Risk Management Toolbox[™] Release Notes

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Risk Management Toolbox[™] Release Notes

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R2020b

Version: 1.8

New Features

Bug Fixes

R2020b

Market Risk: Backtest expected shortfall (ES) models using minimally biased Acerbi-Szekely tests

You can perform minimally biased Acerbi-Szekely tests with the minBiasRelative and minBiasAbsolute functions when using an esbacktestbysim object.

Market Risk: Expected shortfall (ES) model VaR level extended to 99.9%

The VaR level is extended to 99.9% for ES backtesting with an esbacktest object that uses precomputed tables of critical values.

Lifetime Credit Analysis: Probability of default models and examples

You can perform lifetime credit analysis using fitLifetimePDModel to create a Logistic or Probit lifetime probability of default model. You can then use the predict, predictLifetime, modelDiscrimination, and modelAccuracy functions to analyze the probability of default model. For more information, see examples:

- "Compare Logisitic Model for Lifetime PD to Champion Model"
- "Compare Lifetime PD Models Using Cross-Validation"
- "Expected Credit Loss (ECL) Computation"

Insurance Analysis: Chain ladder, expected claims, and Bornhuetter-Fergurson techniques for analyzing insurance claims reserves

You can use a development triangle with the chain ladder technique, expected claims technique, or Bornhuetter-Fergurson technique to calculate important measures of insurance risk.

- The developmentTriangle object supports the following functions:
 - view
 - linkRatios
 - linkRatioAverages
 - cdfSummary
 - ultimateClaims
 - fullTriangle
- The chainLadder object supports the following functions:
 - ibnr
 - unpaidClaims
 - summary
- The expectedClaims object supports the following functions:
 - ultimateClaims
 - ibnr

- unpaidClaims
- summary
- The bornhuetterFerguson object supports the following functions:
 - ultimateClaims
 - ibnr
 - unpaidClaims
 - summary

R2020a

Version: 1.7 New Features Bug Fixes Compatibility Considerations

Consumer Credit Risk: Screen credit scorecard predictors on data that is too big to fit in memory using tall Arrays

Tall variable support for screenpredictors.

Random number generation for credit copula classes has changed

In R2019b and previous releases, when using the creditDefaultCopula and creditMigrationCopula classes to perform nonparallel simulations, the MATLAB[®] global random number generator was used to generate scenarios.

Compatibility Considerations

In R2020a, the random number generator for the creditDefaultCopula and creditMigrationCopula classes is set to Threefry for both parallel and nonparallel code paths.

R2019b

Version: 1.6 New Features Bug Fixes Compatibility Considerations

Market Risk: Backtest expected shortfall (ES) models using Du and Escanciano tests

The following new functions provide support for performing expected shortfall (ES) backtests by Du and Escanciano:

- esbacktestbyde
- summary
- runtests
- unconditionalDE
- conditionalDE
- simulate

New examples demonstrating workflows using expected shortfall (ES) backtests by Du and Escanciano:

- Workflow for Expected Shortfall (ES) Backtesting by Du and Escanciano
- Rolling Windows and Multiple Models for Expected Shortfall (ES) Backtesting by Du and Escanciano

Consumer Credit Risk: Validation of compact credit scorecards using validatemodel

Work with compact credit scorecards using the compactCreditScorecard class and then use validatemodel to validate a compact credit scorecard.

Credit Scorecard: Example comparing credit score using logistic regression and decision trees

A new example compares credit score values using logistic regression and decision trees (Comparison of Credit Scoring Using Logistic Regression and Decision Trees).

Credit Scorecard: Example using reject inference to incorporate credit rejection data into creditscorecard workflow

A new example demonstrates two approaches for using reject inference techniques to incorporate credit rejection data as part of the creditscorecard modeling workflow (Use Reject Inference Techniques with Credit Scorecards).

Consumer Credit Risk: Example comparing probability of default using through-the-cycle and point-in-time models

A new example compares the probability of default using through-the-cycle (TTC) and point-in-time (PIT) models (Comparison of Probability of Default Using Through-the-Cycle and Point-in-Time Models).

Consumer Credit Risk: Example fitting different types of models to loss given default (LGD) data

A new example demonstrates fitting different types of models to loss given default (LGD) data (Model Loss Given Default).

Calculation of p-value for bin has changed

In R2019a and previous releases, the binomial VaR backtest reports the tail probability as the *p*-value and compares the reported *p*-value to half of the *alpha* (1- TestLevel) of the bin test.

In R2019b, the binomial VaR backtest bin calculates the *p*-value contained in the TestResults output using the 2*tail probability convention and the *p*-value can be compared to *alpha*. For more information, see Algorithms.

R2019a

Version: 1.5 New Features

Bug Fixes

Consumer Credit Risk: Predictor screening for credit scorecards

Perform predictor screening for credit scorecards using the screenpredictors function. For more information, see Feature Screening with screenpredictors.

Consumer Credit Risk: Support for compact credit scorecards for easier deployment and reduced memory usage

Work with compact credit scorecards using the compactCreditScorecard class along with associated functions for displaypoints, score, and probdefault. In addition, you can create a compact credit scorecard by using the compact function from Financial Toolbox[™] with a creditscorecard object. For more information, see compactCreditScorecard Object Workflow and Validate the Quality of a Compact Credit Scorecard Model.

R2018b

Version: 1.4 New Features Bug Fixes Compatibility Considerations

Binning Explorer: Bin data automatically using merge and split algorithms

The Binning Explorer app supports the merge and split algorithms. For more information, see **Binning Explorer**.

Binning Explorer: Bin missing data in a separate bin

The **Binning Explorer** app supports binning missing data for a predictor in a separate <missing> bin. For more information, see **Binning Explorer**.

Binning Explorer: Load data from the command line

The Binning Explorer app supports loading data or a creditscorecard object from the command line. For more information, see **Binning Explorer**.

Corporate Credit Risk: Perform parallel simulations using creditDefaultCopula and creditMigrationCopula

The creditDefaultCopula and creditMigrationCopula classes support the 'UseParallel' property for parallel simulations when using the simulate and riskContributions functions. You can set the 'UseParallel' property when creating creditDefaultCopula or creditMigrationCopula objects only if you have Parallel Computing Toolbox[™].

creditCopula object removed

The creditCopula object is removed.

Compatibility Considerations

,	What Happens When You Use It	Use This Instead	Compatibility Considerations
creditCopula	Removed		Replace all instances of creditCopula object with a creditDefaultCopula object using creditDefaultCopula.

R2018a

Version: 1.3 New Features

Bug Fixes

Corporate Credit Risk: Calculate standard deviation and value-at-risk contributions for each counterparty in a credit portfolio

The riskContribution function for creditDefaultCopula and the riskContribution function for creditMigrationCopula support returned information for counterparty contributions for standard deviation of the losses (Std) and value at risk (VaR) at the threshold VaRLevel.

R2017b

Version: 1.2 New Features Bug Fixes Compatibility Considerations

Corporate Credit Risk: Compute regulatory capital and value-at-risk using an asymptotic single risk factor (ASRF) model

The asrf function provides an ASRF model for credit risk analysis. asrf accepts the risk characteristics of a portfolio of credit-sensitive instruments as input and computes the necessary capital using an ASRF model.

Corporate Credit Risk: Perform credit portfolio simulation with random loss given default (LGD)

Support is provided for specifying random LGD (loss given default) for creditDefaultCopula and creditMigrationCopula objects. You can now specify the LGD input argument as a NumCounterparties-by-2 matrix, where the first column contains the LGD mean values and the second column contains the LGD standard deviations. In this case, LGD values are drawn randomly from a beta distribution with the parameters provided for the defaulting counterparty.

Market Risk: Backtest expected shortfall models

The following tools support expected shortfall (ES) backtesting for table-based tests for the unconditional Acerbi-Szekely test.

- esbacktest
- summary
- runtests
- unconditionalNormal
- unconditionalT

The following tools support expected shortfall (ES) backtesting for distribution tests for normal and t distributions.

- esbacktestbysim
- summary
- runtests
- conditional
- unconditional
- quantile
- simulate

Consumer Credit Risk: Specify weights in credit scorecards using Binning Explorer

Specify weights in a credit scorecard when using the Binning Explorer app. For more information on defining weights for a creditscorecard object, see the optional name-value pair argument WeightsVar for creditscorecard.

creditCopula object renamed

The creditCopula object is renamed to the creditDefaultCopula object.

Compatibility Considerations

Object Name	What Happens When You Use This Object	Use This Object Instead	Compatibility Considerations
creditCopula	Errors	creditDefaultCopula	Replace all instances of creditCopula object with creditDefaultCopula object using creditDefaultCopula.
			Note The CounterpartyLosses property of the creditCopula object is removed in the creditDefaultCopula object. To obtain counterparty losses, use the getScenarios function.

R2017a

Version: 2.5 New Features Bug Fixes Compatibility Considerations

Corporate Credit Risk: Estimate the probability of credit rating migration based on multifactor copula model

The following tools support corporate credit portfolio analysis for credit migration simulation using a creditMigrationCopula object for copula-based simulations:

- creditMigrationCopula
- simulate
- portfolioRisk
- riskContribution
- confidenceBands
- getScenarios

Corporate Credit Risk: Quantify credit concentration risk by Herfindahl index and other concentration measures

The concentrationIndices function supports the following concentration indices:

- CR Concentration ratio
- Deciles Deciles of the portfolio weights distribution
- Gini Gini coefficient
- HH Herfindahl-Hirschman index
- HK Hannah-Kay index
- HT Hall-Tideman index
- TE Theil entropy index

Corporate Credit Risk: Model corporate default risk using Merton model

The mertonmodel and mertonByTimeSeries functions estimate the default probability using Merton's model.

creditCopula object renamed

The creditCopula object is renamed to the creditDefaultCopula object.

Compatibility Considerations

Object Name	What Happens When You Use This Object	Use This Object Instead	Compatibility Considerations
creditCopula	Warns	creditDefaultCopula	Replace all instances of creditCopula object with creditDefaultCopula object using the creditDefaultCopula constructor.
			Note The CounterpartyLosses property of creditCopula object is removed in the creditDefaultCopula object. To obtain counterparty losses, use the getScenarios function.

R2016b

Version: 2.4

New Features

Consumer Credit Risk: Binning Explorer for Credit Scorecards

Binning Explorer is an app for developing and modifying binning assignments for a creditscorecard object. For more information, see Binning Explorer.

Corporate Credit Risk: Copula-based simulation framework

The following tools support corporate credit portfolio analysis using a creditCopula object for copula-based simulations:

- creditCopula Creates a creditCopula object.
- simulate Simulates credit defaults using a creditCopula object.
- portfolioRisk Generates portfolio-level risk measurements for a creditCopula object.
- confidenceBands Generates confidence interval bands for a creditCopula object.
- riskContribution Generates risk contributions for each counterparty in the creditCopula object.

Market Risk: Value-at-Risk Backtesting Tools

Value-at-risk (VaR) is an important measure of financial risk. VaR is an estimate of how much value a portfolio can lose in a given time period with a given confidence level. VaR backtesting tools assess the accuracy of VaR models. The following VaR backtesting tools are supported:

- varbacktest Creates a varbacktest object using portfolio outcomes data and corresponding value-at-risk (VaR) data.
- bin Binomial test.
- cc Christoffersen's conditional coverage mixed test.
- cci Christoffersen's conditional coverage independence test.
- pof Kupiec's proportion of failures test.
- tbf Haas's time between exceptions independence test.
- tbfi Haas's mixed time between exceptions (independence and frequency) test.
- tl Traffic light test.
- tuff Kupiec's time until the first failure test.
- summary Summary report on the given varbacktest data.
- runtests Runs all tests and reports the final test results.