J.P.Morgan/Reuters

RiskMetrics™—Technical Document

- J.P. Morgan and Reuters have teamed up to enhance RiskMetrics™. Morgan will continue to be responsible for enhancing the methods outlined in this document, while Reuters will control the production and distribution of the RiskMetrics™ data sets.
- Expanded sections on methodology outline enhanced analytical solutions for dealing with nonlinear options risks and introduce methods on how to account for non-normal distributions.
- Enclosed diskette contains many examples used in this document. It allows readers to experiment with our risk measurement techniques.
- All publications and daily data sets are available free of charge on J.P. Morgan's Web page on the Internet at **http://www.jpmorgan.com/RiskManagement/RiskMetrics/RiskMetrics.html**. This page is accessible directly or through third party services such as CompuServe®, America Online™, or Prodigy®.

Fourth Edition, 1996

New York
December 17, 1996

Morgan Guaranty Trust Company
Risk Management Advisory
Jacques Longerstaey
(1-212) 648-4936
riskmetrics@jpmorgan.com

Reuters Ltd
International Marketing
Martin Spencer
(44-171) 542-3260
martin.spencer@reuters.com

This *Technical Document* provides a detailed description of RiskMetrics™, a set of techniques and data to measure market risks in portfolios of fixed income instruments, equities, foreign exchange, commodities, and their derivatives issued in over 30 countries. This edition has been expanded significantly from the previous release issued in May 1995.

We make this methodology and the corresponding RiskMetrics™ data sets available for three reasons:

1. We are interested in promoting greater transparency of market risks. Transparency is the key to effective risk management.

2. Our aim has been to establish a benchmark for market risk measurement. The absence of a common point of reference for market risks makes it difficult to compare different approaches to and measures of market risks. Risks are comparable only when they are measured with the same yardstick.

3. We intend to provide our clients with sound advice, including advice on managing their market risks. We describe the RiskMetrics™ methodology as an aid to clients in understanding and evaluating that advice.

Both J.P. Morgan and Reuters are committed to further the development of RiskMetrics™ as a fully transparent set of risk measurement methods. We look forward to continued feedback on how to maintain the quality that has made RiskMetrics™ the benchmark for measuring market risk.

RiskMetrics™ is based on, but differs significantly from, the risk measurement methodology developed by J.P. Morgan for the measurement, management, and control of market risks in its trading, arbitrage, and own investment account activities. **We remind our readers that no amount of sophisticated analytics will replace experience and professional judgment in managing risks**. RiskMetrics™ is nothing more than a high-quality tool for the professional risk manager involved in the financial markets and is not a guarantee of specific results.

RiskMetrics™—Technical Document
Fourth Edition (December 1996)

Copyright © 1996 Morgan Guaranty Trust Company of New York.
All rights reserved.

RiskMetrics™ is a registered trademark of J. P. Morgan in the United States and in other countries. It is written with the symbol ™ at its first occurrence in this publication, and as RiskMetrics thereafter.

This book

This is the reference document for RiskMetrics™. It covers all aspects of RiskMetrics and supersedes all previous editions of the *Technical Document*. It is meant to serve as a reference to the methodology of statistical estimation of market risk, as well as detailed documentation of the analytics that generate the data sets that are published daily on our Internet Web sites.

This document reviews

1. The conceptual framework underlying the methodologies for estimating market risks.
2. The statistics of financial market returns.
3. How to model financial instrument exposures to a variety of market risk factors.
4. The data sets of statistical measures that we estimate and distribute daily over the Internet and shortly, the Reuters Web.

Measurement and management of market risks continues to be as much a craft as it is a science. It has evolved rapidly over the last 15 years and has continued to evolve since we launched RiskMetrics in October 1994. Dozens of professionals at J.P. Morgan have contributed to the development of this market risk management technology and the latest document contains entries or contributions from a significant number of our market risk professionals.

We have received numerous constructive comments and criticisms from professionals at Central Banks and regulatory bodies in many countries, from our competitors at other financial institutions, from a large number specialists in academia and last, but not least, from our clients. Without their feedback, help, and encouragement to pursue our strategy of open disclosure of methodology and free access to data, we would not have been as successful in advancing this technology as much as we have over the last two years.

What is RiskMetrics?

RiskMetrics is a set of tools that enable participants in the financial markets to estimate their exposure to market risk under what has been called the “Value-at-Risk framework”. RiskMetrics has three basic components:

- A set of market risk measurement methodologies outlined in this document.
- Data sets of volatility and correlation data used in the computation of market risk.
- Software systems developed by J.P.Morgan, subsidiaries of Reuters, and third party vendors that implement the methodologies described herein.

With the help of this document and the associated line of products, users should be in a position to estimate market risks in portfolios of foreign exchange, fixed income, equity and commodity products.

J.P. Morgan and Reuters team up on RiskMetrics

In June 1996, J.P. Morgan signed an agreement with Reuters to cooperate on the building of a new and more powerful version of RiskMetrics. Since the launch of RiskMetrics in October 1994, we have received numerous requests to add new products, instruments, and markets to the daily volatility and correlation data sets. We have also perceived the need in the market for a more flexible VaR data tool than the standard matrices that are currently distributed over the Internet. The new partnership with Reuters, which will be based on the precept that both firms will focus on their respective strengths, will help us achieve these objectives.

Methodology

J.P. Morgan will continue to develop the RiskMetrics set of VaR methodologies and publish them in the quarterly *RiskMetrics Monitor* and in the annual *RiskMetrics—Technical Document*.

RiskMetrics data sets

Reuters will take over the responsibility for data sourcing as well as production and delivery of the risk data sets. The current RiskMetrics data sets will continue to be available on the Internet free of charge and will be further improved as a benchmark tool designed to broaden the understanding of the principles of market risk measurement.

When J.P. Morgan first launched RiskMetrics in October 1994, the objective was to go for broad market coverage initially, and follow up with more granularity in terms of the markets and instruments covered. This over time, would reduce the need for proxies and would provide additional data to measure more accurately the risk associated with non-linear instruments.

The partnership will address these new markets and products and will also introduce a new customizable service, which will be available over the Reuters Web service. The customizable RiskMetrics approach will give risk managers the ability to scale data to meet the needs of their individual trading profiles. Its capabilities will range from providing customized covariance matrices needed to run VaR calculations, to supplying data for historical simulation and stress-testing scenarios.

More details on these plans will be discussed in later editions of the *RiskMetrics Monitor.*

Systems

Both J.P. Morgan and Reuters, through its Sailfish subsidiary, have developed client-site RiskMetrics VaR applications. These products, together with the expanding suite of third party applications will continue to provide RiskMetrics implementations.

What is new in this fourth edition?

In terms of content, the Fourth Edition of the *Technical Document* incorporates the changes and refinements to the methodology that were initially outlined in the 1995–1996 editions of the *RiskMetrics Monitor*:

- **Expanded framework:** We have worked extensively on refining the analytical framework for analyzing options risk without having to perform relatively time consuming simulations and have outlined the basis for an improved methodology which incorporates better information on the tails of distributions related to financial asset price returns; we've also developed a data synchronization algorithm to refine our volatility and correlation estimates for products which do not trade in the same time zone;

- **New markets:** We expanded the daily data sets to include estimated volatilities and correlations of additional foreign exchange, fixed income and equity markets, particularly in South East Asia and Latin America.

- **Fine-tuned methodology:** We have modified the approach in a number of ways. First, we've changed our definition of price volatility which is now based on a total return concept; we've also revised some of the algorithms used in our mapping routines and are in the process of redefining the techniques used in estimating equity portfolio risk.

- **RiskMetrics products:** While we have continued to expand the list of third parties providing RiskMetrics products and support, this is no longer included with this document. Given the rapid pace of change in the availability of risk management software products, readers are advised to consult our Internet web site for the latest available list of products. This list, which now includes FourFifteen™, J.P. Morgan's own VaR calculator and report generating software, continues to grow, attesting to the broad acceptance RiskMetrics has achieved.

- **New tools to use the RiskMetrics data sets:** We have published an Excel add-in function which enables users to import volatilities and correlations directly into a spreadsheet. This tool is available from our Internet web site.

The structure of the document has changed only slightly. As before, its size warrants the following note: One need not read and understand the entire document in order to benefit from RiskMetrics. The document is organized in parts that address subjects of particular interest to many readers.

Part I: Risk Measurement Framework

This part is for the general practitioner. It provides a practical framework on how to think about market risks, how to apply that thinking in practice, and how to interpret the results. It reviews the different approaches to risk estimation, shows how the calculations work on simple examples and discusses how the results can be used in limit management, performance evaluation, and capital allocation.

Part II: Statistics of Financial Market Returns

This part requires an understanding and interest in statistical analysis. It reviews the assumptions behind the statistics used to describe financial market returns and how distributions of future returns can be estimated.

Part III: Risk Modeling of Financial Instruments

This part is required reading for implementation of a market risk measurement system. It reviews how positions in any asset class can be described in a standardized fashion (foreign exchange, interest rates, equities, and commodities). Special attention is given to derivatives positions. The purpose is to demystify derivatives in order to show that their market risks can be measured in the same fashion as their underlying.

Part IV: RiskMetrics Data Sets

This part should be of interest to users of the RiskMetrics data sets. First it describes the sources of all daily price and rate data. It then discusses the attributes of each volatility and correlation series in the RiskMetrics data sets. And last, it provides detailed format descriptions required to decipher the data sets that can be downloaded from public or commercial sources.

Appendices

This part reviews some of the more technical issues surrounding methodology and regulatory requirements for market risk capital in banks and demonstrates the use of RiskMetrics with the example diskette provided with this document. Finally, Appendix H shows you how to access the RiskMetrics data sets from the Internet.

RiskMetrics examples diskette

💾 This diskette is located inside the back cover. It contains an Excel workbook that includes some of the examples shown in this document. Such examples are identified by the icon shown here.

Future plans

We expect to update this *Technical Document* annually as we adapt our market risk standards to further improve the techniques and data to meet the changing needs of our clients.

RiskMetrics is a now an integral part of J.P. Morgan’s Risk Management Services group which provides advisory services to a wide variety of the firm’s clients. We continue to welcome any suggestions to enhance the methodology and adapt it further to the needs of the market. All suggestions, requests and inquiries should be directed to the authors of this publication or to your local RiskMetrics contacts listed on the back cover.

Acknowledgments

The authors would like to thank the numerous individuals who participated in the writing and editing of this document, particularly Chris Finger and Chris Athaide from J.P. Morgan’s risk management research group, and Elizabeth Frederick and John Matero from our risk advisory practice. Finally, this document could not have been produced without the contributions of our consulting editor, Tatiana Kolubayev. We apologize for any omissions to this list.

Table of contents

Part I Risk Measurement Framework

Chapter 1. Introduction 3
1.1 An introduction to Value-at-Risk and RiskMetrics 6
1.2 A more advanced approach to Value-at-Risk using RiskMetrics 7
1.3 What RiskMetrics provides 16

Chapter 2. Historical perspective of VaR 19
2.1 From ALM to VaR 22
2.2 VaR in the framework of modern financial management 24
2.3 Alternative approaches to risk estimation 26

Chapter 3. Applying the risk measures 31
3.1 Market risk limits 33
3.2 Calibrating valuation and risk models 34
3.3 Performance evaluation 34
3.4 Regulatory reporting, capital requirement 36

Part II Statistics of Financial Market Returns

Chapter 4. Statistical and probability foundations 43
4.1 Definition of financial price changes and returns 45
4.2 Modeling financial prices and returns 49
4.3 Investigating the random-walk model 54
4.4 Summary of our findings 64
4.5 A review of historical observations of return distributions 64
4.6 RiskMetrics model of financial returns: A modified random walk 73
4.7 Summary 74

Chapter 5. Estimation and forecast 75
5.1 Forecasts from implied versus historical information 77
5.2 RiskMetrics forecasting methodology 78
5.3 Estimating the parameters of the RiskMetrics model 90
5.4 Summary and concluding remarks 100

Part III Risk Modeling of Financial Instruments

Chapter 6. Market risk methodology 105
6.1 Step 1—Identifying exposures and cash flows 107
6.2 Step 2—Mapping cash flows onto RiskMetrics vertices 117
6.3 Step 3—Computing Value-at-Risk 121
6.4 Examples 134

Chapter 7. Monte Carlo 149
7.1 Scenario generation 151
7.2 Portfolio valuation 155
7.3 Summary 157
7.4 Comments 159

Part IV RiskMetrics Data Sets

Chapter 8. Data and related statistical issues 163
8.1 Constructing RiskMetrics rates and prices 165
8.2 Filling in missing data 170
8.3 The properties of correlation (covariance) matrices and VaR 176
8.4 Rebasing RiskMetrics volatilities and correlations 183
8.5 Nonsynchronous data collection 184

Chapter 9. Time series sources 197
9.1 Foreign exchange 199
9.2 Money market rates 199
9.3 Government bond zero rates 200
9.4 Swap rates 202
9.5 Equity indices 203
9.6 Commodities 205

Chapter 10. RiskMetrics volatility and correlation files 207
10.1 Availability 209
10.2 File names 209
10.3 Data series naming standards 209
10.4 Format of volatility files 211
10.5 Format of correlation files 212
10.6 Data series order 214
10.7 Underlying price/rate availability 214

Part V Backtesting

Chapter 11. Performance assessment 217
11.1 Sample portfolio 219
11.2 Assessing the RiskMetrics model 220
11.3 Summary 223

Appendices

Appendix A. Tests of conditional normality 227
Appendix B. Relaxing the assumption of conditional normality 235
Appendix C. Methods for determining the optimal decay factor 243
Appendix D. Assessing the accuracy of the delta-gamma approach 247
Appendix E. Routines to simulate correlated normal random variables 253
Appendix F. BIS regulatory requirements 257
Appendix G. Using the RiskMetrics examples diskette 263
Appendix H. RiskMetrics on the Internet 267

Reference

Glossary of terms 271
Bibliography 275

List of charts

Chart 1.1 VaR statistics 6
Chart 1.2 Simulated portfolio changes 9
Chart 1.3 Actual cash flows 9
Chart 1.4 Mapping actual cash flows onto RiskMetrics vertices 10
Chart 1.5 Value of put option on USD/DEM 14
Chart 1.6 Histogram and scattergram of rate distributions 15
Chart 1.7 Valuation of instruments in sample portfolio 15
Chart 1.8 Representation of VaR 16
Chart 1.9 Components of RiskMetrics 17
Chart 2.1 Asset liability management 22
Chart 2.2 Value-at-Risk management in trading 23
Chart 2.3 Comparing ALM to VaR management 24
Chart 2.4 Two steps beyond accounting 25
Chart 3.1 Hierarchical VaR limit structure 33
Chart 3.2 Ex post validation of risk models: DEaR vs. actual daily P&L 34
Chart 3.3 Performance evaluation triangle 35
Chart 3.4 Example: comparison of cumulative trading revenues 35
Chart 3.5 Example: applying the evaluation triangle 36
Chart 4.1 Absolute price change and log price change in U.S. 30-year government bond 47
Chart 4.2 Simulated stationary/mean-reverting time series 52
Chart 4.3 Simulated nonstationary time series 53
Chart 4.4 Observed stationary time series 53
Chart 4.5 Observed nonstationary time series 54
Chart 4.6 USD/DEM returns 55
Chart 4.7 USD/FRF returns 55
Chart 4.8 Sample autocorrelation coefficients for USD/DEM foreign exchange returns 57
Chart 4.9 Sample autocorrelation coefficients for USD S&P 500 returns 58
Chart 4.10 USD/DEM returns squared 60
Chart 4.11 S&P 500 returns squared 60
Chart 4.12 Sample autocorrelation coefficients of USD/DEM squared returns 61
Chart 4.13 Sample autocorrelation coefficients of S&P 500 squared returns 61
Chart 4.14 Cross product of USD/DEM and USD/FRF returns 63
Chart 4.15 Correlogram of the cross product of USD/DEM and USD/FRF returns 63
Chart 4.16 Leptokurtotic vs. normal distribution 65
Chart 4.17 Normal distribution with different means and variances 67
Chart 4.18 Selected percentile of standard normal distribution 69
Chart 4.19 One-tailed confidence interval 70
Chart 4.20 Two-tailed confidence interval 71
Chart 4.21 Lognormal probability density function 73
Chart 5.1 DEM/GBP exchange rate 79
Chart 5.2 Log price changes in GBP/DEM and VaR estimates (1.65σ) 80
Chart 5.3 NLG/DEM exchange rate and volatility 87
Chart 5.4 S&P 500 returns and VaR estimates (1.65σ) 88
Chart 5.5 GARCH(1,1)-normal and EWMA estimators 90
Chart 5.6 USD/DEM foreign exchange 92
Chart 5.7 Tolerance level and decay factor 94
Chart 5.8 Relationship between historical observations and decay factor 95
Chart 5.9 Exponential weights for $T = 100$ 95
Chart 5.10 One-day volatility forecasts on USD/DEM returns 96
Chart 5.11 One-day correlation forecasts for returns on USD/DEM FX rate and on S&P500 96
Chart 5.12 Simulated returns from RiskMetrics model 101
Chart 6.1 French franc 10-year benchmark maps 109

Chart 6.2 Cash flow representation of a simple bond 109
Chart 6.3 Cash flow representation of a FRN 110
Chart 6.4 Estimated cash flows of a FRN 111
Chart 6.5 Cash flow representation of simple interest rate swap 111
Chart 6.6 Cash flow representation of forward starting swap 112
Chart 6.7 Cash flows of the floating payments in a forward starting swap 113
Chart 6.8 Cash flow representation of FRA 113
Chart 6.9 Replicating cash flows of 3-month vs. 6-month FRA 114
Chart 6.10 Cash flow representation of 3-month Eurodollar future 114
Chart 6.11 Replicating cash flows of a Eurodollar futures contract 114
Chart 6.12 FX forward to buy Deutsche marks with US dollars in 6 months 115
Chart 6.13 Replicating cash flows of an FX forward 115
Chart 6.14 Actual cash flows of currency swap 116
Chart 6.15 RiskMetrics cash flow mapping 118
Chart 6.16 Linear and nonlinear payoff functions 123
Chart 6.17 VaR horizon and maturity of money market deposit 128
Chart 6.18 Long and short option positions 131
Chart 6.19 DEM 3-year swaps in Q1-94 141
Chart 7.1 Frequency distributions for and 153
Chart 7.2 Frequency distribution for DEM bond price 154
Chart 7.3 Frequency distribution for USD/DEM exchange rate 154
Chart 7.4 Value of put option on USD/DEM 157
Chart 7.5 Distribution of portfolio returns 158
Chart 8.1 Constant maturity future: price calculation 170
Chart 8.2 Graphical representation 175
Chart 8.3 Number of variables used in EM and parameters required 176
Chart 8.4 Correlation forecasts vs. return interval 185
Chart 8.5 Time chart 188
Chart 8.6 10-year Australia/US government bond zero correlation 190
Chart 8.7 Adjusting 10-year USD/AUD bond zero correlation 194
Chart 8.8 10-year Japan/US government bond zero correlation 195
Chart 9.1 Volatility estimates: daily horizon 202
Chart 11.1 One-day Profit/Loss and VaR estimates 219
Chart 11.2 Histogram of standardized returns 221
Chart 11.3 Standardized lower-tail returns 222
Chart 11.4 Standardized upper-tail returns 222
Chart A.1 Standard normal distribution and histogram of returns on USD/DEM 227
Chart A.2 Quantile-quantile plot of USD/DEM 232
Chart A.3 Quantile-quantile plot of 3-month sterling 234
Chart B.1 Tails of normal mixture densities 238
Chart B.2 GED distribution 239
Chart B.3 Left tail of GED (ν) distribution 240
Chart D.1 Delta vs. time to expiration and underlying price 248
Chart D.2 Gamma vs. time to expiration and underlying price 249

List of tables

Table 2.1 Two discriminating factors to review VaR models 29
Table 3.1 Comparing the Basel Committee proposal with RiskMetrics 39
Table 4.1 Absolute, relative and log price changes 46
Table 4.2 Return aggregation 49
Table 4.3 Box-Ljung test statistic 58
Table 4.4 Box-Ljung statistics 59
Table 4.5 Box-Ljung statistics on squared log price changes (cv = 25) 62
Table 4.6 Model classes 66
Table 4.7 VaR statistics based on RiskMetrics and BIS/Basel requirements 71
Table 5.1 Volatility estimators 78
Table 5.2 Calculating equally and exponentially weighted volatility 81
Table 5.3 Applying the recursive exponential weighting scheme to compute volatility 82
Table 5.4 Covariance estimators 83
Table 5.5 Recursive covariance and correlation predictor 84
Table 5.6 Mean, standard deviation and correlation calculations 91
Table 5.7 The number of historical observations used by the EWMA model 94
Table 5.8 Optimal decay factors based on volatility forecasts 99
Table 5.9 Summary of RiskMetrics volatility and correlation forecasts 100
Table 6.1 Data provided in the daily RiskMetrics data set 121
Table 6.2 Data calculated from the daily RiskMetrics data set 121
Table 6.3 Relationship between instrument and underlying price/rate 123
Table 6.4 Statistical features of an option and its underlying return 130
Table 6.5 RiskMetrics data for 27, March 1995 134
Table 6.6 RiskMetrics map of single cash flow 134
Table 6.7 RiskMetrics map for multiple cash flows 135
Table 6.8 Mapping a 6x12 short FRF FRA at inception 137
Table 6.9 Mapping a 6x12 short FRF FRA held for one month 137
Table 6.10 Structured note specification 139
Table 6.11 Actual cash flows of a structured note 139
Table 6.12 VaR calculation of structured note 140
Table 6.13 VaR calculation on structured note 140
Table 6.14 Cash flow mapping and VaR of interest-rate swap 142
Table 6.15 VaR on foreign exchange forward 143
Table 6.16 Market data and RiskMetrics estimates as of trade date July 1, 1994 145
Table 6.17 Cash flow mapping and VaR of commodity futures contract 145
Table 6.18 Portfolio specification 147
Table 6.19 Portfolio statistics 148
Table 6.20 Value-at-Risk estimates (USD) 148
Table 7.1 Monte Carlo scenarios 155
Table 7.2 Monte Carlo scenarios—valuation of option 156
Table 7.3 Value-at-Risk for example portfolio 158
Table 8.1 Construction of rolling nearby futures prices for Light Sweet Crude (WTI) 168
Table 8.2 Price calculation for 1-month CMF NY Harbor #2 Heating Oil 169
Table 8.3 Belgian franc 10-year zero coupon rate 175
Table 8.4 Singular values for USD yield curve data matrix 182
Table 8.5 Singular values for equity indices returns 182
Table 8.6 Correlations of daily percentage changes with USD 10-year 184
Table 8.7 Schedule of data collection 186
Table 8.7 Schedule of data collection 187
Table 8.8 RiskMetrics closing prices 191
Table 8.9 Sample statistics on RiskMetrics daily covariance forecasts 191
Table 8.10 RiskMetrics daily covariance forecasts 192

Table 8.11 Relationship between lagged returns and applied weights 193
Table 8.12 Original and adjusted correlation forecasts 193
Table 8.13 Correlations between US and foreign instruments 196
Table 9.1 Foreign exchange 199
Table 9.2 Money market rates: sources and term structures 200
Table 9.3 Government bond zero rates: sources and term structures 201
Table 9.4 Swap zero rates: sources and term structures 203
Table 9.5 Equity indices: sources 204
Table 9.6 Commodities: sources and term structures 205
Table 9.7 Energy maturities 205
Table 9.8 Base metal maturities 206
Table 10.1 RiskMetrics file names 209
Table 10.2 Currency and commodity identifiers 210
Table 10.3 Maturity and asset class identifiers 210
Table 10.4 Sample volatility file 211
Table 10.5 Data columns and format in volatility files 212
Table 10.6 Sample correlation file 213
Table 10.7 Data columns and format in correlation files 213
Table 11.1 Realized percentages of VaR violations 220
Table 11.2 Realized “tail return” averages 221
Table A.1 Sample mean and standard deviation estimates for USD/DEM FX 228
Table A.2 Testing for univariate conditional normality 230
Table B.1 Parameter estimates for the South African rand 240
Table B.2 Sample statistics on standardized returns 241
Table B.3 VaR statistics (in %) for the 1st and 99th percentiles 242
Table D.1 Parameters used in option valuation 249
Table D.2 MAPE (%) for call, 1-day forecast horizon 251
Table D.3 ME (%) for call, 1-day forecast horizons 251

Part I

Risk Measurement Framework

Chapter 1. Introduction

1.1 An introduction to Value-at-Risk and RiskMetrics 6
1.2 A more advanced approach to Value-at-Risk using RiskMetrics 7
    1.2.1 Using RiskMetrics to compute VaR on a portfolio of cash flows 9
    1.2.2 Measuring the risk of nonlinear positions 11
1.3 What RiskMetrics provides 16
    1.3.1 An overview 16
    1.3.2 Detailed specification 18

Chapter 1. Introduction

Jacques Longerstaey
Morgan Guaranty Trust Company
Risk Management Advisory
(1-212) 648-4936
riskmetrics@jpmorgan.com

This chapter serves as an introduction to the RiskMetrics product. RiskMetrics is a set of methodologies and data for measuring market risk. By market risk, we mean the potential for changes in value of a position resulting from changes in market prices.

We define risk as the degree of uncertainty of future net returns. This uncertainty takes many forms, which is why most participants in the financial markets are subject to a variety of risks. A common classification of risks is based on the source of the underlying uncertainty:

- Credit risk estimates the potential loss because of the inability of a counterparty to meet its obligations.
- Operational risk results from errors that can be made in instructing payments or settling transactions.
- Liquidity risk is reflected in the inability of a firm to fund its illiquid assets.
- Market risk, the subject of the methodology described in this document, involves the uncertainty of future earnings resulting from changes in market conditions, (e.g., prices of assets, interest rates). Over the last few years measures of market risk have become synonymous with the term Value-at-Risk.

RiskMetrics has three basic components:

- The first is a set of methodologies outlining how risk managers can compute Value-at-Risk on a portfolio of financial instruments. These methodologies are explained in this *Technical Document*, which is an annual publication, and in the *RiskMetrics Monitor*, the quarterly update to the *Technical Document*.
- The second is data that we distribute to enable market participants to carry out the methodologies set forth in this document.
- The third is Value-at-Risk calculation and reporting software designed by J.P. Morgan, Reuters, and third party developers. These systems apply the methodologies set forth in this document and will not be discussed in this publication.

This chapter is organized as follows:

- Section 1.1 presents the definition of Value-at-Risk (VaR) and some simple examples of how RiskMetrics offers the inputs necessary to compute VaR. The purpose of this section is to offer a basic approach to VaR calculations.
- Section 1.2 describes more detailed examples of VaR calculations for a more thorough understanding of how RiskMetrics and VaR calculations fit together. In Section 1.2.2 we provide an example of how to compute VaR on a portfolio containing options (nonlinear risk) using two different methodologies.
- Section 1.3 presents the contents of RiskMetrics at both the general and detailed level. This section provides a step-by-step analysis of the production of RiskMetrics volatility and correlation files as well as the methods that are necessary to compute VaR. For easy reference we provide section numbers within each step so that interested readers can learn more about that particular subject.

Reading this chapter requires a basic understanding of statistics. For assistance, readers can refer to the glossary at the end of this document.

1.1 An introduction to Value-at-Risk and RiskMetrics

Value-at-Risk is a measure of the maximum potential change in value of a portfolio of financial instruments with a given probability over a pre-set horizon. VaR answers the question: how much can I lose with x% probability over a given time horizon. For example, if you think that there is a 95% chance that the DEM/USD exchange rate will not fall by more than 1% of its current value over the next day, you can calculate the maximum potential loss on, say, a USD 100 million DEM/USD position by using the methodology and data provided by RiskMetrics. The following examples describe how to compute VaR using standard deviations and correlations of financial returns (provided by RiskMetrics) under the assumption that these returns are normally distributed. (RiskMetrics provides alternative methodological choices to address the inacurracies resulting from this simplifying assumption).

- **Example 1:** You are a USD-based corporation and hold a DEM 140 million FX position. What is your VaR over a 1-day horizon given that there is a 5% chance that the realized loss will be greater than what VaR projected? The choice of the 5% probability is discretionary and differs across institutions using the VaR framework.

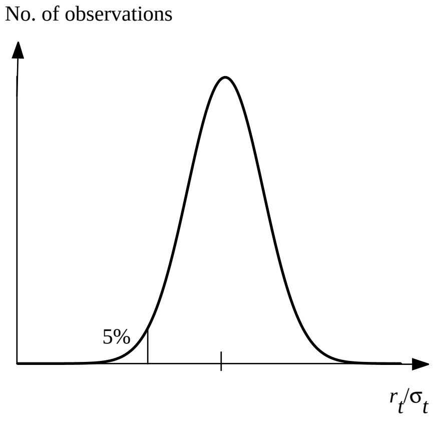
What is your exposure?

The first step in the calculation is to compute your exposure to market risk (i.e., mark-to-market your position). As a USD-based investor, your exposure is equal to the market value of the position in your base currency. If the foreign exchange rate is 1.40 DEM/USD, the market value of the position is USD 100 million.

What is your risk?

Moving from exposure to risk requires an estimate of how much the exchange rate can potentially move. The standard deviation of the return on the DEM/USD exchange rate, measured historically can provide an indication of the size of rate movements. In this example, we calculated the DEM/USD daily standard deviation to be 0.565%. Now, under the standard RiskMetrics assumption that standardized returns ((r_t/σ_t) on DEM/USD are normally distributed given the value of this standard deviation, VaR is given by 1.65 times the standard deviation (that is, 1.65σ) or 0.932% (see Chart 1.1). This means that the DEM/USD exchange rate is not expected to drop more than 0.932%, 95% of the time. **RiskMetrics provides users with the VaR statistics 1.65σ.**

Chart 1.1
VaR statistics



In USD, the VaR of the position[1] is equal to the market value of the position times the estimated volatility or:

FX Risk: \$100 million × 0.932% = \$932,000

What this number means is that 95% of the time, you will not lose more than \$932,000 over the next 24 hours.

---

[1] This is a simple approximation.

- **Example 2:** Let’s complicate matters somewhat. You are a USD-based corporation and hold a DEM 140 million position in the 10-year German government bond. What is your VaR over a 1-day horizon period, again, given that there is a 5% chance of understating the realized loss?

What is your exposure?

The only difference versus the previous example is that you now have both interest rate risk on the bond and FX risk resulting from the DEM exposure. The exposure is still USD 100 million but it is now at risk to two market risk factors.

What is your risk?

If you use an estimate of 10-year German bond standard deviation of 0.605%, you can calculate:

Interest rate risk: \$100 million × 1.65 × 0.605% = \$999,000
FX Risk: \$100 million × 1.65 × 0.565% = \$932,000

Now, the total risk of the bond is not simply the sum of the interest rate and FX risk because the correlation[2] between the return on the DEM/USD exchange rate the return on the 10-year German bond is relevant. In this case, we estimated the correlation between the returns on the DEM/USD exchange rate and the 10-year German government bond to be –0.27. Using a formula common in standard portfolio theory, the total risk of the position is given by:

$$[1.1] \qquad \mathrm{VaR} = \sqrt{\sigma^2_{\text{Interest rate}} + \sigma^2_{\text{FX}} + (2 \times \rho_{\text{Interest rate, FX}} \times \sigma_{\text{Interest rate}} \times \sigma_{\text{FX}})}$$

$$\begin{aligned} \mathrm{VaR} &= \sqrt{(0.999)^2 + (0.932)^2 + (2 \times -0.27 \times 0.999 \times 0.932)} \\ &= \$\,1.168 \text{ million} \end{aligned}$$

To compute VaR in this example, RiskMetrics provides users with the VaR of interest rate component (i.e., 1.65 × 0.605), the VaR of the foreign exchange position (i.e., 1.65 × 0.565) and the correlation between the two return series, –0.27.

1.2 A more advanced approach to Value-at-Risk using RiskMetrics

Value-at-Risk is a number that represents the potential change in a portfolio’s future value. How this change is defined depends on (1) the horizon over which the portfolio’s change in value is measured and (2) the “degree of confidence” chosen by the risk manager.

VaR calculations can be performed without using standard deviation or correlation forecasts. These are simply **one** set of inputs that can be used to calculate VaR, and that RiskMetrics provides for that purpose. The principal reason for preferring to work with standard deviations (volatility) is the strong evidence that the volatility of financial returns is predictable. Therefore, if volatility is predictable, it makes sense to make forecasts of it to predict future values of the return distribution.

---

[2] Correlation is a measure of how two series move together. For example, a correlation of 1 implies that two series move perfectly together in the same direction.

Suppose we want to compute the Value-at-Risk of a portfolio over a 1-day horizon with a 5% chance that the actual loss in the portfolio's value is greater than the VaR estimate. The Value-at-Risk calculation consists of the following steps.

1. Mark-to-market the current portfolio. Denote this value by V_0.
2. Define the future value of the portfolio, V_1, as $V_1 = V_0 e^r$ where[3] r represents the return on the portfolio over the horizon. For a 1-day horizon, this step is unnecessary as RiskMetrics assumes a 0 return.
3. Make a forecast of the 1-day return on the portfolio and denote this value by $\hat{r}$, such that there is a 5% chance that the actual return will be less than $\hat{r}$. Alternatively expressed,

$$\text{Probability } (r < \hat{r}) = 5\%.$$

4. Define the portfolio's future "worst case" value $\hat{V}_1$, as $\hat{V}_1 = V_0 e^{\hat{r}}$. The Value-at-Risk estimate is simply $V_0 - \hat{V}_1$.

   Notice that the VaR estimate can be written as $V_0\left(1 - e^{\hat{r}}\right)$. In the case that $\hat{r}$ is sufficiently small, $e^{\hat{r}} \approx 1 + \hat{r}$ so that VaR is approximately equal to $V_0\hat{r}$. is approximately equal to $V_0\hat{r}$. The purpose of a risk measurement system such as RiskMetrics is to offer a means to compute $\hat{r}$.

Within this more general framework we use a simple example to demonstrate how the RiskMetrics methodologies and data enable users to compute VaR. Assume the forecast horizon over which VaR is measured is one day and the level of "confidence" in the forecast to 5%. Following the steps outlined above, the calculation would proceed as follows:

1. Consider a portfolio whose current marked-to-market value, V_0, is USD 500 million.
2. To carry out the VaR calculation we require 1-day forecasts of the mean $\mu_{1|0}$. Within the RiskMetrics framework, we assume that the mean return over a 1-day horizon period is equal to 0.
3. We also need the standard deviation, $\sigma_{1|0}$, of the returns in this portfolio. Assuming that the return on this portfolio is distributed conditionally normal, $\hat{r} = -1.65\sigma_{1|0} + \mu_{1|0}$. The RiskMetrics data set provides the term $1.65\sigma_{1|0}$. Hence, setting $\mu_{1|0} = 0$ and $\sigma_{1|0} = 0.0321$, we get $V_1 = \text{USD 474.2 million}$.[4]
4. This yields a Value-at-Risk of USD 25.8 million (given by $V_0 - \hat{V}_1$).

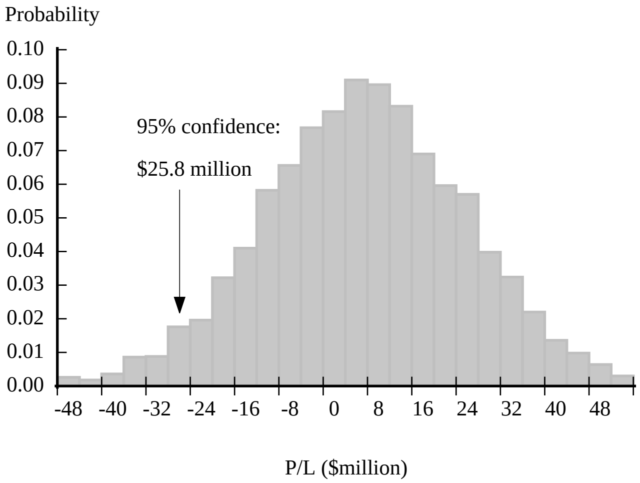
The histogram in Chart 1.2 presents future changes in value of the portfolio. VaR reduces risk to just one number, i.e., a loss associated with a given probability. It is often useful for risk managers to focus on the total distribution of potential gains and losses and we will discuss why this is so later in this document. (See Section 6.3).

---

[3] Where e is approximately 2.27183

[4] This number is computed from $\left(e^{-1.65\sigma}\right)V_0$

Chart 1.2
Simulated portfolio changes

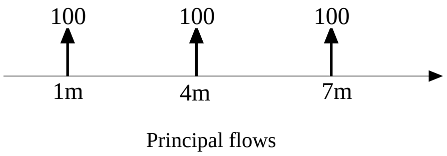


1.2.1 *Using RiskMetrics to compute VaR on a portfolio of cash flows*

Calculating VaR usually involves more steps than the basic ones outlined in the examples above. Even before calculating VaR, you need to estimate to which risk factors a particular portfolio is exposed. The preferred methodology for doing this is to decompose financial instruments into their basic cash flow components. The RiskMetrics methodology and data allow users to compute the VaR on portfolios consisting of a variety of cash flows. We use a simple example (a portfolio consisting of three cash flows) to demonstrate how to compute VaR.

Step 1. Each financial position in a portfolio is expressed as one or more cash flows that are marked-to-market at current market rates. For example, consider an instrument that gives rise to three USD 100 cash flows each occurring in 1, 4, and 7 months' time as shown in Chart 1.3.

Chart 1.3
Actual cash flows



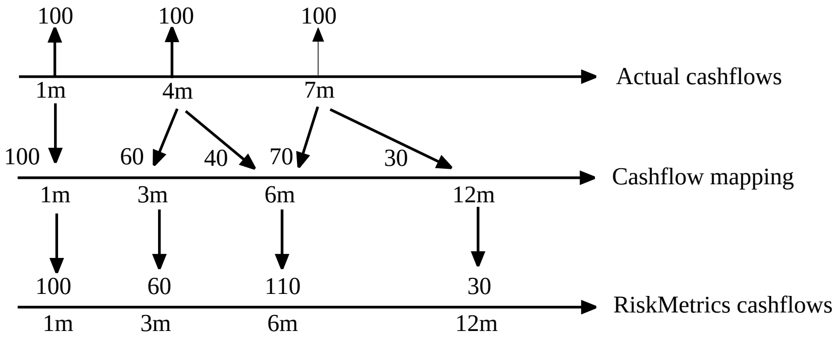
Step 2. When necessary, the actual cash flows are converted to RiskMetrics cash flows by mapping (redistributing) them onto a standard grid of maturity vertices, known as RiskMetrics vertices, which are fixed at the following intervals:

1m 3m 6m 12m 2yr 3yr 4yr 5yr 7yr 9yr 10yr 15yr 20yr 30yr

The purpose of the mapping is to standardize the cash flow intervals of the instrument such that we can use the volatilities and correlations that are routinely computed for the given vertices in the RiskMetrics data sets. (It would be impossible to provide volatility and correlation estimates on every possible maturity so RiskMetrics provides a mapping methodology which distributes cash flows to a workable set of standard maturities). The methodology for mapping cash flows is detailed in Chapter 6.

To map the cash flows, we use the RiskMetrics vertices closest to the actual vertices and redistribute the actual cash flows as shown in Chart 1.4.

Chart 1.4
Mapping actual cash flows onto RiskMetrics vertices



The RiskMetrics cash flow map is used to work backwards to calculate the return for each of the actual cash flows from the cash flow at the associated RiskMetrics vertex, or vertices.

For each actual cash flow, an analytical expression is used to express the relative change in value of the actual cash flow in terms of an underlying return on a particular instrument. Continuing with Chart 1.4, we can write the return on the actual 4-month cash flow in terms of the combined returns on the 3-month (60%) and 6-month (40%) RiskMetrics cash flows:

[1.2] $$r_{4m} = 0.60r_{3m} + 0.40r_{6m}$$

where

r_{4m} = return on the actual 4-month cash flow
r_{3m} = return on the 3-month RiskMetrics cash flow
r_{6m} = return on the 6-month RiskMetrics cash flow

Similarly, the return on the 7-month cash flow can be written as

[1.3] $$r_{7m} = 0.70r_{6m} + 0.30r_{12m}$$

Note that the return on the actual 1-month cash flow is equal to the return on the 1-month instrument.

Step 3. VaR is calculated at the 5th percentile of the distribution of portfolio return, and for a specified time horizon. In the example above, the distribution of the portfolio return, r_p, is written as:

[1.4] $$r_p = 0.33r_{1m} + 0.20r_{3m} + 0.37r_{6m} + 0.10r_{12m}$$

where, for example the portfolio weight 0.33 is the result of 100 divided by the total portfolio value 300.

Now, to compute VaR at the 95th percent confidence level we need the fifth percentile of the portfolio return distribution. Under the assumption that r_p is distributed conditionally normal, the fifth percentile is $-1.65\sigma_p$ where σ_p is the standard deviation of the portfolio return distribution. Applying Eq. [1.1] to a portfolio containing more than two instruments requires using simple matrix algebra. We can thus express this VaR calculation as follows:

$$VaR = \sqrt{\vec{V} R \vec{V}^T} \qquad [1.5]$$

where $\vec{V}$ is a vector of VaR estimates per instrument,

$$\vec{V} = [\,(0.33 \cdot 1.65\sigma_{1m}),\ (0.20 \cdot 1.65\sigma_{3m}),\ (0.37 \cdot 1.65\sigma_{6m}),\ (0.10 \cdot 1.65\sigma_{12m})\,]\ ,$$

and R is the correlation matrix

$$R = \begin{bmatrix} 1 & \rho_{3m,\,1m} & \rho_{6m,\,1m} & \rho_{12m,\,1m} \\ \rho_{1m,\,3m} & 1 & \rho_{6m,\,3m} & \rho_{12m,\,3m} \\ \rho_{1m,\,6m} & \rho_{3m,\,6m} & 1 & \rho_{12m,\,6m} \\ \rho_{1m,\,12m} & \rho_{3m,\,12m} & \rho_{6m,\,12m} & 1 \end{bmatrix} \qquad [1.6]$$

where, for example, $\rho_{1m,\,3m}$ is the correlation estimate between 1-month and 3-month returns.

Note that RiskMetrics provides the vector of information

$$\vec{V} = [\,(1.65\sigma_{1m}),\ (1.65\sigma_{3m}),\ (1.65\sigma_{6m}),\ (1.65\sigma_{12m})\,]$$

as well as the correlation matrix R. What the user has to provide are the actual portfolio weights.

1.2.2 Measuring the risk of nonlinear positions

When the relationship between position value and market rates is nonlinear, then we cannot estimate changes in value by multiplying “estimated changes in rates” by “sensitivity of the position to changing rates;” the latter is not constant (i.e., the definition of a nonlinear position). In our previous examples, we could easily estimate the risk of a fixed income or foreign exchange product by assuming a linear relationship between the value of an instrument and the value of its underlying. This is not a reasonable assumption when dealing with nonlinear products such as options.

RiskMetrics offers two methodologies, an **analytical approximatio**n and a **structured Monte Carlo simulation** to compute the VaR of nonlinear positions:

1. The first method approximates the nonlinear relationship via a mathematical expression that relates the return on the position to the return on the underlying rates. This is done by using what is known as a Taylor series expansion.

   This approach no longer necessarily assumes that the change in value of the instrument is approximated by its delta alone (the first derivative of the option’s value with respect to the underlying variable) but that a second order term using the option’s gamma (the second derivative of the option’s value with respect to the underlying price) must be introduced to measure the curvature of changes in value around the current value. In practice, other "greeks" such as vega (volatility), rho (interest rate) and theta (time to maturity) can also be used to improve the accuracy of the approximation. In Section 1.2.2.1, we present two types of analytical methods for computing VaR—the delta and delta-gamma approximation.

2. The second alternative, structured Monte Carlo simulation, involves creating a large number of possible rate scenarios and revaluing the instrument under each of these scenarios. VaR is then defined as the 5th percentile of the distribution of value changes. Due to the required revaluations, this approach is computationally more intensive than the first approach.

The two methods differ not in terms of how market movements are forecast (since both use the RiskMetrics volatility and correlation estimates) but in how the value of portfolios changes as a result of market movements. The analytical approach approximates changes in value, while the structured Monte Carlo fully revalues portfolios under various scenarios.

Let us illustrate these two methods using a practical example. We will consider throughout this section a portfolio comprised of two assets:

Asset 1: a future cash flow stream of DEM 1 million to be received in one year's time. The current 1-year DEM rate is 10% so the current market value of the instrument is DEM 909,091.

Asset 2: an at-the-money (ATM) DEM put/USD call option with contract size of DEM 1 million and expiration date one month in the future. The premium of the option is 0.0105 and the spot exchange rate at which the contract was concluded is 1.538 DEM/USD. We assume the implied volatility at which the option is priced is 14%.

The value of this portfolio depends on the USD/DEM exchange rate and the one-year DEM bond price. Technically, the value of the option also changes with USD interest rates and the implied volatility, but we will not consider these effects. Our risk horizon for the example will be five days. We take as the daily volatilities of these two assets $\sigma_{FX} = 0.42\%$ and $\sigma_B = 0.08\%$ and as the correlation between the two $\rho = -0.17$.

Both alternatives will focus on price risk exclusively and therefore ignore the risk associated with volatility (vega), interest rate (rho) and time decay (theta risk).

1.2.2.1 Analytical method

There are various ways to analytically approximate nonlinear VaR. This section reviews the two alternatives which we discussed previously.

Delta approximation

The standard VaR approach can be used to come up with first order approximations of portfolios that contain options. (This is essentially the same simplification that fixed income traders use when they focus exclusively on the duration of their portfolio). The simplest such approximation is to estimate changes in the option value via a linear model, which is commonly known as the "delta approximation." Delta is the first derivative of the option price with respect to the spot exchange rate. The value of δ for the option in this example is –0.4919.

In the analytical method, we must first write down the return on the portfolio whose VaR we are trying to calculate. The return on this portfolio consisting of a cash flow in one year and a put on the DEM/call on the USD is written as follows:

[1.7] $$r_p = r_{1y} + r_{\frac{DEM}{USD}} + \delta r_{\frac{DEM}{USD}}$$

where

r_{1p} = the price return on the 1-year German interest rates
$r_{\frac{DEM}{USD}}$ = the return on the DEM/USD exchange rate
δ = the delta of the option

Under the assumption that the portfolio return is normally distributed, VaR at the 95% confidence level is given by

$$\text{VaR} = 1.65 \sqrt{\sigma_{1y}^{2} + (1+\delta)^{2}\sigma_{\frac{DEM}{USD}}^{2} + 2\,(1+\delta)\,\rho_{1y,\frac{DEM}{USD}}\sigma_{1y}\sigma_{\frac{DEM}{USD}}} \qquad [1.8]$$

Using our volatilities and correlations forecasts for DEM/USD and the 1-year DEM rate (scaled up to the weekly horizon using the square root of time rule), the weekly VaR for the portfolio using the delta equivalent approach can be approximated by:

	Market value in USD	VaR(1w)
1-yr DEM cash flow	$591,086	$1,745
FX position - FX hedge	$300,331	$4,654
		Diversified VaR
		$4,684

Delta-gamma approximation

The delta approximation is reasonably accurate when the exchange rate does not change significantly, but less so in the more extreme cases. This is because the delta is a linear approximation of a non linear relationship between the value of the exchange rate and the price of the option as shown in Chart 1.5. We may be able to improve this approximation by including the gamma term, which accounts for nonlinear (i.e. squared returns) effects of changes in the spot rate (this attempts to replicate the convex option price to FX rate relationship as shown in Chart 1.5). The expression for the portfolio return is now

$$r_p = r_{1y} + r_{\frac{DEM}{USD}} + \delta r_{\frac{DEM}{USD}} + 0.5 \cdot \Gamma P_{\frac{DEM}{USD}}\left(r_{\frac{DEM}{USD}}\right)^{2} \qquad [1.9]$$

where

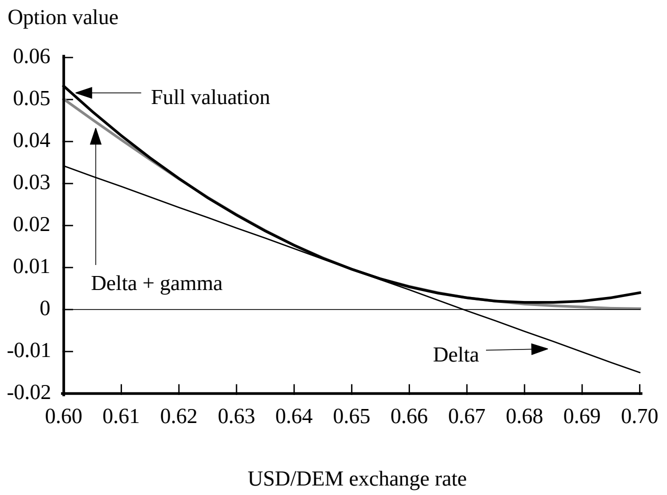
$P_{\frac{DEM}{USD}}$ = the value of the DEM/USD exchange rate when the VaR forecast is made
Γ = the gamma of the option.

In this example, Γ = DEM/USD 15.14.

Now, the gamma term (the fourth term in Eq. [1.9]) introduces skewness into the distribution of r_P (i.e., the distribution is no longer symmetrical around its mean). Therefore, since this violates one of the assumptions of normality (symmetry) we can no longer calculate the 95th percentile VaR as 1.65 times the standard deviation of r_p. Instead we must find the appropriate multiple (the counterpart to –1.65) that incorporates the skewness effect. We compute the 5th percentile of r_p's distribution (Eq. [1.9]) by computing its first four moments, i.e., r_p's mean, variance, skewness and kurtosis. We then find distribution whose first four moments match those of r_p's. (See Section 6.3 for details.)

Applying this methodology to this approach we find the VaR for this portfolio to be USD 3,708. Note that in this example, incorporating gamma reduces VaR relative to the delta only approximation (from USD 5006 to USD 3708).

Chart 1.5
Value of put option on USD/DEM
strike = 0.65 USD/DEM. Value in USD/DEM.



1.2.2.2 Structured Monte-Carlo Simulation

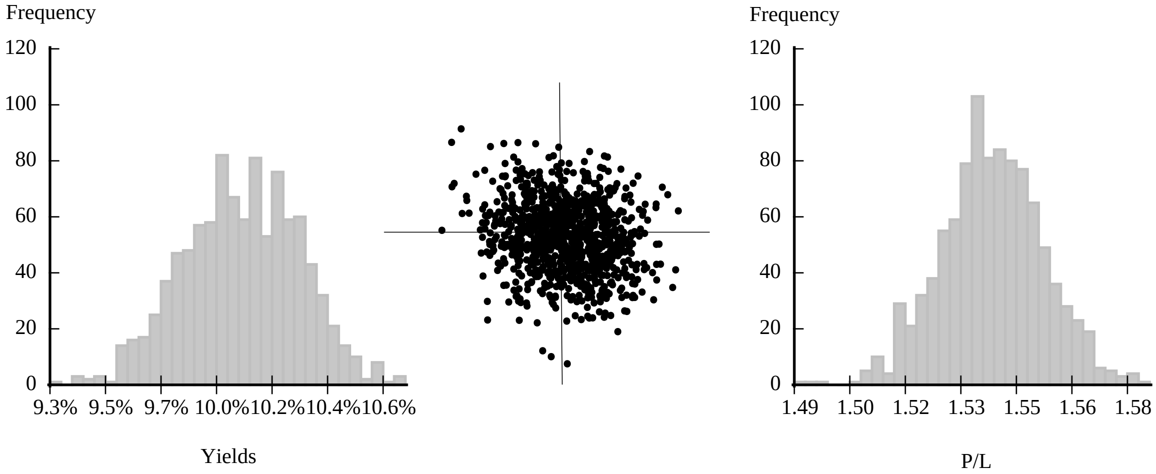
Given the limitations of analytical VaR for portfolios whose P/L distributions may not be symmetrical let alone normally distributed, another possible route is to use a model which instead of estimating changes in value by the product of a rate change (σ) and a sensitivity (δ, Γ), focuses on revaluing positions at changed rate levels. This approach is based on a full valuation precept where all instruments are marked to market under a large number of scenarios driven by the volatility and correlation estimates.

The Monte Carlo methodology consists of three major steps:

1. **Scenario generation** —Using the volatility and correlation estimates for the underlying assets in our portfolio, we produce a large number of future price scenarios in accordance with the lognormal models described previously. The methodology for generating scenarios from volatility and correlation estimates is described in Appendix E.
2. **Portfolio valuation** — For each scenario, we compute a portfolio value.
3. **Summary** — We report the results of the simulation, either as a portfolio distribution or as a particular risk measure.

Using our volatility and correlation estimates, we can apply our simulation technique to our example portfolio. We can generate a large number of scenarios (1000 in this example case) of DEM 1-year and DEM/USD exchange rates at the 1-week horizon. Chart 1.6 shows the actual distributions for both instruments as well as the scattergram indicating the degree of correlation (–0.17) between the two rate series.

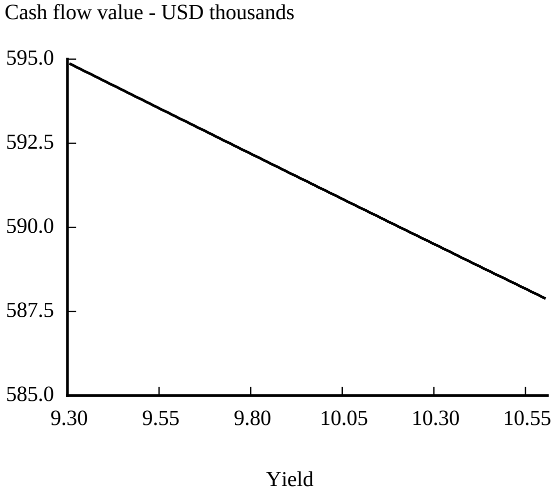
Chart 1.6
Histogram and scattergram of rate distributions
2-yr DEM rate and DEM/USD rate



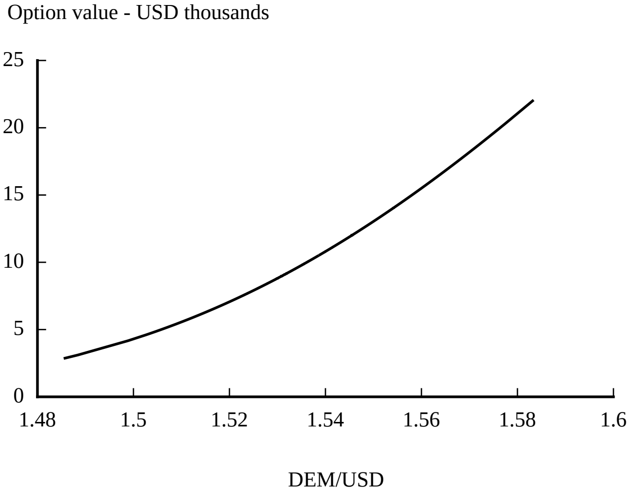
With the set of interest and foreign exchange rates obtained under simulation, we can revalue both of the instruments in our portfolio. Their respective payouts are shown in Chart 1.7.

Chart 1.7
Valuation of instruments in sample portfolio

Value of the cash flow stream

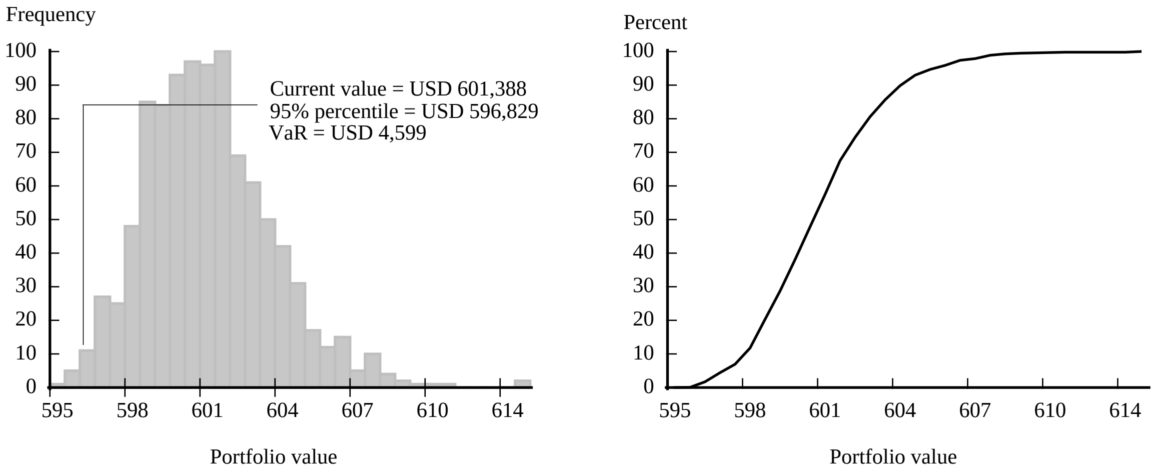


Value of the FX option



The final task is to analyze the distribution of values and select the VaR using the appropriate percentile. Chart 1.8 shows the value of the components of the portfolio at the end of the horizon period.

Chart 1.8
Representation of VaR
Histogram of portfolio values



The charts above provide a visual indication as to why the delta approximation is usually not suitable for portfolios that contain options. The distribution of returns in portfolios that include options is typically skewed. The standard delta equivalent VaR approach expects symmetry around the mean and applies a basic normal distribution approach (i.e., the 95% percentile equates to a 1.65 standard deviation move). In this case, the lack of symmetry in the distribution does not allow us to apply the normal approximation. Furthermore, the distribution's skewness results in a VaR number that is basically position dependent (i.e., the risk is different whether you are long or short the option).

1.3 What RiskMetrics provides

As discussed previously, RiskMetrics has three basic components which are detailed below.

1.3.1 An overview

With RiskMetrics J.P. Morgan and Reuters provide

1. A set of methodologies for statistical market risk measures that are based on, but differ significantly from, the methodology developed and used within J.P. Morgan. This approach was developed so as to enable other financial institutions, corporate treasuries, and investors to estimate their market risks in a consistent and reasonable fashion. Methodology defines how positions are to be mapped and how potential market movements are estimated and is detailed in the following chapters.
2. Daily recomputed data sets which are comprehensive sets of consistently estimated instrument level VaRs (i.e., 1.65 standard deviations) and correlations across a large number of asset classes and instruments. We currently distribute three different data sets over the Internet: one for short term trading risks, the second for intermediate term investment risks and the third for regulatory reporting. These are made available to the market free of charge.

   In the near future, a more customizable version of RiskMetrics where users will be able to create covariance matrices from a large underlying database according to various numerical methods will be made available over the Reuters Web. This product will not replace the data sets available over the Internet but will provide subscribers to the Reuters services with a more flexible tool.

The four basic classes of instruments that RiskMetrics methodology and data sets cover are represented as follows:

- Fixed income instruments are represented by combinations of amounts of cash flows in a given currency at specified dates in the future. RiskMetrics applies a fixed number of dates (14 vertices) and two types of credit standings: government and non-government. The data sets associated with fixed income are zero coupon instrument VaR statistics, i.e., 1.65σ, and correlations for both government and swap yield curves.

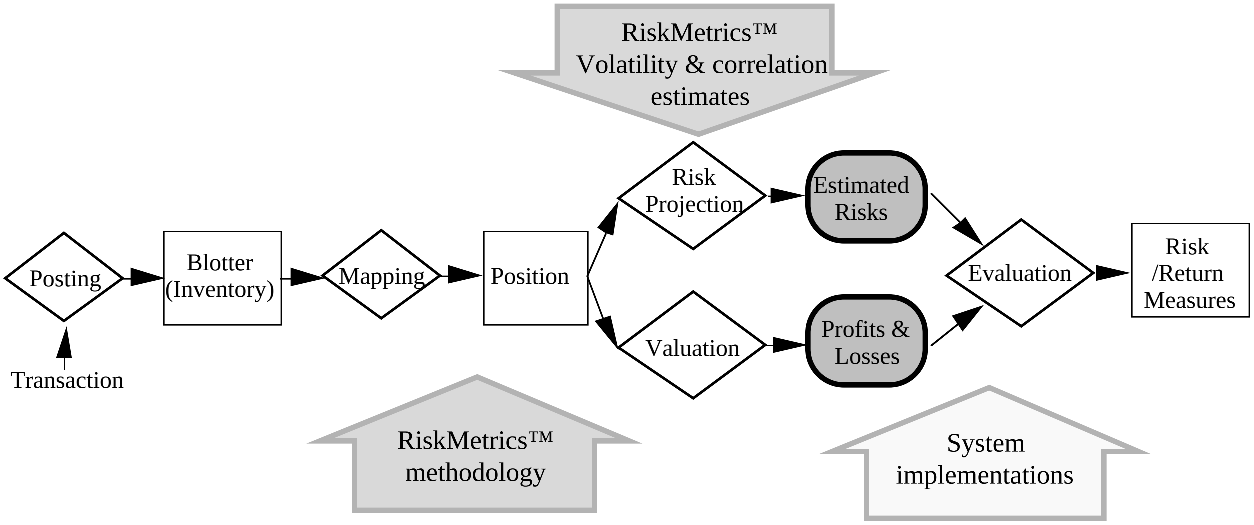
- Foreign exchange transactions are represented by an amount and two currencies. RiskMetrics allows for 30 different currency pairs (as measured against the USD).

- Equity instruments are represented by an amount and currency of an equity basket index in any of 30 different countries. Currently, RiskMetrics does not consider the individual characteristics of a company stock but only the weighted basket of companies as represented by the local index.

- Commodities positions are represented by amounts of selected standardized commodity futures contracts traded on commodity exchanges

3. Software provided by J.P. Morgan, Reuters and third party firms that use the RiskMetrics methodology and data documented herein.

Chart 1.9
Components of RiskMetrics


RiskMetrics™
Volatility & correlation
estimates
Risk
Projection
Estimated
Risks
Posting
Blotter
(Inventory)
Mapping
Position
Evaluation
Risk
/Return
Measures
Valuation
Profits &
Losses
Transaction
RiskMetrics™
methodology
System
implementations

Since the RiskMetrics methodology and the data sets are in the public domain and freely available, anyone is free to implement systems utilizing these components of RiskMetrics. Third parties have developed risk management systems for a wide range of clients using different methodologies. The following paragraphs provide a taxonomy comparing the different approaches.

1.3.2 Detailed specification

The section below provides a brief overview of how the RiskMetrics datasets are produced and how the parameters we provide can be used in a VaR calculation.

1.3.2.1 Production of volatility and correlation data sets

RiskMetrics provides the following sets of volatility and corresponding correlation data files. One set is for use in estimating VaR with a forecast horizon of one day. The other set is optimized for a VaR forecast horizon of one month. The third set is based on the quantitative criteria set by the Bank for International Settlements on the use of VaR models to estimate the capital required to cover market risks. The process by which these data files are constructed are as follows:

1. Financial prices are recorded from global data sources. (In 1997, RiskMetrics will switch to using Reuters data exclusively). For certain fixed income instruments we construct zero rates. See Chapter 9 for data sources and RiskMetrics building blocks.
2. Fill in missing prices by using the Expectation Maximization algorithm (detailed in Section 8.2). Prices can be missing for a variety of reasons, from technical failures to holiday schedules.
3. Compute daily price returns on all 480 time series (Section 4.1).
4. Compute standard deviations and correlations of financial price returns for a 1-day VaR forecast horizon. This is done by constructing exponentially weighted forecasts. (See Section 5.2). Production of the daily statistics also involves setting the sample daily mean to zero. (See Section 5.3). If data is recorded at different times (Step 1), users may require an adjustment algorithm applied to the correlation estimates. Such an algorithm is explained in Section 8.5. Also, users who need to rebase the datasets to account for a base currency other than the USD should see Section 8.4.
5. Compute standard deviations and correlations of financial price returns for 1-month VaR forecast horizon. This is done by constructing exponentially weighted forecasts (Section 5.3). Production of the monthly statistics also involves setting the sample daily mean to zero.

1.3.2.2 RiskMetrics VaR calculation

1. The first step in the VaR calculation is for the user to define three parameters: (1) VaR forecast horizon—the time over which VaR is calculated, (2) confidence level—the probability that the realized change in portfolio will be less than the VaR prediction, and (3) the base currency.
2. For a given portfolio, once the cash flows have been identified and marked-to-market (Section 6.1) they need to be mapped to the RiskMetrics vertices (Section 6.2).
3. Having mapped all the positions, a decision must be made as to how to compute VaR. If the user is willing to assume that the portfolio return is approximately conditionally normal, then download the appropriate data files (instrument level VaRs and correlations) and compute VaR using the standard RiskMetrics approach (Section 6.3).
4. If the user's portfolio is subject to nonlinear risk to the extent that the assumption of conditional normality is no longer valid, then the user can choose between two methodologies—delta-gamma and structured Monte Carlo. The former is an approximation of the latter. See Section 6.3 for a description of delta-gamma and Chapter 7for an explanation of structured Monte Carlo.

Chapter 2. Historical perspective of VaR

2.1 From ALM to VaR 22
2.2 VaR in the framework of modern financial management 24
    2.2.1 Valuation 25
    2.2.2 Risk estimation 25
2.3 Alternative approaches to risk estimation 26
    2.3.1 Estimating changes in value 26
    2.3.2 Estimating market movements 27

Chapter 2. Historical perspective of VaR

Jacques Longerstaey
Morgan Guaranty Trust Company
Risk Management Advisory
(1-212) 648-4936
riskmetrics@jpmorgan.com

Measuring the risks associated with being a participant in the financial markets has become the focus of intense study by banks, corporations, investment managers and regulators. Certain risks such as counterparty default have always figured at the top of most banks' concerns. Others such as market risk (the potential loss associated with market behavior) have only gotten into the limelight over the past few years. Why has the interest in market risk measurement and monitoring arisen? The answer lies in the significant changes that the financial markets have undergone over the last two decades.

1. Securitization: Across markets, traded securities have replaced many illiquid instruments, e.g., loans and mortgages have been securitized to permit disintermediation and trading. Global securities markets have expanded and both exchange traded and over-the-counter derivatives have become major components of the markets.

   These developments, along with technological breakthroughs in data processing, have gone hand in hand with changes in management practices—a movement away from management based on accrual accounting toward risk management based on marking-to-market of positions. Increased liquidity and pricing availability along with a new focus on trading led to the implementation of frequent revaluation of positions, the mark-to-market concept.

   As investments became more liquid, the potential for frequent and accurate reporting of investment gains and losses has led an increasing number of firms to manage daily earnings from a mark-to-market perspective. The switch from accrual accounting to mark-to-market often results in higher swings in reported returns, therefore increasing the need for managers to focus on the volatility of the underlying markets. The markets have not suddenly become more volatile, but the focus on risks through mark-to-market has highlighted the potential volatility of earnings.

   Given the move to frequently revalue positions, managers have become more concerned with estimating the potential effect of changes in market conditions on the value of their positions.

2. Performance: Significant efforts have been made to develop methods and systems to measure financial performance. Indices for foreign exchange, fixed income securities, commodities, and equities have become commonplace and are used extensively to monitor returns within and/or across asset classes as well as to allocate funds.

   The somewhat exclusive focus on returns, however, has led to incomplete performance analysis. Return measurement gives no indication of the cost in terms of risk (volatility of returns). Higher returns can only be obtained at the expense of higher risks. While this trade-off is well known, the risk measurement component of the analysis has not received broad attention.

   Investors and trading managers are searching for common standards to measure market risks and to estimate better the risk/return profile of individual assets or asset classes. Notwithstanding the external constraints from the regulatory agencies, the management of financial firms have also been searching for ways to measure market risks, given the potentially damaging effect of miscalculated risks on company earnings. As a result, banks, investment firms, and corporations are now in the process of integrating measures of market risk into their management philosophy. They are designing and implementing market risk monitoring systems that can provide management with timely information on positions and the estimated loss potential of each position.

Over the last few years, there have been significant developments in conceptualizing a common framework for measuring market risk. The industry has produced a wide variety of indices to measure return, but little has been done to standardize the measure of risk. Over the last 15 years many market participants, academics, and regulatory bodies have developed concepts for measuring market risks. Over the last five years, two approaches have evolved as a means to measure market risk. The first approach, which we refer to as the statistical approach, involves forecasting a portfolio's return distribution using probability and statistical models. The second approach is referred to as scenario analysis. This methodology simply revalues a portfolio under different values of market rates and prices. Note that in stress scenario analysis does not necessarily require the use of a probability or statistical model. Instead, the future rates and prices that are used in the revaluation can be arbitrarily chosen. Risk managers should use both approaches—the statistical approach to monitor risks continuously in all risk-taking units and the scenario approach on a case-by-case basis to estimate risks in unique circumstances. **This document explains, in detail, the statistical approach—RiskMetrics—to measure market risk.**

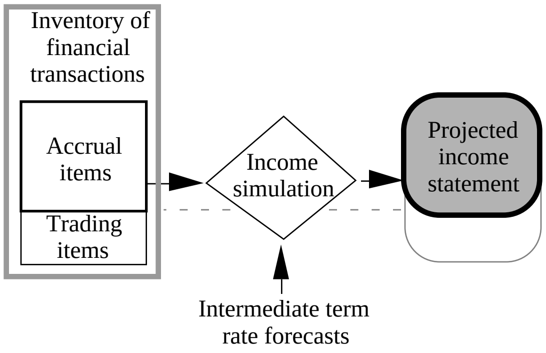
This chapter is organized as follows:

- Section 2.1 reviews how VaR was developed to support the risk management needs of trading activities as opposed to investment books. Though the distinction to date has been an accounting one not an economic one, VaR concepts are now being used across the board.
- Section 2.2 looks at the basic steps of the risk monitoring process.
- Section 2.3 reviews the alternative VaR models currently being used and how RiskMetrics provides end-users with the basic building blocks to test different approaches.

2.1 From ALM to VaR

A well established method of looking at market risks in the banking industry is to forecast earnings under predetermined price/rate market conditions (or scenarios). Earnings here are defined as earnings reported in a firm's Financial Statements using generally accepted accounting principles. For many institutions the bulk of activities are reported on an accrual basis, i.e., transactions are booked at historical costs +/- accruals. Only a limited number of trading items are marked to market. Because changes in market rates manifest themselves only slowly when earnings are reported on an accrual basis, the simulation of income has to be done over extended periods, i.e., until most of the transactions on the books mature. Chart 2.1 illustrates this conventional Asset/Liability Management approach.

Chart 2.1
Asset liability management


Inventory of financial transactions
Accrual items
Trading items
Income simulation
Projected income statement
Intermediate term rate forecasts

There are two major drawbacks to this methodology:

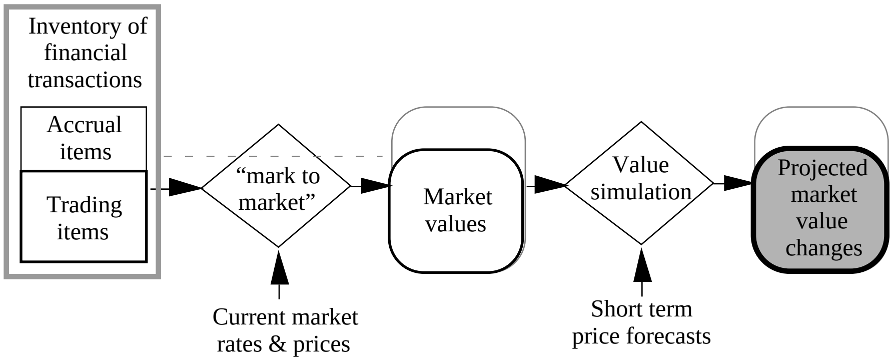
- It requires projecting market rate developments over extended periods into the future.

- It supports the illusion that gains and losses occur at the time they show up in the accrual accounts (i.e., when they are realized following accounting principles). What this means is that return is only defined as net interest earnings, a framework which ignores the change in price component of the return function.

  Every investor would agree that the total return on a bond position is the sum of the interest earned and the change in the value of the bond over a given time horizon. Traditional ALM, as a result of accounting conventions, ignores the change in value of the instrument since positions are not marked to market. This has often lead crafty ALM managers to create positions which look attractive on paper because of high net interest earnings, but which would not perform as well if their change in market value were considered.

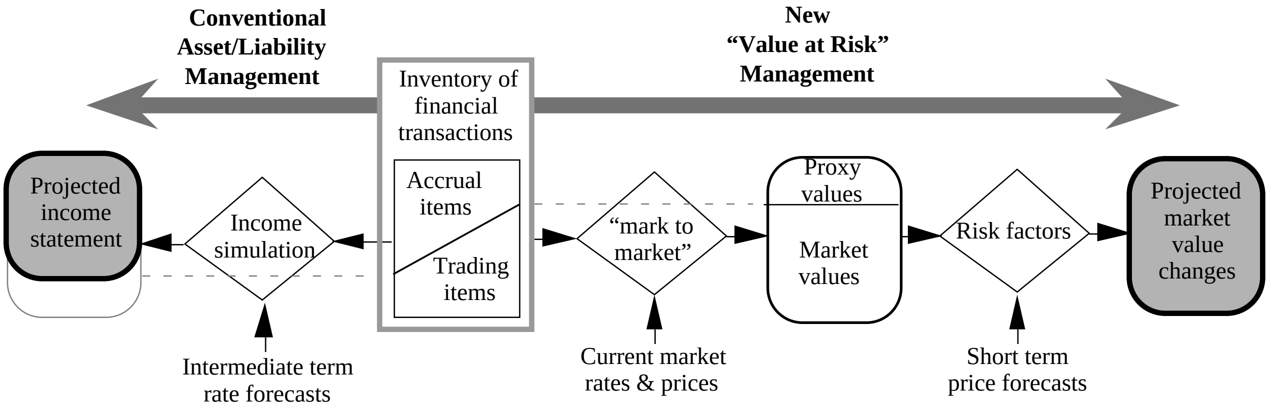
The market risk in trading positions is usually measured differently and managed separately. Trading positions are marked-to-market and the market value is then subjected to projections of changes in short term in rates and prices. This is much less hazardous as rate forecasts are usually limited to short horizons, i.e., the time it should take to close out or hedge the trading position.

Chart 2.2
Value-at-Risk management in trading


Inventory of financial transactions
Accrual items
Trading items
"mark to market"
Market values
Value simulation
Projected market value changes
Current market rates & prices
Short term price forecasts

The distinction between accrual items and trading items and their separate treatment for market risk management has led to significant complications—particularly when transactions classified as "trading items" under generally accepted accounting principles are used to hedge transactions classified as "accrual items". In an effort to overcome this difficulty, many firms – particularly those with relatively large trading books have expanded the market risk approach to also include accrual items, at least for internal risk management reporting. This is done by estimating the fair market value of the accrual items and the changes in their fair value under different short term scenarios. Thus we are witnessing the evolution of an alternative to the conventional approach of Asset/Liability Management, the Value-at-Risk approach. It started in pure trading operations, but is now gaining increased following in the financial industry.

Chart 2.3
Comparing ALM to VaR management


Conventional Asset/Liability Management
New "Value at Risk" Management
Inventory of financial transactions
Accrual items
Trading items
Projected income statement
Income simulation
Intermediate term rate forecasts
"mark to market"
Current market rates & prices
Proxy values
Market values
Risk factors
Short term price forecasts
Projected market value changes

The advantages of VaR Management are that it

- Incorporates the mark-to-market approach uniformly.
- Relies on a much shorter horizon forecast of market variables. This improves the risk estimate as short horizon forecasts tend to be more accurate than long horizon forecasts.

Of course, drawbacks exist. One of them is that it may not be trivial to mark certain transactions to market or even understand their behavior under certain rate environments. This is particularly true for instruments such as demand deposits in a retail banking environment for example. Whatever the difficulties, the aim of getting an integrated picture of a firm's exposure to market risks is worth a number of assumptions, some of which may be reasonable representations of reality. In the case of demand deposits, a recent article by Professor Robert Jarrow outlines how power swaps could be modelled to represent a retail bank's core deposit base risks (RISK, February 1996).

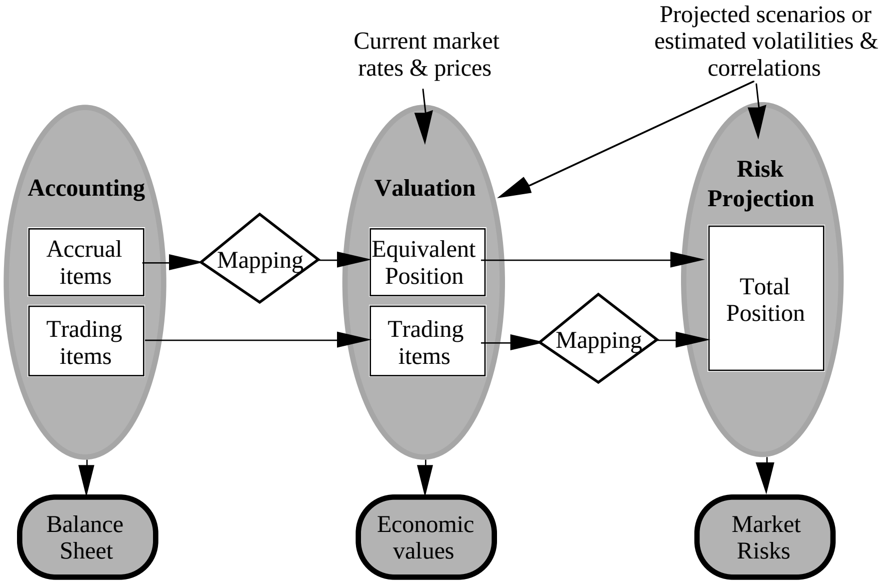
Some critics also argue that marking-to-market all transactions over short time periods creates too much "earnings" or volatility. Looking at risks in this fashion may be misleading. This is the direction of the industry and its accounting regulators however and it will be up to financial analysts to adapt to the new environment. The volatility of earnings will not just appear out of the blue. The changes in accounting practices will ultimately show economic reality as it really is.

Market risk can be absolute or relative. In its absolute form, what is measured is the loss in the value of a position or a portfolio resulting from changes in market conditions. Absolute market risk is what managers of trading operations measure. Corporates who wish to estimate real potential losses from their treasury operations also focus on absolute market risk. Regulators are interested in absolute market risks in relation to a firm's capital. When investment performance is measured against an index, the inherent market risk is relative in the sense that it measures the potential underperformance against a benchmark.

2.2 VaR in the framework of modern financial management

As discussed before there are two steps to VaR measurement. First, all positions need to be marked to market (valuation). Second we need to estimate the future variability of the market value. Chart 2.4 illustrates this point.

Chart 2.4
Two steps beyond accounting


Projected scenarios or estimated volatilities & correlations
Current market rates & prices
Accounting
Valuation
Risk Projection
Accrual items
Mapping
Equivalent Position
Total Position
Trading items
Trading items
Mapping
Balance Sheet
Economic values
Market Risks

2.2.1 *Valuation*

Trading items are valued at their current prices/rates as quoted in liquid secondary markets. To value transactions for which, in the absence of a liquid secondary market, no market value exists, we first map them into equivalent positions, or decompose them into parts for which secondary market prices exist. The most basic such "part" is a single cash flow with a given maturity and currency of the payor. Most transactions can be described as a combination of such cash flows and thus can be valued approximately as the sum of market values of their component cash flows.

Only non-marketable items that contain options cannot be valued in this simple manner. For their valuation we also need expected volatilities and correlations of the prices and rates that affect their value, and we need an options pricing model. Volatilities describe potential movements in rates with a given probability; correlations describe the interdependencies between different rates and prices. Thus, for some valuations, we require volatilities and correlations.

2.2.2 *Risk estimation*

Here we estimate value changes as a consequence of expected changes in prices and rates. The potential changes in prices are defined by either specific scenarios or a set of volatility and correlation estimates. If the value of a position depends on a single rate, then the potential change in value is a function of the rates in the scenarios or volatility of that rate. If the value of a position depends on multiple rates, then the potential change in its value is a function of the combination of rates in each scenario or of each volatility and of each correlation between all pairs of rates.

Generating equivalent positions on an aggregate basis facilitates the simulation. As will be shown later, the simulation can be done algebraically (using statistics and matrix algebra), or exhaustively by computing estimated value changes for many combinations of rate changes.

In the RiskMetrics framework, forecasts of volatilities and correlations play a central role. They are required for valuations in the case of derivatives, the critical inputs for risk estimation.

2.3 Alternative approaches to risk estimation

More than one VaR model is currently being used and most practitioners have selected an approach based on their specific needs, the types of positions they hold, their willingness to trade off accuracy for speed (or vice versa), and other considerations.

The different models used today differ on basically two fronts:

- How the changes in the values of financial instruments are estimated as a result of market movements.

- How the potential market movements are estimated.

What makes the variety of models currently employed is the fact that the choices made on the two fronts mentioned above can be mixed and matched in different ways.

2.3.1 Estimating changes in value

There are basically two approaches to estimating how the value of a portfolio changes as a result of market movements: analytical methods and simulation methods.

2.3.1.1 Analytical methods

One such method is the analytical sensitivity approach based on the following equation:

$$\text{estimated value change} = f\,(\text{position sensitivity, estimated rate/price change})$$

where the position sensitivity factor establishes the relationship between the value of the instrument and of the underlying rate or price, and determines the accuracy of the risk approximation.

In its simplest form, the analytical sensitivity approach looks like this:

$$\text{estimated value change} = \text{position sensitivity} \times \text{estimated rate change}$$

For example, the value change of a fixed income instrument can be estimated by using the instrument's duration. Although this linear approximation simplifies the convex price/yield relationship of a bond, it is extensively used in practice because duration often accounts for the most significant percentage of the risk profile of a fixed income instrument. Similar simplifications can be made for options where the estimated change in value is approximated by the option's delta.

The initial versions of RiskMetrics basically used an analytical VaR approach that assumed that market risk could be estimated by using a simple first-order approximation such as the one outlined above. We have since extended the analytical approach to account for nonlinear relationships between market value and rate changes (e.g., options), which requires accounting for gamma risk in addition to delta risk. The more refined version of the analytical approach looks like this:

$$\begin{aligned}\text{estimated value change} &= (\text{position sensitivity } 1 \times \text{estimated rate change}) \\ &\quad + 1/2\,(\text{position sensitivity } 2) \times (\text{estimated rate change})^2 + \ldots\end{aligned}$$

In the case of an option, the first-order position sensitivity is the delta, while the second-order term is the gamma. Higher order effects can also be estimated using an analytical approach, but the math typically gets more complex.

The analytical approach requires that positions be summarized in some fashion so that the estimated rate changes can be applied. This process of aggregating positions is called mapping and is described in Chapter 6.

The advantages of analytical models is that they are computationally efficient and enable users to estimate risk in a timely fashion.

2.3.1.2 Simulation methods

The second set of approaches, typically referred to as Full Valuation models rely on revaluing a portfolio of instruments under different scenarios. How these scenarios are generated varies across models, from basic historical simulation to distributions of returns generated from a set of volatility and correlation estimates such as RiskMetrics. Some models include user-defined scenarios that are based off of major market events and which are aimed at estimating risk in crisis conditions. This process is often referred to a stress testing.

Full Valuation models typically provide a richer set of risk measures since users are able to focus on the entire distribution of returns instead of a single VaR number. Their main drawback is the fact that the full valuation of large portfolios under a significant number of scenarios is computationally intensive and takes time. It may not be the preferred approach when the goal is to provide senior management with a timely snapshot of risks across a large organization.

2.3.2 Estimating market movements

The second discriminant between VaR approaches is how market movements are estimated. There is much more variety here and the following list is not an exhaustive list of current practice.

RiskMetrics

RiskMetrics uses historical time series analysis to derive estimates of volatilities and correlations on a large set of financial instruments. It assumes that the distribution of past returns can be modelled to provide us with a reasonable forecast of future returns over different horizons.

While RiskMetrics assumes conditional normality of returns, we have refined the estimation process to incorporate the fact that most markets show kurtosis and leptokurtosis. We will be publishing factors to adjust for this effect once the RiskMetrics customizable data engine becomes available on the Reuters Web.

These volatility and correlation estimates can be used as inputs to:

- Analytical VaR models
- Full valuation models. In Appendix E we outline how the RiskMetrics volatility and correlation data sets can be used to drive simulations of future returns.

Historical Simulation

The historical simulation approach, which is usually applied under a full valuation model, makes no explicit assumptions about the distribution of asset returns. Under historical simulation, portfolios are valued under a number of different historical time windows which are user defined. These lookback periods typically range from 6 months to 2 years.

Once the RiskMetrics customizable data engine becomes available on the ReutersWeb, users will be able to access the underlying historical data needed to perform this type of simulation.

Monte Carlo Simulation

While historical simulation quantifies risk by replicating one specific historical path of market evolution, stochastic simulation approaches attempt to generate many more paths of market returns. These returns are generated using a defined stochastic process (for example, assume that interest rates follow a random walk) and statistical parameters that drive the process (for example, the mean and variance of the random variable).The RiskMetrics data sets can be used as inputs to this process.

In addition, the following VaR models add refinements to the results generated by the approaches listed above.

Implied volatilities

Some practitioners will also look to the market to provide them with an indication of future potential return distributions. Implied volatility as extracted from a particular option pricing model is the market's forecast of future volatility. Implied volatilities are often used in comparison to history to refine the risk analysis.

Implied volatilities are not currently used to drive global VaR models as this would require observable options prices on all instruments that compose a portfolio. Unfortunately, the universe of consistently observable options prices is not yet large enough; generally only exchange traded options are reliable sources of prices. In particular, the number of implied correlations that can be derived from traded options prices is insignificant compared to the number of correlations required to estimate risks in portfolios containing many asset types.

User-defined scenarios

Most risk management models add user-defined rate and price movements to the standard VaR number, if only to test the effect of what could happen if historical patterns do not repeat themselves. Some scenarios are subjectively chosen while others recreate past crises events. The latter is referred to as stress testing and is an integral part of a well designed risk management process.

Selecting the appropriate measurement method is not, however, straightforward. Judgment in the choice of methodologies will always be important. Cost benefit trade-offs are different for each user, depending on his position in the markets, the number and types of instruments traded, and the technology available. Different choices can be made even at different levels of an organization, depending on the objectives. While trading desks of a bank may require precise risk estimation involving simulation on relatively small portfolios, senior management may opt for an analytical approach that is cost efficient and timely. It is important for senior management to know whether the risk of the institution is \$10 million or \$50 million. It is irrelevant for them to make the distinction between \$10 million and \$11 million. Achieving this level of accuracy at the senior management level is not only irrelevant, but can also be unachievable operationally, or at a cost which is not consistent with shareholder value.

Since its introduction, RiskMetrics has become the umbrella name for a series of VaR methodologies, from the simple analytical estimation based on the precept that all instruments are linear (the so-called delta approximation) to the structured Monte Carlo simulation.

Not all participants with exposure to the financial and commodities markets will have the resources to perform extensive simulations. That is why we have strived in this update of the *RiskMetrics— Technical Document* to refine analytical approximations of risk for non-linear instruments (the delta-gamma approximations). During 1997, the availability of historical rates and prices under the RiskMetrics customizable data engine will make historical simulation an option for users of our products.

Table 2.1
Two discriminating factors to review VaR models

		How to estimate the change in the value of instruments		
			Analytical	**Full Valuation**
How to estimate rate and price changes	**Full VaR model**	RiskMetrics	Covariance matrices applied to standard instrument maps.	Covariance matrices used to define scenarios for structured Monte Carlo.
		Historical simulation	Not applicable.	Portfolios revalued under historical return distributions (lookback period varies.
		Monte Carlo	Not applicable.	Statistical parameters determine stochastic processes. Sources of data vary (can include RiskMetrics covariance matrices).
	Partial VaR model	Implied volatilities	Covariance matrices applied to standard instrument maps.	Covariance matrices used to define scenarios for structured Monte Carlo.
		User defined	Sensitivity analysis on single instruments.	Limited number of scenarios.

Chapter 3. Applying the risk measures

3.1 Market risk limits 33
3.2 Calibrating valuation and risk models 34
3.3 Performance evaluation 34
3.4 Regulatory reporting, capital requirement 36
  3.4.1 Capital Adequacy Directive 36
  3.4.2 Basel Committee Proposal 37

Chapter 3. Applying the risk measures

Jacques Longerstaey
Morgan Guaranty Trust Company
Risk Management Advisory
(1-212) 648-4936
riskmetrics@jpmorgan.com

The measures of market risk outlined in the preceding sections can have a variety of applications. We will highlight just a few:

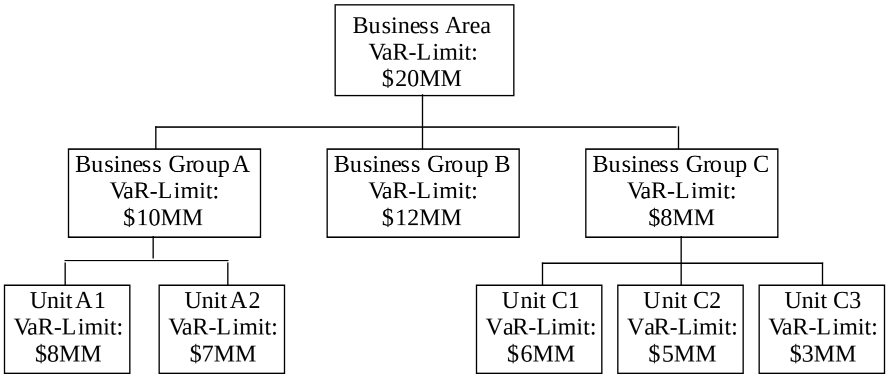
- To measure and compare market risks.
- To check the valuation/risk models.
- To evaluate the performance of risk takers on a return/risk basis.
- To estimate capital levels required to support risk taking.

3.1 Market risk limits

Position limits have traditionally been expressed in nominal terms, futures equivalents or other denominators unrelated to the amount of risk effectively incurred. The manager of a USD bond portfolio will be told for example that he cannot hold more than 100 million USD worth of U.S. Treasury bonds. In most cases, the measure contains some risk constraint expressed in a particular maturity or duration equivalent (if the 100 million limit is in 2-year equivalents, the manager will not be able to invest 100 million in 30-year bonds). Setting limits in terms of Value-at-Risk has significant advantages: position benchmarks become a function of risk and positions in different markets while products can be compared through this common measure. A common denominator rids the standard limits manuals of a multitude of measures which are different for every asset class. Limits become meaningful for management as they represent a reasonable estimate of how much could be lost.

A further advantage of Value-at-Risk limits comes from the fact that VaR measures incorporate portfolio or risk diversification effects. This leads to hierarchical limit structures in which the risk limit at higher levels can be lower than the sum of risk limits of units reporting to it.

Chart 3.1
Hierarchical VaR limit structure

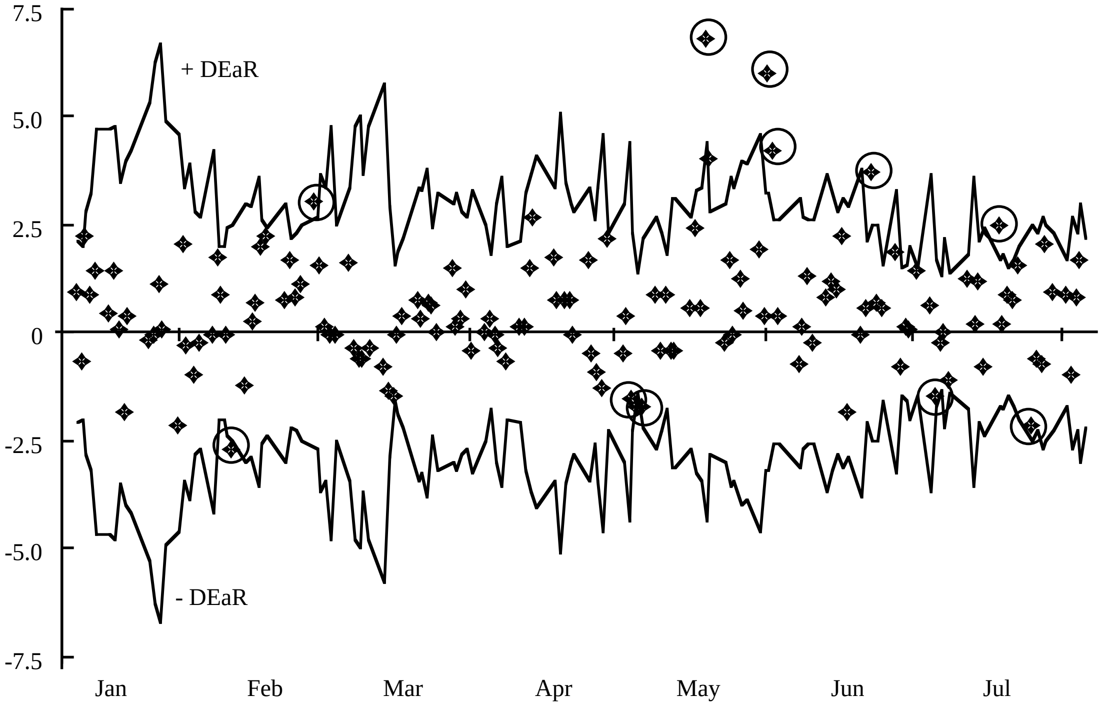


Setting limits in terms of risk helps business managers to allocate risk to those areas which they feel offer the most potential, or in which their firms' expertise is greatest. This motivates managers of multiple risk activities to favor risk reducing diversification strategies.

3.2 Calibrating valuation and risk models

An effective method to check the validity of the underlying valuation and risk models is to compare DEaR estimates with realized daily profits and losses over time. Chart 3.2 illustrates the concept. The stars show the daily P&L of a global trading business during the first 7 months of 1993, the two lines show the Daily Earnings at Risk, plus and minus.

Chart 3.2
Ex post validation of risk models: DEaR vs. actual daily P&L



By definition, the cone delimited by the +/–DEaR lines should contain 90% of all the stars, because DEaR is defined as the maximum amount of expected profit or losses 90% of the time. If there are substantially more than 10% of the stars outside the DEaR-cone, then the underlying models underestimate the risks. If there are no stars outside the DEaR cone and not even close to the lines, then the underlying models overestimate the risks.

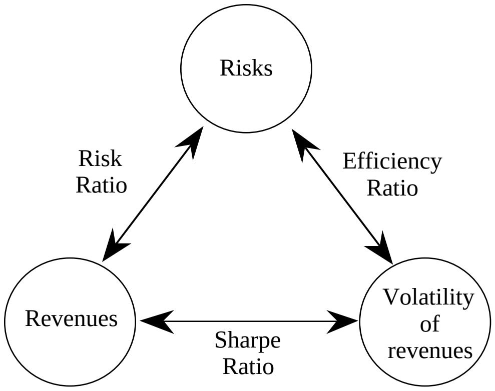
This type of chart is only a reasonable reflection of the risk statistics if the daily profit and losses are derived solely from overnight risk taking and not intraday trading and other activities. Often this is not the case. Then instead of the daily P&L you should plot what is often referred to as the “no-action-P&L”; it describes the hypothetical P&L on the position that would have been incurred if the previous day’s closing position had been kept for the next 24 hours and then revalued. This data is often difficult to collect.

3.3 Performance evaluation

To date, trading and position taking talent have been rewarded to a significant extent on the basis of total returns. Given the high rewards bestowed on outstanding trading talent this may bias the trading professionals towards taking excessive risks.It is often referred to as giving traders a free option on the capital of your firm. The interest of the firm or capital provider may be getting out of line with the interest of the risk taking individual unless the risks are properly measured and returns are adjusted for the amount of risk effectively taken.

To do this correctly one needs a standard measure of risks. Ideally risk taking should be evaluated on the basis of three interlinked measures: revenues, volatility of revenues, and risks. This is illustrated by Chart 3.3:

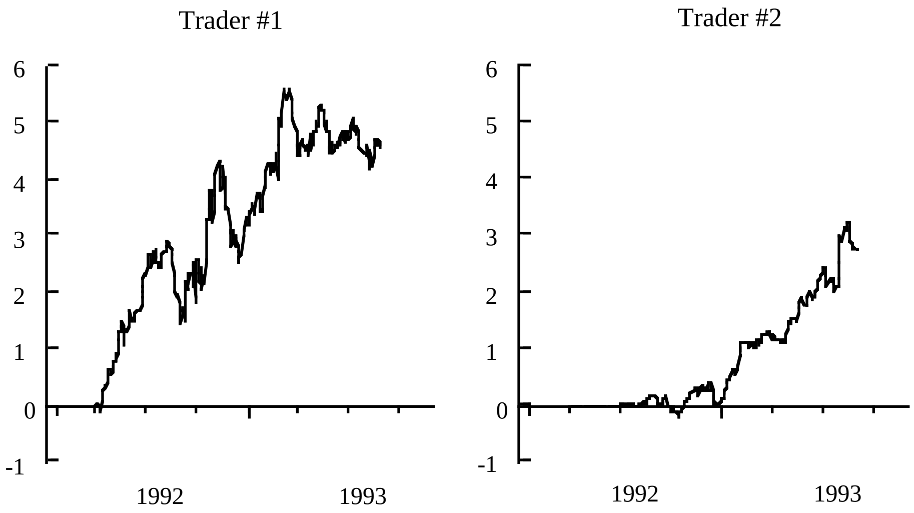
Chart 3.3
Performance evaluation triangle



Including estimated (ex ante) and realized (ex post) volatility of profits adds an extra dimension to performance evaluation. The ratio of P&L over risk (risk ratio) and of P&L over volatility (Sharpe ratio) can be combined into what we define as a trader's efficiency ratio (estimated risk/realized volatility) that measures an individual's capacity to translate estimated risk into low realized volatility of revenues.

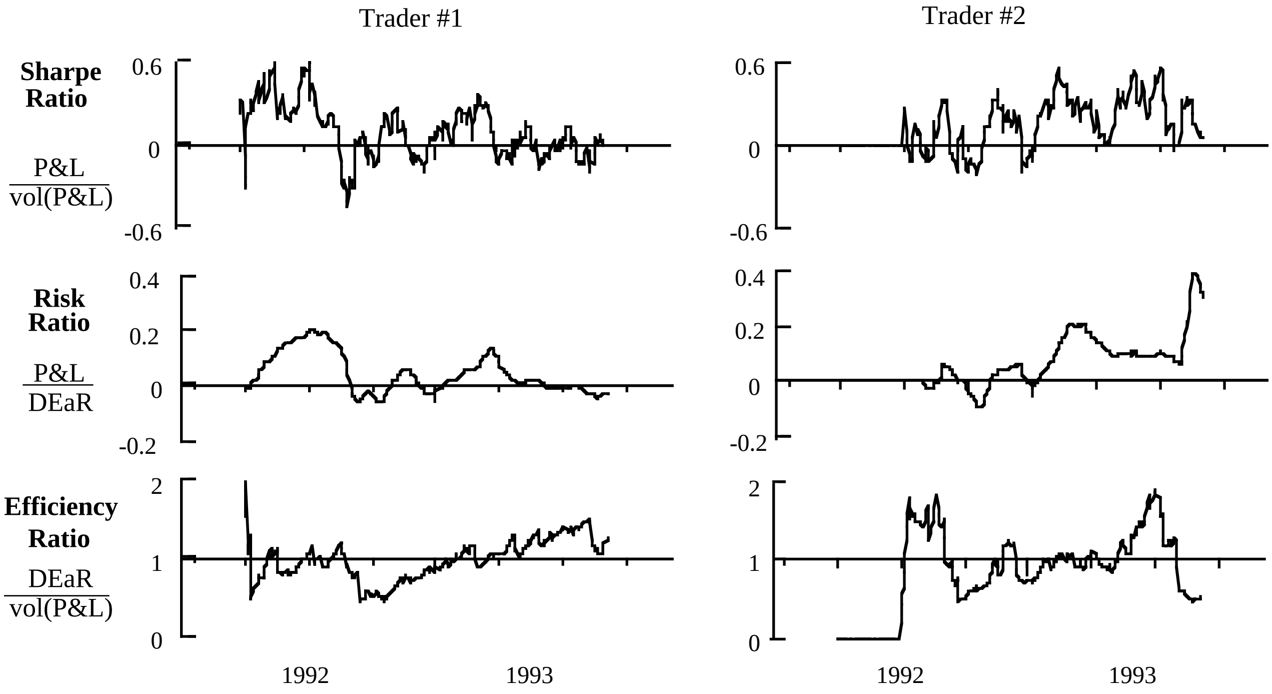
Consider an example to illustrate the issue. Assume you have to evaluate Trader #1 relative to Trader #2 and the only information on hand is the history of their respective cumulative trading revenues (i.e., trading profits). This information allows you to compare their profits and volatility of their profits, but says nothing about their risks.

Chart 3.4
Example: comparison of cumulative trading revenues
cumulative revenues



With risk information you can compare the traders more effectively. Chart 3.5 shows, for the two traders the risk ratio, sharpe ratio, and efficiency ratio over time.

Chart 3.5
Example: applying the evaluation triangle



Note, you have no information on the type of market these traders operate in or the size of positions they have taken. Nevertheless Chart 3.5 provides interesting comparative information which lead to a richer evaluation.

3.4 Regulatory reporting, capital requirement

Financial institutions such as banks and investment firms will soon have to meet capital requirements to cover the market risks that they incur as a result of their normal operations. Currently the driving forces developing international standards for market risk based capital requirements are the European Community which issued a binding Capital Adequacy Directive (EC-CAD) and the Basel Committee on Banking Supervision at the Bank for International Settlements (Basel Committee) which has recently come out with revised proposals on the use of banks internal models. (See Appendix F for more information.)

3.4.1 Capital Adequacy Directive

The European Union’s EEC 93/6 directive mandates banks and investment firms to set capital aside to cover market risks. In a nutshell the EC-CAD computes the capital requirement as a sum of capital requirements on positions of different types in different markets. It does not take into account the risk reducing effect of diversification. As a result, the strict application of the current recommendations will lead to financial institutions, particularly the ones which are active internationally in many different markets, to overestimate their market risks and consequently be required to maintain very high capital levels. While there may be some prudential advantages to this, it is not an efficient allocation of financial resources and could lead certain activities to be moved outside the jurisdiction of the financial regulatory authorities.

3.4.2 Basel Committee Proposal

In January 1996, the Basel Committee on Banking Supervision of the BIS issued a revised consultative proposal on an “Internal Model-Based Approach to Market Risk Capital Requirements” that represents a big step forward in recognizing the new quantitative risk estimation techniques used by the banking industry. These proposals recognize that current practice among many financial institutions has superseded the original guidelines in terms of sophistication, and that banks should be given the flexibility to use more advanced methodologies. This so-called “internal models” approach addresses a number of issues that were raised when banks commented on the original proposal dated April 1993.

Table 3.1 compares the methodologies for estimating market risks as recently proposed by the Basel Committee with the RiskMetrics methodology covered in this document. This comparison focuses exclusively on the so-called quantitative factors that the BIS guidelines will require banks to use. It does not address the qualitative ones related to the risk management process and which are beyond the scope of this document.

While the methodologies outlined in the BIS proposals have come a long way in overcoming important objections to the first set of proposals, there are still a number of issues that will be debated further. In order to facilitate the discussion between regulators and regulated, we have published since mid-1995 in parallel with the existing volatility and correlation data sets, a RiskMetrics Regulatory Data Set. The distribution of this regulatory data set is not an endorsement of the Basel committee proposals and the following paragraphs which explain how the data set can be used do not constitute J.P. Morgan’s official position on the content and scope of the Basel committee proposal.

Consistent with the other RiskMetrics data sets, the Regulatory Data Set contains volatility estimates for a 1-day holding period. Given that the BIS rules require market risk estimates to be calculated over a 10-day holding period and a 99% confidence interval (i.e., 2.33 standard deviations), users will need to rescale the 1-day volatility (see Eq. [3.1]). The Basel proposals allow for this adjustment of data (they actually refer to scaling up VaR estimates but exclude this practice in the case of options since it only works for instruments’ whose pricing formulae are linear). Scaling up volatility estimates is perfectly legitimate, assuming no autocorrelation in the data. Scaling up Value-at-Risk does not work for options, though using scaled up volatilities to estimate the market risks of options with adequate pricing algorithms poses no problem.

As in the other data sets, volatilities and correlations are measured as daily log changes in rates and prices. However, contrary to the exponential weighting schemes used for the other data sets, estimates in the Regulatory Data Set are based on simple moving averages of 1 year of historical data, sampled daily.

To make it comparable to the standard data sets, the RiskMetrics Regulatory Data Set is based on 95% confidence. Including the adjustment for the holding period, users downloading the data sets will need to rescale the volatility estimates according to the following equation, in order to meet the requirements set forth in the Basel proposals (this adjustment assumes a normal distribution. More refined methods incorporating the characteristics of fat tailed distributions are outlined in the statistics section of this document):

$$[3.1] \qquad \begin{aligned} V_{\text{Basel}} &= \frac{2.33}{1.65} \cdot V_{\text{RiskMetrics RD}} \cdot \sqrt{10} \\ &= 4.45 \cdot V_{\text{RiskMetrics RD}} \end{aligned}$$

where

$V_{\text{RiskMetrics RD}}$ = volatilities provided in RiskMetrics Regulatory Dataset
V_{Basel} = volatilities suggested by Basel Committee for use in internal models

Correlations across asset classes (i.e., foreign exchange to government bonds for example) are supplied in the RiskMetrics Regulatory Data Set, despite the fact that actual use of empirical correlations in the VaR estimates is subject to regulatory approval. The BIS has stated that the use of correlations across asset classes would be based on whether the supervisory authority was satisfied with the integrity of the estimation methodology.

Table 3.1
Comparing the Basel Committee proposal with RiskMetrics

Issue	Basel Committee proposal	RiskMetrics
Mapping: how positions are described in summary form	• Fixed Income: at least 6 time buckets, differentiate government yield curves and spread curves.	• Fixed Income: data for 7–10 buckets of government yield curves in 16 markets, 4 buckets money market rates in 27 markets, 4–6 buckets in swap rates in 18 markets.
	• Equities: country indices, individual stocks on basis of beta equivalent.	• Equities: country indices in 27 markets, individual stocks on beta (correction for non-systematic risk).
	• Commodities: to be included, not specified how.	• Commodities: 80 volatility series in 11 commodities (spot and term).
Volatility: how statistics of future price movement are estimated	• Volatility expressed in standard deviation of normal distribution proxy for daily historical observations year or more back. Equal weights or alternative weighting scheme provided effective observation period is at least one year.	• Volatility expressed in standard deviation of normal distribution proxy for exponentially weighted daily historical observations with decay factors of .94 (for trading, 74 day cutoff 1%) and .97 (for investing, 151 day cutoff at 1%).
		• Special Regulatory Data Set, incorporating Basel Committee 1-year moving average assumption.
	• Estimate updated at least quarterly.	• Estimates updated daily.
Adversity: size of adverse move in terms of normal distribution	• Minimum adverse move expected to happen with probability of 1% (2.32 standard deviations) over 10 business days. Permission to use daily statistics scaled up with square root of 10 (3.1). Equivalent to 7.3 daily standard deviations.	• For trading: minimum adverse move expected to happen with probability of 5% (1.65 standard deviation) over 1 business day.
		• For investment: minimum adverse move expected to happen with probability of 5% (1.65 standard deviation) over 25 business days.
Options: treatment of time value and non-linearity	• Risk estimate must consider effect of non-linear price movement (gamma-effect).	• Non-linear price movement can be estimated analytically (delta-gamma) or under simulation approach. Simulation scenarios to be generated from estimated volatilities and correlations.
	• Risk estimate must include effect of changes in implied volatilities (vega-effect).	• Estimates of volatilities of implied volatilities currently not provided, thus limited coverage of options risk.
Correlation: how risks are aggregated	• Portfolio effect can be considered within asset classes (Fixed Income, Equity, Commodity, FX). Use of correlations across asset classes subject to regulatory approval.	• Full portfolio effect considered across all possible parameter combinations.
	• Correlations estimated with equally weighted daily data for more than one year.	• Correlations estimated using exponentially weighted daily historical observations with decay factors of 0.94 (for trading, 74 day cutoff 1%) and 0.97 (for investing, 151 day cutoff at 1%).
Residuals: treatment of instrument specific risks	• Instrument specific risks not covered by standard maps should be estimated.	• Does not deal with specific risks not covered in standard maps.
	• Capital requirements at least equal to 50% of charge calculated under standard methodology.	