

The Iman-Conover Method

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Overview

- Mechanics of IC method
- Examples of IC method
- Choices in IC method
- Software



Mechanics of Generic IC

- Inputs
 - Sample (n x r matrix) from marginal distributions
 - Correlation matrix (r x r matrix)
- Output
 - Sample re-ordered to have the same rank correlation as a reference distribution with desired linear correlation



Mechanics of IC

- Need ability to produce samples from reference multivariate distributions
 - For r lines, sample size n, desired correlation matrix S
 - Let C be Choleski root of S
 - S=C'C, where C' = transpose(C)
 - Let M be n x r matrix of independent variables where each column has mean zero and standard deviation 1
 - Corr(M) = M'M \cong I, identity matrix (simulation error)
 - Let N = MC
 - Corr(N) = N'N = C'M'MC = C'IC = C'C = S(!)
 - Gives an easy to implement algorithm for making reference multivariate distributions



Mechanics of IC

- Fine tune reference distribution
 - If Corr(M) = M'M =: T, let T = D'D be the Choleski decomposition of T and set E=D⁻¹
 - Adjust M to N = MEC
 - N has exact desired correlation
 - Corr(N)= C'E'M'MEC = C'E'D'DEC = C'C = S
- Extra step makes IC more accurate



"Normal Copula" IC

- Apply method above with each column of matrix M equal to a random permutation (shuffling) of normal scores
 - Normal scores {Φ⁻¹(p)}, p=1/(n+1), ..., n/(n+1)
 - Scores ensure mean zero and standard deviation 1
 - Using scores reduces need to simulate normals
- Give input marginals same rank correlation as N, computed above
- Called Normal Copula IC (NCIC) because reference MV distribution is related to sample from normal copula
- NCIC is used by @Risk



Normal Copula Method

- Columns of M are random samples from N(0,1), not scores
 - Simulate random uniform u's in place of p=i/(n+1)
- Do not make adjustment to M for correlation error (no E step)
- For j=1,...,r simulate F_j-¹(u_j) in place of matching rank order with existing sample
 - NCIC approximates percentiles of F_j with order statistics from sample from F_j



"NCIC" vs Normal Copula

- Input marginals:
 - NCIC works with existing sample from marginals
 - Normal copula generates sample from marginals by inverting distribution function
 - NCIC approximate percentiles from F_j but copula computes exactly
- Scores vs. sample from reference normals
- Only NCIC has E adjustment



"NCIC" vs Normal Copula

- Outputs
 - NCIC samples have exactly correct rank correlation matrices and approximately correct linear correlation matrices
 - Normal copula samples have approximately correct rank and linear correlation matrices



Example of IC

- SCARE Tool (see last slide)
- ComprehensiveTester.xls spreadsheet
 - Shuffles sheet basic operation



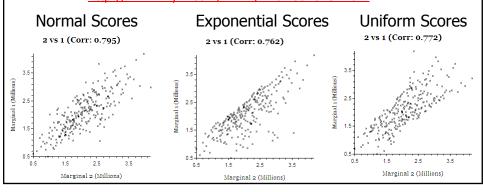
Choices in IC

- Key choice is form multivariate reference distribution; have two options
 - 1) Use Choleski-trick with different score distributions
 - Do no necessarily know distribution of resulting score marginals (e.g. sum of uniforms is not uniform)
 - Do not know form of implied copula
 - 2) Use multivariate reference sample generated in some other way
 - Many tractable alternatives
 - t-copula
 - Elliptically contoured distribution copula
 - Multivariate Laplace copula



IC with Choleski-trick: varying score distribution

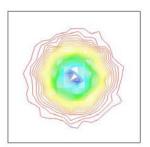
- Three examples with target correlation 80%, same marginal distributions
- http://www.mynl.com/MALT/ImanConover.html

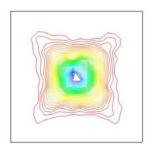




IC: other multivariate reference distribution

- Indep_Uncorr sheet
- Difference between normal and tcopulas







Software

- Working party software http://www.mynl.com/wp/default.html
 - Tools for FFTs, bivariate FFTs, Iman-Conover methods, bootstrap estimates of reserves, graphics and visualization
 - MALT Aggregate Loss Tools, for COOS Call Program 2001
 - GREAT Suite of actuarial tools, includes updated MALT; mostly VBA/Excel add-in
 - SCARE Tools for Iman-Conover
 - See also http://www.mynl.com/varview/home.html
 - VarView Visualization tools, integrates with Excel