



Computing with Bivariate Distributions

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Copulas

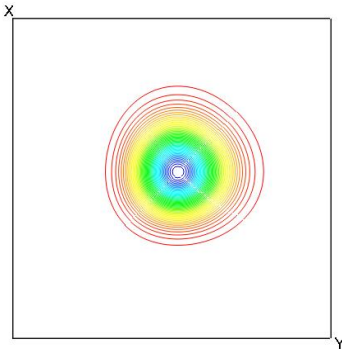
- ✱ Venter Talk and Presentation
- ✱ If $H(x, y)$ is a bivariate distribution then

$$H(x,y) = C(F(x), G(y))$$

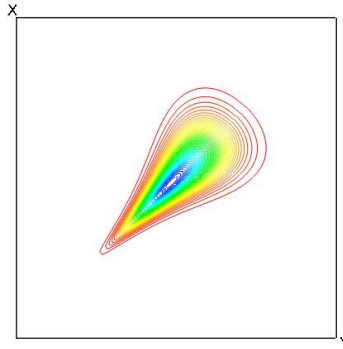
where F, G are the marginals and C is a copula

Examples of Copulas

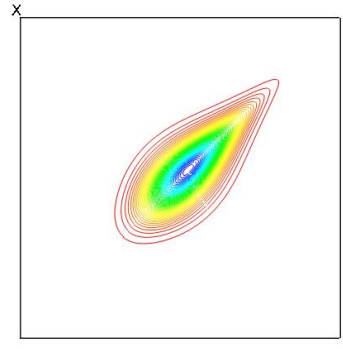
MALT - Bivariate Distribution



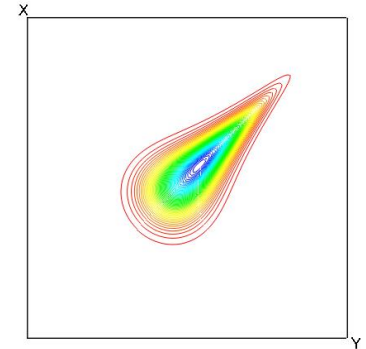
MALT - Bivariate Distribution



MALT - Bivariate Distribution



MALT - Bivariate Distribution



Independent

Clayton

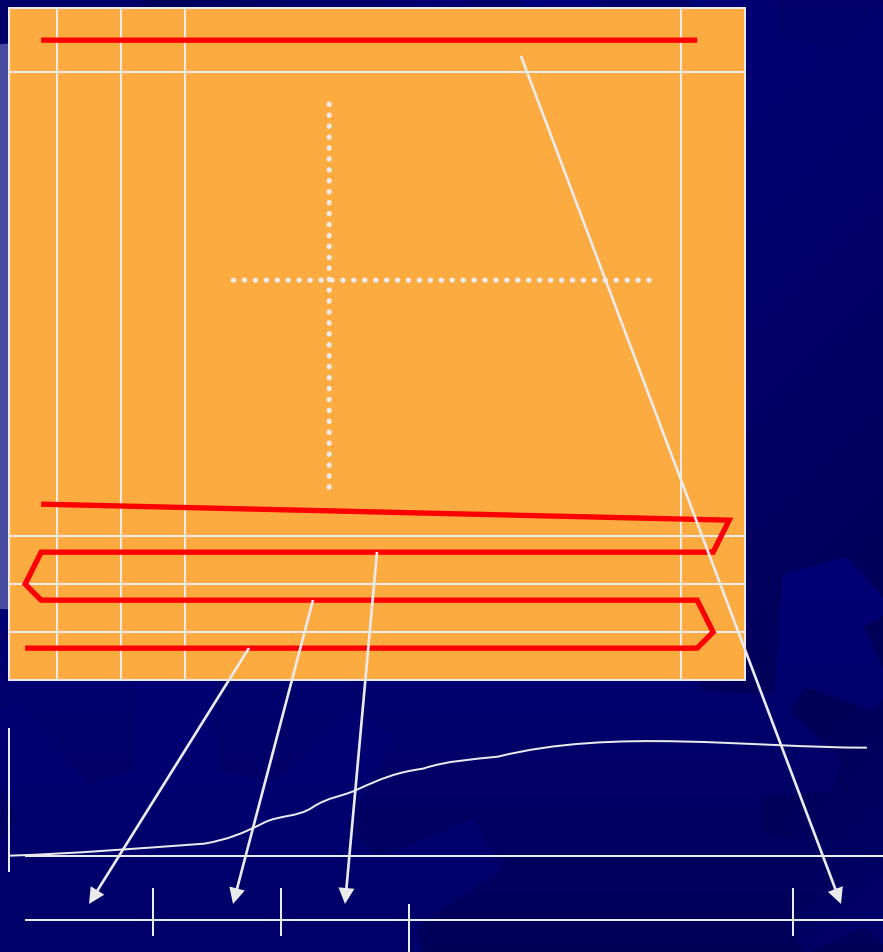
Gumbel

Venter HRT

Simulating From Copulas

- ★ Univariate: $F^{-1}(u)$, $u \sim \text{Uniform}$
- ★ Bivariate: doesn't work
- ★ Moments thought: tricky problem
- ★ Venter: invert conditional and use two step method
- ★ Normal Copula: Choleski decomposition

Simulating From Copulas



- ✦ Use space-filling curve to convert bivariate distribution into univariate distribution
- ✦ Sample off univariate distribution
- ✦ Convert back to bivariate distribution!

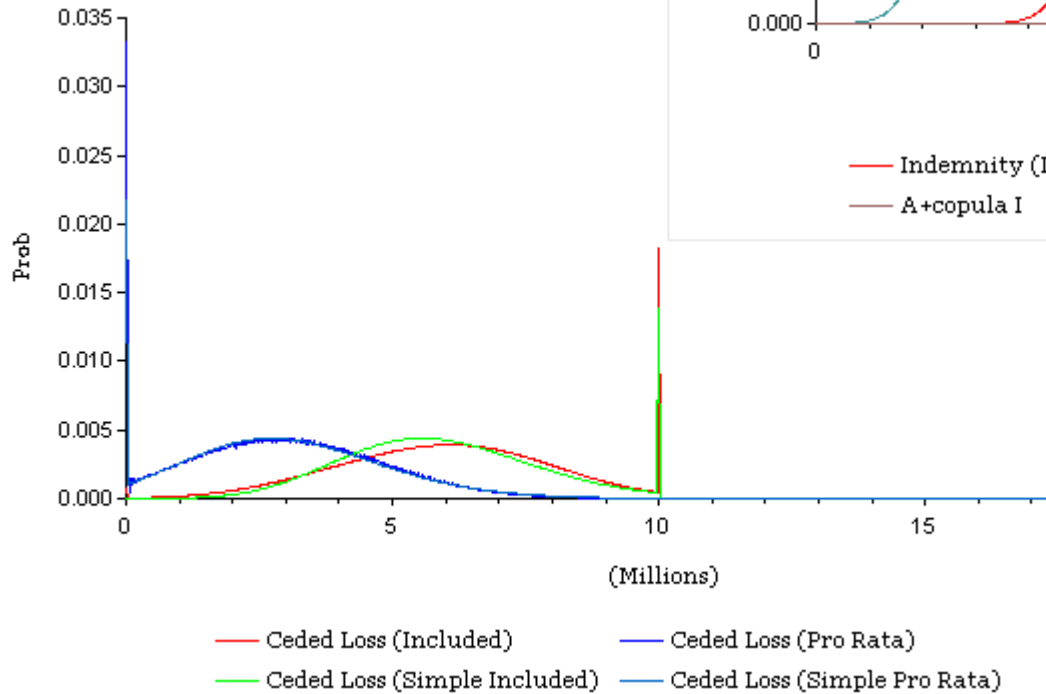
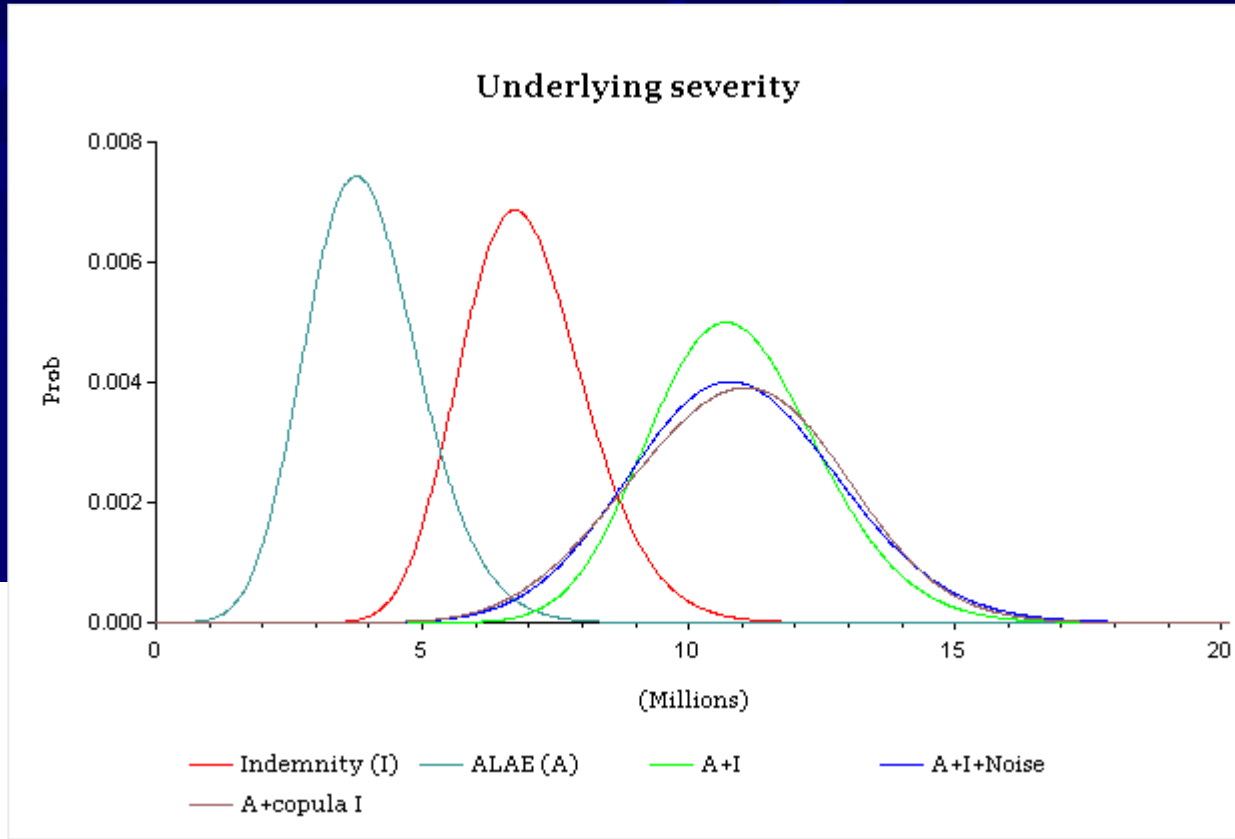
Convolution and Aggregates

- ✱ X, Y random variables with MGFs $M_X(t)$ and $M_Y(t)$, then
- ✱ $X+Y$ has MGF $M_{X+Y}(t)=M_X(t).M_Y(t)$
- ✱ If N is a frequency distribution and
 - ✱ $S = X_1 + \dots + X_N$
- ✱ Then
 - ✱ $M_S(t) = M_N(\log(M_X(t)))$
- ✱ Key Observation:
 - ✱ X, Y need not be 1-dimensional!!

Sum with Copula Dependence

- ★ If $(X, Y) \sim H(x, y)$ is a bivariate distribution
 - ★ Marginals and copula specified
 - Cat losses in D_e, M_d
 - Loss, ALAE
 - ★ M = matrix “bucketed” sample from H
 - ★ $X+Y = \text{IFFT}(\text{Diagonal}(\text{FFT2}(M)))$
 - FFT2 is two dimensional FFT
 - Not sensible, easier to sum diagonals
 - ★ Can also use FFT methods to add white-noise to increase variance

Example: Loss & ALAE



(Loss, Ultimate Loss)

★ General Problem: distribution of

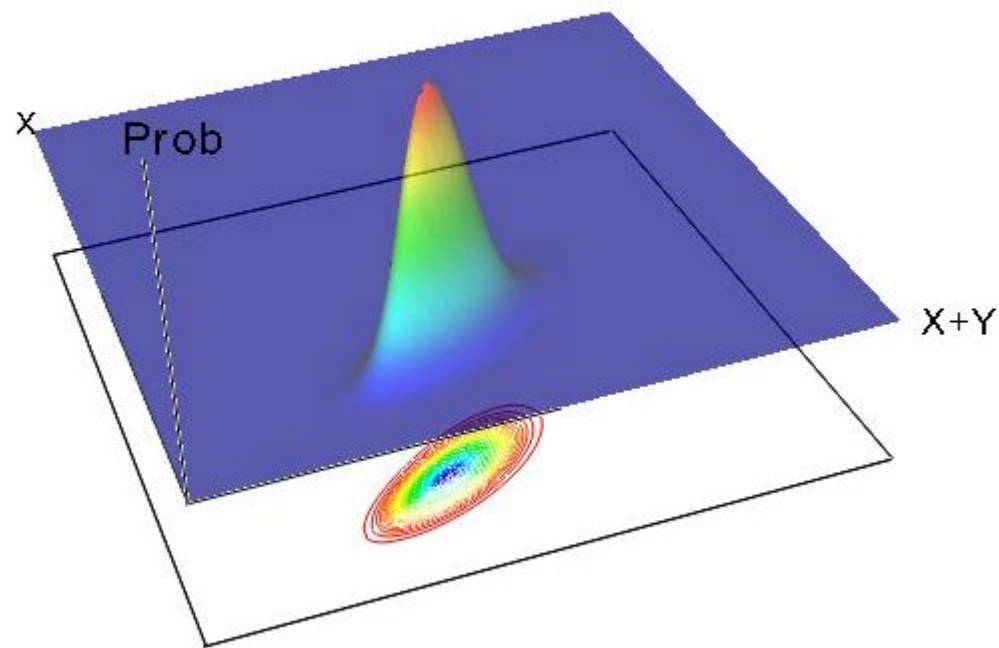
★ $(X, X+Y)$, where the X 's are perfectly correlated

- $X(1,1) + Y(0,1)$
- $X =$ incurred or paid loss
- $Y =$ bulk IBNR

★ Use FFT techniques:

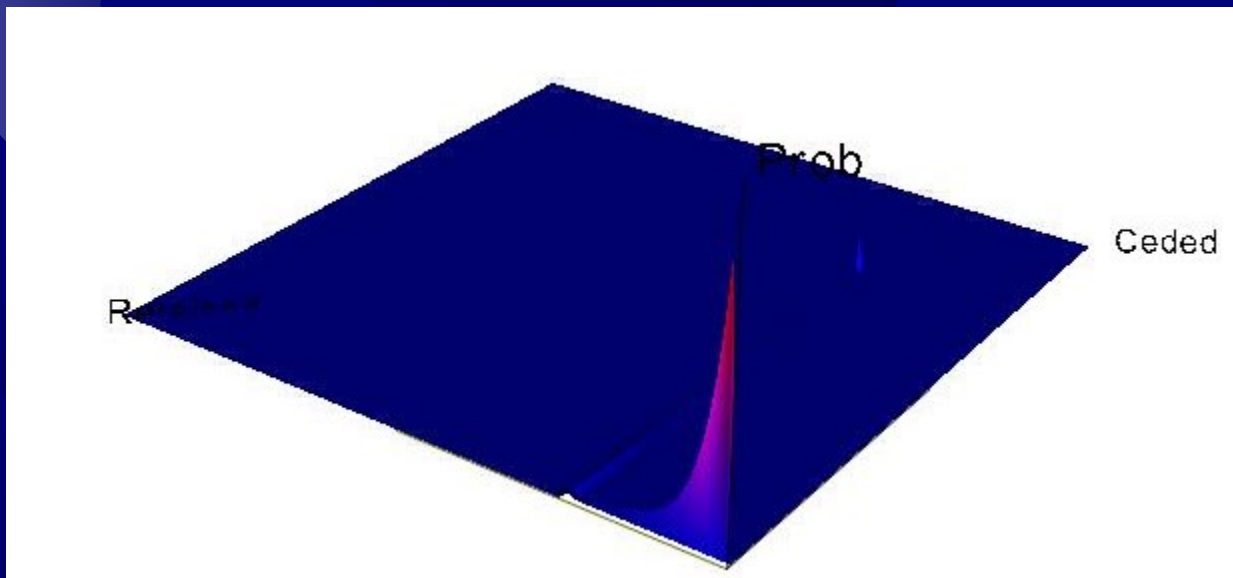
- ★ $K =$ density of X along diagonal (matrix)
- ★ $L =$ density of Y along Y axis (matrix)
- ★ $\text{IFFT2}(\text{FFT2}(K).\text{FFT2}(L))$ is required distribution

Loss, Ultimate



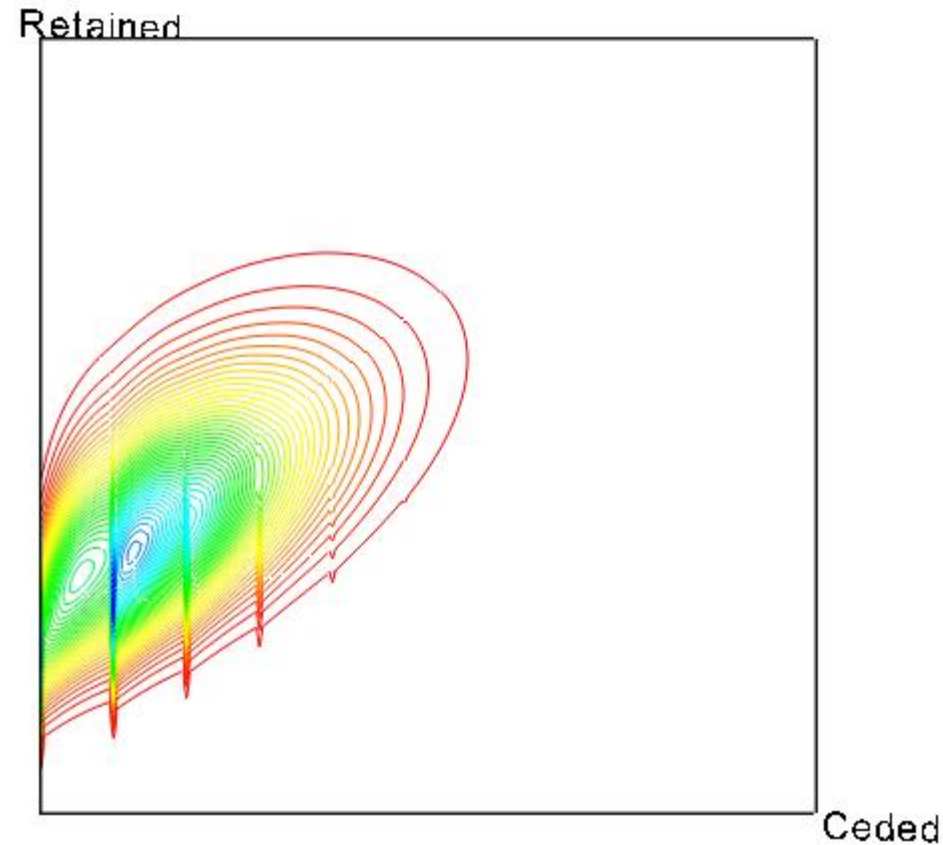
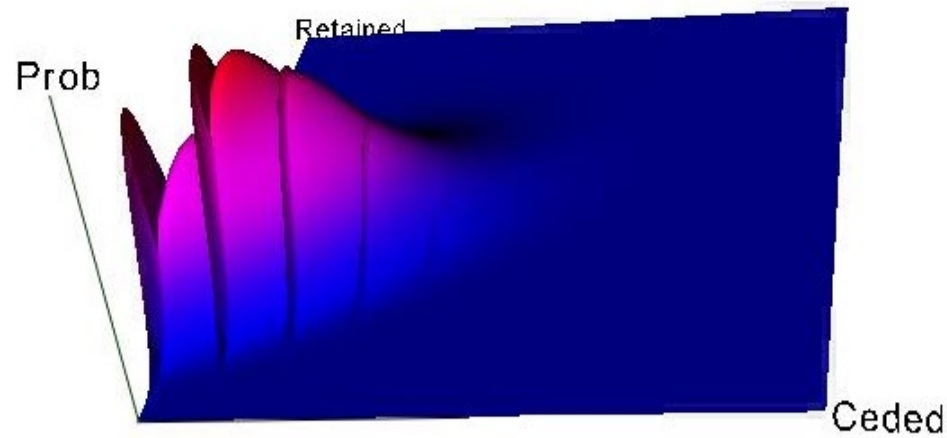
Net and Ceded

- ★ Per occurrence cover: \$1M policy limit, \$50K deductible, 750K xs 250K ceded
- ★ Per claim distribution:



Net and Ceded

- ★ Apply claim count distribution using MGFs
 - ★ 50 claims xs \$50K expected
 - ★ Neg. Binomial distribution, $\text{Var} = 150$



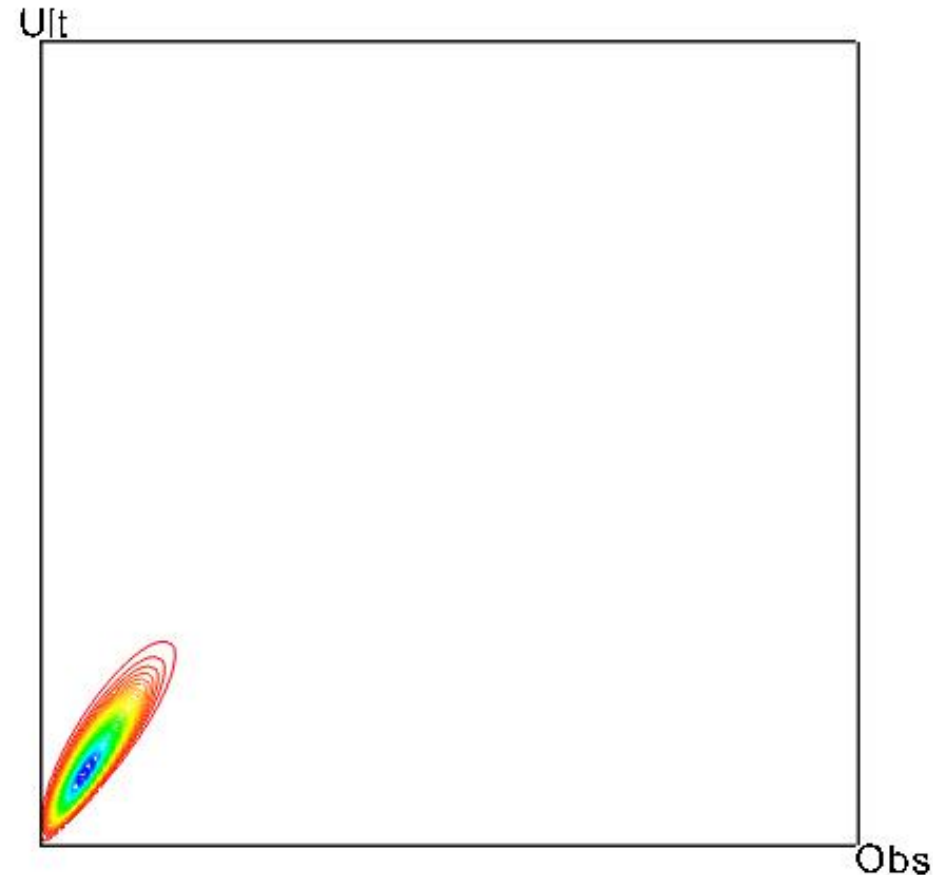
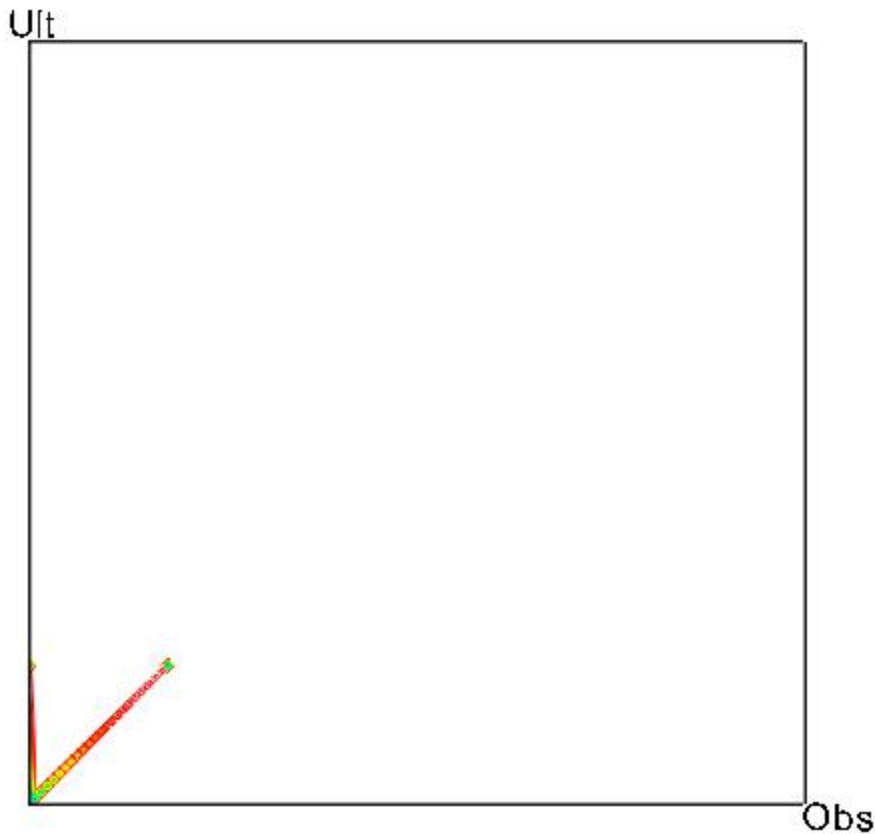
Paid Loss Development or (Loss, Ultimate) Redux

- At time n claim either paid or not paid

Per Claim Distribution

Scales are different!

Convolve with NB(50,300)



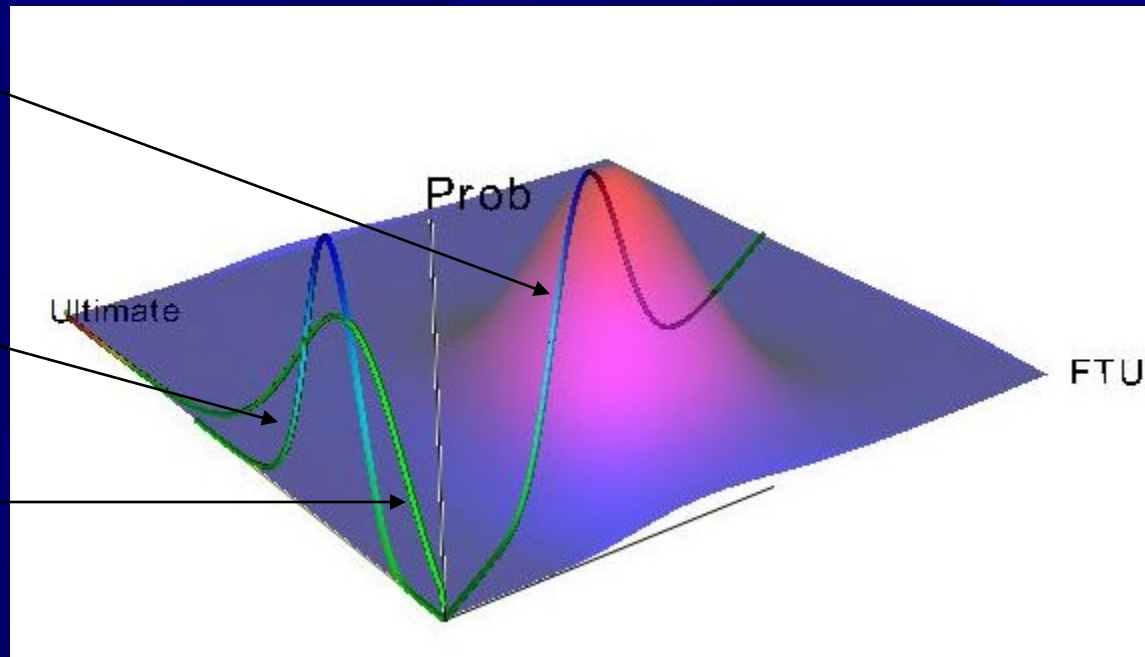
Paid Loss Bayesian Development

- ☀ Transform to Bivariate Dist of Ult vs FTU

Ult = FTU x Observed Loss

Posterior dist of ult losses
given observed losses

Prior dist of ult losses



MALT - Ultimates for 1998

