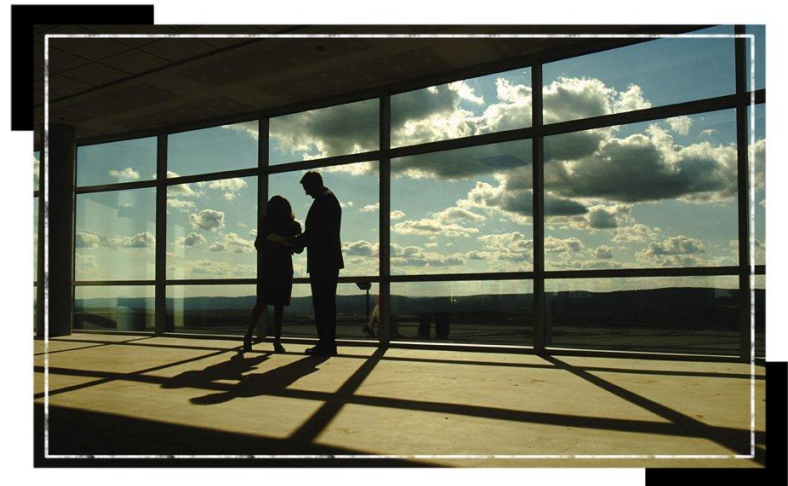


Use of Simulation Models in Pricing Health Insurance and Reinsurance Risk

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Use of Simulation Models in Pricing Health Insurance and Reinsurance Risk

- Health insurance pricing intro
- Why simulation modeling?
- Setting up the models
- Case Study 1: Excess Loss Reinsurance
- Case Study 2: Employer Stop Loss Insurance
- Summary / Questions

Health Insurance Pricing Intro

- We are trying to predict the total cost to the risk taker (e.g. health plan, employer, reinsurer) per covered life, per year
 - How many units of service will be utilized?
 - What will be the cost per unit of service?
 - What mix of services will be utilized?
 - How will utilization and cost vary by demographic category?

- We base our prediction on an analysis of past experience
 - Ideally millions of lives exposed per year analyzed
 - Similar type of risk (Medicare, Medicaid, Commercial)
 - Typically 5% of the population uses 50% of the resources

Health Insurance Pricing Intro

- In reality the risk is carved up among different risk-takers with different characteristics
 - The covered employee or subscriber (e.g. copays, deductibles, non-covered services)
 - The health plan or self-insured employer (what we typically think of as “health insurance”)
 - One or more reinsurers (may cover a share of the total risk, or only claims exceeding some threshold, or both)

- Each risk taker must quantify the expected cost and the variability unique to their share of the risk
 - Claim frequency
 - Claim size

Why Simulation Modeling?

- Given certain data points, “first dollar” healthcare costs are fairly predictable
 - Scope of services to be covered
 - Rates of payment to healthcare providers
 - Size and demographic characteristics of covered population
 - Historical data on population cost and utilization

- BUT when we start carving up the risk there is tremendous variability *within risk cells*
 - By claim amount (i.e. “large” or “excess loss” claims)
 - By disease state, treatment protocol, etc.

- Reinsurers quantify this variability in order to assure that their premium and capital levels are adequate

Why Simulation Modeling?

- Look at the relationships among group size, annual claim dollar limits, and the variability in annual claim costs
- This table shows the *coefficient of variation* in claim costs (ratio of the standard deviation to the mean)

Covered Lives	upper limit on annual claims for any one covered life										
	\$15,000	\$25,000	\$50,000	\$75,000	\$100,000	\$125,000	\$150,000	\$175,000	\$200,000	\$500,000	no limit
50	23.2%	27.7%	30.9%	38.4%	40.9%	48.2%	44.6%	51.4%	48.5%	50.4%	54.4%
100	17.2%	17.0%	23.0%	25.3%	27.7%	29.0%	35.4%	36.2%	32.2%	36.3%	40.6%
250	10.0%	10.9%	15.2%	15.6%	16.0%	16.4%	20.5%	22.1%	19.3%	28.4%	27.9%
500	8.1%	8.3%	9.5%	11.3%	11.9%	13.5%	13.6%	15.1%	14.0%	19.2%	20.6%
750	6.0%	6.5%	7.9%	9.2%	10.4%	12.2%	11.1%	12.6%	11.6%	14.6%	16.8%
1,000	5.3%	5.8%	6.3%	7.5%	9.0%	10.4%	9.0%	10.5%	10.9%	12.3%	15.3%
1,500	4.2%	4.9%	5.4%	6.2%	7.4%	8.4%	7.3%	8.2%	9.4%	9.9%	11.7%
2,000	3.7%	4.4%	4.8%	5.5%	6.3%	6.8%	6.0%	7.5%	8.4%	8.7%	10.0%
3,000	3.1%	3.5%	4.0%	4.7%	5.0%	5.1%	4.6%	5.6%	6.6%	7.0%	8.7%
4,000	2.9%	3.0%	3.4%	4.1%	4.5%	4.1%	3.9%	4.9%	5.4%	5.6%	7.0%
5,000	2.5%	2.7%	3.2%	3.7%	3.7%	3.7%	3.9%	4.7%	4.8%	5.2%	6.3%

Why Simulation Modeling?

Example of variability in cost and utilization for diabetics

Data for members with a diabetic diagnosis										
Two year period ending July 2007										
Class	Members	% of Total Members	# ER Visits	# Office Visits	# Admits	ALOS	Average Diseases per Member	Average Cost per Member	Risk Index	Adjusted Risk Index
7	46	0.6%	3.9	35.0	4.5	20.9	5.3	\$422,829	96.65	154.13
6	166	2.3%	3.6	34.0	2.9	9.4	4.7	\$148,281	64.34	107.73
5	374	5.2%	1.8	25.7	1.6	7.1	4.0	\$67,774	40.92	73.23
4	1,183	16.5%	1.3	21.1	0.8	4.2	3.2	\$30,107	24.43	48.25
3	1,548	21.6%	0.8	14.9	0.3	3.8	2.5	\$14,342	15.61	33.64
2	1,592	22.2%	0.5	10.9	0.1	4.7	2.2	\$7,304	11.56	26.58
1	2,263	31.6%	0.2	6.6	0.1	4.1	1.8	\$2,300	7.37	19.78
	7,172	100.0%	0.8	13.6	0.4	4.6	2.5	\$20,087	16.53	34.66

Why Simulation Modeling?

- The risks we are modeling do not have a single distribution with observable parameters
 - In reality they are a “convolution” – a complex aggregation of distributions
 - So the parameters – *especially the variability* – are difficult or impossible to determine without simulation
- For example, consider the risk for total population claims greater than \$100,000
 - How many ways are there to accumulate \$100,000 in claims for a group?
 - 4 claims at \$25,000? 100 claims at \$1,000? 1 claim at \$100,000?
 - The combinations are almost infinite . . . and this is only one risk cell!
- We solve this problem via simulation by generating an empirical (experience-based) distribution of the risk and then exploring the variability of the modeled risk
 - More on this in the case studies . . .

Setting Up the Simulation Model

- Our goal is to create a mathematical model of the *range of financial results* to the risk-taker(s)
- This is best accomplished through simulation
 - Remember we have a convolution of distributions that cannot be easily represented mathematically
 - Need the variability in results as well as the expected value
 - Provides a means to test the impact of specific model parameters (e.g. coverage terms, pricing assumptions)
 - Capital and surplus targets can be expressed with respect to various percentiles in the distribution of results (e.g. “funding to the 97.5th percentile”)
- Requires appropriate stochastic modeling software: @Risk

Setting Up the Simulation Model: Key Steps

- Describe the anticipated block of business via a series of mathematical models and assumptions (a “model office” approach)
- Using historical data and “manual rates” where necessary, develop realistic empirical distributions of loss results, identifying key risk factors
- Replace the expected value of each key risk factor with a statistical distribution that describes the mean and variability associated with the risk
- Each input distribution will be sampled to arrive at a projection of total loss costs

Setting Up the Simulation Model: Special Considerations

- Process risk vs. parameter risk
 - Process risk reflects the inherent variability in empirical data which is quantified via the selected distributions and risk factors
 - Parameter risk reflects the possibility that the pricing assumptions and underwriting practices fundamentally misrepresent the actual risk as experience emerges
 - Need to model the impact of both!
 - A simple example: replace the mean of the loss distribution with another distribution
- Model efficiency – look for ways to minimize inputs
- Number of simulations
 - Required number will depend on the degree of variability in results
 - Want enough so that results are not significantly different when the process is repeated

Case Study #1: Excess Loss (“XOL”) Reinsurance

■ Definitions

- Generally refers to total claim dollars for one individual in excess of some large annual dollar amount (e.g. \$100K, \$500K, \$1M, etc.)
 - The “tail” of the healthcare claim distribution → variability
 - Even large health plans may lay off some or all of this risk
 - Substantial variability implies substantial capital and surplus requirements
 - Part of overall enterprise risk management strategy
 - Variability can only be reduced by aggregating large numbers of covered lives → This is the role of the Reinsurer
-
- Client was a regional health plan looking to restructure the reinsurance of its XOL risk

Case Study #1: Purpose Of Building the Model

- To understand risks assumed by the new reinsurance entity by simulating large claim experience
- To evaluate the impact of different reinsurance structures on capital needs and long term stability
- To evaluate changes in enterprise risk capital relative to changes in required Risk Based Capital
 - There may be tradeoffs: e.g., big change in enterprise risk capital versus change in RBC with a particular structure
 - May get more risk stability but at the expense of proportionately more RBC
- To develop a net premium target for reinsurer negotiations
- To develop a framework to provide assistance to potential reinsurers

Case Study #1: Stochastic Modeling Basic Steps

- Build a framework for the business to be modeled (simple “model office” structure)
- Develop representative claims distributions for each model office cell
 - e.g., service type, product type, client size, geography
- Consider parameter risk
 - A function of the size of the covered population
- Build a stochastic model using @Risk
- Run sets of simulations under various scenarios
 - Simulating large claim frequency and conditional claim amounts
- Run tests on the stability of the simulation results
 - Did we run enough scenarios?

Case Study #1: Empirical Data

- Obtained detailed claims and eligibility data for three most recent calendar years
- Data were analyzed by service type, risk type, and client size
- Developed claim continuance tables or distributions
- Distributions specify claim frequency and average claim size in 59 different size bands (\$0 claims up to \$1 million plus)

Case Study #1

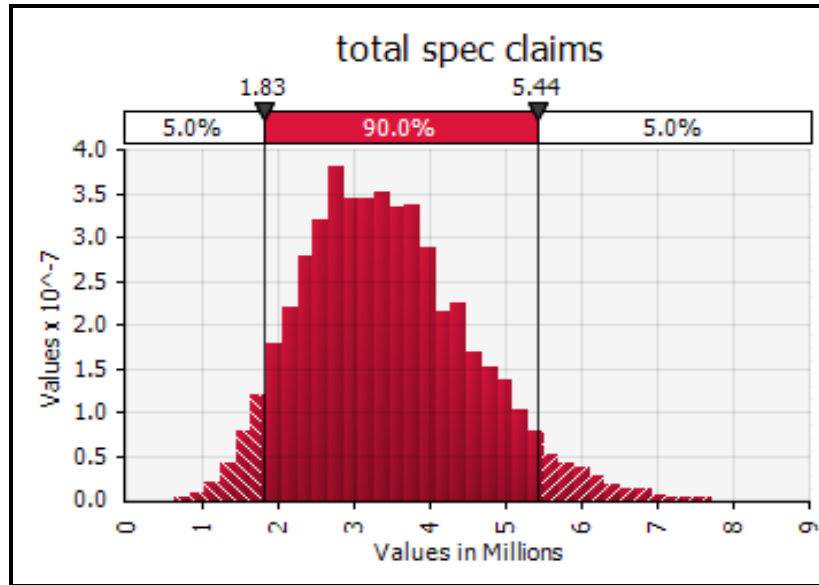
Sample Claim Distributions

claim size band	average claim size	frequency of having annual claims in the given band				
		inpatient hospital	outpatient hospital	total hospital	physician	total all categories
nonclaimant	\$0	0.950399	0.717809	0.699279	0.220507	0.191002
0-100	\$50	0.000068	0.071336	0.067842	0.115684	0.096911
100-200	\$150	0.000190	0.041629	0.038513	0.117807	0.095338
200-300	\$250	0.000063	0.021690	0.019469	0.080555	0.065361
300-400	\$350	0.000127	0.014709	0.013048	0.057464	0.050519
400-500	\$450	0.000118	0.013602	0.012222	0.042218	0.040571
500-600	\$550	0.000811	0.011257	0.010838	0.032361	0.031312
600-700	\$650	0.000260	0.010160	0.009260	0.025546	0.026823
700-800	\$750	0.000087	0.007954	0.007193	0.022470	0.023066
800-900	\$850	0.000327	0.006930	0.006343	0.018495	0.019801
900-1000	\$950	0.001151	0.006298	0.006388	0.016384	0.018139
1000-1250	\$1,125	0.002687	0.012286	0.013095	0.033506	0.038973
1250-1500	\$1,375	0.000842	0.009035	0.008417	0.026909	0.030235
1500-1750	\$1,625	0.001044	0.007182	0.006780	0.020954	0.025865
1750-2000	\$1,875	0.001390	0.005641	0.005869	0.017564	0.020652
2000-2500	\$2,250	0.002323	0.008015	0.009059	0.028516	0.034270
2500-3000	\$2,750	0.002444	0.006704	0.007638	0.020987	0.024875
3000-3500	\$3,250	0.002719	0.005014	0.006567	0.016729	0.018828
3500-4000	\$3,750	0.002466	0.003124	0.004727	0.013025	0.015211
4000-4500	\$4,250	0.002422	0.002297	0.003944	0.009685	0.012781
4500-5000	\$4,750	0.002934	0.002049	0.003778	0.008026	0.011728
5000-6000	\$5,500	0.003518	0.003300	0.006287	0.012347	0.017304
6000-7000	\$6,500	0.002310	0.002138	0.004259	0.009224	0.013363
7000-8000	\$7,500	0.001481	0.001864	0.003305	0.006131	0.010461
8000-9000	\$8,500	0.001270	0.001224	0.002295	0.004359	0.007943
9000-10000	\$9,500	0.001059	0.001001	0.001859	0.003585	0.006691
10000-12500	\$11,250	0.003573	0.001824	0.004582	0.005501	0.011343
12500-15000	\$13,750	0.002665	0.000857	0.003664	0.003174	0.008095
15000-17500	\$16,250	0.001304	0.000765	0.002324	0.002457	0.005760
17500-20000	\$18,750	0.001398	0.000537	0.001884	0.001608	0.004516
20000-25000	\$22,500	0.001481	0.000573	0.002437	0.001949	0.005815
25000-30000	\$27,500	0.000821	0.000268	0.001273	0.001332	0.003961
30000-35000	\$32,500	0.000816	0.000210	0.001067	0.000784	0.002425
35000-40000	\$37,500	0.000568	0.000159	0.000756	0.000599	0.001963
40000-45000	\$42,500	0.000394	0.000077	0.000435	0.000334	0.001376
45000-50000	\$47,500	0.000302	0.000122	0.000505	0.000276	0.000908
50000-60000	\$55,000	0.000600	0.000154	0.000742	0.000337	0.001395
60000-70000	\$65,000	0.000330	0.000094	0.000504	0.000307	0.001069
70000-80000	\$75,000	0.000210	0.000033	0.000315	0.000101	0.000731
80000-90000	\$85,000	0.000220	0.000010	0.000233	0.000053	0.000590
90000-100000	\$95,000	0.000120	0.000013	0.000144	0.000057	0.000404
100000-125000	\$112,500	0.000178	0.000035	0.000231	0.000045	0.000655
125000-150000	\$137,500	0.000154	0.000022	0.000241	0.000029	0.000248
150000-175000	\$162,500	0.000104	0.000000	0.000084	0.000010	0.000216
175000-200000	\$187,500	0.000031	0.000000	0.000054	0.000010	0.000127
200000-225000	\$212,500	0.000059	0.000000	0.000049	0.000000	0.000077
225000-250000	\$237,500	0.000027	0.000000	0.000039	0.000000	0.000073
250000-275000	\$262,500	0.000017	0.000000	0.000027	0.000000	0.000039
275000-300000	\$287,500	0.000010	0.000000	0.000006	0.000000	0.000023
300000-350000	\$325,000	0.000023	0.000000	0.000039	0.000000	0.000053
350000-400000	\$375,000	0.000041	0.000000	0.000041	0.000000	0.000026
400000-450000	\$425,000	0.000019	0.000000	0.000009	0.000000	0.000026
450000-500000	\$475,000	0.000010	0.000000	0.000019	0.000000	0.000015
500000-600000	\$550,000	0.000003	0.000000	0.000003	0.000000	0.000006
600000-700000	\$650,000	0.000006	0.000000	0.000006	0.000000	0.000013
700000-800000	\$750,000	0.000005	0.000000	0.000005	0.000000	0.000009
800000-900000	\$850,000	0.000003	0.000000	0.000003	0.000000	0.000006
900000-1000000	\$950,000	0.000002	0.000000	0.000002	0.000000	0.000003
1000000+	\$1,095,572	0.000000	0.000000	0.000000	0.000000	0.000010

Case Study #1: Detailed Modeling Approach

- Model claims over and under XOL threshold separately
 - This is the convolution of distributions problem to be solved via simulation
- Focus is on simulating XOL or large claim experience
 - Use a Poisson distribution to model large claim frequency
 - Expected value of the Poisson is defined by the claim distributions built from the empirical data
 - Large claim size (given the occurrence of a large claim) is generated from these same distributions
 - Use a conditional distribution for efficiency (only need to model claim amounts in excess of the XOL threshold)
- Fit a normal distribution to claims under the XOL threshold so we can simulate total population claim costs
 - Expected value is again defined by the claim distribution
 - Variance is a function of 1) population size; and 2) the XOL threshold
 - Understand how reinsurance impacts the risk that is retained by the health plan
- Run 10,000 trials for each simulation

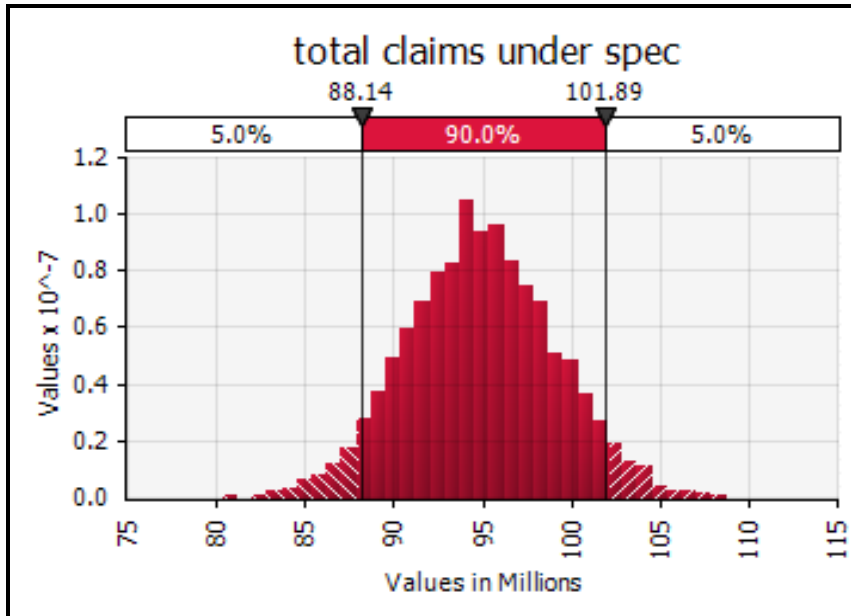
Case Study #1: @RISK Sample Output for a \$150,000 Limit with 35,000 Lives



Simulation Summary Information	
Workbook	distributions_model.xls
Name	
Number of Simulations	1
Number of Iterations	10000
Number of Inputs	252
Number of Outputs	3
Sampling Type	Monte Carlo
Simulation Start Time	9/22/08 10:23:16
Simulation Duration	00:00:44
Random # Generator	Mersenne Twister
Random Seed	128362817

Summary Statistics for total spec claims			
Statistics		Percentile	
Minimum	\$433,830	5%	\$1,828,971
Maximum	\$8,547,406	10%	\$2,102,149
Mean	\$3,448,711	15%	\$2,317,444
Std Dev	\$1,104,909	20%	\$2,492,141
Variance	1.22082E+12	25%	\$2,647,087
Skewness	0.497811821	30%	\$2,782,563
Kurtosis	3.132339054	35%	\$2,919,826
Median	\$3,349,348	40%	\$3,059,680
Mode	\$2,714,632	45%	\$3,205,125
Left X	\$1,828,971	50%	\$3,349,348
Left P	5%	55%	\$3,490,381
Right X	\$5,437,534	60%	\$3,642,509
Right P	95%	65%	\$3,793,436
Diff X	\$3,608,563	70%	\$3,948,887
Diff P	90%	75%	\$4,132,487
#Errors	0	80%	\$4,364,669
Filter Min	Off	85%	\$4,610,239
Filter Max	Off	90%	\$4,950,569
#Filtered	0	95%	\$5,437,534

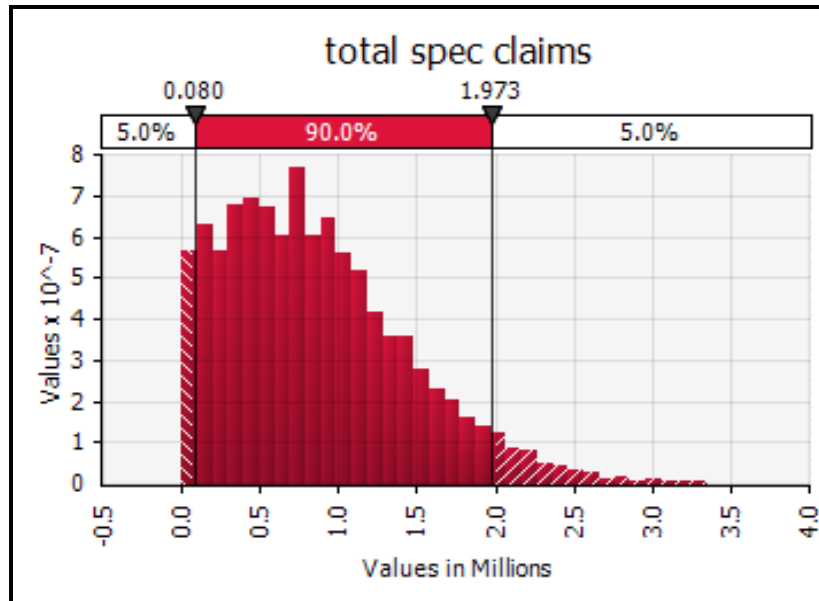
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Simulation Duration	00:00:44
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Random Seed	128362817

Summary Statistics for total claims under spec			
Statistics		Percentile	
Minimum	\$78,683,807	5%	\$88,137,036
Maximum	\$112,103,382	10%	\$89,591,209
Mean	\$94,930,767	15%	\$90,546,483
Std Dev	\$4,197,245	20%	\$91,387,094
Variance	1.76169E+13	25%	\$92,084,074
Skewness	0.021650898	30%	\$92,717,995
Kurtosis	3.009773192	35%	\$93,340,481
Median	\$94,860,472	40%	\$93,902,175
Mode	\$94,134,298	45%	\$94,371,466
Left X	\$88,137,036	50%	\$94,860,472
Left P	5%	55%	\$95,404,077
Right X	\$101,894,226	60%	\$95,931,694
Right P	95%	65%	\$96,502,803
Diff X	\$13,757,190	70%	\$97,088,713
Diff P	90%	75%	\$97,772,748
#Errors	0	80%	\$98,449,413
Filter Min	Off	85%	\$99,316,437
Filter Max	Off	90%	\$100,382,383
#Filtered	0	95%	\$101,894,226

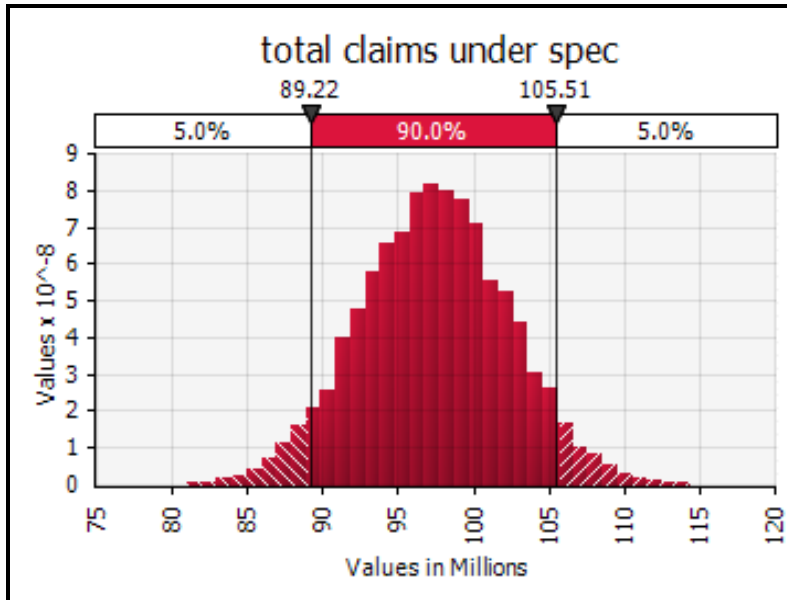
Case Study #1: @RISK Sample Output for a \$400,000 Limit with 35,000 Lives



Simulation Summary Information	
Workbook Name	distributions_model.xls
Number of Simulations	1
Number of Iterations	10000
Number of Inputs	252
Number of Outputs	3
Sampling Type	Monte Carlo
Simulation Start Time	9/30/08 12:51:10
Simulation Duration	00:00:56
Random # Generator	Mersenne Twister
Random Seed	2056025269

Summary Statistics for total spec claims			
Statistics		Percentile	
Minimum	\$0	5%	\$79,580
Maximum	\$3,946,844	10%	\$170,953
Mean	\$864,320	15%	\$264,100
Std Dev	\$585,579	20%	\$331,391
Variance	3.42903E+11	25%	\$401,501
Skewness	0.839407704	30%	\$475,729
Kurtosis	3.664581806	35%	\$560,129
Median	\$774,228	40%	\$632,050
Mode	\$0	45%	\$704,504
Left X	\$79,580	50%	\$774,228
Left P	5%	55%	\$851,501
Right X	\$1,972,535	60%	\$945,572
Right P	95%	65%	\$1,019,800
Diff X	\$1,892,954	70%	\$1,101,501
Diff P	90%	75%	\$1,209,671
#Errors	0	80%	\$1,335,052
Filter Min	Off	85%	\$1,476,653
Filter Max	Off	90%	\$1,672,068
#Filtered	0	95%	\$1,972,535

Case Study #1: @RISK Sample Output for a \$400,000 Limit with 35,000 Lives



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Number of Iterations	10000
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Sampling Type	Monte Carlo
Simulation Start Time	9/30/08 12:51:10
Simulation Duration	00:00:56
Random # Generator	Mersenne Twister
Random Seed	2056025269

Summary Statistics for total claims under spec			
Statistics		Percentile	
Minimum	\$78,046,818	5%	\$89,220,184
Maximum	\$117,443,849	10%	\$91,185,897
Mean	\$97,449,004	15%	\$92,260,402
Std Dev	\$4,943,285	20%	\$93,274,789
Variance	2.44361E+13	25%	\$94,090,670
Skewness	-0.016172513	30%	\$94,843,034
Kurtosis	2.979466945	35%	\$95,575,101
Median	\$97,444,286	40%	\$96,242,341
Mode	\$97,295,398	45%	\$96,855,440
Left X	\$89,220,184	50%	\$97,444,286
Left P	5%	55%	\$98,088,169
Right X	\$105,507,237	60%	\$98,713,995
Right P	95%	65%	\$99,339,628
Diff X	\$16,287,052	70%	\$100,071,667
Diff P	90%	75%	\$100,744,103
#Errors	0	80%	\$101,654,737
Filter Min	Off	85%	\$102,619,887
Filter Max	Off	90%	\$103,779,162
#Filtered	0	95%	\$105,507,237

Case Study #1: Sample Results and Implications

		XOL limit \$150,000	XOL limit \$200,000	XOL limit \$300,000	XOL limit \$400,000	XOL limit \$150,000 capped at \$400,000
Claims over the XOL limit						
	Mean	\$3,448,711	\$2,433,703	\$1,407,004	\$862,065	\$2,597,991
	Standard Deviation	\$1,104,909	\$970,433	\$759,320	\$590,973	\$678,717
	Coefficient of Variation	32.0%	39.9%	54.0%	68.6%	26.1%
	75th percentile	\$4,132,487	\$3,033,870	\$1,859,419	\$1,209,671	\$3,036,001
	85th percentile	\$4,610,239	\$3,443,046	\$2,191,721	\$1,467,330	\$3,306,082
	95th percentile	\$5,437,534	\$4,183,948	\$2,793,596	\$1,965,372	\$3,766,205
	Minimum	\$433,830	\$181,726	\$0	\$0	\$612,969
	Maximum	\$8,547,406	\$7,941,955	\$5,569,368	\$4,110,943	\$5,454,084
Claims below the XOL limit						
	Mean	\$94,930,767	\$95,958,424	\$96,968,427	\$97,587,025	\$94,893,764
	Standard Deviation	\$4,197,245	\$4,558,732	\$4,782,731	\$4,914,901	\$4,254,639
	Coefficient of Variation	4.4%	4.8%	4.9%	5.0%	4.5%
	75th percentile	\$97,772,748	\$99,022,670	\$100,199,718	\$100,865,265	\$97,758,682
	85th percentile	\$99,316,437	\$100,667,285	\$101,918,752	\$102,628,136	\$99,266,819
	95th percentile	\$101,894,226	\$103,455,799	\$104,874,648	\$105,650,456	\$101,921,531
	Minimum	\$78,683,807	\$78,951,060	\$80,178,166	\$81,250,558	\$79,000,934
	Maximum	\$112,103,382	\$112,041,432	\$116,600,400	\$118,820,811	\$111,131,916
<i>modeling is based upon 3 years historical data and assumes 35,000 covered lives</i>						

Case Study #2: Employer Stop Loss Insurance

■ Definitions

- Many employers (nearly all large employers) “self-fund” the healthcare benefits for their employees
 - But they limit their risk exposure by purchasing Employer Stop Loss Insurance
 - Specific Stop Loss – covers the risk that any one individual’s claims exceed a given \$\$ threshold (the “Specific Deductible”) in one year
 - Aggregate Stop Loss – covers the risk that total claims for the group exceed a given \$\$ threshold (the “Aggregate Attachment Point”) in one year
-
- Client was an underwriter of Employer Stop Loss Insurance looking to participate in the risk by forming a Captive Reinsurer

Case Study #2: Purpose Of Building the Model

- Remember our “convolution” problem - a complex aggregation of distributions
 - Risks are intertwined because aggregate risk depends on the specific deductible size
 - We also have a portfolio of groups with different sizes and specific deductibles, and thus different risk distributions
 - Simulation is the only way to quantify the risks and understand the inter-relationships
- Determine capital and surplus requirements which are defined based on outcome likelihoods
- Regulatory requirements
- Provides a benchmark for monitoring performance going forward

Case Study #2: Empirical Data

- Data elements specific to the covered employer groups
 - Range of group sizes and demographics
 - Range of specific deductibles
 - Aggregate as well as specific risk? (not all groups have both)
 - Rate model parameters
 - Historical results including variability across the above factors
- Certain “environmental” or less technical factors
 - Plans for business volume and growth
 - Regulatory environment / requirements
 - Expenses
- Developed customized claim continuance tables or distributions
- Distributions specify claim frequency and average claim size in 198 different size bands (\$0 claims up to \$2.5 million plus)

Case Study #2: Employer Stop Loss Insurance

Sample of customized (empirical) claim distribution used in simulation

distribution cell	claim size range	claim frequency	average claim size	distribution cell	claim size range	claim frequency	average claim size	distribution cell	claim size range	claim frequency	average claim size
1	non-claimant	0.201823	\$0	91	24000-25000	0.002340	\$24,483	183	1250000-1300000	0.000004	\$1,271,040
2	>0-100	0.037605	\$53	92	25000-26000	0.002127	\$25,473	184	1300000-1350000	0.000002	\$1,331,602
3	100-200	0.043446	\$150	93	26000-27000	0.001987	\$26,474	185	1350000-1400000	0.000003	\$1,376,906
4	200-300	0.038102	\$249	94	27000-28000	0.001852	\$27,483	186	1400000-1450000	0.000003	\$1,427,281
5	300-400	0.033471	\$349	95	28000-29000	0.001705	\$28,462	187	1450000-1500000	0.000003	\$1,474,188
6	400-500	0.030051	\$449	96	29000-30000	0.001553	\$29,473	188	1500000-1600000	0.000003	\$1,542,129
7	500-600	0.026787	\$548	97	30000-32500	0.003470	\$31,179	189	1600000-1700000	0.000002	\$1,651,998
8	600-700	0.024324	\$648	98	32500-35000	0.002812	\$33,692	190	1700000-1800000	0.000003	\$1,763,366
9	700-800	0.022083	\$748	99	35000-37500	0.002468	\$36,201	191	1800000-1900000	0.000002	\$1,853,297
10	800-900	0.020319	\$848	100	37500-40000	0.002061	\$38,699	192	1900000-2000000	0.000001	\$1,948,248
11	900-1000	0.018718	\$948	101	40000-42500	0.001799	\$41,217	193	2000000-2100000	0.000002	\$2,054,121
12	1000-1100	0.017265	\$1,047	102	42500-45000	0.001579	\$43,705	194	2100000-2200000	0.000001	\$2,161,123
13	1100-1200	0.016062	\$1,147	103	45000-47500	0.001304	\$46,170	195	2200000-2300000	0.000000	\$2,266,564
14	1200-1300	0.014899	\$1,247	104	47500-50000	0.001222	\$48,674	196	2300000-2400000	0.000001	\$2,313,825
15	1300-1400	0.013983	\$1,347	105	50000-52500	0.001068	\$51,191	197	2400000-2500000	0.000001	\$2,439,469
16	1400-1500	0.013069	\$1,447	106	52500-55000	0.000967	\$53,716	198	>2500000	0.000003	\$3,546,305
*	*	*	*	*	*	*	*				
*	*	*	*	*	*	*	*				
*	*	*	*	*	*	*	*				

Case Study #2: Detailed Modeling Approach

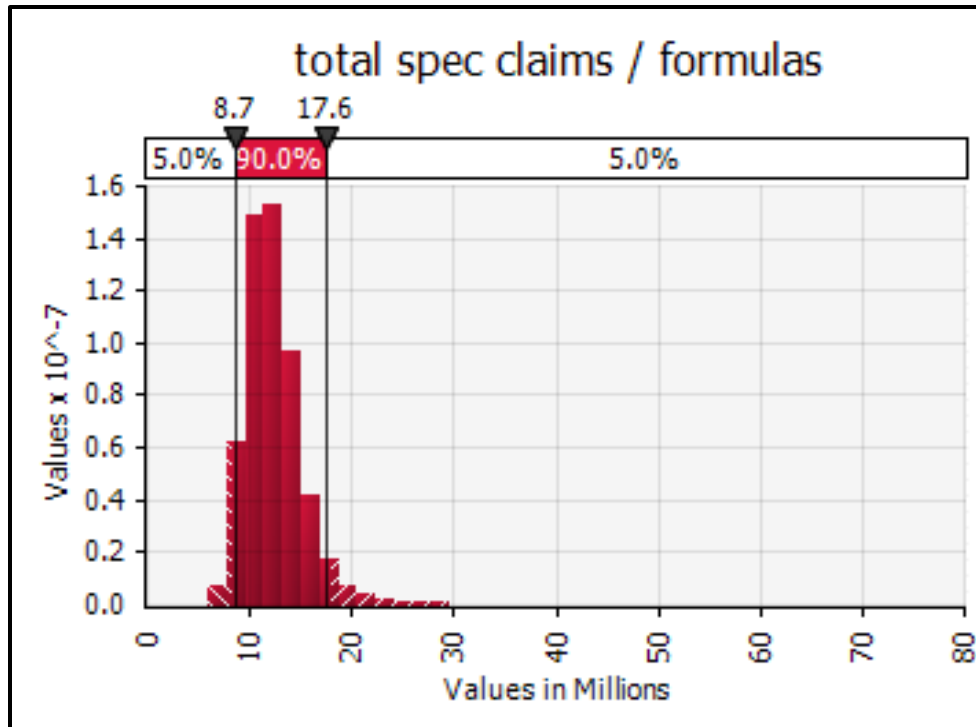
- Model office approach simulates Specific and Aggregate claims *separately for each group* (considers size and specific deductible)

- Specific Stop Loss modeling (individual claims)
 - Use a Poisson distribution to model large claim frequency
 - Expected value of the Poisson is defined by the claim distributions built from the empirical data
 - Large claim size (given the occurrence of a large claim) is generated from these same distributions
 - Use a conditional distribution for efficiency (only need to model claim amounts in excess of the specific deductible)

- Aggregate Stop Loss modeling (group claims)
 - Fit a normal distribution to claims under the specific deductible
 - Expected value is again defined by the claim distribution
 - The Aggregate Attachment Point is defined with respect to this expected value (e.g. 125%)
 - You have an Aggregate claim whenever simulated claims exceed the Attachment Point
 - Variance is a function of 1) population size; and 2) the specific deductible

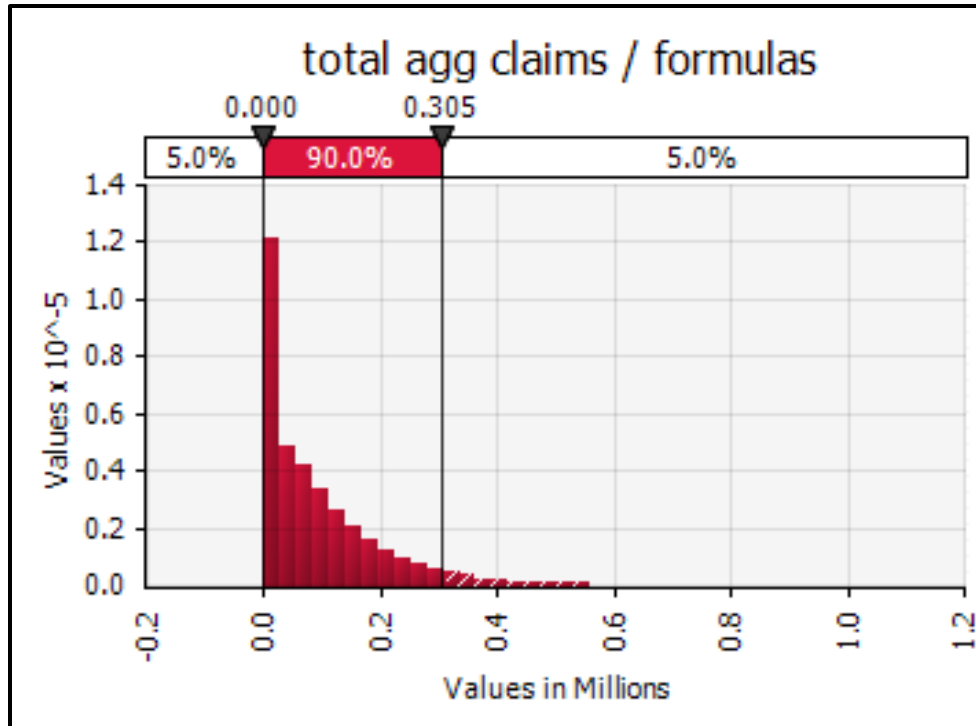
- Run 10,000 trials for each simulation

Case Study #2: @Risk Sample Output for Specific Stop Loss



Summary Statistics for total spec claims / formula			
Statistics		Percentile	
Minimum	\$5,939,536	5%	\$8,682,237
Maximum	\$79,158,970	10%	\$9,302,049
Mean	\$12,496,243	15%	\$9,789,260
Std Dev	\$3,259,370	20%	\$10,177,709
Variance	1.06235E+13	25%	\$10,510,568
Skewness	3.463568093	30%	\$10,823,809
Kurtosis	38.48036088	35%	\$11,117,011
Median	\$12,035,122	40%	\$11,428,281
Mode	\$12,059,220	45%	\$11,740,217
Left X	\$8,682,237	50%	\$12,035,122
Left P	5%	55%	\$12,331,343
Right X	\$17,577,595	60%	\$12,655,016
Right P	95%	65%	\$13,022,898
Diff X	\$8,895,358	70%	\$13,412,477
Diff P	90%	75%	\$13,821,204
#Errors	0	80%	\$14,293,867
Filter Min	Off	85%	\$14,948,585
Filter Max	Off	90%	\$15,922,062
#Filtered	0	95%	\$17,577,595

Case Study #2: @Risk Sample Output for Aggregate Stop Loss



Summary Statistics for total agg claims / formulas			
Statistics		Percentile	
Minimum	\$0	5%	\$0
Maximum	\$1,117,178	10%	\$0
Mean	\$93,130	15%	\$0
Std Dev	\$105,541	20%	\$2,503
Variance	11138884825	25%	\$11,909
Skewness	1.973279933	30%	\$20,992
Kurtosis	9.36901393	35%	\$30,173
Median	\$62,464	40%	\$40,293
Mode	\$0	45%	\$50,919
Left X	\$0	50%	\$62,464
Left P	5%	55%	\$73,512
Right X	\$304,563	60%	\$86,317
Right P	95%	65%	\$100,885
Diff X	\$304,563	70%	\$117,333
Diff P	90%	75%	\$136,258
#Errors	0	80%	\$159,493
Filter Min	Off	85%	\$190,461
Filter Max	Off	90%	\$234,152
#Filtered	0	95%	\$304,563

Case Study #2:

Interpreting the Model Results

- In addition to mean and variance, we examine simulated claim costs at varying percentiles (e.g. 75th, 90th, 95th, 97.5th, 99th)
- Then we can directly determine the financial requirements associated with funding the captive to those levels of claim costs
- For example, the required surplus at a given percentile level is equal to the simulated loss (if any)
- Determination of the premium-to-surplus ratio follows directly from simulation results and pricing assumptions

Case Study #2: Interpreting the Model Results

Captive Program Feasibility Study								
Simulation Results and Recommended Surplus Levels								
<i>hypothetical results - for illustrative purposes only</i>								
		75th	80th	85th	90th	95th	97.5th	
	mean	percentile	percentile	percentile	percentile	percentile	percentile	99th percentile
Specific Stop Loss Results								
Gross Premium	\$15,148,374	\$15,148,374	\$15,148,374	\$15,148,374	\$15,148,374	\$15,148,374	\$15,148,374	\$15,148,374
Ceding Commission	\$3,635,610	\$3,635,610	\$3,635,610	\$3,635,610	\$3,635,610	\$3,635,610	\$3,635,610	\$3,635,610
Claims	\$10,301,607	\$11,624,634	\$12,097,324	\$12,750,441	\$13,722,963	\$15,375,748	\$17,459,119	\$20,954,139
Profit/(Loss)	\$1,211,157	(\$111,870)	(\$584,560)	(\$1,237,676)	(\$2,210,199)	(\$3,862,984)	(\$5,946,355)	(\$9,441,375)
Required surplus to fund claims	--	\$111,870	\$584,560	\$1,237,676	\$2,210,199	\$3,862,984	\$5,946,355	\$9,441,375
Ratio: gross premium to required surplus	--	135.4	25.9	12.2	6.9	3.9	2.5	1.6
Aggregate Stop Loss Results								
Gross Premium	\$537,640	\$537,640	\$537,640	\$537,640	\$537,640	\$537,640	\$537,640	\$537,640
Ceding Commission	\$129,034	\$129,034	\$129,034	\$129,034	\$129,034	\$129,034	\$129,034	\$129,034
Claims	\$93,119	\$136,250	\$159,480	\$190,459	\$234,150	\$304,528	\$371,114	\$458,103
Profit/(Loss)	\$315,487	\$272,356	\$249,127	\$218,148	\$174,456	\$104,078	\$37,492	(\$49,496)
Required surplus to fund claims	--	\$0	\$0	\$0	\$0	\$0	\$0	\$49,496
Ratio: gross premium to required surplus	--	n/a	n/a	n/a	n/a	n/a	n/a	10.9
Specific and Aggregate Combined								
Gross Premium	\$15,686,014	\$15,686,014	\$15,686,014	\$15,686,014	\$15,686,014	\$15,686,014	\$15,686,014	\$15,686,014
Ceding Commission	\$3,764,643	\$3,764,643	\$3,764,643	\$3,764,643	\$3,764,643	\$3,764,643	\$3,764,643	\$3,764,643
Claims	\$10,394,726	\$11,760,884	\$12,256,804	\$12,940,899	\$13,957,114	\$15,680,276	\$17,830,233	\$21,412,241
Profit/(Loss)	\$1,526,644	\$160,487	(\$335,434)	(\$1,019,529)	(\$2,035,743)	(\$3,758,906)	(\$5,908,863)	(\$9,490,871)
Required surplus to fund claims	--	\$0	\$335,434	\$1,019,529	\$2,035,743	\$3,758,906	\$5,908,863	\$9,490,871
Ratio: gross premium to required surplus	--	n/a	46.8	15.4	7.7	4.2	2.7	1.7

- Gross premium and expenses are defined in the model by pricing assumptions
- Simulated claims are the variable in the table
- Required surplus and ratios are calculated directly

Questions?



For More Information:

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