

Pricing Options in Jump Diffusion Models Using Mellin Transforms*

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Received June 19, 2013; revised July 18, 2013; accepted July 27, 2013

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ABSTRACT

This paper is concerned with the valuation of options in jump diffusion models. The partial integro-differential equation (PIDE) inherent in the pricing problem is solved by using the Mellin integral transform. The solution is a single integral expression independent of the distribution of the jump size. We also derive analytical expressions for the Greeks. The results are implemented and compared to other approaches.

Keywords: Jump Diffusion; Mellin Transform; European Option

1. Introduction

Pure diffusion models are in most cases not flexible enough to fit the empirical observations concerning the movements of stock prices. Jump diffusions are a natural extension of pure diffusions since they are able to account for large and sudden changes. This is accomplished by adding a second source of uncertainty into the diffusion dynamics. The second source is discontinuous and models the jumps in the underlying asset. Also, pricing of options in jump diffusions is consistent with the volatility smile often observed in financial markets.

In his seminal paper from 1976 Merton [1] introduced a first jump diffusion process for modeling the behavior of stock prices. Since his work jump diffusions have become a very popular tool in modeling equity, foreign exchange and commodity markets. Under the assumption of log-normally distributed jumps, Merton solved the European option pricing problem explicitly in closed-form in terms of an infinite series of Black-Scholes prices. Other popular pure and non-pure jump diffusion processes are those proposed in [2-6] among others. Kou [7] proposed a double exponential jump-diffusion model where jump sizes are double exponentially distributed. The model allows for an analytical pricing of some path-

*This paper is a revised and shortened version of a working paper that circulated under the title "Option Pricing in Jump Diffusions: A New Integral-Based Framework for European and American Options". Disclaimer: Statements and conclusions made in this article are entirely those of the author. They do not necessarily reflect the views of LBBW.

dependent options, such as continuously monitored barrier and lookback, and perpetual American options [8]. In [9], pricing formulas in Kou's model for double (single) barrier and touch options with time-de- pendent rebates are derived applying Laplace transforms.

In many cases, however, an explicit closed-form valuation of options in jump diffusions is not possible and one is restricted to numerical procedures. These procedures rely on the fact that prices of derivatives in a jump diffusion environment satisfy partial integro-differential equations (PIDEs). The methods discretize the asset-time domain and use either binomial trees, finite difference or finites element methods to solve the PIDEs. [10], for instance, uses an explicit multinomial tree based approach for the valuation. [11] develop another pricing procedure based on a lattice framework. Further recent developments concerning the numerical evaluation include the articles [12-15].

The application of Mellin transforms for the purpose of option pricing was firstly introduced in [16] in the geometric Brownian motion economy. It was extended to the stochastic volatility model of Heston by Frontczak (2011) [17]. This paper extends the Mellin integral transform framework further to the valuation of options in jump-diffusion models. It is organized as follows: Chapter 2 reviews the problem formulation. Chapter 3 provides a solution for European derivatives as a single integral. Also simple formulas for the Greeks are derived. Although our solution is general in the sense that it does

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not depend on the distribution of the jump size, we choose the log-normal jump-diffusion model as an explicit example and show that our solutions for plain vanilla and digital puts may be rewritten as infinite series of Black-Scholes-Merton prices. Some numerical results are presented in Chapter 4. Chapter 5 concludes the paper and indicates possible further research projects.

2. Problem Statement

Let S(t) be the price of a dividend paying stock at time t. The dividend is paid continuously at a rate q and is proportional to the price of the stock. The evolution of the stock is affected by two sources of uncertainty: a continuous part modeled as a standard Brownian motion W(t) and a discontinuous part modeled as jumps with a Poisson arrival process N(t). Therefore, under the physical probability $\mathbb P$ the dynamics of the stock are given by

$$\frac{\mathrm{d}S(t)}{S(t^{-})} = (\mu - q - \lambda k)\mathrm{d}t + \sigma\mathrm{d}W(t) + (Y - 1)\mathrm{d}N(t), \quad (2.1)$$

with initial value $S(0) \in (0,\infty)$ and where $t^- = \lim_{\varepsilon \to 0} t - \varepsilon$ is the time instant before t. The parameters μ and σ are the instantaneous drift and volatility, respectively. The process N(t) is a Poisson process with intensity λ and

$$dN(t) = \begin{cases} 1 & \text{with prob. } \lambda dt \\ 0 & \text{with prob. } (1 - \lambda dt), \end{cases}$$

and Y-1 is the proportional change in the stock price. We also assume that the processes W(t), N(t) and $\{Y_1, Y_2, \dots\}$ are independent. For $\lambda = 0$ the jump diffusion process in (2.1) becomes a geometric Brownian motion as assumed in the Black-Scholes and Merton model. In order to guarantee that the discounted process is a martingale, we have

$$k = E[Y-1] = \int_0^\infty (Y-1) f(Y) dY,$$
 (2.2)

where f(Y) is the probability density function of Y and $E[\cdot]$ is the expectation operator. It is well known that markets with jump diffusions are not complete. Hence, an equivalent martingale measure $\mathbb Q$ is not unique. Nevertheless, if we assume that the jump risk is diversifiable as has been done in [1] we can consider the risk-neutral dynamics of S

$$\frac{\mathrm{d}S(t)}{S(t^{-})} = (r - q - \lambda k)\mathrm{d}t + \sigma\mathrm{d}W(t) + (Y - 1)\mathrm{d}N(t), \quad (2.3)$$

where μ is replaced by the riskless interest rate r. The solution to the process above equals

$$S(t) = S(0)e^{\left(r-q-\lambda k - \frac{1}{2}\sigma^2\right)t + \sigma W(t)} \prod_{i=1}^{N(t)} Y_i.$$

Standard arguments show that a European-style derivative written on S and time t with price F = F(S,t) must solve the backward in time PIDE:

$$F_{t}(S,t) + (r-q-\lambda k)SF_{s}(S,t)$$

$$-rF(S,t) + \frac{1}{2}\sigma^{2}S^{2}F_{ss}(S,t)$$

$$+\lambda \int_{s}^{\infty} (F(SY,t) - F(S,t))f(Y)dY = 0,$$
(2.4)

for all $\mathbb{R}_+ \times [0,T)$ and where subscripts denote partial derivatives. For a derivation see [1,18]. The terminal value of the derivative equals F(S,T) = g(S), where $g(\cdot)$ is the payoff function of the contract under consideration. For a plain vanilla European put with price $P^E(S,t)$ we have

$$g(S) = \max(X - S(T), 0),$$
 (2.5)

where X is the strike price of the put and S(T) denotes the terminal stock price¹. For the jump size distribution different distributions are commonly used in the financial literature. Prominent candidates are among others:

• Log-normally distributed jumps ([1]): The probability density function of Y is given by

$$f(Y) = \frac{1}{Y\sqrt{2\pi\delta^2}} e^{-\frac{1}{2} (\ln Y - \mu)^2},$$
 (2.6)

where Y > 0, and μ and δ are constants.

 Double exponentially distributed jumps ([7]): The probability density function of Y is asymmetric and equals

$$f(Y) = p\eta_1 Y^{-\eta_1 - 1} 1_{\{Y \ge 1\}} + q\eta_2 Y^{\eta_2 - 1} 1_{\{0 < Y < 1\}}, \tag{2.7}$$

where Y > 0, $\eta_1 > 1$, $\eta_2 > 0$, $p,q \ge 0$ with p+q=1. The restriction $\eta_1 > 1$ guarantees that $E[S(t)] < \infty$.

• Gamma distributed jumps: In this case f(Y) equals

$$f(Y) = \frac{1}{\Gamma(\alpha)\beta^{\alpha}} Y^{\alpha-1} e^{\frac{-Y}{\beta}}, \qquad (2.8)$$

where $Y, \alpha, \beta > 0$ and $\Gamma(\cdot)$ denotes the gamma function.

• One-sided exponentially distributed jumps²: The probability density function f(Y) may be seen as a special case of the probability density function of gamma

¹If F is a digital European put option, $F(S,t) = P^{D}(S,t)$ the payoff function equals $g(S) = 1 \cdot 1_{\{S_{T} \leq X\}}$, where X is a prespecified level

²This process was suggested in [19] to incorporate jumps into the stochastic volatility process.

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distributed jumps with $\alpha = 1$. Therefore

$$f(Y) = \frac{1}{\beta} e^{-\frac{Y}{\beta}}.$$
 (2.9)

The particular choice of the distribution of Y is crucial in order to solve the problem analytically in the (S,t)-domain. The only two models where Y is assumed to be continuous and that have been solved explicitly in terms of an infinite series are to the best of our knowledge the Merton (1976) model and the model of Kou (2002).

3. Main Results

The following theorem characterizes the solution to (2.4). The functional form of the solution is simple and is expressed as a single integral.

Theorem 3.1 The solution of the PIDE (2.4) is given by

$$F(S,t) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \theta(\omega) e^{H(\omega)(T-t)} S^{-\omega} d\omega, \qquad (3.1)$$

with S = S(t), $\omega = c + iy$, 0 < c, i is the imaginary unit, $\theta(\omega) = \tilde{F}(\omega, T)$ and where

$$H(\omega) = \frac{1}{2}\sigma^2\omega(\omega+1) - (r+\lambda)(\omega+1) + q\omega + \lambda G(\omega),$$
(3.2)

with

$$G(\omega) = \omega E[Y] + E[Y^{-\omega}]. \tag{3.3}$$

In particular, for the European put $\theta(\omega)$ equals²

$$\theta(\omega) = \frac{X^{\omega+1}}{\omega(\omega+1)},\tag{3.4}$$

Proof: For a locally Lebesgue integrable function $f(x), x \in \mathbb{R}_+$, the Mellin transform $M(f(x), \omega)$ with transformation variable $\omega \in \mathbb{C}$ is defined by the relation

$$M(f(x),\omega) := \tilde{f}(\omega) = \int_0^\infty f(x) x^{\omega-1} dx.$$

Conversely, if $\tilde{f}(\omega)$ is a continuous, integrable function, then the inverse of the Mellin transform is given by

$$f(x) = M^{-1}(\tilde{f}(\omega)) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \tilde{f}(\omega) x^{-\omega} d\omega$$

Let $M(F(S,t),\omega) = \tilde{F}(\omega,t)$ denote the Mellin transform of a financial derivative with price F(S,t). Since the Mellin transform is a linear integral operator it follows that the PIDE from (2.4) can be transformed into a

homogeneous ordinary differential equation (ODE) of the form

$$\tilde{F}_{t}(\omega,t) + H(\omega)\tilde{F}(\omega,t) = 0, \tag{3.5}$$

where

$$H(\omega) = \frac{1}{2}\sigma^{2}\omega(\omega+1) - r(\omega+1) + (q+\lambda k)\omega + \lambda E[Y^{-\omega} - 1].$$
(3.6)

The solution to this ODE equals

$$\tilde{F}(\omega,t) = \theta(\omega) e^{H(\omega)(T-t)}$$

with $\theta(\omega) = \tilde{F}(\omega, T)$. Recalling that k = E[Y-1] the expression for $H(\omega)$ follows immediately. The final expression for the price follows from the inversion theorem for Mellin transforms. \square

The previous theorem gives the analytic solution to (2.4) in a general manner, *i.e.* without specifying the distribution of Y. The solution has a simple functional form and prices are obtained very quickly using standard quadrature routines. It is also flexible to account for different jump size distributions and types of derivatives by specifying $G(\omega)$ and $\theta(\omega)$, respectively, for the contract under consideration. To give explicit examples, we state the expressions for $G(\omega)$ in the four special cases for Y discussed above:

 Log-normally distributed jumps ([1]): Since for each real or complex s we have

$$E\left[Y^{s}\right] = e^{\mu s + \frac{1}{2}s^{2}\delta^{2}},\tag{3.7}$$

it follows that $k = E[Y-1] = e^{\mu + \frac{1}{2}\sigma^2} - 1$ and

$$G(\omega) = \omega e^{\mu + \frac{1}{2}\delta^2} + e^{-\mu\omega + \frac{1}{2}\omega^2\delta^2}.$$
 (3.8)

Double exponentially distributed jumps ([7]): We have

$$E\left[Y^{-\omega}\right] = \frac{p\eta_1}{\eta_1 + \omega} + \frac{q\eta_2}{\eta_2 - \omega},\tag{3.9}$$

with $0 < \text{Re}(\omega) < \eta_2$. Therefore

$$k = E[Y-1] = \frac{p\eta_1}{n_1-1} + \frac{q\eta_2}{n_2+1} - 1$$

and

$$G(\omega) = p\eta_1 \left(\frac{\omega}{\eta_1 - 1} + \frac{1}{\eta_1 + \omega} \right) + q\eta_2 \left(\frac{\omega}{\eta_2 + 1} + \frac{1}{\eta_2 - \omega} \right).$$
(3.10)

• Gamma distributed jumps: In this case $E[Y] = \alpha \beta$

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³For a digital put we have $\theta(\omega) = \frac{X^{\omega}}{\omega}$.

and $k = E[Y-1] = \alpha\beta - 1$. From the identity

$$\int_0^\infty x^{a-1} e^{-bx} dx = \frac{1}{b^a} \Gamma(a), \quad \text{Re}(a) > 0, \text{Re}(b) > 0$$

it follows that

$$G(\omega) = \omega \alpha \beta + \beta^{-\omega} \frac{\Gamma(\alpha - \omega)}{\Gamma(\alpha)}, \tag{3.11}$$

with $0 \le \text{Re}(\alpha - \omega)$.

• One-sided exponentially distributed jumps: Here $k = E[Y-1] = \beta - 1$, and

$$G(\omega) = \omega \beta + \beta^{-\omega} \Gamma(1 - \omega), \tag{3.12}$$

with $0 < \text{Re}(\omega) < 1$.

In each of the above cases, the price F(S,t) can be determined easily by a single integration. Another advantage of the approach is that hedging parameters, commonly known as Greek letters, are easily determined analytically. Popular Greeks widely used for risk management purposes are the delta, the gamma, and the vega of a contract⁴. The results for Greeks are summarized in the next proposition.

Proposition 3.2 Independently of the distribution of the jump size Y, the analytical expressions for the delta, gamma, and vega can be stated as follows:

$$\Delta_F = -\frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \omega \theta(\omega) e^{H(\omega)(T-t)} S^{-(\omega+1)} d\omega, \qquad (3.13)$$

$$\Gamma_F = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \omega(\omega+1) \theta(\omega) e^{H(\omega)(T-t)} S^{-(\omega+2)} d\omega, \quad (3.14)$$

$$\Lambda_{F} = \frac{\sigma(T-t)}{2\pi i} \int_{c-i\infty}^{c+i\infty} \omega(\omega+1) \theta(\omega) e^{H(\omega)(T-t)} S^{-\omega} d\omega
= \sigma(T-t) S^{2} \Gamma_{F},$$
(3.15)

where c > 0 and $H(\omega)$ is given in Equation (3.2). In particular if F(S,t) is a European plain vanilla put option the Greeks are determined as

$$\Delta_{pE} = -\frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \frac{e^{H(\omega)(T-t)}}{\omega+1} \left(\frac{X}{S}\right)^{\omega+1} d\omega, \qquad (3.16)$$

$$\Gamma_{PE} = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \frac{1}{S} e^{H(\omega)(T-t)} \left(\frac{X}{S}\right)^{\omega+1} d\omega, \qquad (3.17)$$

$$\Lambda_{pE} = \frac{\sigma X (T - t)}{2\pi i} \int_{c - i\infty}^{c + i\infty} e^{H(\omega)(T - t)} \left(\frac{X}{S}\right)^{\omega} d\omega.$$
 (3.18)

Proof: The expressions follow directly for Theorem 3.1 since we can differentiate under the integral sign. For the Greeks of a plain vanilla put one uses the Equation for $\theta(\omega)$ from (3.4) and simplifies. \square

Similar to the derivative's price the Greeks are char-

acterized as a single integral which may be evaluated easily. Furthermore, hedge parameters for a range of different types of derivatives may be obtained by specifying the functional form for $\theta(\omega)$.

We complete this section by showing explicitly that Equation (3.1) with the special choices of $F(S,t) = P^{E}(S,t)$ and log-normally distributed jumps is equivalent to Merton's infinite series solution for a European put.

Proposition 3.3 If $F(S,t) = P^E(S,t)$ and jumps are distributed log-normally with parameters μ and δ , $Y \sim LN(\mu, \delta^2)$, then (3.1) is equivalent to Merton's solution which is formulated as an infinite series of BSM-prices:

$$P^{E}(S,t) = \sum_{n=0}^{\infty} \frac{\left(\lambda^{*}(T-t)\right)^{n}}{n!} e^{-\lambda^{*}(T-t)}$$

$$P^{BSM}(S,X,r^{*},q,\sigma^{*},T,t),$$
(3.19)

where $P^{BSM}\left(S,X,r^*,q,\sigma^*,T,t\right)$ denotes the European put price due to Black, Scholes, and Merton and equals

$$P^{BSM}\left(S, X, r, q, \sigma, T, t\right) = Xe^{-r(T-t)}N\left(-d_2\right)$$
$$-S\left(t\right)e^{-q(T-t)}N\left(-d_1\right),$$

with

$$\begin{split} d_1 &= d_1 \left(S, X, r, q, \sigma, T, t \right) \\ &= \frac{\ln \frac{S\left(t\right)}{X} + \left(r - q + \frac{1}{2} \sigma^2 \right) \left(T - t \right)}{\sigma \sqrt{T - t}}, \end{split}$$

$$d_2 = d_2(S, X, r, q, \sigma, T, t) = d_1 - \sigma\sqrt{T - t}.$$

N(x) is the cumulative standard normal distribution function and the adjusted parameters are given by

$$\lambda^* = \lambda (1+k), \ r^* = r - \lambda k + \frac{n \ln(1+k)}{T-t},$$

and
$$\sigma^{*2} = \sigma^2 + \frac{n\delta^2}{T - t}$$
.

We will need the following Lemma. It gives the analytical expressions for two specific Mellin transforms.

Lemma 3.4 Let p(t) be a function of t (time) and the functions $f_1(S,t)$ and $f_2(S,t)$ be defined as

$$f_1(S,t) = \frac{1}{\sqrt{4\pi p(t)}} e^{-\frac{(\ln S(t))^2}{4p(t)}},$$
 (3.20)

and

$$f_2(S,t) = \sum_{n=1}^{\infty} \frac{((T-t)\lambda)^n}{n!} \frac{1}{\sqrt{2\pi n\delta^2}} e^{-\frac{1}{2} \frac{(\ln S(t) + \mu n)^2}{n\delta^2}}.$$
 (3.21)

⁴Others are for instance the theta or the rho. They are determined in the same manner as outlined in the text and therefore omitted.

Then

$$M(f_1(S,t),\omega) = e^{p(t)\omega^2}, \qquad (3.22)$$

and

$$M(f_2(S,t),\omega) = e^{(T-t)\lambda e^{-\mu\omega + \frac{1}{2}\delta^2\omega^2}} - 1.$$
 (3.23)

Proof: Using the definition of Mellin transforms, we have that

$$M\left(f_{1}(S,t),\omega\right) = \int_{0}^{\infty} \frac{1}{\sqrt{4\pi p(t)}} e^{-\frac{\left(\ln S(t)\right)^{2}}{4p(t)}} S\left(t\right)^{\omega-1} dS(t)$$
$$= E\left[S(t)^{\omega}\right],$$

where S(t) is a log-normally distributed random variable with mean 0 and variance 2p(t). The statement follows directly from (3.7). Analogously we have for $f_2(S,t)$ after interchanging summation and integration

$$M\left(f_{2}(S,t),\omega\right) = \sum_{n=1}^{\infty} \frac{\left(\left(T-t\right)\lambda\right)^{n}}{n!}$$

$$\cdot \int_{0}^{\infty} \frac{1}{\sqrt{2\pi n\delta^{2}}} \frac{1}{S(t)} e^{-\frac{1}{2} \frac{\left(\ln S(t) + \mu n\right)^{2}}{n\delta^{2}}} S(t)^{\omega} dS(t),$$

and the integral can be viewed as the ω -th moment of a log-normally distributed random variable with mean $-\mu n$ and variance $n\delta^2$. Therefore

$$M\left(f_{2}\left(S,t\right),\omega\right) = \sum_{n=1}^{\infty} \frac{\left(\left(T-t\right)\lambda\right)^{n}}{n!} e^{-\mu n\omega + \frac{1}{2}n\delta^{2}\omega^{2}},$$

and the assertion follows. \square

Now we are going to prove Proposition 3.3.

Proof: Let $\tilde{g}(\omega,t) = e^{H(\omega)(T-t)}$. Then

$$\tilde{g}(\omega,t) = e^{-\lambda(T-t)} e^{h(\omega)(T-t)}$$

$$+ e^{-\lambda(T-t)} e^{h(\omega)(T-t)} \left(e^{(T-t)\lambda e^{-\mu\omega + \frac{1}{2}\delta^2\omega^2}} - 1 \right),$$
(3.24)

where

$$h(\omega) = \frac{1}{2}\sigma^2\omega(\omega+1) - r(\omega+1) + (q+\lambda k)\omega.$$

Now.

$$h(\omega) = \frac{1}{2}\sigma^2(\omega + a)^2 - \frac{1}{2}\sigma^2(a^2 + b),$$
 (3.25)

with

$$a = \frac{1}{2} \left(1 - \frac{2r - 2(q + \lambda k)}{\sigma^2} \right), \quad b = \frac{2r}{\sigma^2}.$$
 (3.26)

Therefore

$$\tilde{g}(\omega,t) = e^{-\left(\lambda + \frac{1}{2}\sigma^{2}(a^{2} + b)\right)(T-t)} e^{\frac{1}{2}\sigma^{2}(T-t)(\omega+a)^{2}}$$

$$+ e^{-\left(\lambda + \frac{1}{2}\sigma^{2}(a^{2} + b)\right)(T-t)} e^{\frac{1}{2}\sigma^{2}(T-t)(\omega+a)^{2}}$$

$$\cdot \left(e^{(T-t)\lambda e^{-\mu\omega + \frac{1}{2}\delta^{2}\omega^{2}}} - 1\right)$$

$$= \tilde{g}_{1}(\omega,t) + \tilde{g}_{2}(\omega,t).$$

Let $g(S,t) = M^{-1}(\tilde{g}(\omega,t))$. Then it follows from Equations (3.20) and (3.22) that

$$g_{1}(S,t) = e^{-\left(\lambda + \frac{1}{2}\sigma^{2}(a^{2} + b)\right)(T - t)} \cdot \frac{S(t)^{a}}{\sqrt{2\pi\sigma^{2}(T - t)}} e^{-\frac{1}{2}\left(\frac{\ln S(t)}{\sqrt{\sigma^{2}(T - t)}}\right)^{2}}.$$
 (3.27)

For $g_2(S,t)$, we apply the convolution theorem for Mellin transforms and (3.21) and (3.23) to get

$$g_{2}(S,t) = e^{-\left(\lambda + \frac{1}{2}\sigma^{2}(a^{2} + b)\right)(T-t)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} e^{\frac{1}{2}\sigma^{2}(T-t)(\omega + a)^{2}} \cdot \left(e^{(T-t)\lambda e^{-\mu\omega + \frac{1}{2}\delta^{2}\omega^{2}}} - 1\right) S(t)^{-\omega} d\omega$$

$$= e^{-\left(\lambda + \frac{1}{2}\sigma^{2}(a^{2} + b)\right)(T-t)} \sum_{n=1}^{\infty} \frac{\left(\left(T-t\right)\lambda\right)^{n}}{n!}$$

$$\cdot \frac{1}{\sqrt{2\pi\sigma^{2}(T-t)}} \frac{1}{\sqrt{2\pi n\delta^{2}}}$$

$$\times \int_{-\frac{1}{2}}^{\infty} \xi^{a-1} e^{-\frac{1}{2}\left(\frac{\ln\xi}{\sqrt{\sigma^{2}(T-t)}}\right)^{2}} e^{-\frac{1}{2}\left(\frac{\ln\xi-(\ln S(t) + \mu n)}{\sqrt{n\delta^{2}}}\right)^{2}} d\xi.$$
(3.28)

The transformation $\xi = e^z$ gives

$$\begin{split} &\int_0^\infty \xi^{a-1} \mathrm{e}^{-\frac{1}{2} \left(\frac{\ln \xi}{\sqrt{\sigma^2 (T-t)}} \right)^2} \mathrm{e}^{-\frac{1}{2} \left(\frac{\ln \xi - (\ln S(t) + \mu n)}{\sqrt{n \delta^2}} \right)^2} \mathrm{d} \xi \\ &= \mathrm{e}^{-\frac{\tilde{\mu}^2}{2n \delta^2}} \int_{-\infty}^\infty \mathrm{e}^{-P^2 z^2 + Q z} \mathrm{d} z, \end{split}$$

with

$$\tilde{\mu} = \ln S(t) + \mu n, \tag{3.29}$$

$$P^{2} = \frac{2\sigma^{2}(T-t) + 2n\delta^{2}}{2\sigma^{2}(T-t)2n\delta^{2}},$$
(3.30)

$$Q = a + \frac{\tilde{\mu}^2}{2n\delta^2},\tag{3.31}$$

where a is defined in (3.26). From [20] p. 337,

$$\int_{-\infty}^{\infty} e^{-P^2 z^2 + Qz} dz = e^{\frac{Q^2}{4P^2}} \frac{\sqrt{\pi}}{P}, \operatorname{Re}(P^2) > 0.$$

Using this result and simplifying yields after some algebra

$$g_{2}(S,t) = e^{-(\lambda+r)(T-t)} \sum_{n=1}^{\infty} \frac{\left((T-t)\lambda \right)^{n}}{n!} \cdot \frac{1}{\sqrt{2\pi(\sigma^{2}(T-t)+n\delta^{2})}} e^{-\frac{1}{2}\left(\frac{\ln S(t)-(a\sigma^{2}(T-t)-\mu n)}{\sqrt{\sigma^{2}(T-t)+n\delta^{2}}}\right)^{2}}.$$
(3.32)

One observes that the expression for the summands in the Equation for $g_2(S,t)$ is also valid for n = 0. Since

$$-\frac{1}{2} \left(\frac{\ln S(t) - a\sigma^2(T - t)}{\sqrt{\sigma^2(T - t)}} \right)^2$$

$$= -\frac{1}{2} \left(\frac{\ln S(t)}{\sqrt{\sigma^2(T - t)}} \right)^2 + a \ln S(t) - \frac{1}{2} a^2 \sigma^2(T - t),$$

We arrive at the more compact form

$$g(S,t) = e^{-(\lambda+r)(T-t)} \sum_{n=0}^{\infty} \frac{((T-t)\lambda)^{n}}{n!} \cdot \frac{1}{\sqrt{2\pi(\sigma^{2}(T-t)+n\delta^{2})}} e^{-\frac{1}{2}\left(\frac{\ln S(t)-(a\sigma^{2}(T-t)-\mu n)}{\sqrt{\sigma^{2}(T-t)+n\delta^{2}}}\right)^{2}}.$$
 (3.33)

Merton's solution follows now from the identity

$$P^{E}(S,t) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \theta(\omega) \tilde{g}(\omega,t) S(t)^{-\omega} d\omega,$$

where c > 0 and $\theta(\omega)$ is given in Equation (3.4). Since we have been able to derive an analytical expression for the inverse Mellin transform of $\tilde{g}(\omega,t)$, convolution allows to write for the European put price

$$P^{E}(S,t) = \sum_{n=0}^{\infty} \frac{\left(\left(T-t\right)\lambda\right)^{n}}{n!} \frac{e^{-(\lambda+r)(T-t)}}{\sqrt{2\pi\left(\sigma^{2}\left(T-t\right)+n\delta^{2}\right)}}$$

$$\cdot \int_{0}^{X} \left(X-\xi\right) e^{-\frac{1}{2}\left[\frac{\ln\frac{S(t)}{\xi}-\left(a\sigma^{2}\left(T-t\right)-\mu n\right)}{\sqrt{\sigma^{2}\left(T-t\right)+n\delta^{2}}}\right]^{2}} \frac{d\xi}{\xi}.$$
(3.34)

The integrals above can be evaluated straightforwardly. Standard transformations convert both integrals into normal integrals. We get

$$P^{E}(S,t) = e^{-(\lambda+r)(T-t)} X \sum_{n=0}^{\infty} \frac{\left(\left(T-t\right)\lambda\right)^{n}}{n!} N\left(-d_{2}^{*}\right)$$

$$-e^{-q(T-t)-\lambda(1+k)(T-t)} S(t) \sum_{n=0}^{\infty} \frac{\left(\left(T-t\right)\lambda\right)^{n}}{n!} e^{\mu n + \frac{1}{2}n\delta^{2}} N\left(-d_{1}^{*}\right)$$
(3.35)

with the temporary quantities

$$d_{1}^{*} = d_{1}^{*} \left(S, X, r, q + \lambda k, \sigma, T, t, n \right)$$

$$= \frac{\ln \frac{S(t)}{X} + \left(r - \left(q + \lambda k \right) + \frac{1}{2} \sigma^{2} \right) (T - t) + \mu n + n \delta^{2}}{\sqrt{\sigma^{2} (T - t) + n \delta^{2}}}, (3.36)$$

and

$$d_{2}^{*} = d_{2}^{*}(S, X, r, q + \lambda k, \sigma, T, t, n)$$

$$= \frac{\ln \frac{S(t)}{X} + \left(r - (q + \lambda k) - \frac{1}{2}\sigma^{2}\right)(T - t) + \mu n}{\sqrt{\sigma^{2}(T - t) + n\delta^{2}}}.$$
(3.37)

Finally, observing that

$$d_1^* (S, X, r, q + \lambda k, \sigma, T, t, n) = d_1 (S, X, r^*, q, \sigma^*, T, t),$$

and

$$d_{2}^{*}(S, X, r, q + \lambda k, \sigma, T, t, n) = d_{2}(S, X, r^{*}, q, \sigma^{*}, T, t),$$

Merton's solution follows immediately. □

Corollary 3.5 The price of a European digital put $P^{D}(S,t)$ in Merton's model equals

$$P^{D}(S,t) = e^{-(\lambda+r)(T-t)} \sum_{n=0}^{\infty} \frac{\left(\left(T-t\right)\lambda\right)^{n}}{n!}$$

$$\cdot N\left(-d_{2}\left(S,X,r^{*},q,\sigma^{*},T,t\right)\right)$$

$$= \sum_{n=0}^{\infty} e^{-r^{*}(T-t)} \frac{\left(\left(T-t\right)\lambda^{*}\right)^{n}}{n!}$$

$$\cdot e^{-\lambda^{*}(T-t)} N\left(-d_{2}\left(S,X,r^{*},q,\sigma^{*},T,t\right)\right),$$

where the modified parameters r^*, λ^* , and σ^* are defined in Proposition 3.3.

Proof: We have

$$P^{D}(S,t) = \int_{0}^{X} g\left(\frac{S}{\xi},t\right) \frac{\mathrm{d}\xi}{\xi},$$

with g(S,t) given in (3.33). The evaluation of the integral produces the result. \square

4. Numerical Results

This section presents the numerical results from implementing the analytical solutions derived in this paper. We start with options in the log-normal model of Merton (1976). The input parameters can be found in **Table 1**.

The parameters were found in [12] and are based on a calibration using European options on the S&P 500 stock index. Prices are determined for three different stock price levels (in-, out-, and at-the-money options). The solution based on Mellin transforms is given in Theorem 3.1. The integral is evaluated numerically using Gauss-Lobatto quadrature. To get some information about the

convergence of the results we truncate the limits of integration $c \pm i\infty$ gradually at $c \pm i25$, $c \pm i50$, $c \pm i100$, and $c \pm i200$ and choose c = 2. We also compute the "exact" solution using the infinite series representation with n = 30. Tables 2 and 3 show the results for the log-normal model.

The next table summarizes the results for a digital put. Our first and most important finding is that the frame-

work proposed in this article is able to produce highly accurate prices for both types of derivatives. Even if the integration limits are truncated at ±i50 the results are correct up to the third/fourth position after decimal point

Table 1. Input parameters used for the numerical evaluation of the integral for European options in Merton's model.

Input Parameters			
X=100	$\lambda = 0.10$		
r = 0.05	q = 0.00		
$\sigma = 0.15$	T-t=0.25		
$\mu = -0.90$	$\delta = 0.45$		

Table 2. Prices of European put options in Merton's model using the infinite series representation with n = 30 and the integral solution from Theorem 3.1.

	S = 90	S = 100	S = 110			
	Exact Solution					
n = 30	9.28541807	3.14902574	1.40118588			
	Equation (3.1)					
y = 25	9.24635615	3.18309453	1.37553078			
y = 50	9.28542920	3.14905225	1.40120467			
y = 100	9.28541807	3.14902574	1.40118588			
y = 200	9.28541807	3.14902574	1.40118588			

Table 3. Prices of European digital put options in Merton's model using the infinite series representation with n = 30and the integral solution from Theorem 3.1.

	S = 90	S = 100	S = 110		
Exact Solution					
n = 30	0.85489802	0.38715332	0.07792321		
Equation (3.1)					
y = 25	0.85108779	0.39006445	0.07835271		
y = 50	0.85491701	0.38716273	0.07791555		
y = 100	0.85489802	0.38715332	0.07792321		
y = 200	0.85489802	0.38715332	0.07792321		

in each of the stock price scenarios. For $y \ge 100$ the prices coincide up to the eighth decimal place and there are no noteworthy differences between the two frameworks.

We conclude the numerical analysis by considering put options in Kou's double exponential model. The "exact" solution in form of an infinite series can by found in the original article of Kou. The input parameters are the same as above except for the jump component: $\eta_1 = 3.0465, \, \eta_2 = 3.0775$ and p = 0.3445. **Table 4** shows the results for the double exponential model.

Similar to the log-normal model, the numerical approximation of the integral solution produces satisfactory results. The convergence rate is comparable to the previous model.

In summary, the numerical experiments carried out demonstrate that the new framework besides its theoretical insights produces prices with a convincing degree of precision. The Mellin transform approach must be regarded as a capable alternative to existing methods.

5. Conclusions

The purpose of this paper is to explore the valuation of options where the underlying follows a jump-diffusion. The Mellin integral transform approach was applied to solve the homogeneous PIDE analytically. The proposed solution has a very compact form as one single integral which may be evaluated easily and quickly by quadrature. The final pricing formula offers a powerful and flexible tool for handling a variety of jump size distributions and payoff structures. For log-normally distributed jumps, it was shown explicitly that in case plain vanilla puts our integral expression equivalent to the infinite series solution of BSM-prices.

The numerical results are convincing and logically consistent. They come along with the theoretical findings.

Future research projects are multifaceted. An extension of the proposed approach to American options

Table 4. Prices of European put options in Kou's model using the infinite series representation with n = 30 and the integral solution from Theorem 3.1.

	S = 90	S = 100	S = 110			
	Exact Solution					
n = 30	9.43045738	2.73125890	0.55236304			
	Equation (3.1)					
y = 25	9.38984781	2.76604378	0.52982866			
y = 50	9.43049005	2.73129034	0.55236644			
y = 100	9.43045738	2.73125890	0.55236304			
y = 200	9.43045738	2.73125890	0.55236304			

seems to be possible. Also multi-asset options in jump-diffusions could be valued within this framework.

REFERENCES

- [1] R. Merton, "Option Pricing When Underlying Stock Returns Are Discontinuous," *Journal of Financial Economics*, Vol. 3, No. 1-2, 1976, pp. 125-144. doi:10.1016/0304-405X(76)90022-2
- [2] C. Ahn and H. Thompson, "Jump-Diffusion Processes and the Term Structure of Interest Rates," *Journal of Fi*nance, Vol. 43, No. 1, 1988, pp. 155-174. doi:10.1111/j.1540-6261.1988.tb02595.x
- [3] D. Bates, "Jumps and Stochastic Volatility: Exchange Rate Process Implicit in Deutsche Mark Options," *Review of Financial Studies*, Vol. 9, No. 1, 1996, pp. 69-107. doi:10.1093/rfs/9.1.69
- [4] D. Madan, P. Carr and E. Chang, "The Variance Gamma Process and Option Pricing," *European Finance Review*, Vol. 2, 1998, pp. 79-105.
- [5] L. Andersen and J. Andreasen, "Jump-Diffusion Processes: Volatility Smile Fitting and Numerical Methods for Option Pricing," *Review of Derivatives Research*, Vol. 4, No. 3, 2000, pp. 231-262. doi:10.1023/A:1011354913068
- [6] D. Duffie, J. Pan and K. Singleton, "Transform Analysis and Option Pricing for Affine Jump-Diffusions," *Econometrica*, Vol. 68, No. 6, 2000, pp. 1343-1376. doi:10.1111/1468-0262.00164
- [7] S. Kou, "A Jump-Diffusion Model for Option Pricing," *Management Science*, Vol. 48, No. 8, 2002, pp. 1086-1101. doi:10.1287/mnsc.48.8.1086.166
- [8] S. Kou and H. Wang, "Option Pricing under a Double Exponential Jump Diffusion Model," *Management Science*, Vo. 50, No. 9, 2004, pp. 1178-1192. doi:10.1287/mnsc.1030.0163
- [9] A. Sepp, "Analytical Pricing of Double-Barrier Options under a Double-Exponential Jump-Diffusion Process: Applications of Laplace Transform," *International Journal* of Theoretical and Applied Finance, Vol. 7, No. 2, 2004, pp. 151-175. doi:10.1142/S0219024904002402

[10] K. Amin, "Jump Diffusion Option Valuation in Discrete Time," *Journal of Finance*, Vol. 48, No. 5, 1993, pp. 1833-1863. doi:10.1111/j.1540-6261.1993.tb05130.x

- [11] J. Hilliard and A. Schwartz, "Pricing European and American Derivatives under a Jump-Diffusion Process: A Bivariate Tree Approach," *Journal of Financial and Quantitative Analysis*, Vol. 40, No. 3, 2005, pp. 671-691. doi:10.1017/S0022109000001915
- [12] Y. D'Halluin, P. Forsyth and K. Vetzal, "Robust Numerical Methods for Contingent Claims under Jump Diffusion Processes," *IMA Journal of Numerical Analysis*, Vol. 25, No. 1, 2005, pp. 87-112. doi:10.1093/imanum/drh011
- [13] P. Carr and A. Mayo, "On the Numerical Evaluation of Option Prices in Jump Diffusion Processes," *European Journal of Finance*, Vol. 13, No. 4, 2007, pp. 353-372.
- [14] A. Mayo, "Methods for the Rapid Solution of the Pricing PIDEs in Exponential and Merton Models," *Journal of Computational and Applied Mathematics*, Vol. 222, No. 1, 2008, pp. 128-143. doi:10.1016/j.cam.2007.10.017
- [15] S. Clift and P. Forsyth, "Numerican Solution of Two Asset Jump Diffusion Models for Option Valuation," *Applied Numerical Mathematics*, Vol. 58, No. 6, 2008, pp. 743-782. doi:10.1016/j.apnum.2007.02.005
- [16] R. Panini and R. P. Srivastav, "Option Pricing with Mellin Transforms," *Mathematical and Computer Modelling*, Vol. 40, No. 1-2, 2004, pp. 43-56. doi:10.1016/j.mcm.2004.07.008
- [17] R. Frontczak, "Valuing Options in Heston's Stochastic Volatility Model: Another Analytical Approach," *Journal of Applied Mathematics*, Vol. 2011, 2011, Article ID: 198469. doi:10.1155/2011/198469
- [18] R. Cont and P. Tankov, "Financial Modelling with Jump Processes," Chapman & Hall/CRC, Boca Raton, 2004.
- [19] B. Eraker, M. Johannes and N. Polson, "The Impact of Jumps in Volatility and Returns," *Journal of Finance*, Vol. 58, No. 3, 2003, pp. 1269-1300.
- [20] I. Gradshteyn and I. Ryzhik, "Table of Integrals, Series, and Products," 7th Edition, Elsevier Academic Press, Amsterdam, 2007.