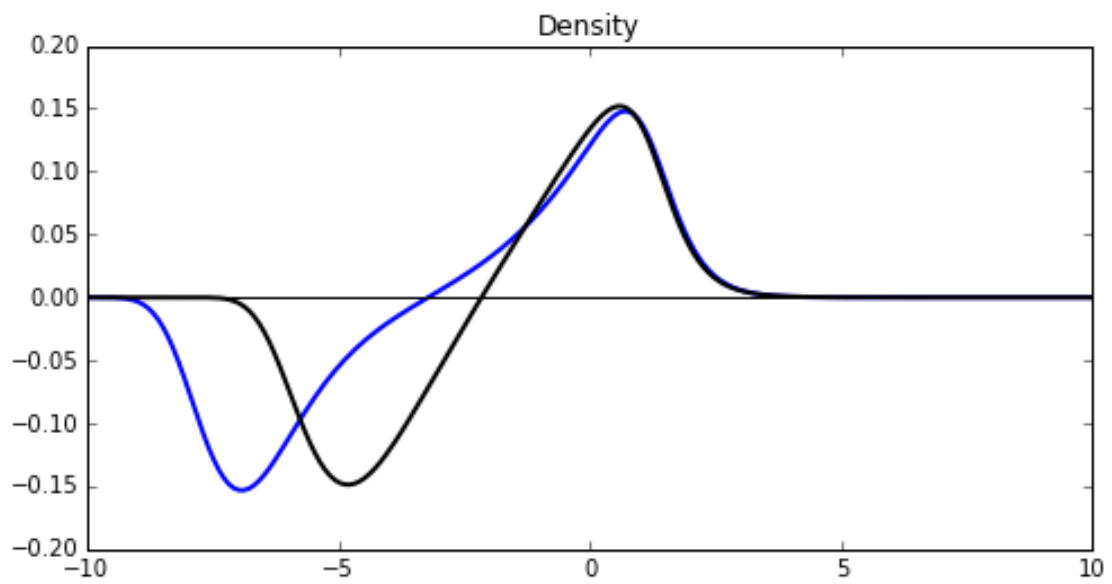
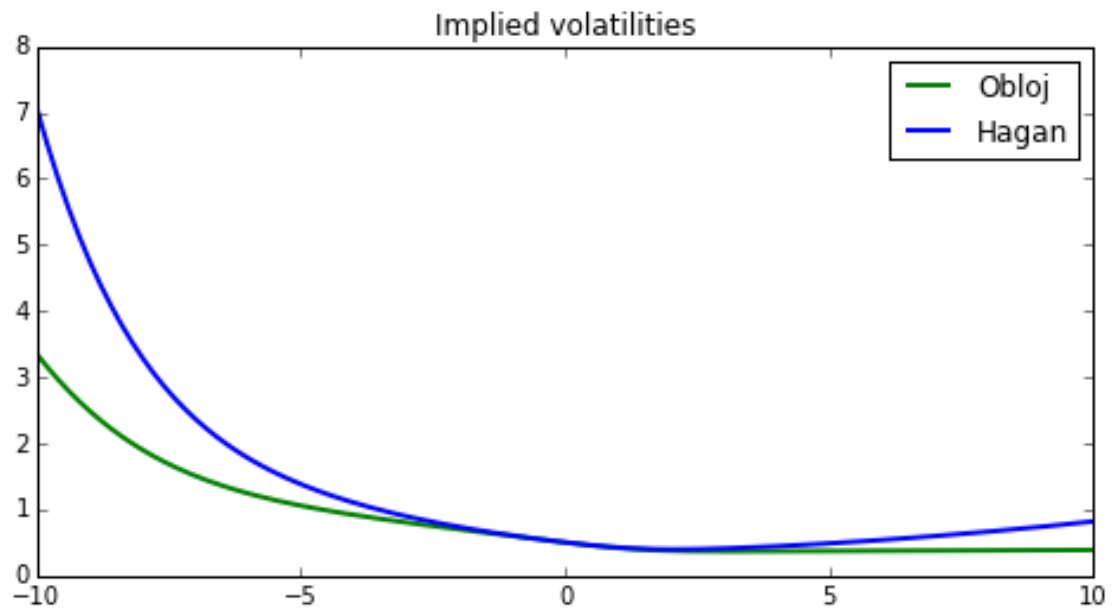
plt.figure(figsize=(8, 4))
plt.plot(
    logStrikes, oblojformula, 'g', linewidth=2, label="Obloj")
plt.plot(
    logStrikes, hklwformula, 'b', linewidth=2, label="Hagan")
plt.title('Implied volatilities')
plt.legend(loc=1)
plt.show()

eps = 1E-4
oblojSmile, oblojSmileP, oblojSmileM, haganSmile, haganSmileP, haganSmileM = [
], [], [], [], [], []

for logStrike in logStrikes:
    oblojSmile.append(
        sabr_vol_Obloj(y0, nu, beta, rho, logStrike, t))
    oblojSmileP.append(
        sabr_vol_Obloj(y0, nu, beta, rho, (logStrike + eps), t))
    oblojSmileM.append(
        sabr_vol_Obloj(y0, nu, beta, rho, (logStrike - eps), t))
    haganSmile.append(
        sabr_vol_HKLW(y0, nu, beta, rho, logStrike, t))
    haganSmileP.append(
        sabr_vol_HKLW(y0, nu, beta, rho, (logStrike + eps), t))
    haganSmileM.append(
        sabr_vol_HKLW(y0, nu, beta, rho, (logStrike - eps), t))

plt.figure(figsize=(8, 4))
plotDensity(
    oblojSmile, oblojSmileP, oblojSmileM, t, eps, logStrikes, 'b')
plotDensity(
    haganSmile, haganSmileP, haganSmileM, t, eps, logStrikes, 'k')
plt.plot(logStrikes, [0. for k in logStrikes], 'k')
plt.title('Density')
plt.show()

```



In [36]: