KURTOSIS IN BLACK-SCHOLES MODEL WITH GARCH VOLATILITY

Muhammad SHERAZ\textsuperscript{1}, Vasile PREDA\textsuperscript{2}

The famous Black-Scholes option pricing model is a mathematical description of financial market and derivative investment instruments [3]. In Black-Scholes model volatility is a constant function, where trading option is indeed risky due to random components such as volatility. The notion of non constant volatility was introduced in GARCH processes [6]. Recently a Black-Scholes model with GARCH volatility has been presented [10]. In this article we derive the kurtosis formula for underlying financial time series using BS-Model with GARCH volatility for the case of at the money option. We present the kurtosis formula in terms of the model's parameters. Also we compare our computational results by using another measure of kurtosis for different values of volatilities.

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1. Introduction

Fischer Black and Myron Scholes published an option valuation formula in their 1973's article [3] that today is known as Black-Scholes model. The model has some restrictions for example; a constant risk free interest rate $r$ and a constant volatility $\sigma$ which do not seem to be realistic. Trading option is risky due to the possibly high random components such as volatility. The concept of non constant volatility was introduced by GARCH models [6]. The study of stock price models under the GARCH volatility is a new horizon in derivative investment instruments. Duan was the first to provide a solid theoretical foundation for GARCH option pricing [7]. Recently a Black-Scholes model with GARCH volatility has been introduced [10]. The volatility measures, the variation of price of financial instrument over time and implied volatility can be derived from the market price of a traded derivative. In financial literature researchers use GARCH models frequently in order to forecast the volatility of underlying stock market. The GARCH models have been used to investigate their performance and consistency of the parameters [17]. Taylor series approximations

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have been frequently used in option pricing. In Risk management particularly first and second order Taylor approximations are crucial. Taylor approximations have been also used frequently in Black-Scholes option pricing formula \([5,13]\).

In this article we consider Black-Scholes model with GARCH volatility \([10]\). In Section 2 we provide fundamental theory and tools. In Section 3 we present our results of kurtosis formula for the case of at the money option (ATM). In section 4 we present our computational results and we compare our results with a new measure of kurtosis \([2]\). Section 5 concludes our results.

2. Preliminaries

Let \((\Omega, \mathcal{F}, \mathbb{P})\) be the probability space then price of an asset \(S\) at time \(t\) is a Geometric Brownian Motion (GBM).
\[
dS_t = rS_t dt + \sigma S_t dW_t
\]
where \(\{W_t\}\) is a standard Brownian motion and \(\sigma\) is the volatility. We know that according to Black-Scholes option pricing model \([3]\), A European call option for Black-Scholes model is given by
\[
C_{BS} = S\phi(d_1) - Ke^{-r\tau}\phi(d_2)
\]
where \(d_1 = \frac{\log(S/K) + (r + \sigma^2/2)\tau}{\sigma\sqrt{\tau}}\) and \(d_2 = d_1 - \sigma\sqrt{\tau}\) and \(\phi(.)\) is a cumulative distribution function for standardized Normal random variable and \(\tau = T - t\) and \(S\) is a price of an asset, \(K\) is the strike price, \(r\) is the interest rate and \(T\) denotes the time to expiry.

**Definition 1** \([3]\) If \(S\) is the stock price, \(r\) is risk free interest rate then \(C\) is a European call option that, gives its holder the right, but not the obligation to buy the one unit of underlying asset for a predetermined price \(K\) at the maturity date \(T\).

When variance of the log of stock returns changes with time i.e. \(\sigma = \theta_t\) then a Black-Scholes model with GARCH volatility for a financial time series let say \(y_t\), is given by \([10]\).
\[
dS_t = rS_t dt + \theta_t S_t dW_t
\]
\[
y_t = \log\left(\frac{S_t}{S_{t-1}}\right) - E\left(\log\left(\frac{S_t}{S_{t-1}}\right)\right) = \theta_t Z_t
\]
where \(\{W_t\}\) is a standard Brownian motion and \(\{\theta_t\}\) is a volatility process. The call option for the model is given by:
Theorem 1 [10] For a twice differentiable functions \( f(x) \) and \( g(x) \), the call price (2.3) can be written as:

\[
C_{BSG} = SE_\theta [\phi(d_1)] - Ke^{-rT}E_\theta [\phi(d_2)]
\]

(2.3)

where \( k(y) = \frac{E(y^4)}{E(y^2)} \) is the kurtosis of the observed logreturns \( y \), and

\[
f(E(\theta^2)) = \phi(d_1) = \phi \left( \frac{\log \left( \frac{S}{K} \right) + rT + \frac{1}{2} E(\theta^2)}{\sqrt{E(\theta^2)}} \right)
\]

(2.5)

\[
g(E(\theta^2)) = \phi(d_2) = \phi \left( \frac{\log \left( \frac{S}{K} \right) + rT - \frac{1}{2} E(\theta^2)}{\sqrt{E(\theta^2)}} \right)
\]

(2.6)

where, \( \theta \) is a stationary GARCH process having mean \( \mu_\theta \) and variance \( \sigma^2_\theta \).

Option pricing based on GARCH models have been studied under the assumption that the innovations are standard normal (i.e. under normal GARCH).

Remark 1 The second derivatives of \( f(E(\theta^2)) \) and \( g(E(\theta^2)) \) are given by

\[
f^*(E(\theta^2)) = \frac{1}{\sqrt{2\pi}} \exp \left\{ -\frac{\left(2H + E(\theta^2)\right)^2}{8E(\theta^2)} \right\}.
\]

\[
\left[ \frac{6H - E(\theta^2)}{8E^2(\theta^2)\sqrt{E(\theta^2)}} \right] - \left[ \frac{E^2(\theta^2) - 4H^2}{8E^2(\theta^2)} \right] \left[ \frac{E^2(\theta^2) - 2H}{8E(\theta^2)\sqrt{E(\theta^2)}} \right]
\]
where

\[ H = \log \left( \frac{S}{K} \right) + rT \]

**Definition 2** [6,11] A process of the form

\[ \theta_t^2 = \omega + \sum_{i=1}^{p} \alpha_i \epsilon_{t-i} + \sum_{j=1}^{q} \beta_j \theta_{t-j}^2 \]

is called GARCH (p,q) process where \( \epsilon_t = \theta_t Z_t \) and \( Z_t \sim N(0, \sigma^2) \), \( \omega > 0 \), \( \alpha_i \geq 0, \beta_j \geq 0 \). Similarly a process of the form

\[ \theta_t^2 = \omega + \sum_{i=1}^{p} \alpha_i (\epsilon_{t-i} + \gamma)^2 + \sum_{j=1}^{q} \beta_j \theta_{t-j}^2 \]

is called AGARCH-(I)-(p,q) process and AGARCH-(I)-(0,1) is given by

\[ \theta_t^2 = \omega + \sum_{i=1}^{p} \alpha_i (\epsilon_{t-i} + \gamma)^2 \], where \( \gamma \) is an extra parameter of asymmetry.

### 3. Main Results

An option is called at the money (ATM) if the strike price \( K \) is the same as the current spot price \( S \) of the underlying asset. We know the Black-Scholes call price when volatility is a GARCH process. The following theorem presents the general results for Black-Scholes model with GARCH volatility when option is at the money.

**Proposition 1** For the call option \( C_{BSG} \) we obtain the following relations.

i. \( \frac{\partial d_1}{\partial \theta_1} = 1 - \frac{d_1}{\theta_1}, \frac{\partial d_2}{\partial \theta_1} = -1 - \frac{d_2}{\theta_1} \) and \( d_1 - d_2 = \theta_1 \)

ii. \( \frac{\partial d_1}{\partial S} = \frac{\partial d_2}{\partial S}, \frac{\partial d_1}{\partial K} = \frac{\partial d_2}{\partial K}, \frac{\partial d_1}{\partial r} = \frac{\partial d_2}{\partial r}, \frac{\partial d_1}{\partial T} = \frac{\partial d_2}{\partial T} \) and \( \frac{\partial d_1}{\partial \theta_1} - \frac{\partial d_2}{\partial \theta_1} = 1 \)
iii. $Ke^{-rT} \frac{\partial \phi(d_2)}{\partial d_2} - S \frac{\partial \phi(d_1)}{\partial d_1} = 0$

**Proof** We know that the value of Call option for BS-model with GARCH volatility is given by equation (2.3). The relations i and ii can be easily obtained by following the values of $d_1$ and $d_2$ where the relation iii can be obtained using the value of $K$ and $\frac{\partial \phi(d_1)}{\partial d_1} = \frac{1}{\sqrt{2\pi}} e^{-\frac{d_1^2}{2}}, \frac{\partial \phi(d_2)}{\partial d_2} = \frac{1}{\sqrt{2\pi}} e^{-\frac{d_2^2}{2}}$

$$Ke^{-rT} \frac{\partial \phi(d_2)}{\partial d_2} - S \frac{\partial \phi(d_1)}{\partial d_1} = Se \left\{ \left( d_1 \theta_1 - rT - \frac{\theta_1^2}{2} \right) \right\} e^{-rT}, \frac{1}{\sqrt{2\pi}} e^{-\frac{d_2^2}{2}} - S \frac{1}{\sqrt{2\pi}} e^{-\frac{d_1^2}{2}}$$

$$= \frac{S}{\sqrt{2\pi}} e \left\{ \left( d_1 \theta_1 - rT - \frac{\theta_1^2}{2} \right) \right\} e^{-rT}, \frac{1}{\sqrt{2\pi}} e^{-\frac{(d_1 - \theta_1)^2}{2}} - S \frac{1}{\sqrt{2\pi}} e^{-\frac{d_1^2}{2}} = 0$$

**Theorem 2** Consider the equations (2.5) and (2.6) when stock price $S$ and strike price $K$ are identical i.e. $S = K$ and $r = 0$, then we can write as:

$$f\left( E\left( \theta_1^2 \right) \right) = \phi(d_1) = \phi\left( \frac{\sqrt{E\left( \theta_1^2 \right)}}{2} \right)$$

and

$$g\left( E\left( \theta_1^2 \right) \right) = \phi(d_2) = \phi\left( -\frac{\sqrt{E\left( \theta_1^2 \right)}}{2} \right)$$

Then we have following results:

i. $f^*\left( E\left( \theta_1^2 \right) \right) = \frac{1}{\sqrt{2\pi}} e^{-\frac{E(\theta_1^2)}{8}} \left[ -\frac{1}{8E(\theta_1^2)\sqrt{E(\theta_1^2)}} - \frac{1}{32\sqrt{E(\theta_1^2)}} \right]$  

ii. $g^*\left( E\left( \theta_1^2 \right) \right) = \frac{1}{\sqrt{2\pi}} e^{-\frac{E(\theta_1^2)}{8}} \left[ \frac{1}{8E(\theta_1^2)\sqrt{E(\theta_1^2)}} + \frac{1}{32\sqrt{E(\theta_1^2)}} \right]$
**Proof:** Following the given condition for \( f\left(E\left(\theta_i^2\right)\right) = \frac{\sqrt{E\left(\theta_i^2\right)}}{2} \) and 
\[ g\left(E\left(\theta_i^2\right)\right) = \phi\left(\frac{-\sqrt{E\left(\theta_i^2\right)}}{2}\right) \]
we obtain:

\[ f^*\left(E\left(\theta_i^2\right)\right) = \frac{-1}{8E\left(\theta_i^2\right)\sqrt{E\left(\theta_i^2\right)}} \varphi\left(\frac{\sqrt{E\left(\theta_i^2\right)}}{2}\right) + \frac{1}{16E\left(\theta_i^2\right)} \varphi'\left(\frac{\sqrt{E\left(\theta_i^2\right)}}{2}\right) \]

\[ g^*\left(E\left(\theta_i^2\right)\right) = \frac{1}{8E\left(\theta_i^2\right)\sqrt{E\left(\theta_i^2\right)}} \varphi\left(-\frac{\sqrt{E\left(\theta_i^2\right)}}{2}\right) + \frac{1}{16E\left(\theta_i^2\right)} \varphi'\left(-\frac{\sqrt{E\left(\theta_i^2\right)}}{2}\right) \]

where \( \varphi\left(E\left(\theta_i^2\right)\right) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}} \), therefore

\[ \varphi'\left(E\left(\theta_i^2\right)\right) = -E\left(\theta_i^2\right)\varphi\left(E\left(\theta_i^2\right)\right) \]

and thus we obtain the required form of the \( f^*\left(E\left(\theta_i^2\right)\right) \) and \( g^*\left(E\left(\theta_i^2\right)\right) \).

**Theorem 3** Consider the call option of BS-model with GARCH volatility

\[ C_{BSG} = S \left( f\left(E\left(\theta_i^2\right)\right) + \frac{1}{2} f^*\left(E\left(\theta_i^2\right)\right) \left( \frac{1}{3} k^{(y)} - 1 \right) E^2\left(\theta_i^2\right) \right) - Ke^{-rT} \]

\[ g\left(E\left(\theta_i^2\right)\right) + \frac{1}{2} g^*\left(E\left(\theta_i^2\right)\right) \left( \frac{1}{3} k^{(y)} - 1 \right) E^2\left(\theta_i^2\right) \]

Then for the case of ATM the value of the kurtosis is in terms of the models parameters is given by

\[
    k^{(y)} = 768 \left( \frac{2\pi}{\sqrt{E\left(\theta_i^2\right)}} \cdot \frac{C_{BSG}}{S} - 1 \right) \left( \frac{1}{E\left(\theta_i^2\right) + 4} \right) \left( \frac{1}{E\left(\theta_i^2\right) - 8} \right) + 3
\]

where \( C_{BSG} \) is the value of call option for BS-model with GARCH volatility, \( S \) is the stock price and \( \theta_i \) is the GARCH volatility.
Proof Using Theorem 1 for the case of ATM option and interest rate \( r = 0 \) we obtain

\[
C_{BSG} = S \left[ f \left( E \left( \theta^2_1 \right) \right) - g \left( E \left( \theta^2_1 \right) \right) + \frac{k_1}{2} \left( f'' \left( E \left( \theta^2_1 \right) \right) - g'' \left( E \left( \theta^2_1 \right) \right) \right) E^2 \left( \theta^2_1 \right) \right]
\]  

(3.1)

In the above equation (3.1) \( k_1 = \frac{k^{(y)}}{3} - 1 \). Using the property, \( d_1 - d_2 = E \left( \theta^2_1 \right) \)

and expansion \( \phi \left( E \left( \theta^2_1 \right) \right) \approx \frac{1}{2} + \frac{1}{\sqrt{2\pi}} \left[ E \left( \theta^2_1 \right) - \frac{E^3 \left( \theta^2_1 \right)}{6} + \frac{E^5 \left( \theta^2_1 \right)}{40} - \ldots \right] \), we find

\[
f \left( E \left( \theta^2_1 \right) \right) - g \left( E \left( \theta^2_1 \right) \right) = \left[ \frac{1}{2} + \frac{d_1}{\sqrt{2\pi}} \right] - \left[ \frac{1}{2} + \frac{d_2}{\sqrt{2\pi}} \right]
\]

\[
\approx \frac{1}{\sqrt{2\pi}} \left( d_1 - d_2 \right) = \frac{1}{\sqrt{2\pi}} \sqrt{E \left( \theta^2_1 \right)}
\]  

(3.2)

Now we find \( f'' \left( E \left( \theta^2_1 \right) \right) \) and \( g'' \left( E \left( \theta^2_1 \right) \right) \). Let us suppose \( H = \left( \log \left( \frac{S}{K} \right) + rT \right) \), then we can write as :

\[
f'' \left( E \left( \theta^2_1 \right) \right) = \frac{1}{\sqrt{2\pi}} \left[ e^{-E \left( \theta^2_1 \right)} \right] \left[ \frac{-1}{8E \left( \theta^2_1 \right) \sqrt{E \left( \theta^2_1 \right)} - \frac{1}{32E \left( \theta^2_1 \right)} \sqrt{E \left( \theta^2_1 \right)} \right]
\]

Using the expansion \( e^{-E \left( \theta^2_1 \right) / 8} = 1 - \frac{E \left( \theta^2_1 \right) / 8} + \ldots \) we obtain

\[
f'' \left( E \left( \theta^2_1 \right) \right) \approx \frac{1}{\sqrt{2\pi}} \left[ 1 - \frac{E \left( \theta^2_1 \right)}{8} \right] \left[ \frac{-4 - E \left( \theta^2_1 \right)}{32E \left( \theta^2_1 \right) \sqrt{E \left( \theta^2_1 \right)} \sqrt{E \left( \theta^2_1 \right)} \right] = \frac{1}{\sqrt{2\pi}} \left[ \frac{E \left( \theta^2_1 \right) - 8 \left( E \left( \theta^2_1 \right)^2 + 4 \right)}{256E \left( \theta^2_1 \right) \sqrt{E \left( \theta^2_1 \right)} \sqrt{E \left( \theta^2_1 \right)} \right]
\]

i.e.

\[
f'' \left( E \left( \theta^2_1 \right) \right) \approx \frac{1}{\sqrt{2\pi}} \left[ \frac{\left( E \left( \theta^2_1 \right)^2 - 4 \left( E \left( \theta^2_1 \right)^2 - 32 \right)}{256E \left( \theta^2_1 \right) \sqrt{E \left( \theta^2_1 \right)} \sqrt{E \left( \theta^2_1 \right)} \right]
\]

Similarly we can find
\[
\begin{align*}
g^*(E(\theta_t^2)) & \equiv \frac{1}{\sqrt{2\pi}} \left( 1 - \frac{E(\theta_t^2)}{8} \right) \left[ \frac{4 + E(\theta_t^2)}{32E(\theta_t^2) \sqrt{E(\theta_t^2)}} \right] \\
i.e. \quad g^*(E(\theta_t^2)) & \equiv -\frac{1}{\sqrt{2\pi}} \left[ \frac{(E(\theta_t^2))^2 - 4E(\theta_t^2) - 32}{256E(\theta_t^2) \sqrt{E(\theta_t^2)}} \right] \\
\text{Now we can obtain} \\
f^*(E(\theta_t^2)) - g^*(E(\theta_t^2)) & = \frac{1}{\sqrt{2\pi}} \left[ \left( \frac{(E(\theta_t^2))^2 - 4E(\theta_t^2) - 32}{256E(\theta_t^2) \sqrt{E(\theta_t^2)}} \right) + \left( \frac{(E(\theta_t^2))^2 - 4E(\theta_t^2) - 32}{256E(\theta_t^2) \sqrt{E(\theta_t^2)}} \right) \right] \\
& = \frac{1}{\sqrt{2\pi}} \left[ \left( \frac{(E(\theta_t^2))^2 - 4E(\theta_t^2) - 32}{128E(\theta_t^2) \sqrt{E(\theta_t^2)}} \right) \right] \\
\text{Now we put the values of } f\left(E(\theta_t^2)\right) - g\left(E(\theta_t^2)\right) \text{ and } f^*\left(E(\theta_t^2)\right) - g^*\left(E(\theta_t^2)\right) \text{ in equation (3.1) and we obtain:} \\
C_{BSG} & = S \left[ \frac{1}{\sqrt{2\pi}} \sqrt{E(\theta_t^2)} + \frac{k_1}{2} \left( \frac{1}{\sqrt{2\pi}} \left( \frac{(E(\theta_t^2))^2 - 4E(\theta_t^2) - 32}{128E(\theta_t^2) \sqrt{E(\theta_t^2)}} \right) \right) \right] E^2(\theta_t^2) \\
& = S \frac{256E(\theta_t^2) + k_1 \left( \frac{(E(\theta_t^2))^3 - 4E(\theta_t^2)^2 - 32E(\theta_t^2)}{256 \sqrt{E(\theta_t^2)}} \right)}{256 \sqrt{E(\theta_t^2)}} \\
Sk_1 \left( E(\theta_t^2)^3 - 4(E(\theta_t^2))^2 - 32E(\theta_t^2) \right) + 256SE(\theta_t^2) - 256\sqrt{2\pi}C_{BSG} \sqrt{E(\theta_t^2)} & = 0
\end{align*}
\]
which is equivalent to

\[
k_1 = \frac{256\sqrt{2\pi} C_{BSG} \sqrt{E(\theta_t^2)} - 256SE(\theta_t^2)}{S\left(\left(E(\theta_t^2)\right)^3 - 4\left(E(\theta_t^2)\right)^2 - 32E(\theta_t^2)\right)}
\]

\[
k^{(y)} = \frac{256}{3} \left(\sqrt{\frac{2\pi}{E(\theta_t^2)\left(C_{BSG}-1\right)}}\right) - 1
\]

\[
= \frac{768}{3} \left(\sqrt{\frac{2\pi}{E(\theta_t^2)\left(C_{BSG}-1\right)}}\right) - 1 + 3
\]

and we have obtained finally our result.

**Proposition 2** If \( \theta_t \) is a stationary GARCH process having mean \( \mu_\theta \) and variance \( \sigma_\theta^2 \), then we

i. For the case of GARCH(1,1), \( E(\theta_t^2) = \frac{\omega}{1 - (\alpha_1 \sigma_z^2 + \beta_1)} \)

ii. For the case of AGARCH(I)-(0,1), \( E(\theta_t^2) = \frac{\omega}{1 - \alpha_1 \sigma_z^2} \)

**Proof** We know that for GARCH(1,1) process

\[
\theta_t^2 = \omega + \alpha_1 \varepsilon_{t-1}^2 + \beta_1 \theta_{t-1}^2 \\
E(\theta_t^2) = \omega + \alpha_1 E(\theta_{t-1}^2) + \beta_1 E(\theta_{t-1}^2)
\]

Using the second order stationarity condition we obtain the required result where for the case of AGARCH(I)-(0,1) we can follow the similar procedure.
4. Computational Results

We consider different values of Call option \( C_{BS_G} \) and different stock prices \( S \) for BS-model with GARCH volatility proposed in Gong et al. (2010). We consider the kurtosis formula for different values of volatility. The kurtosis measure has three types in the literature, which are defined in terms of the standardized variable, quintile measures and order statistics. The classical definition of kurtosis belongs to the first type. Since the introduction of kurtosis [14], various new measures of kurtosis have been introduced see [2]. In 2003 Blest introduced a new measure of kurtosis that makes an explicit adjustment for the skewness of the distribution.

\[
k_1^* = k_1 - 3 \left( 1 + f^* \right)^2 - 1
\]  
(4.1)

where

\[
f^* = \left( \sqrt{0.25\gamma_1^2 + 1 + 0.5\gamma_1} \right)^{1/3} - \left( \sqrt{0.25\gamma_1^2 + 1 - 0.5\gamma_1} \right)^{1/3}
\]  
(4.2)

and \( \gamma_1 \) is the standardized third moment given by \( \gamma_1 = \frac{\mu_3}{\sigma^3} \), where \( \mu \) is the mean and \( \sigma \) denotes standard deviation. Aggregate stock market returns display negative skewness, A large body of literature has aimed to explain this stylized fact about the distribution of aggregate stock returns see [4, 8, 9, 10]. Firm stock returns display positive skewness [1]. We consider the new measure of kurtosis \( k_1^* \) as given above in equation (4.1) in order to compare our results of kurtosis \( k_1 \). We consider \( \gamma_1 = 0.5 \) and 1 for calculating \( k_1^* \).

<table>
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<tr>
<th>Volatility</th>
<th>Call Price</th>
<th>Kurtosis ( k_1 )</th>
<th>Stock Price</th>
<th>Kurtosis ( k_1^* )</th>
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<td></td>
<td>Classical Stock Price</td>
<td>( \gamma_1 = 0.5 )</td>
<td>( \gamma_1 = 1 )</td>
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From our computational results we can see that we obtain different value of \( k_1^* \) when \( \gamma_1 \) have different values for \( k_1 \) with the same value. In addition when call price is high, then value of \( k_1 \) is less. We can see when volatility is increased
then corresponding values of $k_1$ are also increased and $k_1^*<k_1$ always because $k_1^*$ is adjusted with skewness $\gamma_1$.

5. Conclusions

In this article, some extensions of the Black-Scholes model with GARCH volatility have been derived. We have used Taylor approximation to obtain the kurtosis formula. The proposed kurtosis formula contains the parameters of BS-models with GARCH volatility and particularly in the case of GARCH volatility we can use various extensions of GARCH models. The calibration results show that kurtosis value for the underlying stock process is greater than 3 in the case of at the money option, which strongly affirms the non Gaussian, return distribution. A comparison with another new measure of kurtosis with adjusted skewness is also given.

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