A trinomial Markov tree model is studied for pricing options in which the dynamics of the stock price are modeled by the first-order Markov process. Firstly, we construct a trinomial Markov tree with recombining nodes. Secondly, we give an algorithm for estimating the risk-neutral probability and provide the condition for the existence of a validation risk-neutral probability. Thirdly, we propose a method for estimating the volatilities. Lastly, we analyze the convergence and sensitivity of the pricing method implementing trinomial Markov tree. The result shows that, compared to binomial Markov tree, the proposed model is a natural combining tree and, while changing the probability of the node, it is still combining, so the computation is very fast and very easy to be implemented.

1. Introduction

Pricing options have attracted a lot of scholars to research about pricing option by use of tree methods. Black and Scholes put forward the famous pricing option model [1]. However, the knowledge of mathematics of this model is too deep and difficult to understand, and it is not widely known by the general readers. Merton published a paper titled “theory of rational option pricing” so as to achieve a major breakthrough in the field of pricing option for they developed the model known as “Black-Scholes” formula or “Black-Scholes-Merton” [2]. Later, Cox et al. proposed a binomial pricing option model that is widely understood and easy to accept due to its brief mathematical methods and the implicit economic importance, so it is widely used in financial markets [3]. However, because the model only allows two possible states: rise and fall when the underlying asset price changes at a certain time interval, which is up, down, and equal. This is more realistic than the binomial model and makes trinomial tree model for pricing option more accurate in the solution and faster in convergence speed than the binary option, which makes it widely used in pricing more complex option models [5–8]. Zhang solved the pricing option problem under the framework of the uncertain volatility model proposed by Avellaneda; Levy and Par. A trinomial tree can be used to solve the pricing problem for arithmetic average Asian option calculated based on the single stock model [9].

Han raised the trinomial tree model to price options for specific cases in numerical methods and drew relevant results: compared with the binomial model, the trinomial tree model can better approximate to the continuous distribution of the underlying asset price movements with more states and has higher accuracy [10].

Liu et al. assumed interest rates follows a Markov process and derived a different pricing option formula [11]. He compared the rate of convergence between trinomial tree model and binary tree model based on the number of nodes produced, computer time used, and the approximation error and provided the examples to explain that the accuracy of the trinomial tree model was better than that of the binary tree model through Visual Basic program [12].
Xiong presented a binomial pricing option model based on the MCMC method and concluded that it is more accurate than the usual binomial pricing option model although they both underestimate the option price of market [13]. An algorithm for pricing barrier options in one-dimensional Markov models is presented by Mijatovic and Pistorius [14].

Xiong proposed a trinomial pricing option model based on Bayesian Markov Chain Monte Carlo Method which compared the classical binomial tree model, the classical trinomial tree model, the BS model, and the warrant price by using the actual data of the Chinese warrant market; the result shows that the price deviation of the trinomial tree pricing option model based on Bayesian MCMC method is smaller than any other models, although they all underestimate the market price [15].

Yuen and Yang put forward a fast and simple tree model to price simple and exotic options in Markov regime switching model (MRSM) with multiregime. They modified the trinomial tree model of Boyle [4] by controlling the risk neutral probability measure in different regime states to ensure that the tree model can accommodate the data of all different regimes and at the same time preserving its combining tree structure [16]. Bhat and Kumar (2012) proposed the Markov tree (MT) model for pricing option by a non-IID process, a modification of the standard binomial pricing options model, that takes this first-order Markov behavior into account [17]. Then pricing option under a normal mixture distribution derived from the Markov tree model is shown and concludes that the mixture of the two normal distributions fits much better than a single normal.

The existing studies on pricing option by tree methods mainly focus on binomial Markov tree or just trinomial but not considering the first-order Markov process. By construction, the trinomial model has advantage of simplified paths; on the other hand, the probability of the trinomial tree's node is not unique. Hence, the trinomial Markov tree can be seen as combining the strengths of non-IID (independent and identically distributed) models, and trinomial tree methods all within the framework of risk-neutral pricing. In this paper, the main contribution is threefold. (1) A trinomial Markov tree for pricing American options with recombining nodes is proposed. (2) The condition for the existence of a validation risk-neutral probability is provided. (3) An algorithm for estimating the volatilities is given. The essential difference between the trinomial Markov tree and the traditional trinomial tree is that the next period stock price depends not only on the current stock price but also on the history stock price in the trinomial Markov tree model proposed, but the next period stock price only depends on the current stock price in the traditional trinomial tree model.

2. First-Order Markov Process

At first, introduce the definition of first-order Markov process. Set a sequence \( \{Y_n\}_{n \geq 1} \) of random variables. We define

\[
P(Y_n | Y_{n-1}, \ldots, Y_1) = P(Y_n | Y_{n-1}). \tag{1}
\]

This implies that the current node \( Y_n \) is only dependent on the past \( n - 1 \) nodes. Applied to the stock, the stock price in this period is only decided by the last period price.

3. Model

Since the pricing of put option and call option is similar, we take call options as an example to describe trinomial Markov tree method. Let \( S_n \) be the stock's spot price at time step \( n \). \( S_0 \) is the stock's spot price at the beginning, \( K \) is the strike price. When \( n = 0 \), we use one step of the standard trinomial tree:

\[
P(S_1 = uS_0) = P_1
\]
\[
P(S_1 = mS_0) = P_2
\]
\[
P(S_1 = dS_0) = P_3,
\]
as shown in Figure I(a).

For \( n \geq 1 \), we define three events:

\[
S^n_+ = \{S_n > S_{n-1}\}
\]
\[
S^n_0 = \{S_n = S_{n-1}\}
\]
\[
S^n_- = \{S_n < S_{n-1}\},
\]

where the event \( S^n_+ \) is the event that the stock price increased from time step \( n - 1 \) to time step \( n \), the event \( S^n_- \) is the complement of \( S^n_0 \), that is, the event that the stock price unchanged from time step \( n - 1 \) to time step \( n \), the event \( S^n_0 \) is the event that the stock price decreased from time step \( n - 1 \) to time step \( n \).

Next our model for the evolution of \( S_n \) is as follows for \( n \geq 1 \):

\[
P(S_{n+1} = uS_n | S^n_+) = P^n_1
\]
\[
P(S_{n+1} = mS_n | S^n_+) = P^n_2
\]
\[
P(S_{n+1} = dS_n | S^n_+) = P^n_3
\]
\[
P(S_{n+1} = uS_n | S^n_0) = P^n_1
\]
\[
P(S_{n+1} = mS_n | S^n_0) = P^n_2
\]
\[
P(S_{n+1} = dS_n | S^n_0) = P^n_3
\]
\[
P(S_{n+1} = uS_n | S^n_-) = P^n_1
\]
\[
P(S_{n+1} = mS_n | S^n_-) = P^n_2
\]
\[
P(S_{n+1} = dS_n | S^n_-) = P^n_3,
\]

where three symbols: \( u, m, \) and \( d \) represent the different factors by which the stock price at one node in a step could change as three prices of three nodes in the next step. According to our model, if the stock price increased from step \( n - 1 \) to step \( n \), then the stock price at step \( n + 1 \) is \( uS_n \) with probability \( P^n_1 \), if equal, it is \( mS_n \) with probability \( P^n_2 \), and if decreased, it is \( dS_n \) with probability \( P^n_3 \). They are shown in Figures I(b), I(c), and I(d).
We remark that we think of \( \{P_1, P_2, P_3\} \), \( \{P_u^1, P_u^2, P_u^3\} \), \( \{P_m^1, P_m^2, P_m^3\} \), \( \{P_d^1, P_d^2, P_d^3\} \) as, respectively, risk-neutral versions of the empirical probabilities \( \{p(u), p(m), p(d)\} \) \( \{p(u | u), p(m | u), p(d | u)\} \) \( \{p(u | m), p(m | m), p(d | m)\} \) \( \{p(u | d), p(m | d), p(d | d)\} \).

Figure 1 will help us understand these complex probabilities.

Figure 2 is the trinomial tree established. Figure 3 is a trinomial tree with combination nodes and can be used to describe the movement of the stock price.

The calculating process of trinomial tree is similar to that of binary tree. It is pushing down option value from the tail of the tree to the root of the tree. The strike price and the value of holding the option are necessary to calculate at every node. By calculating, the value of holding the option is

\[
    e^{-r\Delta t}(P_u f_u + P_m f_m + P_d f_d).
\]

The value of a node is acquired from three events: \( S_{n+1}^+ \), \( S_n^- \), and \( S_n^0 \). And there are three types of risk-neutral probability because the stock price follows a first-order Markov process.

What’s more, we give the three corresponding option prices as follows:

\[
    f^u_{kij}, f^m_{kij}, f^d_{kij}, \quad (5)
\]

where \( k(k = 0, 1, \ldots, n_1) \) is \( k \) steps (i.e., time \( k\Delta t \)) and \( i(i = -n_1, \ldots, -1, 0, 1, \ldots, n_1) \) is vertical position.

When \( k = 0, f_{0,0} = e^{-r\Delta t}(P_u f^u_{1,1} + P_m f^m_{1,0} + P_d f^d_{1,-1}) \);

When \( 1 \leq k \leq n_1 - 1 \), for the quantity of the paths to reach each node in the trinomial tree, there are three possibilities: three paths, two paths, and one path. Here, we provide the formulas, respectively, for the three possibilities.

For the possibility of three paths, \( 2 - k \leq i \leq k - 2 \),

\[
    f^u_{k,i} = e^{-r\Delta t}(P_u f^u_{k+1,i+1} + P_m f^m_{k+1,i} + P_d f^d_{k+1,i-1}) \quad (6)
\]
For the possibility of two paths, if \( i = k - 1 \),
\[
\begin{align*}
    f_{k,j}^u &= e^{-r \Delta t} \left( p_{1}^u f_{k+1,j+1}^u + p_{2}^u f_{k+1,j}^m + p_{3} f_{k+1,j-1}^d \right) \\
    f_{k,j}^m &= e^{-r \Delta t} \left( p_{1}^m f_{k+1,j+1}^u + p_{2}^m f_{k+1,j}^m + p_{3} f_{k+1,j-1}^d \right) \\
    f_{k,j}^d &= e^{-r \Delta t} \left( p_{1}^d f_{k+1,j+1}^u + p_{2}^d f_{k+1,j}^m + p_{3} f_{k+1,j-1}^d \right).
\end{align*}
\]
(7)

For the possibility of one path, if \( i = k \),
\[
\begin{align*}
    f_{n,i}^u &= f_{n,i}^m = f_{n,i}^d = \max \left( S_0 u^i - K, 0 \right).
\end{align*}
\]
(9)

3.1. Risk-Neutral Probabilities. For the character of the natural recombination, we set \( ud = m^2 = 1 \). Compared with binomial tree, trinomial tree has the advantage of additional freedom degrees.

As a matter of fact, when we construct the trinomial tree to describe the change of the option price, it is to make the shape of the tree and the stock price volatility consistent, by choosing the value of \( P_1, P_2, P_3 \).

Firstly, set
\[
P_1 + P_2 + P_3 = 1.
\]
(12)

Obviously, we can get an analytic solution for (12), (13), and (14) via solving the ternary equations, just as follows:
\[
\begin{align*}
P_1 &= \frac{(S + R^2 - R) u - (R - 1)}{(u - 1)(u^2 - 1)} \\
P_2 &= \frac{Ru^2 - (S + R^2 + 1) u + R}{(u - 1)^2} \\
P_3 &= \frac{(S + R^2 - R - Ru + u) u^2}{(u - 1)(u^2 - 1)}
\end{align*}
\]
(15)

Similarly, there are three volatilities \( \sigma^+, \sigma^-, \sigma^- \) according to the three events, and risk-neutral probabilities can be calculated by formula (15).

Usually, we suppose \( u = e^{\sigma \sqrt{\Delta t}} \).

\( \{P_1, P_2, P_3\} \) are risk-neutral probabilities which should satisfy \( 0 \leq P_1, P_2, P_3 \leq 1 \). However, we perhaps get unreasonable solutions for \( \{P_1, P_2, P_3\} \) if \( u = e^{\sigma \sqrt{\Delta t}} \). So we should take some measures to make sure that the risk-neutral probabilities are reasonable. Thus, \( u \) should satisfy the following conditions corresponding to different volatilities:
\[
\begin{align*}
    (S + R^2 - R) u - (R - 1) > 0 \\
    Ru^2 - (S + R^2 + 1) u + R > 0 \\
    S + R^2 - R - Ru + u > 0.
\end{align*}
\]
(16)

The following descriptions are solutions for (16). And we suppose that
\[
\begin{align*}
    A &= \min \left[ \frac{(S + R^2 - R)}{(R - 1)} \right]_{\sigma^+, \sigma^-, \sigma^-} \\
    B_1 &= \min \left[ \frac{(S + R^2 + 1) - \sqrt{(S + R^2 + 1)^2 - 4R^2}}{2R} \right]_{\sigma^+, \sigma^-, \sigma^-} \\
    B_2 &= \max \left[ \frac{(S + R^2 + 1) + \sqrt{(S + R^2 + 1)^2 - 4R^2}}{2R} \right]_{\sigma^+, \sigma^-, \sigma^-} \\
    C &= \frac{1}{A}
\end{align*}
\]
(17)

If \( B_2 < A \), we suppose \( u = B_2 + \lambda (A - B_2), \lambda \in (0, 1) \). If \( C < B_1 \), we suppose \( u = (1 + B_1)/2 \).

Otherwise there are no reasonable risk-neutral probabilities.

3.2. Volatilities. For each date on which we wish to value an option, we start with the time series of one prior year’s worth of adjusted closing daily returns for the stock. We
scan through this time series and form three disjoint time series: when \( \ln(S_{n+1}/S_n) > a \), we add that return to series 1 (\( a \) is a threshold which is always greater than or equal to 0 and determines the market environment); when \( -a \leq \ln(S_{n+1}/S_n) \leq a \), we add that return to series 2; when \( \ln(S_{n+1}/S_n) < -a \), we add that return to series 3. We then take the logarithm of all returns in series 1, series 2, and series 3 and also in the original time series.

Set \( \sigma_+ \) and \( l_+ \) as the standard deviation and length of log return series, and set \( \sigma_- \) and \( l_- \) as the standard deviation and length of log return series. Let \( \sigma \) be the standard deviation of the entire log return series. The standard deviations are, then, converted to volatilities \( \sigma_+ \), \( \sigma_- \), and \( \sigma \), using \( \sigma = \sqrt{252} \sigma_+ = \sqrt{252} \sigma_- \) and \( \sigma = \sqrt{252} \sigma \).

The parameters \( \sigma_+ \) are calculated precisely in the same way as \( \sigma \), except that, for \( \sigma_- \), we take the standard deviation of log returns on days when the stock’s return increased, and the same way for \( \sigma \), when the stock’s return decreased. This is discussed in greater detail above.

### 4. Comparison with Binomial Markov Tree

In the binomial Markov tree, when the number of states is large, the degree of efficiency of the tree models mentioned above is not high. But, in this paper, the proposed model is a combining tree, with the idea that under the first-order Markov process, we change the probability, the tree is still combining. Since it’s a combining tree, the nodes of the trinomial Markov tree are only \( 2n+1 \) when \( k = n \) and its total node quantity are \( \sum_{k=0}^{n} (2k+1) \), while that of binomial tree is, respectively, \( n^2 - n + 2 \) and \( \sum_{k=0}^{n} (k^2 - k + 2) \). From this, it can be concluded that the computational complexity of trinomial Markov tree is \( o(n^2) \) while it is \( o(n^3) \) in the binomial Markov tree; on the other hand, for each node, the binomial Markov tree only has one value, but there is more than one value in the trinomial Markov tree which is decided by the path which comes to this node; that is to say, the computation of the proposed model is very fast and uncomplicated to implement.

What’s more, the path of the binomial Markov tree is more complex while using the backtracking method to handle. And it is very difficult in the programming application. However, because of the natural recombination, the trinomial Markov tree has its advantage, which can use the same backtracking method with the general trinomial tree, especially in American option.

In addition, there exists a weak point, in the binomial Markov tree; European option has the analytical solution which can be expressed by the formula. Yet the trinomial Markov tree perhaps has one but it is too complex to express.

### 5. Computing Case

#### 5.1. Collecting Data

We estimate the risk-free rate using the no-arbitrage future pricing formula \( F = Se^{rt} \), where \( F \) is the future price, \( S \) is the spot price, and \( t \) is the time horizon until expiration of the future contract. On August 24, 2009, we found that \( S = 75.43 \) and \( F = 75.658 \) for the AI future expiring in December 2009, which also gave \( t = 84 \) trading days = 0.33 years. This yields an annualized risk-free rate of \( r = 0.0090543 \).

We find that Air Liquide (Euronext: AI), a French company for estimating stock prices, uses data from August 25, 2008 to August 24, 2009. On August 24, 2009, we obtained it from euronext.com the end-of-day market prices for European call options of Air Liquide (symbol: AI) expiring in September 2010.

#### 5.2. Convergence

In Figure 4, \( x \)-axis is \( n_1 \), and \( y \)-axis is the option price calculated. We can find from the figure that the value of European call option is increasing and convergent as \( n \) increases.

#### 5.3. Sensitivity of Volatilities

Since the event that stock price remain the same from the day \( n \) to the day \( n+1 \) is almost impossible in the actual market, in order to get a positive volatility of this event, we always let the volatility of district \(-a \leq \ln(S_{n+1}/S_n) \leq a \) as the volatility of this event. So we
Table 1: Sensitivity of volatilities.

<table>
<thead>
<tr>
<th>Strike</th>
<th>Market</th>
<th>B-S</th>
<th>Trinomial</th>
<th>$a = 0.005$</th>
<th>$a = 0.006$</th>
<th>$a = 0.007$</th>
<th>$a = 0.008$</th>
<th>$a = 0.009$</th>
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<td>21.10</td>
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<tr>
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<td>10.27</td>
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<td>0.26</td>
<td>0.24</td>
<td></td>
</tr>
</tbody>
</table>

Note: the first column is the strike price, the second is the market price, and the third is the price calculated by B-S model while the forth is calculated by the trinomial tree. The latter columns are the prices calculated by the proposed model when $a$ is the corresponding value.

Figure 6: The changes of option price as $\lambda$ increases.

The specific reason will be described in the sensitive analysis of $u$ in Section 5.4.

5.4. Sensitivity of $u$. $u = B_2 + \lambda(A - B_2)$, $\lambda \in (0, 1)$ have been described in Section 3.1; therefore, we can analyze the sensitivity of $\lambda$ when $n = 100$.

As seen from Figure 5, the option price has a small change as $\lambda$ varies among $(0.1, 1)$; that is to say, the sensitivity of $u$ is weak. In other words, when calculating ITM, the option price stays stable no matter how $\lambda$ changes among $(0.1, 1)$. This is also the reason why option prices in Markov trinomial tree model are much closer to market prices than that in ordinary trinomial tree model.

In Figure 6, x-axis is $\lambda$ and y-axis is the option price calculated. It can spot from Figure 6 that the option price changes much as $\lambda$ changes when $\lambda$ is between $(0, 1)$. What’s worse, as the strike price of out-of-the-money option becomes higher and higher, the wave character becomes more and more obvious. That is to say, when calculating
out-of-the-money option, the option price is not reliable due to its high sensibility unless we can find a method to determine a reasonable $u$. Therefore, we think that making $u$ equal to $e^{\sigma\sqrt{3\Delta t}}$ is reasonable when calculating out-of-the-money options by numerical experiment.

6. Conclusions

We proposed a trinomial Markov tree for pricing options with recombining nodes. An algorithm for estimating the risk-neutral probability was given, and the condition of a validation risk-neutral probability existing was also provided. And then a method for estimating the volatilities was proposed. We analyze the convergence and sensitivity of the pricing method implementing trinomial Markov tree. Compared to binomial Markov tree, the proposed model is a natural combining tree; when changing the probability of the node, it is still combining, so we can draw the conclusion that the pricing method of trinomial Markov tree is very fast and very easy to implement.

Conflict of Interests

The authors declare that there is no conflict of interests regarding the publication of this paper.

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