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# Using Information Technology in Determining Yearly Aggregate Loss Distribution in an Insurance Portfolio

Key Words: Determining Yearly Aggregate Loos Distribution through Information Technology

Sherif Mohamad Mohsen

PHD in Insurance Faculty of Commerce Menofia University Egypt Sherifmohsen2023@outlook.com

WhatsApp: +201223971785

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#### Abstract

This article shows the using of information technology in preparing the claims aggregate loss table in an insurance portfolio or insurance policy and and will compare the results with both the four moments for one insurance policy introduced by HON SHIANG LAU and the for moments of all insurance portfolio introduced by THOMAS AUIPPA, hence the measurements of skewness and kurtosis.

## Introduction

In general insurance work, we have to get the claim frequency distribution and claim size distribution which express the severity of claims in order to estimate the maximum probable yearly aggregate loss (MPY) of an insurance portfolio. We cant's use the traditional approach for estimating MPY because of the large portfolio size, the following equation will clear this:

$$SS = \sum_{i=0}^{n} C^{i} \quad -----(1)$$

Where SS denotes to the sample size, C is denoting the number of claim size distribution classes and n denotes the maximum number of claims. For example, if we have a claim frequency table consists of four rows and two columns like this: [3, 39]

Number of Claims	Number of policyholders making Relative Freque				
0	17353	0.893931589			
1	1414	0.072841541			
2	620	0.031939007			
3	25	0.001287863			
Sum	19412	1			

Also, we have the following claim size frequency distribution: [3, 53]

Mean Claim Size \$	Number of policyholders making Relative Frequen				
1000	488	0.579572447			
3000	115	0.136579572			
5000	92	0.109263658			
7000	54	0.064133017			
9000	33	0.039192399			
11000	19	0.022565321			
13000	15	0.017814727			
15000	15	0.017814727			
17000	7	0.008313539			

19000	4	0.004750594
SUM	842	1

The sample space of aggregate loss distribution for one insurance policy will be according to equation (1):

$$SS = \sum_{i=0}^{n} C^{i}$$
  
$$SS = \sum_{i=1}^{3} 10^{i} = 10^{0} + 10^{1} + 10^{2} + 10^{3} = 1 + 10 + 100 + 1000 = 1111$$

Through A computer program, we can summarize the previous aggregate loss table into 49 rows table, eventually we calculate the sample space of an insurance portfolio consists of N policies according to the following equation:

 $SSP = M^{N}$  -----(2)

Where M denotes the number of aggregate loss table rows, while N is the number of polices in an Insurance Portfolio, Applying the previous equation for ten policies:

 $SSP = 49^{10} = 7.97923E+16$ 

Of course, We can't build a loss aggregate table contains 7.97923E+16 rows.

Fortunately, in 1984 HON SHINAG LAU discovered the four moments of aggregate loss distribution for one insurance policy from the ordinary loss frequency moments and claim size moments.

Also in 1988, THOMAS A. AIUPPA discovered the four moments of the entire insurance portfolio from the ordinary loss frequency moments through these moments only, we can use Pearson Family Curves as a good approximation to Maximum Probable Yearly Aggregate Loss (MPY) that's because Pearson Family Curves approximation to MPY methodology depends upon the moments of an insurance portfolio.

In this research I will introduce a computer program used in preparing the annual aggregate loss table for one insurance policy and also for a number of insurance policies.

# **Current Study**

I developed a computer used in preparing aggregate loss table for one insurance policy in an insurance portfolio and for a number of policies in order to determine an exact loss distribution and I will verify the results obtained from the program through HON SHIANG LAU and THOMAS A. AUIPPA equations of moments.

Lets' assume have the following observed numbers of policyholders making 0, 1, 2 claims which is called claim frequency table: [1, 259]

Number of claims	<b>Relative Frequency</b>
0	0.8
1	0.15
2	0.05

SUM 1

The four moments of the claims frequency for one policy can be calculated as follows:

$$\mu_{\rm n} = 0.25, \quad \mu_2(n) = 0.2875, \quad \mu_3(n) = 0.31875, \quad \mu_4(n) = 0.51953125$$

Let's assume we have the following claim size frequency distribution resulting from those policyholders: [1, 236]

Claims size \$	<b>Relative Frequency</b>
10000	0.6
20000	0.3
40000	0.1
SUM	1

The four moments of the claims frequency for one policy can be calculated as follows:

 $\mu_x = 16000, \quad \mu_2(x) = 84000000, \quad \mu_3(x) = 1.272E + 12, \quad \mu_4(x) = 3.4032E + 16$ 

From the previous two tables we can prepare the claims aggregate distribution manually for one policy as follows: [1, 238]

Claims	Aggregate	Probabilities and how they happen
numbers	Losses	
0	0	No claim probability = <b>0.8</b>
1	10000	Occurring of one claim with a size of 10000 = 0.15 x 0.6 = <b>0.09</b>
1	20000	Occurring of one claim with a size of 20000 = 0.15 x 0.3 = <b>0.045</b>
1	40000	Occurring of one claim with a size of 40000 = 0.15 x 0.1 = <b>0.015</b>
2	20000	Occurring of two claims, the size of the first 10000 and the size of the
		second is 10000 = 0.05 x 0.6 x 0.6 = <b>0.018</b>
2	30000	Occurring of 2 claims, the size of the first is 10000 and the size of the
		second is 20000 = 0.05 x 0.6 x 0.3 = <b>0.009</b>
2	50000	Occurring of 2 claims, the size of the first is 10000 and the size of the
		second is 40000 = 0.05 x 0.6 x 0.1 = <b>0.003</b>
2	40000	Occurring of 2 claims, the size of the first is 20000 and the size of the
		second is 20000 = 0.05 x 0.3 x 0.3 = <b>0.0045</b>
2	30000	Occurring of 2 the size of the first is 10000 and the size of the second is
		20000 = 0.05 x 0.6 x 0.3 = <b>0.009</b>
2	60000	Occurring of 2 claims, with a first claims, the size of the first is20000 and
		the size of the second is 40000 = 0.05 x 0.3 x 0.1 = <b>0.0015</b>
2	80000	Occurring of 2 claims, the size of the first is 40000 and the size of the
		second is 40000 = 0.05 x 0.1 x 0.1 = <b>0.0005</b>
2	50000	Occurring of 2 claims, the size of the first is 40000 and the size of the
		second is 10000 = 0.05 x 0.6 x 0.1 = <b>0.003</b>
2	60000	Occurring of 2 claims, size of the first is 40000 and the size of the second

	is 20000 = 0.05 x 0.3 x 0.1 = <b>0.0015</b>
	Sum of Probabilities = 1

The sample space of the previous preliminary claims aggregate loss can be calculated according to the equation (1):

$$SS = \sum_{I=0}^{n} C^{n}$$

Where C denotes the number of claim size distribution classes and n denotes the maximum number of claims

$$SS = \sum_{I=0}^{2} 3^{i}$$
$$SS = 3^{0} + 3^{1} + 3^{2}$$
$$SS = 13$$

We will aggregate the probabilities according to each aggregate loss from the previous table to get the following yearly aggregate loss distribution per one unit.

0 0	1		
	Aggregate Losses \$	Probability	
	0	0.80	
	10000	0.09	
	20000	0.063	6 I
	30000	0.018	
	40000	0.0195	
	50000	0.006	
	60000	0.003	
	80000	0.005	
	Sum	1	

Because the calculations are long and complex, I developed a computer program to do the yearly loss distribution per one unit and I got the following same result:

The Preliminary aggregate loss distribution of one policy

	AC	-		AP	-	
		0	C	.800000	0119	
		10000	C	.090000	0036	
		20000	C	.045000	0018	
		40000	C	.015000	0006	
		20000	C	.018000	0011	
		30000	C	.009000	0005	
		50000	C	.003000	0003	
		30000	C	.009000	0005	
		40000	C	.004500	0003	
		60000	C	.001500	0001	
		50000	C	.003000	0003	
		60000	C	.001500	0001	
		80000	C	.000500	0000	
Rec	ord: I4	1 of 13	;	► ►I ►S		N

The Final losses aggregate distribution for one policy



# The aggregate Loss distribution of one policy

AC SumOfAP

0	0.8
10000	0.09
20000	0.063
30000	0.018
40000	0.0195
50000	0.006
60000	0.003
80000	0.0005
Sum	1

Where the variable AC denotes the aggregate losses and the variable SumOFAP denotes the probability

The four moments of claims aggregate distribution can be calculated as follows:

 $\mu_1 = 4000, \quad \mu_2(l) = 94,600,005, \quad \mu_3(l) = 2.7828E + 12, \quad \mu_4(l) = 1.18478E + 17$ 

Hon Shiang Lau presented the following equations to calculate the four moments of aggregate loss distribution for one unit (policy): [4, 24]

$$\mu_{l} = \mu_{x}\mu_{n}$$

$$\mu_{2}(l) = \mu_{x}^{2}\mu_{2}(n) + \mu_{n}\mu_{2}(x)$$

$$\mu_{3}(l) = \mu_{x}^{3}\mu_{3}(n) + \mu_{n}\mu_{3}(x) + 3\mu_{x}\mu_{2}(x)\mu_{2}(n)$$

$$\mu_{4}(l) = \mu_{x}^{4}\mu_{4}(n) + \mu_{n}\mu_{4}(x) + 4\mu_{x}\mu_{3}(x)\mu_{2}(n) + 6\mu_{x}^{2}\mu_{2}(x)[\mu_{n}\mu_{2}(n) + \mu_{3}(n)] + 3[\mu_{2}(x)]^{2}[\mu_{n}^{2} - \mu_{n} + \mu_{2}(n)]$$
(3)

And now we will recall the previously calculated four moments for claims frequency and claim size to apply these equations on them:

 $\mu_n = 0.25, \quad \mu_2(n) = 0.2875, \quad \mu_3(n) = 0.31875, \quad \mu_4(n) = 0.51953125$   $\mu_x = 16000, \quad \mu_2(x) = 84000000, \quad \mu_3(x) = 1.272E + 12, \quad \mu_4(x) = 3.4032E + 16$   $\mu_l = 4,000.00, \quad \mu_2(l) = 94,600,000.00, \quad \mu_3(l) = 2,782,800,000,000.00, \quad \mu_4(l) = 1.18478E + 17$ As we saw, the results were identical from both LAU's equations or one policy loss aggregate distribution

It's very hard to prepare the annual loss aggregate losses table for the hole insurance portfolio for a reason due to the sample according to equation (2) as follows:

$$SS_2 = M^N$$

Where, M denotes to the number of claims aggregate losses distribution table rows for one policy and n denotes to the total number of policies in the insurance portfolio

For example, if we calculate the sample space for 10 polices under the last table of losses aggregate distribution for one policy it will be:

 $SS_2 = 8^{10} = 1,073,741,824$  , and for 8 policies

$$SS_2 = 8^8 = 16,777,216$$

We can calculate the four moments of an insurance portfolio as follows:

First, we calculate the four moments of claims frequency for the total number of policies (m) or insured units according to Thomas Auippa equations as follows: [2, 430]

$$\mu_{N} = m\mu_{n}$$

$$\mu_{2}(N) = m\mu_{2}(n)$$

$$\mu_{3}(N) = m\mu_{3}(n)$$

$$\mu_{4}(N) = m(\mu_{4}(n) - 3\mu_{2}^{2}(n)) + 3m^{2}\mu_{2}^{2}(n)$$
(4)

Second, we calculate the four moments of all policies in the portfolio as follows:

$$\mu_{L} = \mu_{x}\mu_{N}$$

$$\mu_{2}(L) = \mu_{x}^{2}\mu_{2}(N) + \mu_{N}\mu_{2}(x)$$

$$\mu_{3}(L) = \mu_{x}^{3}\mu_{3}(L) + \mu_{N}\mu_{3}(x) + 3\mu_{x}\mu_{2}(x)\mu_{2}(N) - (5)$$

$$\mu_{4}(L) = \mu_{x}^{4}\mu_{4}(N) + \mu_{N}\mu_{4}(x) + 4\mu_{x}\mu_{3}(x)\mu_{2}(N) + 6\mu_{x}^{2}\mu_{2}(x)[\mu_{n}\mu_{2}(N) + \mu_{3}(N)] + 3[\mu_{2}(x)]^{2}[\mu_{N}^{2} - \mu_{N} + \mu_{2}(N)]$$

Let's apply these equations (4), (5) in our case assuming we have 8 polices or units of risk N = 8 just as an example:

$$\begin{split} \mu_N &= 2, \quad \mu_2(N) = 2.3, \quad \mu_3(N) = 2.55, \quad \mu_4(N) = 18.0425 \\ \mu_x &= 16000, \quad \mu_2(x) = 84000000, \quad \mu_3(x) = 1.272E + 12, \quad \mu_4(x) = 3.4032E + 16 \\ \mu_L &= 32,000.00, \quad \mu_2(L) = 756,800,000.00, \quad \mu_3(L) = 2.22624E + 13, \quad \mu_4(L) = 2.45128E + 18 \end{split}$$

I developed another computer program to prepare the aggregate annual loss distribution for the insurance portfolio Based on the rule of determining the sample space for the process of throwing a number of dice, but this program had limited success up to 8 polices only due to the large sample space size which needs to a super computer to perform the calculation but the main idea behind the program still the same, and this was the output of my computer program:

	Dore del meet		records		14			i cate i oriniateni	19	14	
	CV6_4.AC - CV6_4.Sum(	✓ CV6_5.AC ✓	CV6_5.Sum( -	CV6_6.AC 👻	Expr1013 •	Expr1014 -	CV6_6.Sum( +	CV6_7.AC -	CV6_7.Sum( +	AC1 🔹	AP 🔺
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.8000000119	0	0.16777218
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	10000	0.01887437
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	20000	0.01321205
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	30000	0.00377487
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	40000	0.00408944
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	50000	0.00125829
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	60000	0.00062914
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	80000	0.00010485
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	10000	0.01887437
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	20000	0.00212336
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	30000	0.00148635
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	40000	0.00042467
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.8000000119	50000	0.00046006
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	60000	0.00014155
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	70000	7.07789E-
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	90000	1.179648E-
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	20000	0.01321205
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	30000	0.00148635
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	40000	0.00104044
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	50000	0.00029727
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	60000	0.0003220
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	70000	9.909045E-
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	80000	4.954523E-
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	100000	8.257538E-
	0 0.80000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	30000	0.00377487
	0 0.8000001	19 (	0.800000119	0	0.800000119	0	0.800000119	0	0.800000119	40000	0.00042467 🚽
Re	cord: 14 4 1 of 16777216 + H	🛤 🌄 No Filter	Search	•							

It's noted that the number of records equals to the 16,777,216 according to the equation (2)  $SS_2 = M^n = 8^8$ The aggregate Loss distribution of eight policies

AC SumOfAP
0 0.167772
10,000 0.150995
20,000 0.165151
30,000 0.126812
40,000 0.115603
50,000 0.085604
60,000 0.065531
70,000 0.043463
80,000 0.030298
90,000 0.019138
100,000 0.012303
110,000 0.00732
120,000 0.004412
130,000 0.002509
140,000 0.001435
150,000 0.000779
160,000 0.000424
170,000 0.000221
180,000 0.000116
190,000 5.81E-05
200,000 2.91E-05
210,000 1.41E-05
220,000 6.82E-06
230,000 3.19E-06

260,000	3.04E-07	
270,000	1.33E-07	
280,000	5.8E-08	
290,000	2.46E-08	
300,000	1.04E-08	
310,000	4.25E-09	
320,000	1.74E-09	
330,000	6.9E-10	
340,000	2.74E-10	
350,000	1.05E-10	
360,000	4.02E-11	
370,000	1.49E-11	
380,000	5.53E-12	
390,000	1.98E-12	
400,000	7.09E-13	
410,000	2.44E-13	
420,000	8.46E-14	
430,000	2.79E-14	
440,000	9.35E-15	
450,000	2.95E-15	
460,000	9.53E-16	
470,000	2.87E-16	
480,000	8.9E-17	
490,000	2.53E-17	
500,000	7.55E-18	
510,000	2.01E-18	
520,000	5.78E-19	
530,000	1.41E-19	
540,000	3.93E-20	
550,000	8.62E-21	
560,000	2.32E-21	
570,000	4.39E-22	
580,000	1.18E-22	
590,000	1.69E-23	
600,000	5.16E-24	
610,000	3.75E-25	
620,000	1.88E-25	
640,000	3.91E-27	
Sum	1	

. ]

10

1.49E-06 6.73E-07

240,000

250,000



According to the well-known rule, the sum of all probabilities equals one The symbol **AC** denotes to aggregate losses and **SumOfAP** denotes to probability The four moments of this distribution will be:

 $\mu_L = 32,000.05365, \quad \mu_2(L) = 756,805,935.3, \quad \mu_3(L) = 2.22629E + 13, \quad \mu_4(L) = 2.45134E + 18$ 

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As we saw the results is very near to that of Thomas Auippa equations (5) which were:

 $\mu_L = 32,000$   $\mu_2(L) = 756,800,000$ ,  $\mu_3(L) = 2.22624E + 13$ ,  $\mu_4(L) = 2.45128E + 18$ 

The Coefficient of Skewness and Coefficient of Kurtosis according to the computer program will be:

 $\beta_1 = \mu_3^2 / \mu_2^3, \quad \beta_2 = \mu_4 / \mu_2^2$ -----(6)

 $\beta_1 = 1.1434, \quad \beta_2 = 4.27991$ 

While that of Auippa's Equations:

$$\beta_1 = 1.1434, \quad \beta_2 = 4.27987$$

It's clear that Auippa's equations treated all of insurance portfolio the same treatment notwithstanding with the heterogeneity of portfolio assumed that any policy in the portfolio has the same maximum probable aggregate loss.

My point of view in respect this subject is to not to treat the insurance portfolio as one unit that's because in the real world there are heterogeneous subpopulations, for example in case of motor insurance portfolio there are several brands ranging from popular to luxury, so we can't group them in one frequency or claim size distributions tables because each one has its maximum probable yearly aggregate losses i.e. we may have a luxury vehicle evaluated half million \$ and other one evaluated 50,000 \$, in this case we have to group them in two separate frequency and severity tables for pricing each kind of vehicle according to its risk factor because each of them has its own maximum probable aggregate loss.

Lau proposed the following equations to calculate the aggregate moments for grouped insurance portfolios to merge the splinted insurance portfolio into one portfolio for pricing excess of loss reinsurance treaty: [4, 28]

For w = x + y  $\mu_w = \mu_x + \mu_y$   $\mu_2(w) = \mu_2(x) + \mu_2(y)$  -----(7)  $\mu_3(w) = \mu_3(x) + \mu_3(y)$  $\mu_4(w) = \mu_4(x) + 6\mu_2(x)\mu_2(y) + \mu_4(y)$ 

Applying the computer program on example state in page 1, we got the following results:

The Preliminary aggregate loss distribution of one policy

$\angle$	AC	<b>.</b>	AP	*
		10	0.893931	5677
		1000	0.042216	9492
		3000	0.009948	6662
		5000	0.007958	9328
		7000	0.004671	5480
		9000	0.002854	8348
		11000	0.001643	6927
		13000	0.001297	6521
		15000	0.001297	6521
		17000	0.000605	5710
		19000	0.000346	0406
		2000	0.010728	4468
		4000	0.002528	2202
		6000	0.002022	5761
		8000	0.001187	1642
		10000	0.000725	4893
		12000	0.000417	7059
		14000	0.000329	7678
		16000	0.000329	7678
		18000	0.000153	8917
		20000	8.7938	LE-05
		4000	0.002528	2202
		6000	0.000595	7896
		8000	0.000476	6317
		10000	0.000279	7621
		12000	0.000170	9657
		14000	9.843480	DE-05
		16000	7.771169	9E-05

It's noted that the program found the number of records = 1111 matches the output of equation (1).

$$SS = \sum_{i=0}^{3} C^{i}$$
  
$$SS = \sum_{i=1}^{3} 10^{i} = 10^{0} + 10^{1} + 10^{2} + 10^{3} = 1 + 10 + 100 + 1000 = 1111$$

It's noted that we can't prepare the previous table manually such as due to the large sample space. The Final aggregate loss distribution of one policy

AC       ▼       SumOfAP         1000       0.0422169492         2000       0.0107284468         3000       0.0101993885         4000       0.0050564404         5000       0.0081361852         6000       0.0046409417         7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.0016249626         13000       0.0013748627         16000       0.001330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0000333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05				
0       0.8939315677         1000       0.0422169492         2000       0.0107284468         3000       0.00101993885         4000       0.0050564404         5000       0.0081361852         6000       0.0046409417         7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.0016249626         13000       0.0013748627         16000       0.001330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0000333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		🖂 AC 👻	SumOfAP 🝷	
1000       0.0422169492         2000       0.0107284468         3000       0.00101993885         4000       0.0050564404         5000       0.0081361852         6000       0.0046409417         7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.001261317         15000       0.0013748627         16000       0.001330139         17000       0.0006731606         18000       0.0003964474         20000       0.000502202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		0	0.8939315677	
2000       0.0107284468         3000       0.0101993885         4000       0.0050564404         5000       0.0081361852         6000       0.0046409417         7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.001261317         15000       0.0013748627         16000       0.001330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.000233153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		1000	0.0422169492	
3000       0.0101993885         4000       0.0050564404         5000       0.0046409417         7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.001261317         15000       0.0013748627         16000       0.0013748627         16000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		2000	0.0107284468	
4000       0.0050564404         5000       0.0081361852         6000       0.0046409417         7000       0.0048551205         8000       0.003275918         9000       0.003081804         10000       0.0023918081         11000       0.0017683924         12000       0.0016249626         13000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.0013748627         16000       0.0007290483         19000       0.0000502202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		3000	0.0101993885	
5000       0.0081361852         6000       0.0046409417         7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.0013748627         16000       0.0007290483         19000       0.0003964474         20000       0.00003964474         20000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		4000	0.0050564404	
6000       0.0046409417         7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.0016249626         13000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.001330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         27000       1.055769E-05		5000	0.0081361852	
7000       0.0048551205         8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.0016249626         13000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.0013748627         16000       0.0007290483         19000       0.0003964474         20000       0.000502202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         27000       1.055769E-05		6000	0.0046409417	
8000       0.0033275918         9000       0.0030081804         10000       0.0023918081         11000       0.0017683924         12000       0.0016249626         13000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.0013748627         16000       0.0006731606         18000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0001374934         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		7000	0.0048551205	
9000 0.0030081804 10000 0.0023918081 11000 0.0017683924 12000 0.0016249626 13000 0.0013932144 14000 0.001261317 15000 0.0013748627 16000 0.0013748627 16000 0.0007290483 19000 0.00005022202 21000 3.822396E-05 22000 0.0002333153 22000 0.0001374934 23000 2.446505E-05 22000 1.642421E-05 24000 0.0001374934 25000 1.642421E-05 26000 7.608988E-05 27000 1.055769E-05 8 ccord: 14 ≤ 1 of 49 → N → S		8000	0.0033275918	
10000 0.0023918081     11000 0.0017683924     12000 0.0016249626     13000 0.0013932144     14000 0.001261317     15000 0.0013748627     16000 0.0013748627     16000 0.0006731606     18000 0.0007290483     19000 0.0003964474     20000 0.0005022202     21000 3.822396E-05     22000 0.0002333153     23000 2.446505E-05     22000 0.0001374934     25000 1.642421E-05     26000 7.608988E-05     27000 1.055769E-05     Record: H ≤ 1 of 49		9000	0.0030081804	
11000       0.0017683924         12000       0.0016249626         13000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.0011330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0001374934         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         27000       1.055769E-05		10000	0.0023918081	
12000       0.0016249626         13000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.0011330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0001374934         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         27000       1.055769E-05		11000	0.0017683924	
13000       0.0013932144         14000       0.001261317         15000       0.0013748627         16000       0.0011330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		12000	0.0016249626	
14000       0.001261317         15000       0.0013748627         16000       0.0011330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         27000       1.055769E-05		13000	0.0013932144	
15000       0.0013748627         16000       0.0011330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		14000	0.001261317	
16000       0.0011330139         17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		15000	0.0013748627	
17000       0.0006731606         18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		16000	0.0011330139	
18000       0.0007290483         19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05		17000	0.0006731606	
19000       0.0003964474         20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         Record: H       1 of 49		18000	0.0007290483	
20000       0.0005022202         21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         Record: H       1 of 49		19000	0.0003964474	
21000       3.822396E-05         22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         Record: H       1 of 49	( )	20000	0.0005022202	
22000       0.0002333153         23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         Record: H       1 of 49		21000	3.822396E-05	
23000       2.446505E-05         24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         Record: H       1 of 49		22000	0.0002333153	
24000       0.0001374934         25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         Record: H       1 of 49       ▶ H ▶ S		23000	2.446505E-05	
25000       1.642421E-05         26000       7.608988E-05         27000       1.055769E-05         Record: I4 ≤ 1 of 49       ► H ► 🖾 🐝		24000	0.0001374934	
26000       7.608988E-05         27000       1.055769E-05         Record: I4 ≤ 1 of 49       ► ► ► ►		25000	1.642421E-05	
27000 1.055769E-05 Record: I4 ≤ 1 of 49 → H → 🖾 🌾		26000	7.608988E-05	
Record: II of 49 + H + K		27000	1.055769E-05	
		Record: I4 4 1 of 49	► H →	

## The aggregate loss distribution of one policy

AC	SumOfAP
0	0.893932
1000	0.042217
2000	0.010728
3000	0.010199
4000	0.005056
5000	0.008136
6000	0.004641
7000	0.004855
8000	0.003328
9000	0.003008

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	10000	0.002392	
	11000	0.001768	
	12000	0.001625	
	13000	0.001393	
	14000	0.001261	
	15000	0.001375	
	16000	0.001133	
	17000	0.000673	
	18000	0.000729	
	19000	0.000396	
	20000	0.000502	
	21000	3.82E-05	
	22000	0.000233	
	23000	2.45E-05	
	24000	0.000137	
	25000	1.64E-05	
	26000	7.61E-05	
	27000	1.06E-05	
	28000	4.41E-05	
	29000	6.84E-06	
	30000	2.64E-05	
	31000	4.45E-06	
	32000	1.49E-05	
	33000	2.82E-06	
	34000	7.61E-06	
	35000	1.73E-06	
	36000	2.52E-06	
	37000	9.48E-07	
	38000	7.21E-07	
	39000	5.13E-07	
	41000	2.59E-07	
	43000	1.38E-07	
	45000	7.18E-08	
	47000	3.69E-08	
	49000	1.8E-08	
	51000	7.73E-09	
	53000	2.82E-09	
	55000	7.25E-10	
	57000	1.38E-10	

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The four moments of aggregate loss distribution for one policy can be calculated as follows:

Sum

 $\mu_1 = 466.1617$ ,  $\mu_2(l) = 4,045,856.238$ ,  $\mu_3(l) = 50,262,934,045$ ,  $\mu_4(l) = 8.00914E + 14$ 

Comparing previously calculated four moments of claims frequency and claim size with that of Lau's equations (3), we found are very near of them as follows:

 $\mu_{n} = 0.140583144, \quad \mu_{2}(n) = 0.192424717, \quad \mu_{3}(n) = 0.279192454, \quad \mu_{4}(n) = 0.5079749$  $\mu_{x} = 3315.914489, \quad \mu_{2}(x) = 13729176.66, \quad \mu_{3}(x) = 98187137416, \quad \mu_{4}(x) = 1.21662E + 15$  $\mu_{1} = 466.16, \quad \mu_{2}(l) = 4,045,856.18, \quad \mu_{3}(l) = 50,262,932,691.02, \quad \mu_{4}(l) = 8.00914E + 14$ 

They are very near to each other.

 $\mu_N = 0.562332578$ ,  $\mu_2(N) = 0.769698866$ ,  $\mu_3(N) = 1.116769818$ ,  $\mu_4(N) = 3.364881375$ Noting that the above moments for 4 policies were calculated from equation (4).

 $\mu_x = 3315.914489$ ,  $\mu_2(x) = 13729176.66$ ,  $\mu_3(x) = 98187137416$ ,  $\mu_4(x) = 1.21662E + 15$ The four moments of Auippa's equations (5) for 4 policies will be:

 $\mu_{L} = 1,864.65, \quad \mu_{2}(L) = 16,183,424.70, \quad \mu_{3}(L) = 2.01052E + 11, \quad \mu_{4}(L) = 3.79294E + 15$ 

The preliminary aggregate loss distribution according to my computer program for 4 policies is:

CV6.AC 👻	CV6.SumOf/ -	CV6_1.AC -	CV6_1.Sum( +	CV6_2.AC -	CV6_2.Sum( -	CV6_3.AC -	CV6_3.Sum( -	AC1 -	AP
0	0.8939315677	0	0.8939315677	0	0.8939315677	0	0.8939315677	0	0.638582
1000	0.0422169492	0	0.8939315677	0	0.8939315677	0	0.8939315677	1000	0.030157
2000	0.0107284468	0	0.8939315677	0	0.8939315677	0	0.8939315677	2000	0.007663
3000	0.0101993885	0	0.8939315677	0	0.8939315677	0	0.8939315677	3000	0.007285
4000	0.0050564404	0	0.8939315677	0	0.8939315677	0	0.8939315677	4000	0.003612
5000	0.0081361852	0	0.8939315677	0	0.8939315677	0	0.8939315677	5000	0.005812
6000	0.0046409417	0	0.8939315677	0	0.8939315677	0	0.8939315677	6000	0.003315
7000	0.0048551205	0	0.8939315677	0	0.8939315677	0	0.8939315677	7000	0.003468
8000	0.0033275918	0	0.8939315677	0	0.8939315677	0	0.8939315677	8000	0.002377
9000	0.0030081804	0	0.8939315677	0	0.8939315677	0	0.8939315677	9000	0.002148
10000	0.0023918081	0	0.8939315677	0	0.8939315677	0	0.8939315677	10000	0.001708
11000	0.0017683924	0	0.8939315677	0	0.8939315677	0	0.8939315677	11000	0.001263
12000	0.0016249626	0	0.8939315677	0	0.8939315677	0	0.8939315677	12000	0.001160
13000	0.0013932144	0	0.8939315677	0	0.8939315677	0	0.8939315677	13000	0.000995
14000	0.001261317	0	0.8939315677	0	0.8939315677	0	0.8939315677	14000	0.000901
15000	0.0013748627	0	0.8939315677	0	0.8939315677	0	0.8939315677	15000	0.000982
16000	0.0011330139	0	0.8939315677	0	0.8939315677	0	0.8939315677	16000	0.000809
17000	0.0006731606	0	0.8939315677	0	0.8939315677	0	0.8939315677	17000	0.000480
18000	0.0007290483	0	0.8939315677	0	0.8939315677	0	0.8939315677	18000	0.000520
19000	0.0003964474	0	0.8939315677	0	0.8939315677	0	0.8939315677	19000	0.000283
20000	0.0005022202	0	0.8939315677	0	0.8939315677	0	0.8939315677	20000	0.000358
21000	3.822396E-05	0	0.8939315677	0	0.8939315677	0	0.8939315677	21000	2.73053
22000	0.0002333153	0	0.8939315677	0	0.8939315677	0	0.8939315677	22000	0.000166
23000	2.446505E-05	0	0.8939315677	0	0.8939315677	0	0.8939315677	23000	1.74766
24000	0.0001374934	0	0.8939315677	0	0.8939315677	0	0.8939315677	24000	9.8218
25000	1.642421E-05	0	0.8939315677	0	0.8939315677	0	0.8939315677	25000	1.17326

It's noted that the number of records equals to the 5,764,801 according to the equation  $SS_2 = M^n = 49^4$ 

The aggregate Loss distribution of four policies

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AC1	SumOfAP
0	0.638583
1000	0.120631
2000	0.039201
3000	0.033756
4000	0.019338
5000	0.026595
6000	0.017778
7000	0.017352
8000	0.013108
9000	0.011546
10000	0.009642
11000	0.007484
12000	0.006754
13000	0.005874
14000	0.005302
15000	0.005473
16000	0.004735
17000	0.003279
18000	0.003193
19000	0.002163
20000	0.002268
21000	0.000902
22000	0.001205
23000	0.000596
24000	0.000765
25000	0.000411
26000	0.000473
27000	0.000273
28000	0.000301
29000	0.000183
30000	0.000196
31000	0,000123
32000	0.000125
33000	8.12E-05
34000	7,69F-05
35000	5.23F-05
36000	4.28F-05
37000	3 14F-05
37000	2 46F-05
39000	1 88F-05
22000	T.00E-02

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40000	1.42E-05
41000	1.1E-05
42000	8.88E-06
43000	6.62E-06
44000	5.56E-06
45000	3.99E-06
46000	3.44E-06
47000	2.41E-06
48000	2.11E-06
49000	1.44E-06
50000	1.28E-06
51000	8.39E-07
52000	7.56E-07
53000	4.84E-07
54000	4.39E-07
55000	2.78E-07
56000	2.5E-07
57000	1.62E-07
58000	1.43E-07
59000	9.55E-08
60000	8.12E-08
61000	5.66E-08
62000	4.65E-08
63000	3.33E-08
64000	2.65E-08

66000

67000

68000

69000

70000

71000

72000 73000

74000

75000

76000

77000

78000

79000

80000

1.93E-08

1.5E-08

1.11E-08

8.42E-09

6.3E-09

4.7E-09

3.53E-09 2.62E-09

1.96E-09

1.47E-09

1.08E-09

8.28E-10

5.98E-10

4.66E-10

3.3E-10

2.61E-10

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(	,)

81000	1.82E-10	
82000	1.45E-10	
83000	9.95E-11	
84000	7.93E-11	
85000	5.43E-11	
86000	4.3E-11	
87000	2.95E-11	
88000	2.31E-11	
89000	1.6E-11	
90000	1.23E-11	
91000	8.71E-12	
92000	6.53E-12	
93000	4.73E-12	
94000	3.46E-12	
95000	2.56E-12	
96000	1.83E-12	
97000	1.38E-12	
98000	9.62E-13	
99000	7.33E-13	1
100000	5.05E-13	
101000	3.86E-13	
102000	2.64E-13	
103000	2.02E-13	
104000	1.38E-13	
105000	1.04E-13	
106000	7.2E-14	
107000	5.37E-14	
108000	3.75E-14	
109000	2.75E-14	
110000	1.95E-14	
111000	1.4E-14	
112000	1.01E-14	
113000	7.12E-15	
114000	5.2E-15	
115000	3.61E-15	
116000	2.65E-15	
117000	1.82E-15	
118000	1.33E-15	
119000	9.18E-16	
120000	6.64E-16	
121000	4.61E-16	

(C)

1220	000	3.28E-16	
1230	000	2.31E-16	
1240	000	1.6E-16	
1250	000	1.16E-16	
1260	000	7.8E-17	
1270	000	5.76E-17	
1280	000	3.78E-17	
1290	000	2.85E-17	
1300	000	1.82E-17	
1310	000	1.4E-17	
1320	000	8.75E-18	
1330	000	6.79E-18	
1340	000	4.19E-18	
1350	000	3.25E-18	
1360	000	2.01E-18	
1370	000	1.54E-18	
1380	000	9.62E-19	
1390	000	7.15E-19	
1400	000	4.61E-19	
1410	000	3.29E-19	
1420	000	2.21E-19	
1430	000	1.5E-19	
1440	000	1.05E-19	10
1450	000	6.73E-20	
1460	000	5.01E-20	
1470	000	3E-20	
1480	000	2.36E-20	
1490	000	1.33E-20	
1500	000	1.1E-20	
1510	000	5.89E-21	
1520	000	5.05E-21	
1530	000	2.6E-21	
1540	000	2.28E-21	
1550	000	1.15E-21	
1560	000	1.01E-21	
1570	000	5.13E-22	
1580	000	4.38E-22	
1590	000	2.31E-22	
1600	000	1.87E-22	
1610	000	1.04E-22	
1620	000	7.82E-23	

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		1

163000	4.74E-23	
164000	3.21E-23	
165000	2.15E-23	
166000	1.29E-23	
167000	9.69E-24	
168000	5.12E-24	
169000	4.31E-24	
170000	2E-24	
171000	1.88E-24	
172000	7.72E-25	
173000	8.06E-25	
174000	2.99E-25	
175000	3.38E-25	
176000	1.17E-25	
177000	1.38E-25	
178000	4.72E-26	
 179000	5.5E-26	
 180000	1.95E-26	
181000	2.13E-26	
 182000	8.28E-27	
 183000	7.96E-27	
 184000	3.56E-27	
 185000	2.87E-27	
 186000	1.53E-27	
 187000	9.93E-28	
 188000	6.56E-28	
 189000	3.26E-28	
190000	2.76E-28	
191000	1.01E-28	
192000	1.14E-28	
193000	2.93E-29	
194000	4.57E-29	
195000	7.84E-30	
196000	1.79E-29	
197000	1.91E-30	
198000	6.84E-30	
199000	4.14E-31	
200000	2.53E-30	
201000	7.84E-32	
202000	9.02E-31	
203000	1.25E-32	

204000	3.09E-31
205000	1.59E-33
206000	1.01E-31
207000	1.46E-34
208000	3.15E-32
209000	7.59E-36
210000	9.19E-33
212000	2.5E-33
214000	6.27E-34
216000	1.43E-34
218000	2.9E-35
220000	5.12E-36
222000	7.6E-37
224000	8.98E-38
226000	7.63E-39
228000	3.63E-40
SUM	1

According to the well-known rule, the sum of all probabilities must equals one

The symbol AC denotes to aggregate losses and SumOfAP denotes to probability

The four moments of this distribution will be:

 $\mu_L = 1,864.647, \quad \mu_2(L) = 16,183,423.83, \quad \mu_3(L) = 2.01052E + 11, \quad \mu_4(L) = 3.79294E + 15$ 

These results almost match that of Thomas Auippa equations which were:

 $\mu_L = 1,864.65$   $\mu_2(L) = 16,183,424.70,$   $\mu_3(L) = 2.01052E + 11,$   $\mu_4(L) = 3.79294E + 15$ 

The Coefficient of Skewness and Coefficient of Kurtosis according to the computer program will be:

$$\beta_1 = \mu_3^2 / \mu_2^3, \quad \beta_2 = \mu_4 / \mu_2^2$$

 $\beta_1 = 9.536837987, \quad \beta_2 = 14.48221549$ 

Matches that of Auippa's Equations:

 $\beta_1 = 9.536836511, \quad \beta_2 = 14.48221389$ 

## Conclusion

This paper presented the output of a new computer program prepared by me, developed for finding the yearly aggregate loss distribution for one insurance policy and for a number of policies and compared the four moments of these aggregate losses with that of computed from Lau and Auippa equations. The result of these comparison was good, because I found the results were almost identical. The program was a complete success in setting up the aggregate loss distribution table for one insurance policy, but encounter a limited success in setting up the aggregate loss distribution for a large number of policies because the large sample space which leads to what's known The Curse of Dimensionality, but it's a step on the road. We don't have to apply Auippa's equations (4), (5) on the whole insurance portfolio because of the heterogeneity of components of the portfolio, so instead of this, we must split the portfolio according to risk factor into sub-portfolio sectors for direct insurance pricing purposes and merge the splinted sub portfolios according to Lau's equations (7) for non-proportional reinsurance treaties pricing purposes.

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