# An Empirical Law of the Stock Option Market 

By U. Abel and G. Boing


#### Abstract

The investigation of the laws governing the prices of stock options is a problem of theoretical and practical interest. The paper empirically studies the relationship between the market prices of options out of the money and the difference between stock and exercise prices, the former being fixed. We use linear regression with subsequent analysis of the residuals. The results are compared with those obtained by the BlackScholes model. Several applications of the findings are suggested.


## I. Introduction

Option prices depend on several parameters, such as stock price, striking (exercise) price, expiration time, dividends paid on stock before the expiration of the option, interest rates etc. Many attempts have been made to theoretically derive option or warrant prices under more or less restrictive model assumptions (e.g. References (1), (3), (4), (5), (7), (12), (14)).

The most famous valuation formula is undoubtedly the one proposed by Black and Scholes in 1973 for non dividend-paying stocks (3). It has prompted an extensive discussion of the model assumptions and empirical studies of the model fit. Today the formula (perhaps in an extended form allowing for dividends (16)) appears to be widely accepted.

Its greatest shortcoming is that it assumes a constant variance rate ("volatility") $v^{2}$ which is an explicit model parameter. In reality the volatility of a stock can hardly be regarded as constant over the time to maturity of the option, and, in any way, determining the variance rate poses a practical problem. Clearly, historical estimates of $v$ (2), (8), (9) can be unreliable and dangerous if money is at stake. Implied estimates are more satisfactory as they are derived from the present market. Their rationale is as follows:

Let $k$ options of a stock be in the market priced at $O_{i}, i=1, \ldots, k$. Putting

$$
O_{i}=\hat{O}_{i}(v),
$$

where $\hat{O}_{i}(v)$ are the prices predicted by the Black/Scholes formula for unspecified $v$, estimated values $v_{1}, \ldots, v_{k}$ are obtained which can be weighted and averaged to yield a weighted implied variance rate of the stock (6), (11), (13), (15).

Various weghting schemes have been used in the literature, all more or less arbitrary or chosen on empirical grounds, so that, in principle, the application of the Black/Scholes model shares some features with empirical valuation formulas such as the one given by Kassouf (10).

## II. An empirical law of the option market

We focus on the special problem of the relationship between the prices of options which are out of the money and the difference between the stock and exercise prices, the former being fixed. We contend that this relationships is approximately loglinear.

Fig. $1 \mathrm{a} / \mathrm{b}$ shows that the hypothesis of loglinearity is not farfetched.


Fig. 1a

The empirical study was based on option prices as published in the Wall Street Journal. Only informative prices were taken into consideration, that is

1. Option prices of $1 / 16$ were excluded because at this level (the lowest possible) numerous anomalies arise. E.g., on Sept. 21st, the October \$ 15, 20 and 25 puts of Homestake were all priced at $1 / 16$.


Fig. 1b
2. Stocks with less than three prices above $1 / 16$ of options out of the money were excluded since they carried no information as to our hypothesis.
We were slightly more restrictive in that three option prices $>1 / 16$ for successive exercise prices had to be available in order to qualify the stock for the analysis.
For puts and calls different days had to be chosen for we failed to find one single day where the criteria of selection were met by a satisfactorily large sample of both calls and puts. The data bases for the analysis were the following:

$$
\left.\begin{array}{ll}
\text { Calls } & \text { expiration in October } 1982 \\
& \text { prices of June 23rd, 1982 } \\
& 20 \text { eligible stocks }
\end{array}\right\} \begin{array}{ll}
\text { Puts } & \text { expiration in January 1983 } \\
& \begin{array}{l}
\text { prices of September 22nd, } 1982 \\
\\
22 \text { eligible stocks. }
\end{array}
\end{array}
$$

For each selected stock a simple linear regression of

$$
\log (O / S) \text { vs } \quad \Delta=100 \cdot|S-E| / S
$$

was performed ( $O, E, S$ denoting the option, exercise and stock prices respectively). Of course, $\log (O / S)$ and $\Delta$ are linearly related if and only if $\log (O)$ and $E-S$ are. The standardization was introduced in the hope that the result would prove independent of $S$ and possibly even of the stock.

Unfortunately, no general test for linearity of a regression exists unless there are more degrees of freedoms than abscissa values, and this is not the case, here. It is, however, reasonable to assume that any alternative to linearity is either concavity or convexity. In both cases a systematic effect must show in the successive differences $d_{i}=r_{i+1}-r_{i}$ of the residuals belonging to increasing abscissa values for each stock. In case of convexity the $d_{i}, i=1,2, \ldots$ should increase, in case of concavity they should decrease.

Tables $1 \mathrm{a} / \mathrm{b}$ and $2 \mathrm{a} / \mathrm{b}$ show the results. The $d_{i}$ are very small compared with the change in the $\log (O / S)$-values as predicted by the regression. This indicates that the linear model fits well. In two cases the $\log (O / S)$ lie even on straight lines.

The Hodges-Lehmann estimators of the median differences of the $d_{i}$, $i=1,2,3$ were

$$
-0.001 \text { and }-0.11 \text { for calls }
$$

$$
\text { and } \quad 0.009 \text { and } 0.19 \text { for puts . }
$$

While there is no perceptible monotonic trend in the $d_{i}$ of calls, such a trend, though very slight, can be ascertained in the samples of puts (Jonckheere test against ordered alternatives, $p<0.01$ ). However, there were only 10 out of 22 stocks with a strictly montonic increase. On the other hand, there were 4 stocks with a strictly monotonic decrease and 2 stocks with equality of the $d_{i}$. Summarizing, deviations from linearity, if any, were small and not systematic in the majority of the stocks.

Table $1 \mathrm{a} / \mathrm{b}$ shows that the intercepts of the regression lines were densely packed, while the differences in the slopes were rather marked. Since calls and puts had about the same time to maturiy, it makes sense to compare their regression parameters, especially for those stocks appearing in both analyses. Slopes and intercepts were smaller in puts than in calls (the arithmetic means were -2.39 and -6.99 in puts versus -2.27 and -6.24 in calls). The parameters of puts of a stock were strikingly similar to those of the calls of the same stock, and, as far as were differences, there seems to be no rule for their sign.

We have seen that the regression lines for different stocks were not equal. It might still, however, be true that, for a given stock, they do not depend on the stock price $S$. An empirical check of this hypothesis is difficult and must rely on few data because large changes of stock prices require some time to

Table 1 a

| STOCK | INTERC. | SLOPE |
| :--- | :--- | :--- |
| FEDEXP | -1.944107 | -7.94420 |
| FLUOR | -2.531504 | -2.95640 |
| HALBTN | -2.366208 | -4.91840 |
| HOMESTK | -2.027928 | -2.89740 |
| MERCK | -2.868246 | -8.36761 |
| MONSAN | -2.497658 | -10.60607 |
| PENNZ | -2.221898 | -3.03217 |
| STORTEC | -2.240172 | -3.81818 |
| TELDYN | -2.052711 | -8.53185 |
| AMEXP | -2.368962 | -9.64208 |
| DIGEQ | -2.251492 | -9.15036 |
| HUTTON | -1.815444 | -3.78057 |
| LILLY | -2.540796 | -11.78133 |
| MERRIL | -2.077980 | -5.36468 |
| COMSAT | -2.177308 | -9.29057 |
| OAK | -2.164429 | -2.73604 |
| WESTUN | -2.208561 | -6.73506 |
| BALDUN | -2.388497 | -4.81473 |
| DSHAM | -2.546658 | -3.36863 |
| WANGB | -2.023090 | -5.05841 |

Table 1 b

| STOCK | INTERC. | SLOPE |
| :--- | :--- | :--- |
| BURLN | -2.493704 | -6.51496 |
| EASTKD | -2.626853 | -12.17279 |
| FEDEXP | -2.203536 | -7.34584 |
| HOMESTK | -2.031629 | -4.37037 |
| IBM | -2.900605 | -6.53806 |
| JOHNJ | -2.761150 | -9.10545 |
| MMM | -2.575750 | -8.86155 |
| MONSAN | -2.894036 | -9.44603 |
| PEPSI | -3.237197 | -8.24209 |
| TELDYN | -2.019670 | -6.77224 |
| TEXIN | -2.251436 | -5.65140 |
| AMEXP | -1.981937 | -10.68347 |
| DIGEQ | -2.248021 | -8.94600 |
| DUPONT | -2.821536 | -8.84224 |
| HUTTON | -1.941320 | -4.62676 |
| MERRIL | -1.840691 | -5.77109 |
| MOTORLA | -2.365029 | -9.87831 |
| PROCG | -2.953498 | -8.93818 |
| WESTNG | -2.585875 | -7.68488 |
| WESTUN | -2.081289 | -10.57049 |
| BALDUN | -1.844625 | -6.32225 |
| WANGB | -2.111259 | -6.38418 |

Table 2 a
Successive Differences of Residuals

| NO | D 1 | D 2 | D 3 | D 4 |
| :--- | ---: | :---: | :---: | :---: |
| FEDEXP | -.235175 | -.006917 | .244398 | - |
| FLUOR | .202733 | -.202733 | - | - |
| HALBTN | .020637 | .017681 | .002938 | - |
| HOMESTK | .160299 | .077412 | .263515 | - |
| MERCK | .026622 | .026622 | - | - |
| MONSAN | .095310 | .095310 | - | - |
| PENNZ | .071381 | .053032 | -.120991 | .0305587 |
| STORTEC | .127967 | .127967 | - | - |
| TELDYN | -.062600 | .065234 | -.024378 | - |
| AMEXP | -.052680 | .052680 | - | - |
| DIGEQ | -.113763 | .048756 | .048756 | - |
| HUTTON | .220916 | .220916 | - | - |
| LILLY | .309520 | .309520 | - | - |
| MERRIL | .314304 | -.314304 | - | - |
| COMSAT | .038481 | -.038481 | - | - |
| OAK | .111572 | -.111572 | - | - |
| WESTUN | -.235002 | .235002 | - | - |
| BALDUN | .009475 | .343481 | -.467449 | - |
| DSHAM | .412088 | -.412088 | - | - |
| WANGB | -235002 | .235002 | - | - |

Table $2 b$
Successive Differences of Residuals

| NO | D1 | D2 | D 3 | D4 | D5 |
| :--- | ---: | :---: | :---: | :---: | :---: |
| BURLN | -.018870 | .018870 | - | - | - |
| EASTKD | -.105532 | .042684 | .048619 | - | - |
| FEDEXP | -.078327 | .078327 | - | - | - |
| HOMESTK | -.022607 | -.059962 | .102557 | - | - |
| IBM | -.155301 | .020430 | .128061 | - | - |
| JOHNJ | -.111572 | .111572 | - | - | - |
| MMM | -.008687 | .072049 | -.087379 | - | - |
| MONSAN | .426039 | -.380436 | .377249 | -.421258 | - |
| PEPSI | -.020411 | .020411 | - | - | - |
| TELDYN | -.142430 | -.291465 | .253466 | .194043 | -.157934 |
| TEXIN | -.221645 | -.012083 | -.237756 | - | - |
| AMEXP | -.011236 | -.011236 | - | - | - |
| DIGEQ | .033605 | -.257272 | .161438 | .110145 | - |
| DUPONT | -.053815 | .053815 | - | - | - |
| HUTTON | -.000000 | .000000 | - | - | - |
| MERRIL | -.031143 | .006598 | .022346 | - | - |
| MOTORLA | -.020469 | -.168685 | -.184189 | .549780 | - |
| PROCG | .159227 | -.159227 | - | - | - |
| WESTNG | .235002 | .235002 | - | - | - |
| WESTUN | .000000 | .000000 | - | - | - |
| BALDUN | -.154809 | .154809 | - | - | - |
| WANGB | .018184 | -.018184 | - | - | - |

occur so that the change in time to maturity exerts influence on the option prices. The examples given in Table 3 show that the parameters of the linear regression remain constant even after pronounced short-term changes of the stock price. This stability is remarkable in view of the extreme sensitivity of the parameters to small variations of the option prices.

Table 3: Changes of the parameters of the linear regression of $\log (\boldsymbol{O} / \boldsymbol{S})$ vs. $|\boldsymbol{E}-\boldsymbol{S}|$ after sharp movements of stock prices.

|  | Motorola maturity Apr. 1984 |  | St. Oil Ohio maturity March 1984 |  | Coleco maturity Apr. 1984 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| date ${ }^{(a)}$ | Dec. 2nd | Dec. 14th | Dec. 2nd | Dec. 14th | Dec. 16th | Dec. 23rd |
| stock price | $1421 / 4$ | 1353/4 | $453 / 4$ | 413/4 | 253/8 | 205/8 |
| calls: intercept slope | (b) | $\begin{aligned} & -2.56 \\ & -0.055 \end{aligned}$ | $\begin{aligned} & -2.61 \\ & -0.154 \end{aligned}$ | $\begin{aligned} & -2.86 \\ & -0.165 \end{aligned}$ | $\begin{aligned} & -1.86 \\ & -0.115 \end{aligned}$ | $\begin{aligned} & -1.88 \\ & -0.1 \end{aligned}$ |
| puts: <br> intercept slope | $\begin{aligned} & -2.88 \\ & -0.065 \end{aligned}$ | $\begin{aligned} & -2.74 \\ & -0.065 \end{aligned}$ | $\begin{aligned} & -2.94 \\ & -0.254 \end{aligned}$ | (b) | $\begin{aligned} & -1.92 \\ & -0.159 \end{aligned}$ | $\begin{aligned} & -1.64 \\ & -0.16 \end{aligned}$ |

(a) Year 1983. - (b) No calculation possible.

Without presenting a detailed analysis we finally note that for options in the money a loglinear relation between the premia and $\Delta$ holds, too, though outliers are more frequent.

## III. Discussion

The empirical investigation has shown that for fixed stock prices $S$ a loglinear relationship between the prices $O$ of options out of the money and the difference between $S$ and the exercise price $E$ very approximately holds. Observe that this is not a consequence of the interval $E-S$ being small (so that it necessarily follows by Taylor expansion approximation). For a stock priced at about $\$ 100$, a $\$ 40$-interval (such as in Fig. 1) is not small as any investor knows, and, moreover, linearity between $S-E$ and some other function of $O$ (say $O$ or $O^{2}$ ) definitely does not hold.

The relationship put forward in this article is not incompatible with Black/Scholes's formula as Fig. 2 shows. However, we have been unable, so far, to deduce approximate loglinearity from this formula.

Our finding gives a strong support to a special estimator for implied volatility of a stock, viz. the value $v$ which minimizes

$$
\sum_{i}\left(\log \left(O_{i}\right)-\log \left(\hat{O}_{i}\right)\right)^{2} .
$$



Fig. 2

Apart from this, it can be applied in several ways.
First, it can be used for predictions of option prices for future stock prices. If changes in stock prices do not occur rapidly, prognosis will require interpolation between the prices predicted for the available times to maturity.

Second, it allows the easy detection of options which are under- or overvalued with respect to other options of the same stock.

Third, it can be exploited for practical investment. We suggest the following (winning?) strategy:

Determine the intercept and slope for the puts and calls of the same stock. Calculate the value $S_{\min }$ such that a given spraddle, i.e. a combination of one put and one call of the stock with

$$
E_{\text {call }}>S>E_{\text {put }},
$$

assumes its minimum ${ }^{1}$. Invest when $S=S_{\min }$. Take profit on any movement of the stock price.

[^0]
## Summary

It is shown that the relationship between the prices of options out of the money and the difference between stock and exercise prices, the former being fixed, is approximately loglinear. This finding can be applied in various ways.

## Zusammenfassung

Es wird gezeigt, daß eine approximativ loglineare Beziehung besteht zwischen den Preisen von Optionen, die nicht im Geld sind, und der Differenz zwischen Basispreis und Aktienkurs, sofern letzterer konstant gehalten wird. Dieses Resultat besitzt eine Reihe von Anwendungen.

## References

(1) Baumol, W. J., B. G. Malkiel and R. E. Quandt (1966), The Valuation of Convertible Securities. Quarterly Journal of Economics 80, 48-59.
(2) Black, F. and M. Scholes (1972), The Valuation of Option Contracts and a Test of Market Efficiency. Journal of Finance 27, 399-418.
(3) Black, F. and M. Scholes (1973), The Pricing of Options and Corporate Liabilities. Journal of Political Economy 81, 637-54.
(4) Boness, J. A. (1964), Elements of a Theory of Stock-Option Values. Journal of Political Economy 72, 163-75.
(5) Chen, A. H. Y. (1970), A Model of Warrant Pricing in a Dynamic Market. Journal of Finance 25, 1041-60.
(6) Chiras, D. P. and S. Manaster (1978), The Informational Contents of Option Prices and a Test of Market Efficiency. Journal of Financial Economics 6, 213 34.
(7) Cootner, P. A. (1967), The Random Character of Stock Market Prices. Cambridge, Mass.
(8) Finnerty, J. E. (1978), The Chicago Board Options Exchange and Market Efficieny. Journal of Financial and Quantitative Analysis 13, 29 - 38.
(9) Galai, D. (1977), Tests of Option Market Efficiency of the Chicago Boards Options Exchange. Journal of Business 50, 167-97.
(10) Kassouf, S. I. (1969), Evaluation of Convertible Securities. New York.
(11) Latané, H. and R. J. Rendleman (1976), Standard Deviation of Stock Price Ratios Implied in Option Premia. Journal of Finance 31, 369-82.
(12) Merton, R. C. (1973), Theory of Rational Option Pricing. Bell Journal of Economics and Management Science, 141-83.
(13) Patell, S. M. and M. A. Wolfson (1979), Anticipated Information Releases Reflected in Call Option Prices. Journal of Accounting and Economics 1, 117-40.
(14) Rendleman, R. and B. Bartter (1979), Two-State Option Pricing. Journal of Finance 34, 1093-1110.
(15) Schmalensee, R. and R. R. Trippi (1978), Common Stock Volatility Expectations Implied by Option Premia. Journal of Finance 33, 129-47.
(16) Whaley, R. E. (1982), Valuation of American Call Options on Dividend-Paying Stocks. Journal of Financial Economics 10, 29-58.


[^0]:    ${ }^{1}$ Given the regression parameters, $S_{\min }$ can be determined analytically by standard calculus.

