

# **"No-Good-Deal" Bounds**

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# A Short History of Bounds

Finding bounds on the values of derivatives is an old "art form":

- Merton (1973),
  - no arbitrage bounds,
- Perrakis and Ryan (1984), Levy (1985), Ritchken and Kuo (1989), Basso and Pianca (1994),
  - bounds based on stochastic dominance (or similar).

Interest in this topic has intensified, with more interest in:

- Levy processes, and other work related to
- incomplete markets

# "No-Good-Deal" Bounds

"No-Good-Deal" Bounds were:

- introduced by Cochrane and Saá-Requejo in 1996,
- modified by Hodges (Generalized Sharpe Ratio) in 1997,
- generalized to a more abstract setting by Cerny and Hodges, 1998 (presented at Bachalier 2000),
- related to Artzner *et al* "coherent risk measures" by Jaschke and Kuchler (2001) (also anticipated in Mejía-Pérez, 1998, and Hodges 1998).

We examine these four themes in more detail

# Cochrane and Saá-Requejo

Pricing bounds are constructed relative to a Sharpe Ratio (expected excess return / standard deviation).

Two cases are provided:

1. Unconstrained: the dual pricing vector is linear in wealth (and must go negative somewhere, unless it is a constant)
2. Constrained: the dual pricing vector is piece-wise linear in wealth, and is set equal to zero where it would otherwise go negative.

# The Analysis

The first case is pure mean-variance analysis.  
[See Cochrane (2001) for a clear exposition].  
The formulation is:

$$\underline{C} = \min_m E(mx^c) \text{ s.t. } \mathbf{p} = E(m\mathbf{x}), E(m^2) \leq A^2.$$

The payoff  $x^c$  is decomposed into its projection in the space of traded assets (the approximate hedge) and an orthogonal residual,  $w$ .

$$x^c = \hat{x}^c + w, \text{ where } \hat{x}^c = E(x^c \mathbf{x}') E(\mathbf{x}\mathbf{x}')^{-1} \mathbf{x}.$$

We can get further insights using the Treynor-Black (1973) analysis:

## Treynor-Black (1973) analysis:

The square of the Sharpe Ratio is the sum of the squares of the Sharpe Ratios of each separate orthogonal bet.

If we let  $h_0$  denote the Sharpe Ratio attainable from the basis assets, then in the notation of the paper it follows immediately that

$$SR^2 = h_0^2 + \frac{FV^2 (\underline{c} - E[m\hat{x}^c])^2}{\sigma^2(w)} = h^2,$$

which enables us to solve for the bounds as:

$$\frac{(\underline{c} - E[m\hat{x}^c])^2}{FV^2} = \frac{\sigma^2(w)}{FV^2} (h^2 - h_0^2) = \frac{\sigma^2(w)}{FV^2} (A^2 - E[x^{*2}]) \text{ as in Proposition 4.}$$

## Extensions

Optimization subject to the pricing vector  $m$  being non-negative is similar but slightly more complicated.

Essentially, it now becomes necessary to search numerically for the shadow prices of the two constraints.

In a multiperiod context, these bounds can be calculated recursively, (but the numerical implementation is non-trivial).

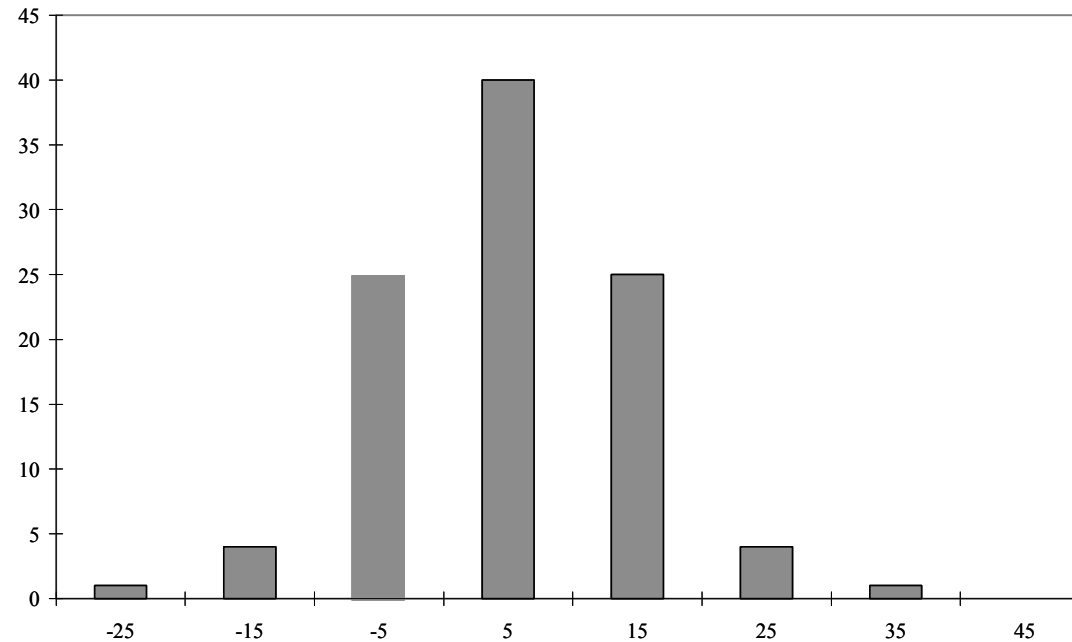
Note that, although the solution for  $m > 0$  is general, the criterion of maximizing the Sharpe Ratio was arbitrary.

# Generalized Sharpe Ratio

- What's wrong with the Sharpe Ratio
- The Generalisation (GSR) and some of its properties
- Applications to:
  - Valuation bounds in Incomplete Markets
  - Value at Risk
  - Performance Measurement



# A Sharpe Ratio Paradox



$$\begin{aligned} \mu &= 5.00 \rightarrow 5.10 \\ \sigma &= 10.00 \rightarrow 10.34 \\ \text{S.R.} = \mu / \sigma &= 0.50 \rightarrow 0.493 \\ \text{GSR} &= 0.498 \rightarrow 0.500 \end{aligned}$$

# Generalized Sharpe Ratio

We propose a new measure where an investor with CARA utility can choose the quantity of the prospect to hold:

- we obtain the usual value for Normal distributions
- for non-Normal distributions, we provide a generalization based on equating expected utility.

For normal distributions we find

$$U^* = \underset{x}{\text{Maximise}} E[U] = -e^{-\frac{1}{2}\frac{\mu^2}{\sigma^2}T}$$

A Generalization of the Sharpe Ratio  $\mu/\sigma$  is obtained as

$$GSR = \sqrt{\frac{-2}{T} \ln(-U^*)}.$$

# Computation

$$\text{Max } E[U] = \sum p_s \exp(-y r_s).$$

First order Condition:

$$\sum p_s r_s \exp(-y r_s) = 0 = f(y).$$

Solve using Newton - Raphson iteration for y with

$$f'(y) = - \sum p_s r_s^2 \exp(-y r_s).$$

We can do this on a spreadsheet.

# Valuation and Hedging

Even where exact replication of derivatives is impossible, the price of a contingent claim may be “cheap” or “dear”.

We solve the choice problem for an investor who maximizes

$$E[U(w)] \text{ with } U = -e^{-\lambda w} .$$

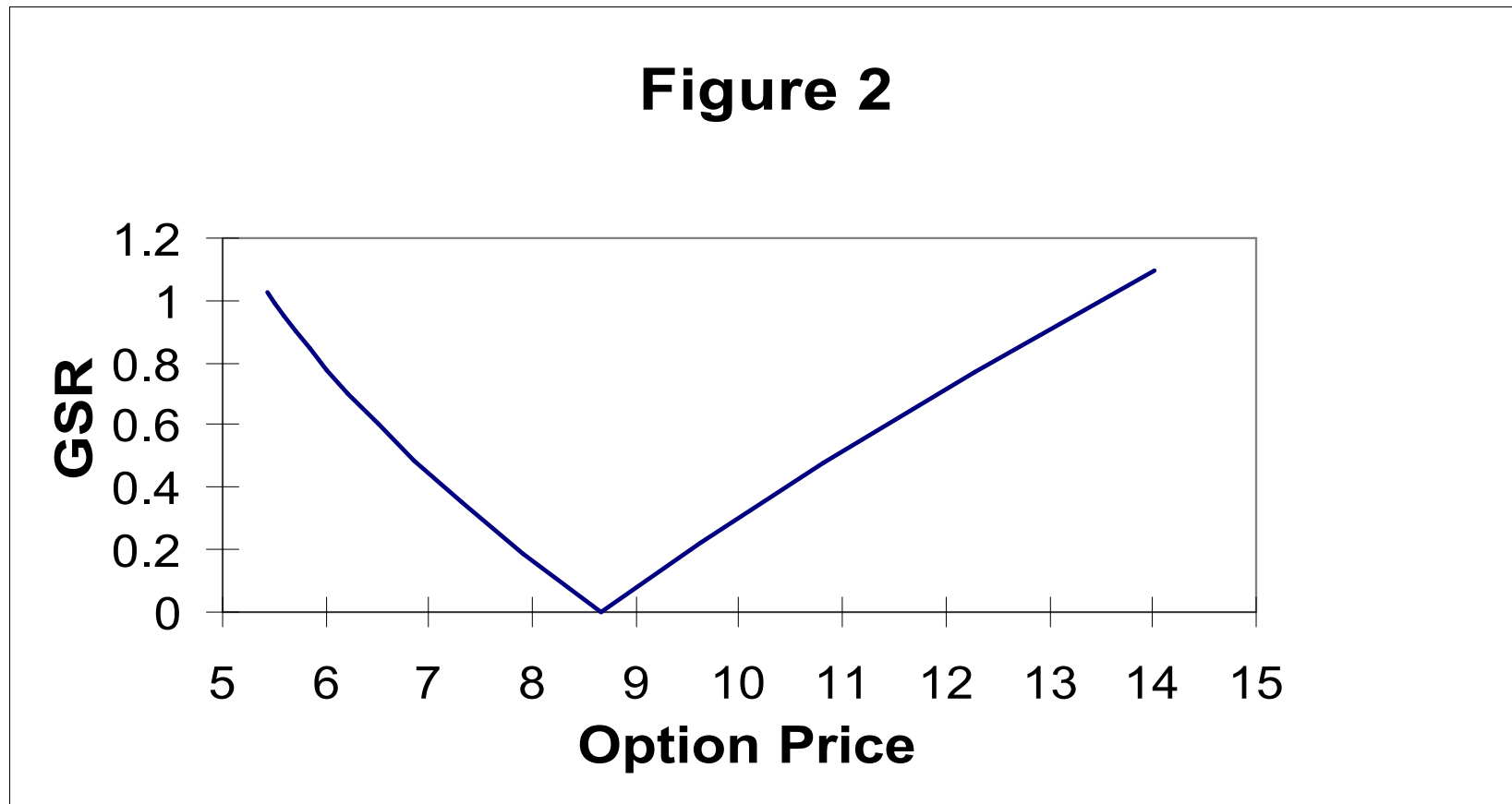
The investor buys  $y$  units of the contingent claim, and hedges with  $x$  units of the underlying:

$$\text{Maximise}_{x,y} E[U] = -E e^{-\lambda \left( \int_0^T x_t dS_t + y(C_T - C_0) \right)}$$

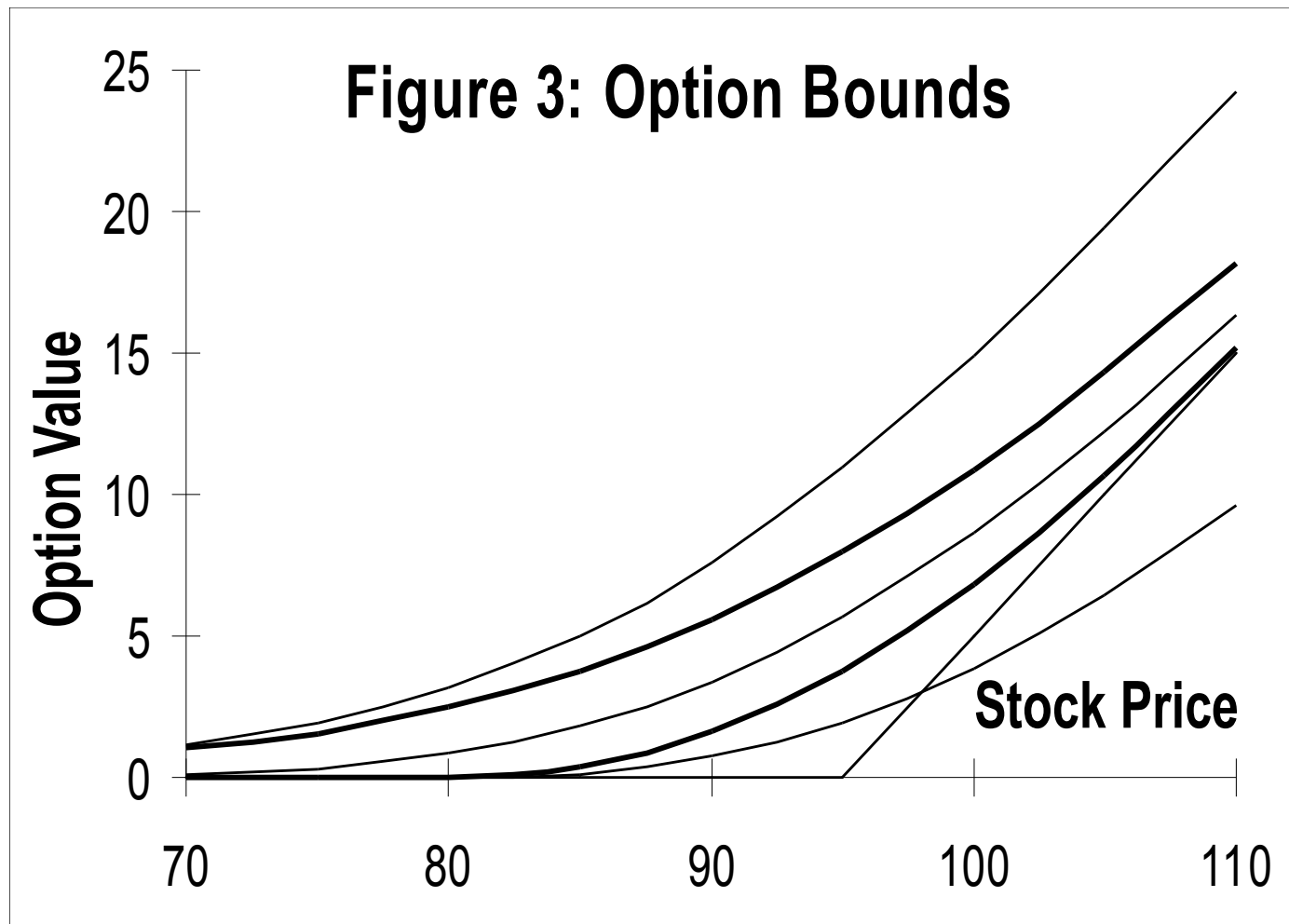
The value of the expected utility provides a GSR measure of the market opportunity provided by any particular  $C_0$ .

## Conditional Bounds

We obtain valuation bounds which are much tighter than could be obtained by riskless arbitrage arguments.



# Bounds at Different Asset Price Levels (GSR = $\frac{1}{2}$ )



## Other Properties

These GSR bounds defined by the class of negative exponential utility functions have a number of advantages and disadvantages:

- The bounds do not explicitly depend
  - on risk aversion, or
  - on wealth levels.
- Losses (negative wealth) is not ruled out
  - as it would be for power or log utility.
- Some claims have very weak (and in some cases infinite) bounds.
  - in particular, any finite certain loss is preferred to a short position in a log-normal distribution, which makes the expected utility infinitely negative .

# Performance Measurement

Under a continuous diffusion process with a constant price of risk  $\mu/\sigma$ , a CARA investor will have constant risk exposure.

The terminal distribution is Normal.

Hence, odd shaped distributions are **not** preferred.

The Generalised Sharpe Ratio is robust in the sense that the maximum *ex ante* GSR **equals** the conventional Sharpe Ratio.



# General Theory of Good-Deal Pricing

Cerny and Hodges (2001) have proposed a more general framework of "no-good-deal" pricing which places

- no-arbitrage, and
- representative agent equilibrium

at the two ends of a spectrum of possibilities.

A **desirable claim** is one which provides a specific level of von Neumann-Morgenstern expected utility.

A **good deal** is a desirable claim with zero or negative price.

## Extension Theorem

In an incomplete market it is often convenient to suppose that the market is augmented in such a way that the resulting complete market contains no arbitrages.

We can more powerfully augment the market so that the complete market contains no arbitrages.

We obtain a set of pricing functionals which form a subset of those which simply preclude arbitrage.

# Pricing Theorem

The link between no arbitrage and strictly positive pricing rules carries over to good deals, and enables price restrictions to be placed on non-marketed claims.

Under suitable technical assumptions (see C&H):

- The no-good-deal price region  $P$  for a set of claims is a convex set,
- Redundant assets have unique good-deal prices

# Coherent Bounds

*GSR* and *G-NGD* bounds satisfy the properties advocated by Artzener et al, 1997 for coherent risk measures (*SR* ones don't):

Linearity:

$$B[\alpha \tilde{C}] = \alpha B[\tilde{C}], \text{ and} \\ B[\beta + \tilde{C}] = \beta + B[\tilde{C}]$$

Subadditivity:

$$LB[\tilde{C}] + LB[\tilde{D}] \leq LB[\tilde{C} + \tilde{D}] \\ UB[\tilde{C} + \tilde{D}] \leq UB[\tilde{C}] + UB[\tilde{D}]$$

Monotonicity

$$\tilde{C} \leq \tilde{D} \Rightarrow B[\tilde{C}] \leq B[\tilde{D}]$$

(where  $B$  denotes any bound,  $LB$  lower bound,  $UB$  upper bound).

# Jaschke and Küchler

There is a one-to-one correspondence between:

1. "coherent risk measures"
2. Cones of "desirable claims"
3. Partial orderings
4. Valuation bounds
5. Sets of "admissible" price systems.

# Tail Areas

The GSR tail area is always strictly less than  $-U^*$ .

This makes it suitable as an alternative coherent substitute for VaR to the "downside" risk measure which has also been suggested.

# Conclusions

The no-good-deal bound framework has been considerably extended from its original Sharpe Ratio definition.

It provides a powerful method for obtaining:

- Valuation bounds in incomplete markets
- Coherent risk measures for Value at Risk

It is computationally attractive, for example:

- Values can be characterized in terms of the attractiveness of different prices (Generalized Sharpe Ratio).
- We can solve under suitable Markov processes or add as a heuristic to Monte Carlo simulations.

# References

- Artzner, P, F Delbaen, J Eber and D Heath, 1997, "Definition of Coherent Measures of Risk", Working Paper, Cornell University, March 1997.
- Cerny, A and S D Hodges, 2000, "The Theory of Good-Deal Pricing in Financial Markets", in Geman, Madan, Pliska, Vorst (eds.): *Selected Proceedings of the First Bachelier Congress*, held in Paris 2000, Springer.
- J H Cochrane and J Saá Requejo, 1996, "Beyond Arbitrage: 'Good-Deal' Asset Price Bounds in Incomplete Markets" Working Paper, February 1996, Graduate School of Business, University of Chicago.
- Cochrane, J H , 2001, *Asset Pricing*, Princeton.
- Hodges, S D, 1998, "A Generalization of the Sharpe Ratio and its Applications to Valuation Bounds and Risk Measures", FORC Preprint 1998/88, University of Warwick.
- Jaschke, S and U Küchler, 2001, "Coherent Risk Measures and Good-Deal Bounds", *Finance and Stochastics*, 5, 181-200.
- Mejía-Pérez, 1998, "Quasi-coherent risk measures and its relation to option pricing bounds in incomplete markets", Working Paper, March 1998, University of Warwick.
- Merton, R C, 1973, "Theory of Rational Option Pricing", *Bell Journal of Economics*, 4, 141-183.
- Sharpe, W F, 1994, "The Sharpe Ratio", *Journal of Portfolio Management*, 21, 49-59.
- Treynor, J Land F Black, 1973, "How to Use Security Analysis to Improve Security Selection", *Journal of Business*, 46, 66-86.