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Prospect theory: An application to European option pricing
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December 2012

Abstract
Empirical studies on quoted options highlight deviations from the theoretical model of Black and Scholes; this is due to different causes, such as assumptions regarding the price dynamics, markets frictions and investors' attitude toward risk. In this contribution, we focus on this latter issue and study how to value European options within the continuous cumulative prospect theory. According to prospect theory, individuals do not always take their decisions consistently with the maximization of expected utility. Decision makers have biased probability estimates; they tend to underweight high probabilities and overweight low probabilities. Risk attitude, loss aversion and subjective probabilities are described by two functions: a value function and a weighting function, respectively. In our analysis, we use alternative probability weighting functions. We consider the pricing problem both from the writer's and holder's perspective, obtaining an interval for the prices of call and put options.

Keywords
Behavioral Finance, Cumulative Prospect Theory, European Option Pricing.

JEL Codes
C63, D81, G13.

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1 Introduction

Black and Scholes model is considered as a milestone in the option pricing literature, it is widely applied in financial markets, and has been developed in many directions. Nevertheless, empirically one observes systematic differences between the market prices of financial options and the corresponding theoretical prices; this is due to different causes, such as assumptions regarding the price dynamics, and volatility in particular, the presence of markets frictions, information imperfections, and investors’ attitude toward risk. Normally one tries to improve the performance of models considering more complex dynamics for the prices of the underlying assets, but leaving unchanged decision maker’s preferences. An alternative approach is to price options considering behavioral aspects of the operators. In this contribution we focus on this issue and study how to evaluate options within the cumulative prospect theory developed by Tversky and Kahneman (1992).

According to the prospect theory, individuals do not always take their decisions consistently with the maximization of expected utility. Decision makers are risk averse when considering gains and risk-seeking with respect to losses. They are loss averse; people are much more sensitive to losses than they are to gains of comparable magnitude. People evaluate gambles based on potential gains and losses relative to a reference point, rather than in terms of final wealth. Individuals have also biased probability estimates; they tend to underweight high probabilities and overweight low probabilities. Kahneman and Tversky (1979) provide empirical evidence of such behaviors. Risk attitude, loss aversion and subjective probabilities are described by two functions: a value function, which is typically concave for gains and convex for losses, and a weighting function, which models probability perception.

In many financial application, and in particular when dealing with options, prospects involve infinitely many values or a continuum of values; hence, prospect theory cannot be applied directly in its original or cumulative versions. Davis and Satchell (2007) provide results for the cumulative prospect value in order to deal with continuous outcomes.

The literature on behavioral finance (see e.g. Barberis and Thaler, 2003, for a survey) and prospect theory is huge, whereas a few studies in this field focus on financial options. A first contribution which applies prospect theory to options valuation is the work of Shefrin and Statman (1993), who consider covered call options in a one period binomial model. A list of paper on this topic should include: Poteshman and Serbin (2003), Abbink and Rockenbach (2006), Breuer and Perst (2007), and more recently Wolff et al. (2010). Following this direction, we apply the cumulative prospect theory in the continuous case in order to evaluate European plain vanilla options, extending the model to the European put option. We consider both the positions of the writer and the holder and obtain an interval.
estimation for the option price. We also use alternative probability weighting functions, and discuss some numerical examples. Empirical evidence suggests that the prospect theory can yield appreciable results when applied to the pricing of option contracts.

The rest of the paper is organized as follows. Section 2 synthesizes the main features of prospect theory and introduce the value and the weighting functions; Section 3 focuses on the application of continuous cumulative prospect theory to European option pricing; in Section 4 numerical results are provided and discussed.

2 Prospect Theory

Prospect theory (PT), in its formulation proposed by Kahneman and Tversky (1979), is based on the subjective evaluation of prospects. Originally PT deals only with a limited set of prospects.

With a finite set of potential future states of nature \( S = \{s_1, s_2, \ldots, s_N\} \), a prospect is a vector

\[
(\Delta x_1, p_1; \Delta x_2, p_2; \ldots; \Delta x_N, p_N)
\]

of pairs \((\Delta x_i, p_i)\), \(i = 1, 2, \ldots, N\). Assume \(\Delta x_i \leq \Delta x_j\) for \(i < j\), \(i, j = 1, 2, \ldots, N\), and \(\Delta x_i \leq 0\) (\(i = 1, 2, \ldots, k\)) and \(\Delta x_i > 0\) (\(i = k + 1, \ldots, N\)). Prospects assign to any possible state of nature \(s_i\) a subjective probability \(p_i\) and an outcome \(\Delta x_i\). Outcome \(\Delta x_i\) is defined relative to a certain reference point \(x^*\); being \(x_i\) the absolute outcome, we have \(\Delta x_i = x_i - x^*\).

2.1 Cumulative Prospect Theory

Cumulative prospect theory (CPT) developed by Tversky and Kahneman (1992) overcomes some drawbacks of the original prospect theory. Risk attitude, loss aversion and subjective probabilities are described by two functions: a value function, which is typically concave for gains and convex for losses, and a weighting function, which models probability perception.

Let us denote now with \(\Delta x_i\), for \(-m \leq i < 0\) negative outcomes and with \(\Delta x_i\), for \(0 < i \leq n\) positive outcomes, with \(\Delta x_i \leq \Delta x_j\) for \(i < j\). According to CPT, an investor’s subjective value for a prospect is

\[
V = \sum_{i=-m}^{n} \pi_i \cdot v(\Delta x_i)
\]
with decision weights $\pi_i$ and value function $v(\Delta x_i)$, which is based on relative outcomes. The function $v$ is typically convex in the range of losses and concave in the range of gains.

In the CPT, subjective values $v(\Delta x_i)$ are not multiplied by probabilities $p_i$. Decision weights $\pi_i$ are computed on the basis of a probability weighting function $w^-$ for losses and $w^+$ for gains, considering cumulative probabilities instead of $p_i$:

$$\pi_i = \begin{cases} 
  w^-(p_{-m}) & i = -m \\
  w^-(\sum_{j=-m}^{j-1} p_j) - w^-(\sum_{j=-m}^{j-1} p_j) & i = -m + 1, \ldots, -1 \\
  w^+(\sum_{j=1}^{n} p_j) - w^+(\sum_{j=1}^{n} p_j) & i = 0, \ldots, n - 1 \\
  w^+(p_n) & i = n. 
\end{cases}$$

(2)

### 2.2 Framing and Mental Accounting

Prospect theory postulates that decision makers evaluate outcomes with respect to deviations from a reference point rather than with respect to net final wealth. The definition of such a reference point is crucial due to the fact that individuals evaluate results through a value function which gives more weight to losses than to gains of comparative magnitude. Individual’s framing of decisions around a reference point is of great importance in prospect theory.

People tend to segregate outcomes into separate mental accounts, these are then evaluated separately for gains and losses. Thaler (1995) argues that, when combining such accounts to obtain overall result, typically individuals do not simply sum up all monetary outcomes, but use hedonic frame, such that the combination of the outcomes appears best possible.

Consider a combination of two sure outcomes $\Delta x$ and $\Delta y$, the hedonic frame can be described as follows (see Thaler, 1999):

$$V = \max\{v(\Delta x + \Delta y), v(\Delta x) + v(\Delta y)\}.$$  

(3)

Outcomes $\Delta x$ and $\Delta y$ are aggregated, and in such a case we have $v(\Delta x + \Delta y)$, or segregated $v(\Delta x) + (\Delta y)$, depending on what yields the highest prospect value.

An extension of the hedonic frame rule is (see also Breuer and Perst, 2007):

$$V = \sum_{i=1}^{N} \pi_i \cdot \max\{v(\Delta x_i + \Delta y), v(\Delta x_i) + v(\Delta y)\} +$$

$$+ \left(1 - \sum_{i=1}^{N} \pi_i\right) \cdot \max\{v(0 + \Delta y), v(0) + v(\Delta y)\},$$

(4)
where $\Delta x_i$ are possible results with subjective probabilities $\pi_i$ and $\Delta y$ is a sure result.

Regarding the valuation of financial options, different aggregation or segregation of the results are possible. One can consider a single option position (narrow framing) or a portfolio, a naked or a covered position. It is also possible to segregate results across time: e.g. one can evaluate separately the premium paid for the option and its final payoff.

### 2.3 Continuous Cumulative Prospect Theory

In order to apply prospect theory in its cumulative version to option valuation, one has to deal with continuous results. Davis and Satchell (2007) provide the continuous cumulative prospect value $V$:

$$V = \int_{-\infty}^{0} \Psi - \left[F(x)\right] f(x) v^-(x) \, dx + \int_{0}^{\infty} \Psi + \left[1 - F(x)\right] f(x) v^+(x) \, dx,$$

where $\Psi = \frac{d w(p)}{dp}$ is the derivative of the weighting function $w$ with respect to the probability variable, $F$ is the cumulative distribution function and $f$ is the probability density function of the outcomes $x$; $v^-$ and $v^+$ denote the value function for losses and gains, respectively.

### 2.4 The value function

Specific parametric forms have been suggested in the literature for the value function; some examples are reported in table 1. A function which is used in many empirical studies is the following value function

$$v^- = -\lambda (-x)^b \quad x < 0 \quad v^+ = x^a \quad x \geq 0,$$

with positive parameters which control risk attitude ($0 < a \leq 1$ and $0 < b \leq 1$) and loss aversion ($\lambda > 1$). Function (6) has zero as reference point; it satisfies the properties required by Kahneman and Tversky (1979) for a value function: it is concave for positive outcomes and convex for negative outcomes, it is steeper for losses. Figure 1 shows an example of the value function defined by (6). Parameters values equal to one imply risk and loss neutrality.

### 2.5 The weighting function

The weighting function models risk attitude towards probabilities. Empirical evidence suggests a particular shape for probability weighting: small probabilities
Figure 1: Value function (6) with parameters $\lambda = 2.25$ and $a = b = 0.88$

Table 1: Alternative value functions

<table>
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<tr>
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<th>formula</th>
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</thead>
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<tr>
<td>Linear</td>
<td>$v(x) = x$</td>
</tr>
<tr>
<td>Logarithmic</td>
<td>$v(x) = \ln(a + x)$</td>
</tr>
<tr>
<td>Power</td>
<td>$v(x) = x^a$</td>
</tr>
<tr>
<td>Quadratic</td>
<td>$v(x) = ax - x^2$</td>
</tr>
<tr>
<td>Exponential</td>
<td>$v(x) = 1 - e^{-ax}$</td>
</tr>
<tr>
<td>Bell</td>
<td>$v(x) = bx - e^{-ax}$</td>
</tr>
<tr>
<td>HARA</td>
<td>$v(x) = -(b + x)^a$</td>
</tr>
</tbody>
</table>

are overweighted whereas individuals tend to underestimate large probabilities. This turns out in a typical inverse S-shape weighting function: the function is concave (probabilistic risk seeking) in the interval $(0, p^*)$ and convex (probabilistic risk aversion) in the interval $(p^*, 1)$, for a certain value of $p^*$. In many empirical studies, the intersection between the weighting function and the linear weighting function $w(p) = p$ (elevation) is for $p^*$ in the interval $(0.3, 0.4)$. Different parametric forms for the weighting function with the above mentioned features have been proposed\(^3\); some examples are reported in table 2.

In this contribution, we first consider the functional form suggested by Tversky and Kahneman (1992):

\[
w^-(p) = \frac{p_\gamma}{(p_\gamma + (1-p)_\gamma)^{1/\gamma}} \quad w^+(p) = \frac{p_\gamma^+}{(p_\gamma^+ + (1-p)_\gamma^+)^{1/\gamma^+}},
\]

\(^3\)See Diecidue et al. (2009) for a discussion and references therein.
Table 2: Alternative weighting functions.

<table>
<thead>
<tr>
<th>function</th>
<th>formula</th>
</tr>
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<tr>
<td>Linear</td>
<td>$w(p) = p$</td>
</tr>
<tr>
<td>Power</td>
<td>$w(p) = p^\gamma$</td>
</tr>
<tr>
<td>Lattimore et al.</td>
<td>$w(p) = \frac{\delta p^\gamma}{\delta p^\gamma + (1-p)^\gamma}$</td>
</tr>
<tr>
<td>Tversky-Kahneman</td>
<td>$w(p) = \frac{p^\gamma}{p^\gamma + (1-p)^\gamma}$</td>
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<tr>
<td>Wu-Gonzales</td>
<td>$w(p) = \frac{(p^\gamma + (1-p)^\gamma)^\delta}{p^\gamma}$</td>
</tr>
<tr>
<td>Prelec-I</td>
<td>$w(p) = e^{-(-\ln p)^\gamma}$</td>
</tr>
<tr>
<td>Prelec-II</td>
<td>$w(p) = e^{-\delta(-\ln p)^\gamma}$</td>
</tr>
</tbody>
</table>

Figure 2: Weighting function (7) for different values of the parameter $\gamma$. As $\gamma$ approaches the value 1, the $w$ tends to the linear function.

where $\gamma^-$ and $\gamma^+$ are positive constants (with some constraint in order to have an increasing function). Note that $w(0) = 0$ and $w(1) = 1$. Hence in equation (5) we have:

$$
\Psi = \frac{dw(p)}{dp} = \gamma p^{\gamma - 1} [p^\gamma + (1-p)^\gamma]^{-1/\gamma} - p^\gamma [p^{\gamma - 1} - (1-p)^{\gamma - 1}] [p^\gamma + (1-p)^\gamma]^{-(\gamma + 1)/\gamma}.
$$

(8)

The parameter $\gamma$ captures the the degree of sensitivity to changes in probabilities from impossibility (zero probability) to certainty; the lower the parameter, the higher is the curvature of the function.
Figure 3: Prelec’s weighting function (9) for different values of the parameter $\gamma$

As an alternative, we also consider the Prelec’s (1998) weighting function

$$w(p) = \exp[-(-\ln p)^\gamma] \quad p \in (0, 1),$$

for which one easily obtains

$$\Psi(p) = \frac{\gamma}{p} (-\ln p)^{\gamma-1} \exp[-(-\ln p)^\gamma].$$

More generally, Prelec suggests the two parameter function

$$w(p) = e^{-\delta(-\ln p)^\gamma}.$$

The weighting function $w$ may be one of the main causes of the options’ mispricing through its effect to the prospect value (see Shiller, 1999). Figures 2 and 3 show some examples of weighting functions defined by (7) and (9) for different values of the parameters. As the parameters tend to the value 1, the weight tends to the objective probability and the function $w$ approaches the 45° line. One can assume different parameters for probabilities when the outcome is in the domain of gains or losses. In order to use a more parsimonious model, one assumes $\gamma^+ = \gamma^-$.

An interesting parametric function is the switch-power weighting function proposed by DiCicdue et al. (2009), which consists in a power function for probabilities below a certain value $\bar{p}$ and a dual power function for probabilities above $\bar{p}$:

$$w(p) = \begin{cases} 
  cp^a & p \leq \bar{p} \\
  1 - d(1 - p)^b & p > \bar{p},
\end{cases}$$

(12)
Such a function has five parameters: $a$, $b$, $c$, $d$, and $\bar{p}$, which reduce to three by assuming continuity of $w(p)$ at $\bar{p}$ and differentiability. All the parameters are positive, due to continuity and monotonicity. For $a,b \leq 1$ the function $w$ has an inverse $S$-shaped form, while for $a,b \geq 1$ the weighting function in convex for $p < \bar{p}$ and concave for $p > \bar{p}$. As $\bar{p}$ tends to 0 one obtains a power function, whereas as $\bar{p}$ approaches 1 the function $w$ reduces to a dual power function. Both parameter $a$ and $b$ govern the curvature of $w$ when $a \neq b$. In such a case $\bar{p}$ may not lie on the $45^\circ$ line, hence it has not the meaning of dividing the region of over- and underweighting of the probability.

3 European options valuation

Let $S_t$ be the price at time $t$, with $t \in [0,T]$, of the underlying asset of a European option with maturity $T$; in a Black-Scholes setting, the underlying price dynamics is driven by a geometric Brownian motion. As a result, for the probability density function (pdf) in (5) we have

$$f(S_T) = \frac{1}{S_T \sigma \sqrt{2\pi T}} \exp \left( -\frac{[\ln(S_T/S_0)-(\mu-\sigma^2/2)T]^2}{2\sigma^2 T} \right),$$

where $\mu$ and $\sigma > 0$ are constants. The cumulative distribution function (cdf) of $S_T$ is

$$F(S_T) = \Phi \left( \frac{\ln(S_T/S_0)-(\mu-\sigma^2/2)T}{\sigma \sqrt{T}} \right),$$

where $\Phi(\cdot)$ is the cumulative distribution function of a standard Gaussian random variable.

Wolff et al. (2010) provide the prospect value of writing $L$ call options and compare their results for the call option with those obtained applying Heston (1993) stochastic volatility model.

Let $c$ be the option premium with strike price $X$. At time $t = 0$, the option’s writer receives $c$ and can invest the premium at the risk-free rate $r$, obtaining $L c e^{rT}$. At maturity, he has to pay the amount $L(S_T - X)$ if the option expires in-the-money. In what follows, we will consider a single contract, $L = 1$.

Considering zero as a reference point, the prospect value of the writer’s position in the time segregated case is

$$V_s = v^+ (c e^{rT}) + \int_{X}^{\infty} \Psi^- (1 - F(S_T)) f(S_T) v^- (X - S_T) \, dS_T,$$

with $f(S_T)$ and $F(S_T)$ being the pdf and the cdf defined in (13) and (14), respectively, of the future underlying price $S_T$, and $v$ is defined as in (6).
In equilibrium, we equate \( V \) at zero and solve for the price \( c \):

\[
c_s = e^{-rT} \left( \lambda \int_{X}^{+\infty} \Psi^-(1-F(S_T)) f(S_T) (S_T - X)^b dS_T \right)^{1/a},
\]

which requires numerical approximation of the integral.

When considering the *time aggregated* prospect value, one obtains:

\[
V_a = w^+ (F(X)) v^+ (c e^{rT}) + \\
+ \int_{X}^{X+c \exp(rT)} \Psi^+ (F(S_T)) f(S_T) v^+ (c \exp(rT) - (S_T - X)) dS_T + \\
+ \int_{X+c \exp(rT)}^{+\infty} \Psi^- (1-F(S_T)) f(S_T) v^- (c \exp(rT) - (S_T - X)) dS_T.
\]

In this latter case, the equilibrium option price has to be determined numerically.

In the case of a put option one cannot use put-call parity arguments; the prospect value of the writer’s position in the time segregated case is

\[
V_s = v^+ (p e^{rT}) + \int_{0}^{X} \Psi^- (F(S_T)) f(S_T) v^- ((S_T - X)) dS_T,
\]

and one obtains

\[
p_s = e^{-rT} \left( \lambda \int_{0}^{X} \Psi^- (F(S_T)) f(S_T) (X - S_T)^b dS_T \right)^{1/a}.
\]

In the time aggregated case the put option value is implicitly defined by the following equation:

\[
V_a = \int_{0}^{X-p e^{rT}} \Psi^- [F(S_T)] f(S_T) v^- [p e^{rT} - (X - S_T)] dS_T + \\
+ \int_{X-p e^{rT}}^{X} \Psi^+ [1-F(S_T)] f(S_T) v^+ [p e^{rT} - (X - S_T)] dS_T + \\
+ w^+ [1-F(X)] v^+ [p e^{rT}].
\]

### 3.1 Option valuation from holder’s perspective

When one considers the problem from the holder’s viewpoint, the prospect values both in the time segregated and aggregated cases changes. Holding zero as reference point, the prospect value of the holder’s position for a call option in the time segregated case is

\[
V_h^b = v^- (-c e^{rT}) + \int_{X}^{+\infty} \Psi^+ (1-F(S_T)) f(S_T) v^+ ((S_T - X)) dS_T,
\]

and for a put option

\[
V_h^p = v^+ (-p e^{rT}) + \int_{0}^{X} \Psi^- (1-F(S_T)) f(S_T) v^- ((S_T - X)) dS_T.
\]
with \( f(S_T) \) and \( F(S_T) \) being the pdf and the cdf defined in (13) and (14) of the future underlying price \( S_T \), and \( v \) is defined as in (6).

We equate \( V_s^h \) at zero and solve for the price \( c \), obtaining:

\[
c_s^h = e^{-rT} \left( \frac{1}{\lambda} \int_X^{+\infty} \Psi^+ (1 - F(S_T)) f(S_T) (S_T - X)^a dS_T \right)^{1/b}.
\] (22)

In the time aggregated case, the prospect value has the following integral representation:

\[
V_a^h = w - (F(X)) v^- (-c e^{rT}) + \int_X^{X+c \exp(rT)} (F(S_T)) f(S_T) v^- ((S_T - X) - c \exp(rT)) dS_T + \int_{X+c \exp(rT)}^{+\infty} (1 - F(S_T)) f(S_T) v^+ ((S_T - X) - c \exp(rT)) dS_T.
\] (23)

In order to obtain the call option price in equilibrium, one has to solve numerically the problem.

In an analogous way one can derive the put option prospect values for the holder’s position.

### 4 Numerical Results

We have calculated the options prices both in the time segregated and aggregated case. Tables 3 and 4 report the results for the European calls and puts, respectively, for different strikes and values of the parameters. Consider a single contract \((L = 1)\). When we set \( \mu = r \), \( a = b = 1 \), \( \lambda = 1 \), and \( \gamma = 1 \), we obtain the same results as in the Black-Scholes model (BS prices are reported in the second column). We compare BS premia with prices obtained considering the parameters used in Tversky and Kahneman (1992). Segregated values combined with TK sentiment yield too high option prices. We then used the moderate sentiment parameters as in Wolff et al. (2010) and compare the prices obtained considering different weighting functions: in particular, we applied (7) and Prelec’s function (9); results are reported in the last two columns of the tables.

Table 5 reports the results for the call option evaluated by the holder\(^4\). The prices are below the writer’s results. In all cases, segregated prospect values combined with TK sentiment provide too low option prices to be used in practice.

If one considers the pricing problem both from the writer’s and holder’s perspective, it is possible to obtain an interval for the prices of call and put options.

\(^4\)Similar results are obtained for the put option.
In table 6 we report some results for the call option: Black-Scholes price always lies in the interval bounded by the holder’s price from below and the writer’s price from above. The range of such an interval depends on the value of the parameters which govern investor’s sentiment (attitude toward risk and loss aversion and probability bias). More moderate sentiment implies smaller estimate intervals.

As already observed, the weighting function \( w \) may be one of the main causes of the mispricing. Option prices are sensitive to the choice of the parameters values. In order to analyze the sensitivity of the option values to prospect theory parameters, we performed some numerical experiments. We isolate the effect of the probability bias, considering a linear value function, hence assuming risk neutrality over gains and losses (with parameters \( a = b = 1 \) and \( \lambda = 1 \)), and letting varying the parameter \( \gamma \) of the weighting function.

In the examples we used the TK weighting function, distinguishing the cases with sensitivity to probability risk both for positive and negative outcomes, assuming \( \gamma^+ = \gamma^- \) in order to have a more parsimonious model (see figure 4), then holding one of the parameter equal to 1 as in the results shown in figure 5 (\( \gamma^+ = 1 \)) and figure 6 (\( \gamma^- = 1 \)). We computed the option prices for different values of the volatility on the underlying asset price (the other option parameters are \( S_0 = 100, X = 100, \mu = r = 0.01, T = 1 \)), both in the time segregated and aggregated cases; note that the option values in the segregated prospect for the writer do not depend on the parameter \( \gamma^+ \).

Finally figure 7 compares the results for two volatility values; we make different hypothesis about probability weighting when considering gains or losses. All the prices in figures 4 to 7 converge decreasing to the Black and Scholes values as \( \gamma \) approaches to 1 (and when \( a = b = 1 \) and \( \lambda = 1 \)). It is also interesting to note that, when considering the writer’s payoff, the major effect on the option value is explained by the parameter \( \gamma^- \), as is also intuitive (the segregated prospect value does depend only on the parameter \( \gamma^- \)).

5 Concluding remarks

Prospect theory has recently begun to attract attention in the literature on financial options valuation; when applied to option pricing in its continuous cumulative version, it seems a promising alternative to other models proposed in the literature, for its potential to explain option mispricing with respect to Black and Scholes model. In particular, we focus on the probability bias and compared the results obtained with two alternative weighting functions.

It is worth noting that hypothesis on the segregation of the results are also important: in particular, results obtained in the time aggregated case with moderate sentiment do not deviate too far from BS prices.
As a final comment, note that option prices are sensitive to the choice of the values of the parameters. Calibrating model parameters to market data and studying market sentiment is an important issue which requires further investigation.

References


Table 3: Call option prices in the Black-Scholes model and in the segregated and aggregated prospects. The parameters are: $S_0 = 100, \mu = 0.05, r = 0.05, \sigma = 0.2, T = 1$, and $L = 1$. In the BS model: $a = b = 1$, $\lambda = 1$, and $\gamma = 1$. The Tversky-Kahneman parameters are: $a = b = 0.88$, $\lambda = 2.25$, $\gamma^+ = 0.61$, and $\gamma^- = 0.69$. The moderate sentiment parameters are: $a = b = 0.988$, $\lambda = 1.125$, $\gamma^+ = 0.961$, and $\gamma^- = 0.969$.

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<th>$X$</th>
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Table 4: Put option prices in the Black-Scholes model and in the segregated and aggregated prospects. The parameters are: $S_0 = 100, \mu = 0.05, r = 0.05, \sigma = 0.2, T = 1$, and $L = 1$. In the BS model: $a = b = 1$, $\lambda = 1$, and $\gamma = 1$. The Tversky-Kahneman parameters are: $a = b = 0.88$, $\lambda = 2.25$, $\gamma^+ = 0.61$, and $\gamma^- = 0.69$. The moderate sentiment parameters are: $a = b = 0.988$, $\lambda = 1.125$, $\gamma^+ = 0.961$, and $\gamma^- = 0.969$.

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Table 5: Call option prices in the Black-Scholes model and in the segregated and aggregated prospects from the holder’s viewpoint. The parameters are: $S_0 = 100$, $\mu = 0.05$, $r = 0.05$, $\sigma = 0.2$, $T = 1$, and $L = 1$. In the BS model: $a = b = 1$, $\lambda = 1$, and $\gamma = 1$. The Tversky-Kahneman parameters are: $a = b = 0.88$, $\lambda = 2.25$, $\gamma^+ = 0.61$, and $\gamma^- = 0.69$. The moderate sentiment parameters are: $a = b = 0.988$, $\lambda = 1.125$, $\gamma^+ = 0.961$, and $\gamma^- = 0.969$.

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Table 6: Call option prices (Black-Scholes formula) and lower (holder’s position) and upper (writer’s position) estimates in the CPT. The parameters are: $S_0 = 100$, $\mu = r = 0.1$, $\sigma = 0.2$, $T = 1$. In the BS model: $a = b = 1$, $\lambda = 1$, and $\gamma = 1$. The Tversky-Kahneman parameters are: $a = b = 0.88$, $\lambda = 2.25$, $\gamma^+ = 0.61$, and $\gamma^- = 0.69$. The moderate sentiment parameters are: $a = b = 0.988$, $\lambda = 1.125$, $\gamma^+ = 0.961$, and $\gamma^- = 0.969$.

<table>
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<th>$c_a$</th>
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<td>4.38–5.48</td>
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</table>
Figure 4: Sensitivity of the call option prices (writer’s position) to probability weighting in the CPT. The parameters are: $S_0 = 100, X = 100, \mu = r = 0.01$, for different values of the volatility $\sigma = 0.1, 0.2, 0.3, 0.4$, $T = 1$; with linear value function ($a = b = 1, \lambda = 1$), TK weighting function with parameters varying in the interval $[0.7, 1.0]$ and $\gamma^+ = \gamma^-$, in the segregated (left figure) and aggregate (right figure) cases.

Figure 5: Sensitivity of the call option prices (writer’s position) to probability weighting in the CPT. The parameters are: $S_0 = 100, X = 100, \mu = r = 0.01$, for different values of the volatility $\sigma = 0.1, 0.2, 0.3, 0.4$, $T = 1$; with linear value function ($a = b = 1, \lambda = 1$), TK weighting function with parameter $\gamma^-$ varying in the interval $[0.7, 1.0]$ and $\gamma^+ = 1$, in the segregated (left figure) and aggregate (right figure) cases.
Figure 6: Sensitivity of the call option prices (writer’s position) to probability weighting in the CPT. The parameters are: $S_0 = 100$, $X = 100$, $\mu = r = 0.01$, for different values of the volatility $\sigma = 0.1, 0.2, 0.3, 0.4$, $T = 1$; with linear value function ($a = b = 1$, $\lambda = 1$), TK weighting function with parameter $\gamma^+$ varying in the interval $[0.7, 1.0]$ and $\gamma^- = 1$, in the segregated (left figure) and aggregate (right figure) cases.

Figure 7: Sensitivity of the call option prices to probability weighting in the CPT (writer’s position in the time aggregated case). The parameters are: $S_0 = 100$, $X = 100$, $\mu = r = 0.01$, with $\sigma = 0.2$ (left figure) and $\sigma = 0.3$ (right figure); with linear value function ($a = b = 1$, $\lambda = 1$), TK weighting function with parameters $\gamma^+$ and $\gamma^-$ varying in the interval $[0.7, 1.0]$.