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A study of <sup>a</sup> family of equivalent martingale measures to price an option with an application to the Swiss market

## <sup>1</sup> Introduction

In this paper we consider the problem of pricing <sup>a</sup> European option in the context of incomplete markets.

Let us first consider a complete market, i.e. a market where every contingent claim is attainable. Black and Scholes (1973) have shown that under some ideal conditions, it is possible to create <sup>a</sup> hedged position, consisting of <sup>a</sup> long position in the stock and <sup>a</sup> short position in the option. Moreover it is possible to maintain the hedge continuously and as <sup>a</sup> consequence the return on the hedged position becomes certain. Hence the unique rational price of <sup>a</sup> contingent claim can be obtained as if it existed in <sup>a</sup> risk-neutral world, this price being equal to the expected value of the discounted payoff according to an appropriate probability measure  $Q$ . This probability measure is equivalent to  $P<sup>1</sup>$ , the *physical* probability measure and such that it makes the discounted price process  $\{e^{-rt}S(t)\}_{t>0}$  a martingale. This measure  $Q$  is called the *equivalent martingale measure*.

Let  $S(t)$  denote the price of a non-dividend paying stock at time  $t \geq 0$ . We assume that there is a stochastic process,  $\{X(t)\}_{t>0}$ , with stationary and independent increments,  $X(0) = x_0 = \ln S(0)$ , such that

$$
S(t) = e^{X(t)}, \qquad t \ge 0. \tag{1}
$$

We may interpret the random variable  $X(t) - x_0$  as the continuously compounded rate of return over the time interval  $[0, t]$ . We suppose throughout this paper that there exists <sup>a</sup> risk-free asset whose rate of return is known and constant through time, and we denote this risk-free rate by  $r$ . The Fundamental Theorem of Asset Pricing tells us that the absence of arbitrage opportunities implies essentially the existence of an equivalent martingale measure. However this equivalent martingale measure is unique if and only if the market is complete. So, when considering incomplete

<sup>1</sup> i.e. we have  $P(A) = 0 \Leftrightarrow Q(A) = 0$ 

models for the stock price process, there exist many equivalent martingale measures. In that sense the price of <sup>a</sup> European option is not unique.

We place ourselves in the context of incomplete markets and we are interested in studying the effect of changing the martingale measure when taking the discounted expected value of the payoff. In order to achieve this goal, we consider the pricing of <sup>a</sup> European option in an incomplete market.

We can describe a European option by a payoff function  $\Pi(s) \geq 0$  and a maturity date  $T$ . At time  $T$ , the holder of the option receives the amount  $II(S(T))$ . We know from financial theory that the price at time  $0 \le t < T$ is calculated as <sup>a</sup> discounted expected value of the payoff. The rate used to perform the discounting is the risk-free rate and expectation is taken according to an equivalent martingale measure. Thus the price of the option at time  $t$  is

$$
e^{-r(T-t)}E_Q[H(S(T)) | \mathfrak{F}_t], \qquad 0 \le t < T. \tag{2}
$$

Here  $E_{\mathcal{O}}[\cdot]$  denotes expectation taken with respect to the equivalent martingale measure Q. This measure must be such that the pricing formula (2) is compatible with the observed price of the stock, i. e. we require that

$$
S(t) = e^{-r(T-t)} E_Q[S(T) | \mathfrak{F}_t], \qquad 0 \le t < T. \tag{3}
$$

## 2 An incomplete model

Intuitively, to have <sup>a</sup> model for the stock price process that implies an incomplete market, more than two outcomes must be possible in an infinitesimal time interval.

As before let  $S(t)$  denote the stock price at time t. Here we will assume that  $\{S(t)\}_{t>0}$  is a geometric compound Poisson process with two possible jump heights, i.e.

$$
S(t) = e^{X(t)},
$$

where

$$
X(t) = x_0 + k_1 N_1(t) + k_2 N_2(t), \qquad t \ge 0.
$$
\n<sup>(4)</sup>

Here,  ${N_1(t)}_{t>0}$  and  ${N_2(t)}_{t>0}$  are independent Poisson processes with respective parameters  $\lambda(k_1)$  and  $\lambda(k_2)$ ,  $k_1$  and  $k_2$  are two constants and  $x_0$  is the initial value of the process  $\{X(t)\}_{t>0}$ . The initial stock price is  $S(0) = e^{x_0}$ . It is the simplest incomplete model we can think of. To avoid the existence of arbitrage opportunities, we assume that at least one of the two constants  $k_1$  or  $k_2$  is positive. Otherwise, selling short the stock and investing the proceeds in the risk-free asset yields  $S(0)(e^{rt} - e^{k_1N_1(t) + k_2N_2(t)})$ , which is positive whether a jump occurs or not at any time  $t$ . In order to shorten the notation in what follows, we will write  $\lambda_i$  for  $\lambda(k_i)$ .

Let  $\mu$ ,  $\sigma^2$ ,  $\gamma$  and  $\eta$  denote the first four cumulants per unit time of the process  $\{X(t)\}_{t\geq 0}$ . Thus

$$
E[X(t)] = k_1 \lambda_1 t + k_2 \lambda_2 t = \mu t,
$$
  
\n
$$
Var[X(t)] = k_1^2 \lambda_1 t + k_2^2 \lambda_2 t = \sigma^2 t,
$$
  
\n
$$
E\left[ (X(t) - \mu t)^3 \right] = k_1^3 \lambda_1 t + k_2^3 \lambda_2 t = \gamma t
$$

and

$$
E [(X(t) - \mu t)^4] - 3(\text{Var}[X(t)])^2 = k_1^4 \lambda_1 t + k_2^4 \lambda_2 t = \eta t. \tag{5}
$$

Since  $k_1$  and  $k_2$  are observable parameters of the process, they must remain unchanged under any equivalent measure. Only the Poisson parameters  $\lambda_1$ and  $\lambda_2$  can be modified. We denote those modified parameters by  $\lambda_1^*$  and  $\lambda_2^*$ . Writing down the martingale condition, we obtain

$$
S(0) = E_Q \left[ e^{-rt} S(t) \right]
$$

$$
= e^{-rt} E_Q[S(t)].
$$

By (1), the parameters  $\lambda_1^*$  and  $\lambda_2^*$  are solutions of the equation

$$
0 = -r + \lambda_1^*(e^{k_1} - 1) + \lambda_2^*(e^{k_2} - 1).
$$
 (6)

At that point, we propose to study the following family of methods to price an option: Set (for  $i = 1, 2$ )

$$
\lambda_i^* = \exp\left[h\left(\frac{e^{ck_i} - 1}{c}\right)\right]\lambda_i, \qquad 0 < c < 1. \tag{7}
$$

Taking the limit as  $c \to 0$  in (7), we obtain the modified Poisson parameters

$$
\lambda_i^* = e^{hk_i} \lambda_i, \qquad i = 1, 2. \tag{8}
$$

For the case  $c = 1$ , we have

$$
\lambda_i^* = e^{h(e^{k_i}-1)}\lambda_i, \qquad i = 1, 2. \tag{9}
$$

For a reason that will be made clear in section 6, we are motivated to study whether pricing an option according to the two different measures corresponding to (8) and (9) leads to two significantly different prices for this one. We want to determine, for every value of c, the value of  $h$ , written  $h^*(c)$ , such that the process

$$
\{e^{-rt}S(t)\}_{t\geq 0}
$$

is <sup>a</sup> martingale with respect to the probability measure corresponding to  $h^*(c)$ . That is  $h^*(c)$  solves (6). We obtain so the following family of implicit equations for  $h^*(c)$ :

$$
0 = -r + e^{h \frac{e^{ck_1} - 1}{c}} \lambda_1 (e^{k_1} - 1) + e^{h \frac{e^{ck_2} - 1}{c}} \lambda_2 (e^{k_2} - 1), \ \ 0 \le c \le 1. (10)
$$

Now the question arises what value for  $c$  should we use, when pricing an option in this incomplete model? What is a "good" value for  $c$ ? Because this question cannot be answered directly in <sup>a</sup> theoretical way, we propose to explore this question by means of real data. In that order, we examine observed prices of European calls. Hence, we have payoff functions of the form  $\Pi(S(T)) = (S(T) - K)_+$ , where K denote the strike price. In that case, formula (2) becomes

$$
e^{-rT} E_Q [(S(T) - K)_+ | S(0)], \qquad (11)
$$

or, in our model,

$$
e^{-rT} \sum_{x \ge \kappa} \left( e^{x_0 + x} - K \right) q(x, T), \tag{12}
$$

where we have considered  $t = 0$ , and defined  $\kappa = \ln\left(\frac{K}{S(0)}\right)$ .  $q(x,T)$  is the probability that  $k_1N_1(T) + k_2N_2(T) = x$ , under the equivalent martingale

measure Q. The distribution of  $N_i$  is of parameter  $e^{h^*(c)(\frac{c^*(c)}{c})}\lambda_i$  for  $i = 1,2$ . Here c is considered to be fixed,  $k_1$ ,  $k_2$ ,  $\lambda_1$  and  $\lambda_2$  are the parameters of the physical probability measure and are determined by solving (5) with  $\mu$ ,  $\sigma^2$ ,  $\gamma$  and  $\eta$  replaced by their estimates.

## 3 Esscher transforms and equivalent martingale measures

In an incomplete model, there are many equivalent martingale measures. A priori, it is not clear which martingale measure should be chosen to calculate the price of an option. Gerber and Shiu (1994) suggested that, in order to obtain <sup>a</sup> unique answer, the choice of the equivalent martingale measure could be limited to the family of Esscher transforms (see Esscher (1932)). A justification in term of minimal relative entropy with respect to the physical probability measure  $P$  has been given by Chan (1997). Let  $M(z,t) = E[e^{zX(t)}]$  denote the moment generating function of  $X(t)$ . Because  $M(z, t)$  is continuous at  $t = 0$ , it can be proved that

$$
M(z,t) = M(z,1)^t, \qquad t > 0.
$$
\n(13)

The process

$$
\left\{ e^{hX(t)}M(h,1)^{-t}\right\}
$$

is a positive martingale and they used it to define <sup>a</sup> change of probability measure. That is, it is used to define the Radon-Nikodym derivative  $dQ/dP$ , where  $P$  denotes as before the original probability measure and  $Q$  is the Esscher measure of parameter h. They call the risk-neutral Esscher measure the Esscher measure of parameter  $h = h^*$  such that the process

$$
\left\{e^{-rt}S(t)\right\}_{t\geq 0}
$$

is <sup>a</sup> martingale with respect to the probability measure corresponding to  $h^*$ .

Gerber and Shiu (1994) apply the Esscher transform (parameter  $h$ ) to the process  $\{X(t)\}_{t>0}$ . Let  $M(z,t;h)$  denote the moment-generating function of the modified distribution of  $X(t)$ . It is easily verified that

$$
M(z, t; h) = \frac{M(z + h, t)}{M(h, t)}.
$$
\n(14)

Writing down the martingale condition, we obtain

$$
S(0) = E_Q \left[ e^{-rt} S(t) \right]
$$

$$
= e^{-rt} E_Q[S(t)].
$$

By (1), the parameter  $h^*$  is the solution of the equation

$$
1 = e^{-rt} E_Q \left[ e^{X(t) - X(0)} \right],
$$

or, using (14),

$$
e^{rt} = E_Q \left[ e^{X(t) - X(0)} \right] = \frac{M(1 + h, t)}{M(h, t)} \,. \tag{15}
$$

From  $(13)$  we see that the solution does not depend on t, and we may set  $t = 1$ :

$$
e^r = \frac{M(1+h,1)}{M(h,1)},
$$
\n(16)

or

$$
r = \ln\left(\frac{M(1+h,1)}{M(h,1)}\right). \tag{17}
$$

For the model given by (4), (14) becomes

$$
M(z, t; h) = \frac{E_P \left[e^{(z+h)X(t)}\right]}{E_P \left[e^{hX(t)}\right]}
$$
  
= 
$$
\frac{\exp\left\{x_0 + \lambda_1 t \left(e^{(z+h)k_1} - 1\right) + \lambda_2 t \left(e^{(z+h)k_2} - 1\right)\right\}}{\exp\left\{x_0 + \lambda_1 t \left(e^{hk_1} - 1\right) + \lambda_2 t \left(e^{hk_2} - 1\right)\right\}}.
$$

After simplification, we obtain

$$
M(z, t; h) = \exp\left\{\lambda_1 e^{hk_1} t \left(e^{z k_1} - 1\right) + \lambda_2 e^{hk_2} t \left(e^{z k_2} - 1\right)\right\}.
$$
 (18)

Hence, the Esscher transform (parameter h) of the process  $\{X(t)\}_{t>0}$  is again <sup>a</sup> compound Poisson process, with the same two possible jump heights  $k_1$  and  $k_2$ , but with modified Poisson parameters  $\lambda_i e^{h\hat{k}_i}$ ,  $i = 1,2$ . From (17) and (18) we see that the parameter  $h^*$  of the risk-neutral Esscher measure is implicitly defined by the equation

$$
r = \lambda_1 e^{h^* k_1} \left( e^{k_1} - 1 \right) + \lambda_2 e^{h^* k_2} \left( e^{k_2} - 1 \right).
$$
 (19)

Then the risk-neutral Esscher parameters are given by  $\lambda_1^* = \lambda_1 e^{h^* k_1}$  and  $\lambda_2^* = \lambda_2 e^{h^* k_2}$ . Hence, we see that in section 2, the case  $c = 0$  corresponds to applying the method of Esscher transforms.

## 4 Two other incomplete models

In order to make numerical comparisons, we present in this section two other incomplete models: the shifted gamma process and the shifted inverse Gaussian process. The modeling of the stock-price movements by means of these two models was first introduced by Gerber and Shiu (1994, section 4). For these two incomplete models, we use the method of Esscher transforms in order to get <sup>a</sup> unique answer for the price of an option. Hence, the price of an option is defined to be the discounted expectation of the payoff where expectation is taken according to the Esscher transform of parameter  $h^*$ , where  $h = h^*$  is determined so that (17) is satisfied. Remember that in section 2, the case  $c = 0$  corresponds to applying the method of Esscher transforms.

## 4.1 The shifted gamma process

Here it is assumed that

$$
X(t) = Y(t) - \nu t,
$$

where  ${Y(t)}$  is a gamma process with shape parameter  $\alpha$  and scale parameter  $\beta$ , and the positive constant  $\nu$  is a third parameter. The moment generating function of  $X(t)$  is

$$
M(z,t) = \left(\frac{\beta}{\beta - z}\right)^{\alpha t} e^{-\nu tz}, \qquad z < \beta.
$$
 (20)

For given values of  $\mu$ ,  $\sigma$  and  $\gamma$ , the three parameters are chosen to match the first three cumulants per unit time, i.e., to solve

$$
E[X(1)] = \frac{\alpha}{\beta} - \nu = \mu,
$$

$$
Var[X(1)] = \frac{\alpha}{\beta^2} = \sigma^2,
$$

$$
E\left[\left(X(1) - \mu t\right)^3\right] = \frac{2\alpha}{\beta^3} = \gamma.
$$

Hence we set

$$
\alpha = \frac{4\sigma^6}{\gamma^2}, \qquad \beta = \frac{2\sigma^2}{\gamma}, \qquad \nu = \frac{2\sigma^4}{\gamma} - \mu. \tag{21}
$$

From  $(14)$  and  $(20)$ , we obtain

$$
M(z,t;h) = \left(\frac{\beta - h}{\beta - h - z}\right)^{\alpha t} e^{-\nu tz}, \qquad z < \beta - h,
$$
 (22)

which shows that the Esscher transform of  $\{X(t)\}\)$  is again a shifted gamma process with unchanged values of  $\alpha$  and  $\nu$  but  $\beta$  replaced by

$$
\beta(h)=\beta-h.
$$

From (16), we obtain the following condition for the martingale measure

$$
\beta = \beta(h^*) = \frac{1}{1 - e^{-(\nu + r)/\alpha}}.
$$

## 4.2 The shifted inverse Gaussian process

Here, it is also assumed that

$$
X(t) = Y(t) - \nu t,
$$

but with  ${Y(t)}$  being an inverse Gaussian process with parameters a and b. The moment generating function of  $X(t)$  is

$$
M(z,t) = e^{at(\sqrt{b-\sqrt{b-z}}) - \nu tz}, \qquad z < b.
$$
 (23)

Again, for given values of  $\mu$ ,  $\sigma$  and  $\gamma$ , the three parameters are chosen to match the first three cumulants per unit time, or to solve

$$
E[X(1)] = \frac{a}{2b^{1/2}} - \nu = \mu,
$$
  

$$
Var[X(1)] = \frac{a}{4b^{3/2}} = \sigma^2,
$$
  

$$
E\left[ (X(1) - \mu t)^3 \right] = \frac{3a}{8b^{5/2}} = \gamma.
$$

Hence we set

$$
a = 3\sigma^5 \sqrt{\frac{6}{\gamma^3}}, \quad b = \frac{3\sigma^2}{2\gamma}, \quad \nu = \frac{3\sigma^4}{\gamma} - \mu. \tag{24}
$$

From (14) and (23), we obtain

$$
M(z, t; h) = \exp\left[at(\sqrt{b - h} - \sqrt{b - h - z}) - \nu t z\right], \qquad z < \beta - h, (25)
$$

which shows that the Esscher transform of  $\{X(t)\}\$ is again a shifted inverse Gaussian process with unchanged values of  $a$  and  $\nu$  but  $b$  replaced by

$$
b(h)=b-h.
$$

From (17), we obtain the following condition for the martingale measure

$$
r = a\left(\sqrt{b - h^*} - \sqrt{b - h^* - 1}\right) - \nu,
$$

or equivalently

$$
\sqrt{b(h^*)} - \sqrt{b(h^*) - 1} = \frac{\nu + r}{a}
$$
,

which is an implicit equation for  $b^* = b(h)$ .

## 5 Numerical examples

In this section we examine the family of equivalent martingale measures given implicitly by equation (10). We are interested in examining the option's price sensitivity to the change of measure involved by <sup>a</sup> change in the parameter c.

The stock prices are obtained from the data base DATASTREAM. Those prices are closing market prices. We consider daily data. From those daily prices, we compute the continuously compounded daily rate of return according to

$$
X(t) = \ln\left(\frac{S(t)}{S(t-1)}\right). \tag{26}
$$

The DATASTREAM's prices are adjusted for operations like splits or increases of capital but not for the payment of dividend. We had to modify the data for the days where dividend payments occurred in order to cancel the jumps (anticipated on the market) due to dividends. DATASTREAM provides us with dividends series. Flence it is possible to correct the rates of return at the dividend payment dates.

As examples we have chosen to consider American call options on stocks ALUSUISSE <sup>R</sup> ("nominative") and SWISS BANK CO <sup>B</sup> ("porteur"). We have selected derivatives for which the volume of transactions was sufficiently high, so that the prices are real market prices. We consider times to maturity between 10 days and 158 days. We made use of observed data from the SOFFEX (Swiss Options and Financial Futures Exchange). The options at the SOFFEX are American options. We have considered only options on stocks for which there was no dividend payment until the date of maturity. For these options the price is identical to the price of European options.

On the Swiss Option Exchange, the expiration date is always the third Friday of the relevant month. Quoted prices for options and traded volumes have also been obtained from DATASTREAM. The options are quoted in Swiss francs with the minimum quoted price fluctuations *(ticks)* given in Table 1.

We had two kinds of daily quoted prices: *last price paid (lpp)* and *settlement* price (sp). As a general rule, the settlement price corresponds to the last price paid, unless there was no exchange during the last hour of quotation or

Option's price				Tick
		From Fr. $-10$ to	Fr. 9.90	$Fr. -10$
	From Fr. $10-$	to	Fr. 19.80	$Fr. -20$
From	$Fr. 20 -$	to	Fr.99.50	$Fr = 50$
From	$Fr. 100 -$			$Fr. 1 -$

Table 1. Minimum quoted price fluctuation (tick) at the SOFFEX

the last price paid was not anymore corresponding to the current situation of the market. In those two cases, the SOFFEX determines the option prices. We have to remember this when making comparisons between observed prices and theoretical ones. For the risk-free rate, we have chosen CURRENCY (SWISS FR.) from London for one, two, six and twelve months. For time to maturity of four and five months, we have considered linear interpolation of the preceding rates.

For each case considered, we have first calculated  $\hat{\mu}$ ,  $\hat{\sigma}^2$ ,  $\hat{\gamma}$  and  $\hat{\eta}$ , estimates of the first four cumulants per unit time (here one day) of the process  $\{X(t)\}_{t>0}$ . Then we have computed  $\hat{\lambda}_1$ ,  $\hat{\lambda}_2$ ,  $\hat{k}_1$  and  $\hat{k}_2$ , estimates of the parameters of our pure jump model by solving the system given by (5). See Tables 2 to 5. We have also computed estimates of  $\alpha$ ,  $\beta$  and  $\nu$ , the parameters of the shifted gamma process, using (21) and finally estimates of a, b and  $\nu$ , the parameters of the shifted inverse Gaussian process, using (24).

To obtain twenty-one different options prices in the first model, we have computed expression (12) for  $c = 0, 0.05, 0.10, \ldots, 1$ . At this stage, the computations are time-consuming (we obtained up to 160,000 probability masses for each distribution given by different values of  $c$ ). We then computed the option prices for the two other incomplete models (see formulas  $(4.1.7)$  and  $(4.2.7)$  given in Gerber and Shiu  $(1994)$ ). See Tables A.1 to A.7 in appendix A. Here because of the high-valued parameters numerical difficulties arise. For example, in the case  $T = 158$  days we had to calculate a gamma distribution function with shape parameter  $\alpha = 1,951.69$ and scale parameter  $\beta = 271.46$  or an inverse Gaussian distribution function with shape parameter  $a = 305.93$  and scale parameter  $b = 202.78$ .

In the first model, we see that option prices are monotone functions of the parameter c. Whether it is an increasing or decreasing function of <sup>c</sup> depends on the case considered. In every example we observe that the differences between the prices obtained with  $c = 0$  and the prices obtained with  $c = 1$ 

		$k_i$
	0.250587820	0.022434177
$\mathcal{L}$	0.237835455	$-0.015715156$

Table 2: Estimates of daily parameters for ALUSUISSE R over the period January <sup>4</sup> - June 30, <sup>1992</sup>

	$k_i$	
2.651383917	0.006081860	
6.976036534	$-0.002422898$	

Table 3: Estimates of daily parameters for SWISS BANK CO B over the period June <sup>29</sup> - November 19, <sup>1992</sup>

	$\kappa_i$	
0.316605487	0.015705559	
0.204049188	$-0.011705723$	

Table 4. Estimates of daily parameters for SWISS BANK CO B over the period January <sup>4</sup> - June 7, 1993

	$\kappa_i$		
0.849021090	0.011786903		
2.422493362	$-0.004344630$		

Table 5. Estimates of daily parameters for SWISS BANK CO B over the period October 18, <sup>1993</sup> - April 18, <sup>1994</sup>

are very small. Figures A.l to A.7 in appendix A show for each value of  $c$  the difference between the price obtained with that particular value of  $c$ and the price obtained with  $c = 0$  in percentage of this latter. Surprisingly, we observe in every of our cases that for any given value of  $c$ , the higher the strike price, the higher this percentage in absolute value. The maximal difference computed between those two prices is of 0.245 % of the price given by  $c = 0$  (see Table A.3 and Figure A.3). In fact, it appears that the range of equivalent martingale measures obtained is in some sense very narrow in the pure jump model considered.

#### <sup>6</sup> A more general jump model

In this section we give <sup>a</sup> justification for section 2. Consider, for the process  ${X(t)}_{t>0}$ , a more general model, specified as follows. The conditional distribution of the amount of <sup>a</sup> jump is of <sup>a</sup> discrete nature. We use the symbol  $\lambda_t(\cdot)$  for the measure of the jump frequencies of the process  ${X(t)}_{t\geq0}$ , i.e.  $\lambda_t(x)dt$  is the probability of a jump of amount x between times t and  $t + dt$ . We adopt a similar notation for the process  $\{S(t)\}_{t>0}$ , with  $\lambda_t(\cdot)$  replaced by  $\tilde{\lambda}_t(\cdot)$ . Because  $\{S(t)\}_{t>0}$ , is adapted, the following equality holds:

$$
\lambda_t(x) = \tilde{\lambda}_t(S(t)(e^x - 1)).
$$
\n(27)

We write  $\tilde{\lambda}_t^{\alpha_t}(\cdot)$  and  $\lambda_t^{h_t}(\cdot)$  to indicate that we are working with the Esscher transform (parameter  $\alpha_t$  for  $\{S(t)\}_{t\geq 0}$  and  $h_t$  for  $\{X(t)\}_{t\geq 0}$ ). So we have

$$
\widetilde{\lambda}_t^{\alpha_t}(y) = e^{\alpha_t y} \widetilde{\lambda}_t(y) = \lambda_t^{h_t}(x) = e^{h_t x} \lambda_t(x), \qquad (28)
$$

where  $y = S(t)(e^x - 1)$ .

We apply now, as in Gerber and Shiu (1994), the Esscher transform to the process  $\{X(t)\}_{t>0}$ . The condition that the process  $\{e^{-rt}S(t)\}_{t\geq 0}$  is a martingale resumes to

$$
-r + \sum_{x} (e^x - 1)e^{h_t x} \lambda_t(x) = 0.
$$
 (29)

Consider now applying the Esscher transform directly to the process  $\{S(t)\}\$ . This is the Esscher method in the sense of Bühlmann (1995). Then, the martingale condition is

$$
-r+\sum_{y}ye^{\alpha_{t}y}\widetilde{\lambda}_{t}(y)=0\,,
$$

which can be rewritten in term of the "parameters" of the process  ${X(t)}_{t>0}:$ 

$$
-r + \sum_{x} (e^x - 1)e^{\alpha_t^*(e^x - 1)} \lambda_t(x) = 0,
$$
\n(30)

with  $\alpha_t^* = \alpha_t S(t)$ .

In comparing conditions (29) and (30), we see that, unless  $\lambda_t(\cdot)$  is concentrated on one point  $\tilde{x}$  (in which case we have to choose  $\alpha_t S(t)(e^{\tilde{x}} - 1) =$  $h_t\tilde{x}$ ), the resulting equivalent martingale measures for  $\{e^{-rt}S(t)\}_{t>0}$  are different. For that reason, we are motivated to study whether pricing an option according to those two different measures leads to two significantly different prices for this one. Hence, we propose the following interpolation formula for the modified jump measure:

$$
\lambda_t(x) \ \longmapsto \ e^{h_t(\frac{e^{cx}-1}{c})}\lambda_t(x)\,, \qquad 0 < c \le 1\,.
$$

Taking the limit as  $c \rightarrow 0$ , we see that the above expression is the modified jump measure obtained when applying the Esscher transform (parameter  $h_t$ ) to the process  $\{X(t)\}_{t>0}$ . The other extreme case where  $c = 1$  corresponds to the modified jump measure obtained when applying the Esscher transform (parameter  $\alpha_t^*$ ) directly to the process  $\{S(t)\}_{t>0}$ . The model given by (4) is time homogeneous, so we can leave the subscript t. We want to determine, for every value of c, the value of h, written  $h^*(c)$ , such that the process

$$
\left\{e^{-rt}S(t)\right\}_{t\geq 0}
$$

is <sup>a</sup> martingale with respect to the probability measure corresponding to  $h^*(c)$ . In fact, we obtain a family of implicit equations for  $h^*(c)$ :

$$
-r + \sum_{x} (e^x - 1)e^{h(\frac{e^{cx} - 1}{c})}\lambda(x) = 0, \qquad 0 < c \le 1. \tag{31}
$$

We can rewrite this family of implicit equations for our model (4) of the process  $\{X(t)\}_{t>0}$  and obtain so (10).

## 7 Examination of an approximation formula

In this section, we examine the linear approximation formula introduced by Gerber and Landry (1997) by means of real data. The examination makes use, as in section 5, of observed data from the SOFFEX.

They have considered models where  $\gamma$ , the third cumulant per unit time of the process  $\{X(t)\}_{t>0}$ , is different from zero. In order to examine the effect of skewness, they proposed to replace the exact density of  $X(1)$  by its first order expansion and obtained <sup>a</sup> linear approximation for the price of <sup>a</sup> European option. It is remarkable that the approximation formula obtained does not depend on the underlying model, as long as option prices are calculated by the Esscher method. The interested reader is referred to their paper for further details. Here is their formula

$$
e^{-r} E_Q[S(1) - K]_+] \simeq e^{-r} \int_{\kappa}^{\infty} (e^{x_0 + x} - K) f_0(x) dx + \frac{\gamma}{\sigma^2} e^{-r} \int_{\kappa}^{\infty} (e^{x_0 + x} - K) f_1(x) dx, \qquad (32)
$$

where  $\kappa = \ln\left(\frac{K}{S(0)}\right)$  and without loss of generality the maturity considered is of 1. Here K is the strike price,  $f_0(x)$  and  $f_1(x)$  are given by

$$
f_0(x) = \frac{1}{\sigma} \phi \left( \frac{x - \mu^*}{\sigma} \right) \tag{33}
$$

and

$$
f_1(x) = \frac{1}{2}(r - \mu) \left( \frac{1}{\sigma^3} \phi'' \left( \frac{x - \mu^*}{\sigma} \right) + \frac{1}{\sigma^2} \phi' \left( \frac{x - \mu^*}{\sigma} \right) \right) - \frac{1}{12} \phi' \left( \frac{x - \mu^*}{\sigma} \right) - \frac{1}{4\sigma} \phi'' \left( \frac{x - \mu^*}{\sigma} \right) - \frac{1}{6\sigma^2} \phi''' \left( \frac{x - \mu^*}{\sigma} \right),
$$
(34)

where  $\mu^* = r - \frac{1}{2}\sigma^2$ . Here  $\phi(\cdot)$  denotes the standard normal probability density function.  $f_0(x)$  is the martingale density of  $X(1)$  in the classical Black-Scholes model. Hence, the approximation consists of the Black-Scholes price combined with an adjustment for skewness.

The prices obtained by this method are displayed under the heading "Linear" Approximation" in Tables A.1 to A.7 in appendix A. In most of our cases the difference between the settlement price and the price obtained by this method is negative. For example (see Table 6) we consider <sup>a</sup> call option with strike price  $K = 400$  and 64 days to maturity on a stock (SWISS BANK CO.) that is selling at present time at 374 Swiss francs. The third cumulant

per unit time  $\gamma$  of the stock is estimated at 0.0000011916643 per day. We obtain a rate of change at  $\gamma = 0$  of 3632.480. Hence the Black-Scholes price must be adjusted by 3632.4797 per  $\gamma$ . The first order effect of the skewness on the price of this option is given for different strike prices in Figure 1. We note that this adjustment can be positive or negative, depending on the strike price K.

Exercise Price $K$	lpp	Observed Prices sp	<b>Black</b> Scholes	Rate of Change at $\gamma = 0$	Linear Approximation
350	31.50	33.00	31.596	815.967	31.658
360	27.00	26.50	24.639	1549.226	24.757
370	19.00	21.00	18.671	2325.263	18.849
380	15.00	15.00	13.736	3001.996	13.965
390	11.50	11.50	9.805	3459.197	10.069
400	8.50	8.50	6.789	3632.480	7.066
425	4.00	3.40	2.377	2963.007	2.603

Table 6: Call option (on SWiSS BANK CO B) prices with  $S(0) = 374$ ,  $T = 64$  days on August 18, 1994 ( $\gamma = 0.000 001 191 664 3$ )



Figure 1: Adjustment of the call option price per  $\gamma$  (see Table 6)

# 8 Implied parameters and the linear approximation formula

In sections <sup>5</sup> and <sup>7</sup> we computed the Black-Scholes formula and expression  $(32)$  in taking parameter values estimated from historical data and substituting them into those two formulas. From a practical point of view, the one parameter in the Black-Scholes formula that cannot be observed rectly is the volatility. By using the historical standard deviation to estimate the volatility, we assume that the past variability of the stock's returns is invariant through time. It is not obvious that volatility is constant for long periods of time and that the historical volatility is independent of the time series from which it is calculated. It is therefore difficult to measure directly the volatility in practice.

However, option prices are quoted in the market. An alternative concept, implied volatility, consists of estimating the volatility of stock returns implicitly reflected in current option prices. A call option price increases monotonically with volatility, so there is <sup>a</sup> one-to-one correspondence between the volatility and the option price. The idea is to invert the Black-Scholes formula from the currently observed price of <sup>a</sup> call option. In this way we obtain the market's opinion of the value of the volatility over the remaining life of the option. This method was originally proposed by Latane and Rendleman (1976). The implied volatility derived from several options written on the same stock will generally not be equal.

Now the problem is to take <sup>a</sup> suitable weighted average of the individual implied volatilities. One can think about taking the arithmetic average or even to weight each option's implied volatility according to its degree of price elasticity with respect to the volatility. Here we mention Beckers' empirical study (1981) of stock returns' future variability estimates. He suggests the use of only one call option price, the one whose price is most sensitive to  $\sigma$ . We measure the sensitivity of an option with respect to  $\sigma$  by the partial derivative of its price with respect to  $\sigma$ , that is

$$
\frac{\partial C}{\partial \sigma} = S(0)\sqrt{T}\Phi'\left(\frac{rT - \kappa + \frac{1}{2}\sigma^2 T}{\sigma\sqrt{T}}\right)
$$
  
=  $S(0)\sqrt{\frac{T}{2\pi}}\exp\left(-\frac{1}{2}\frac{(rT + \ln(S(0)/K) + \frac{1}{2}\sigma^2 T)^2}{\sigma^2 T}\right).$ 

This expression is maximal for

$$
\widetilde{K} = S(0)e^{rT + \frac{1}{2}\sigma^2 T}.
$$
\n(35)

Hence the call option whose strike price is the nearest to the one given by (35) will be chosen.

The parameters in the linear approximation formula (32) that cannot be observed directly are the first three cumulants per unit time of the stock price. Now the idea is to invert this approximation formula by observing the current call option's price. In this way we obtain the market's opinion of the value for the drift, the volatility and the third cumulant per unit time over the remaining life of the option. To do this, we apply the following algorithm. Choose arbitrary initial values for  $\mu$  and  $\gamma$ , say  $\mu_0$  and  $\gamma_0$  (a good idea is to choose historical estimates of them). Then compute  $\sigma_1$ , the value of  $\sigma$  that makes the approximation formula meet exactly the last observed price. Now use  $\sigma_1$  and  $\gamma_0$  to compute  $\mu_1$ , the value of  $\mu$  that makes the approximation formula meet exactly the last observed price. Repeat these steps to obtain  $\gamma_1$ ,  $\sigma_2$ ,  $\mu_2$ ,  $\gamma_2$ ,  $\sigma_3$ ,  $\mu_3$ ,  $\gamma_3$ , ... until convergence is observed.

Tables B.l to B.7 in appendix B show the prices obtained using implied volatility to compute Black-Scholes prices and implied  $\mu$ ,  $\sigma$  and  $\gamma$  to compute the linear approximation formula. For example, consider Table B.5, which shows the prices obtained for <sup>a</sup> call option on stocks of the Swiss Bank Corporation on August 18, 1994, using implied parameters. In this particular example we found an annual implied  $\sigma$  of 0.29321 for the Black-Scholes formula. For the linear approximation formula, we obtained the following implied annual parameters:  $\mu = -0.18889$ ,  $\sigma = 0.29122$ ,  $\gamma = 0.00043496$ . In comparing settlement prices and the prices given by the Black-Scholes formula using implied volatility we see that settlement prices are overforecasted in almost every of our cases. Except for the prices given by Table B.6, we observe that using implied parameters with both formulas leads to differences of identical signs. For both formula and for <sup>a</sup> particular choice of  $S(0)$  and T, the largest difference is obtained for the more out-of-the-money call option. In Table <sup>7</sup> are displayed the mean absolute differences between the theoretical prices and the settlement prices and the sum of absolute differences for all of our cases. Both for the Black-Scholes formula and the linear approximation we remark that the mean absolute

spread decreases significantly while using implied parameters. However this effect is the strongest for the Black-Scholes formula. Examining all our cases together, we remark that 23 times out of 37 the Black-Scholes formula with volatility estimated implicitly leads to better results than using the Black-Scholes formula with volatility estimated historically. We remark also that 26 times out of 37 using implied parameters in the linear approximation formula leads to better results than without using implied parameters. Figure B.l in appendix B shows valuation errors in percent of the settlement prices for the Black-Scholes and linear approximation formulas, using for both formula implied parameters. Figure B.2 in appendix B shows valuation errors in percent of the settlement prices for the linear approximation formula with and without the use of implied parameters. Moneyness is defined as

$$
\frac{S(0)}{K}-1\,.
$$

Options whose *absolute moneyness*  $\left| \frac{S(0)}{K} - 1 \right|$ , is greater than ten percent are not taken into account. These options have little trading activity and price quotes are generally not supported by actual trades.



Table 7: Valuation errors statistics

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# Appendix A



Table A.1: Call option (on ALUSUISSE R) prices with  $S(0) = 508$  and  $T = 94$  days on July 13, 1993

 $\alpha$ 



Table A.2: Call option (on SWISS BANK CO B) prices with  $S(0) = 263.5$  and  $T = 28$ days on November 20, 1992



Table A.3: Call option (on SWISS BANK CO B) prices with  $S(0) = 325$  and  $T = 37$  days on January 13, 1993



Table A.4: Call option (on SWISS BANK CO B) prices with  $S(0) = 380$  and  $T = 50$  days on September 9, 1994

		$K = 350$	360	370	380	390	400	425
Observed	lpp	31.50	27.00	19.00	15.00	11.50	8.50	4.00
Prices	sp	33.00	26.50	21.00	15.00	11.50	8.50	3.40
<b>Black</b> Scholes		31.596	24.639	18.671	13.736	9.805	6.789	2.377
Linear Approximation		31.658	24.757	18.849	13.965	10.069	7.066	2.603
Shifted Gamma		32.745	25.737	19.703	14.683	10.651	7.523	2.820
Shifted Inverse Gaussian		31.656	24.755	18.847	13.963	10.067	7.066	2.606
$c=0$		31.88106	24.93058	18.98086	14.06309	10.13975	7.11750	2.62631
0.05		31.88109	24.93061	18.98089	14.06312	10.13978	7.11753	2.62633
0.10		31.88111	24.93064	18.98092	14.06316	10.13981	7.11756	2.62634
0.15		31.88114	24.93067	18.98095	14.06319	10.13984	7.11758	2.62636
0.20		31.88116	24.93070	18.98099	14.06322	10.13987	7.11761	2.62638
0.25		31.88118	24.93073	18.98102	14.06325	10.13990	7.11764	2.62640
0.30		31.88121	24.93076	18.98105	14.06329	10.13994	7.11767	2.62642
0.35		31.88124	24.93079	18.98108	14.06332	10.13997	7.11770	2.62643
0.40		31.88126	24.93081	18.98111	14.06335	10.14000	7.11772	2.62645
0.45		31.88128	24.93084	18.98114	14.06339	10.14003	7.11775	2.62647
0.50		31.88131	24.93087	18.98118	14.06342	10.14006	7.11778	2.62649
0.55		31.88133	24.93090	18.98121	14.06345	10.14009	7.11781	2.62650
0.60		31.88136	24.93093	18.98124	14.06348	10.14012	7.11784	2.62652
0.65		31.88138	24.93096	18.98127	14.06352	10.14015	7.11787	2.62654
0.70		31.88141	24.93099	18.98130	14.06355	10.14019	7.11789	2.62656
0.75		31.88143	24.93102	18.98134	14.06358	10.14022	7.11792	2.62657
0.80		31.88146	24.93105	18.98137	14.06361	10.14025	7.11795	2.62659
0.85		31.88148	24.93108	18.98140	14.06365	10.14028	7.11798	2.62661
0.90		31.88151	24.93110	18.98143	14.06368	10.14031	7.11801	2.62663
0.95		31.88153	24.93113	18.98146	14.06371	10.14034	7.11803	2.62664
$c=1$		31.88155	24.93116	18.98149	14.06374	10.14037	7.11806	2.62666

Table A.5: Call option (on SWISS BANK CO B) prices with  $S(0) = 374$  and  $T = 64$  days on August 18, 1994



Table A.6: Call option (on SWISS BANK CO B) prices with  $S(0) = 373$  and  $T = 122$ days on June 21, 1994

 $\alpha$ 

 $\mathbf{x}$ 



Table A.7: Call option (on SWISS BANK CO B) prices with  $S(0) = 410$ ,  $T = 158$  days on May 16, 1994



value of c

Figure A.1: Change in price from  $c = 0$ ,  $S(0) = 508$ ,  $T = 94$  days (ALUSUISSE)



Figure A.2: Change in price from  $c = 0$ ,  $S(0) = 263.5$ ,  $T = 28$  days (SBC)



Figure A.3: Change in price from  $c = 0$ ,  $S(0) = 325$ ,  $T = 37$  days (SBC)



Figure A.4: Change in price from  $c = 0$ ,  $S(0) = 380$ ,  $T = 50$  days (SBC)

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Figure A.5: Change in price from  $c = 0$ ,  $S(0) = 374$ ,  $T = 64$  days (SBC)



Figure A.6: Change in price from  $c = 0$ ,  $S(0) = 373$ ,  $T = 122$  days (SBC)



Figure A.7: Change in price from  $c = 0$ ,  $S(0) = 410$ ,  $T = 158$  days (SBC)

# Appendix B



Table B.1: Call option (on ALUSUISSE R) prices with  $S(0) = 508$ ,  $T = 94$  days on July 13, 1993  $(\gamma = 0.000\,001\,906\,308\,2)$ 

Exercise Price K	Observed Prices lpp sp	Bs-iv	$LA-1$
240	26.50 25.50	25.364	25.231
260	9.00 8.50	10.056	9.751
280	2.40 2.40	2.392	2.196
300	1.00 1.00	0.323	0.276

Table B.2: Call option (on SWISS BANK CO B) prices with  $S(0) = 263.5$ ,  $T = 28$  days on November 20, 1992 ( $\gamma = 0.000\,000\,497\,237\,5$ )



Table B.3: Call option (on SWISS BANK CO B) prices with  $S(0) = 325$ ,  $T = 37$  days on May 12, 1993 ( $\gamma = 0.0000008992438$ )



Table B.4: Call option (on SWISS BANK CO B) prices with  $S(0) = 380$ ,  $T = 50$  days on September 1, 1994 ( $\gamma = 0.0000011916643$ )

Exercise Price $K$	Observed Prices lpp sp	$Bs-iv$	$LA - i$
350	31.50 33.00	33.963	33.903
360	27.00 26.50	27.381	27.339
370	19.00 21.00	21.649	21.633
380	15.00 15.00	16.783	16.794
390	11.50 11.50	12.756	12.793
400	8.50 8.50	9.506	9.565
425	4.00 3.40	4.191	4.273

Table B.5: Call option (on SWISS BANK CO B) prices with  $S(0) = 374$ ,  $T = 64$  days on August 18, 1994 ( $\gamma = 0.000\,001\,191\,664\,3$ )



Table B.6: Call option (on SWISS BANK CO B) prices with  $S(0) = 373$ ,  $T = 122$  days on June 21, 1994 ( $\gamma = 0.0000011916643$ )

Exercise Price $K$	Observed Prices lpp sp	$Bs-iv$	$LA-i$
350	71.50 73.50	73.436	73.370
360	65.50 66.00	65.938	65.876
390	44.50 45.00	46.215	46.184
400	39.50 39.00	40.617	40.601
425	27.00 27.00	28.756	28.780
448	18.50 18.50	20.367	20.425

Table B.7: Call option (on SWISS BANK CO B) prices with  $S(0) = 410$ ,  $T = 158$  days on May 16, 1994 ( $\gamma = 0.000\,001\,191\,664\,3$ )



Figure B.l: Valuation errors in percent of the settlement prices for the Black-Scholes and the linear approximation formula, using for both implied parameters



Figure B.2: Valuation errors in percent of the settlement prices for the linear approximation with (LA-i) and without (LA) implied parameters

#### Summary

This paper studies <sup>a</sup> one parameter family of equivalent martingale measures to price an option in <sup>a</sup> particular incomplete model We also examine by means of real data an approximation formula introduced by Gerber and Landry We propose to estimate the parameters in an implicit way in order to compute this formula The study makes use of observed data from the SOFFEX (Swiss Options and Financial Futures Exchange)

#### Zusammenfassung

Der vorliegende Artikel betrachtet eine einparametrige Familie von äquivalenten Martmgalmaßen zur Bewertung von Optionen in einem unvollständigen Markt Überdies wird die Näherungsformel von Gerber und Landry anhand von wirklichen Daten untersucht. Schließlich wird eine implizite Schätzung der unbekannten Parameter vorgeschlagen. Die Studie stützt sich auf Daten der SOFFEX.

## Résumé

Cet article etudie une famille <sup>ä</sup> un parametre de mesures de martingales equivalentes pour evaluer une option dans un modele incomplet particuher On examine aussi <sup>ä</sup> l'aide de données réelles une formule d'approximation introduite par Gerber et Landry. On propose finalement d'estimer les paramètres de manière implicite afin d'évaluer cette formule. L'étude fait usage de données provenant de la SOFFEX.