

The background features abstract blue geometric shapes, including triangles and polygons, in various shades of blue, creating a modern and professional look. The shapes are layered and overlap, with some appearing as solid colors and others as semi-transparent.

Notes on Using Property Catastrophe Model Results

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Agenda

- Popular Cat Models
- OEP, Return Period, AEP and PML
- OEP and the Collective Risk Model
- Simulation of Cat Losses from RMS-style ELT
- Model Blending
- When is the AEP like the OEP?

Popular Cat Models



AIR

Year Event Loss Table (YELT)

| Year | Event ID | Loss |
|------|----------|------|
| 1 | 1 | 100 |
| 3 | 2 | 500 |
| 3 | 3 | 300 |
| 4 | 4 | 100 |

$$\text{Mean}=250=(100+500+300+100)/4$$

$$\text{Std}=320=\{ (100)^2+(500+300)^2+(100)^2/4-250^2 \}^{0.5}$$

RMS

Event Loss Table (ELT)

| Event ID | Rate | Mean | Sdi | Sdc | Exposure |
|----------|------|------|-----|-----|----------|
| 1 | .10 | 500 | 500 | 500 | 10,000 |
| 2 | .10 | 300 | 400 | 800 | 5,000 |
| 3 | .50 | 200 | 300 | 400 | 4,000 |

The total standard deviation for each event is split into two additive components. This is an approximation used by RMS to facilitate combining the correlated pieces of the event.

Individual events are approximated assuming:
 Poisson event count with Poisson Mean=[Rate]
 Beta event size with Size mean=[Mean] and Size std=[Sdi]+[Sdc]

$$\text{Mean}=180=(.1)(500)+(.1)(300)+(.5)(200)$$

$$\text{Std}=737=\{ \begin{aligned} & (.1)((500+500)^2+500^2) \\ & (.1)((400+800)^2+300^2) \\ & (.5)((300+400)^2+200^2) \end{aligned} \}^{0.5}$$

OEP, Return Period, AEP, and PML

The right side of the slide features a decorative graphic composed of several overlapping, semi-transparent blue triangles and polygons. The colors range from a light sky blue to a deep navy blue. The shapes are arranged in a way that creates a sense of depth and movement, with some shapes appearing to be in front of others. The overall effect is a modern, abstract design element.

The Occurrence Exceedance Probability (OEP) describes the distribution of the largest event in a year.

Year, Event Loss Detail

| Year | Event ID | Loss |
|------|----------|------|
| 1 | 1 | 100 |
| 3 | 2 | 500 |
| 3 | 3 | 300 |
| 4 | 4 | 100 |

Year, Largest Event

| Year | Largest Event |
|------|---------------|
| 1 | 100 |
| 2 | 0 |
| 3 | 500 |
| 4 | 100 |

The OEP, $O(x)$ is the probability that the largest event in a year exceeds x .

PML is the dollar amount associated with the OEP

Return Period is the expected number of years between events that exceed x .

| Year | Largest Event |
|------|---------------|
| 1 | 100 |
| 2 | 0 |
| 3 | 500 |
| 4 | 100 |

| PML x | OEP $O(x)$ | Return Period $r=1/O(x)$ |
|------------|---------------|-----------------------------|
| 0 | 75% | 1.33 |
| 100 | 25% | 4.00 |
| 500 | 0% | Inf |

The columns pairs {PML,OEP} or {PML, Return Period} are often referred to as the PML curve or the OEP curve.

The PML curve or the OEP curve

| PML x | OEP $O(x)$ | Return Period $r=1/O(x)$ |
|------------|---------------|-----------------------------|
| 0 | 75% | 1.33 |
| 100 | 25% | 4.00 |
| 500 | 0% | Inf |

Aggregate Exceedance Probability (AEP) describes the distribution of the annual event claim total

Year, Event Loss Detail

| Year | Event ID | Loss |
|------|----------|------|
| 1 | 1 | 100 |
| 3 | 2 | 500 |
| 3 | 3 | 300 |
| 4 | 4 | 100 |

Year, Event Total

| Year | Annual Total |
|------|---------------|
| 1 | 100 |
| 2 | 0 |
| 3 | 800=(500+300) |
| 4 | 100 |

Event Severity, Largest Annual Event and Annual Event Total

| Random Variable | Cumulative distribution function (cdf) | Note | Notation |
|----------------------|--|--|---------------------------|
| Event Severity | $F(x)$ | Claim size component of a frequency-severity model | $X_i, i=1, \dots, N$ |
| Largest Annual Event | $F_M(x)$ | $OEP=O(x)=1-F_M(x)$ | $M=\max(X_1, \dots, X_N)$ |
| Annual Event Total | $F_Z(x)$ | $AEP=A(x)=1-F_Z(x)$ | $Z=X_1 + \dots + X_N$ |

OEP and the Collective Risk Model



OEP, Event Size and Event Counts

$O(x) = \text{OEP}$

$$\begin{aligned} O(x) &= \Pr(M > x) \text{ where } M = \max(X_1, \dots, X_N) \\ &= 1 - F_M(x) \\ &= 1 - \Pr(X_i \leq x \text{ for } i = 1, \dots, N) \\ &= 1 - E_N(F_X(x)^N) = 1 - \text{PGF}(F_X(x)) \end{aligned}$$

$F_M(x)$ = Cumulative distribution function (cdf) for the largest annual event

$F_X(x)$ = Cdf for the Event Size distribution

PGF = Probability Generating function for the Event Count distribution N . $\text{PGF}_N(t) = E(t^N)$

OEP, Event Size and Event Counts

$$F_M(x) = \text{PGF}(F_X(x))$$

$$O(x) = 1 - \text{PGF}(F_X(x))$$

We can write the Event Size distribution as a function of the OEP. When the Event Counts are Poisson we can do this explicitly

With independent Event Counts and independent identically distributed Event Sizes

$$F_X(x) = \text{PGF}^{-1}(1 - O(x))$$

For the Poisson Event Count distribution:

$$\text{PGF}(t) = \exp(\lambda(1 - t))$$

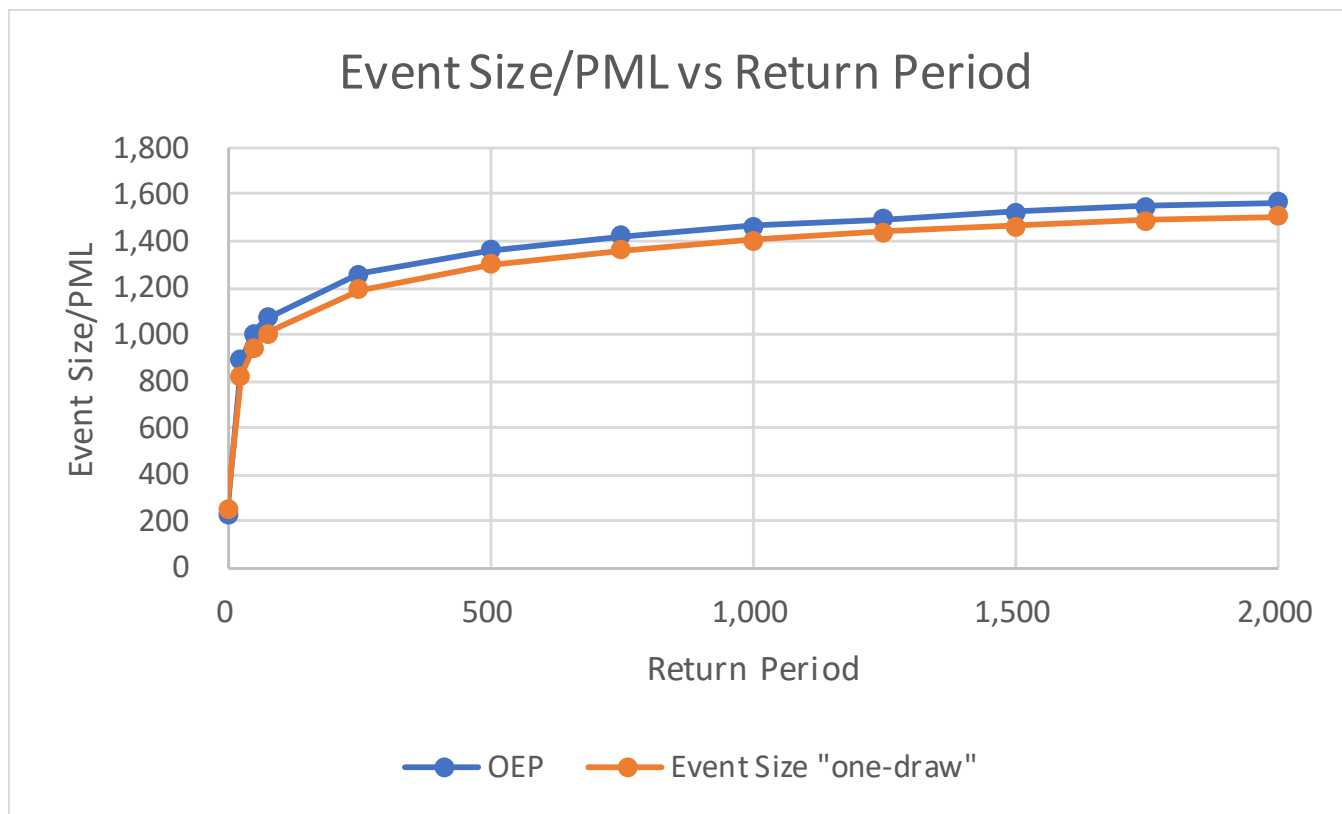
$$\text{PGF}^{-1}(s) = 1 - \frac{\log(s)}{\lambda} \quad \text{and,}$$

$$F_X(x) = 1 + \frac{\log(1 - O(x))}{\lambda}$$

Sample OEP Conversion to Event Size Cumulative Distribution Function (cdf)

| Return Period | Event Size or PML | OEP | Event Count | Cdf |
|---------------|-------------------|------------|-------------|------------------------------|
| R | x | $O(x)=1/R$ | λ | $F(x)=1+\ln(1-O(x))/\lambda$ |
| 1.5 | 224 | 66.67% | 1.5 | 26.76% |
| 2.0 | 354 | 50.00% | | 53.79% |
| 5.0 | 592 | 20.00% | | 85.12% |
| 10.0 | 728 | 10.00% | | 92.98% |
| 25.0 | 889 | 4.00% | | 97.28% |
| 50.0 | 1,003 | 2.00% | | 98.65% |
| 75.0 | 1,068 | 1.33% | | 99.11% |
| 100.0 | 1,114 | 1.00% | | 99.33% |
| 250.0 | 1,255 | 0.40% | | 99.73% |
| 500.0 | 1,360 | 0.20% | | 99.87% |
| 1,000.0 | 1,464 | 0.10% | | 99.93% |
| 2,000.0 | 1,566 | 0.05% | | 99.97% |

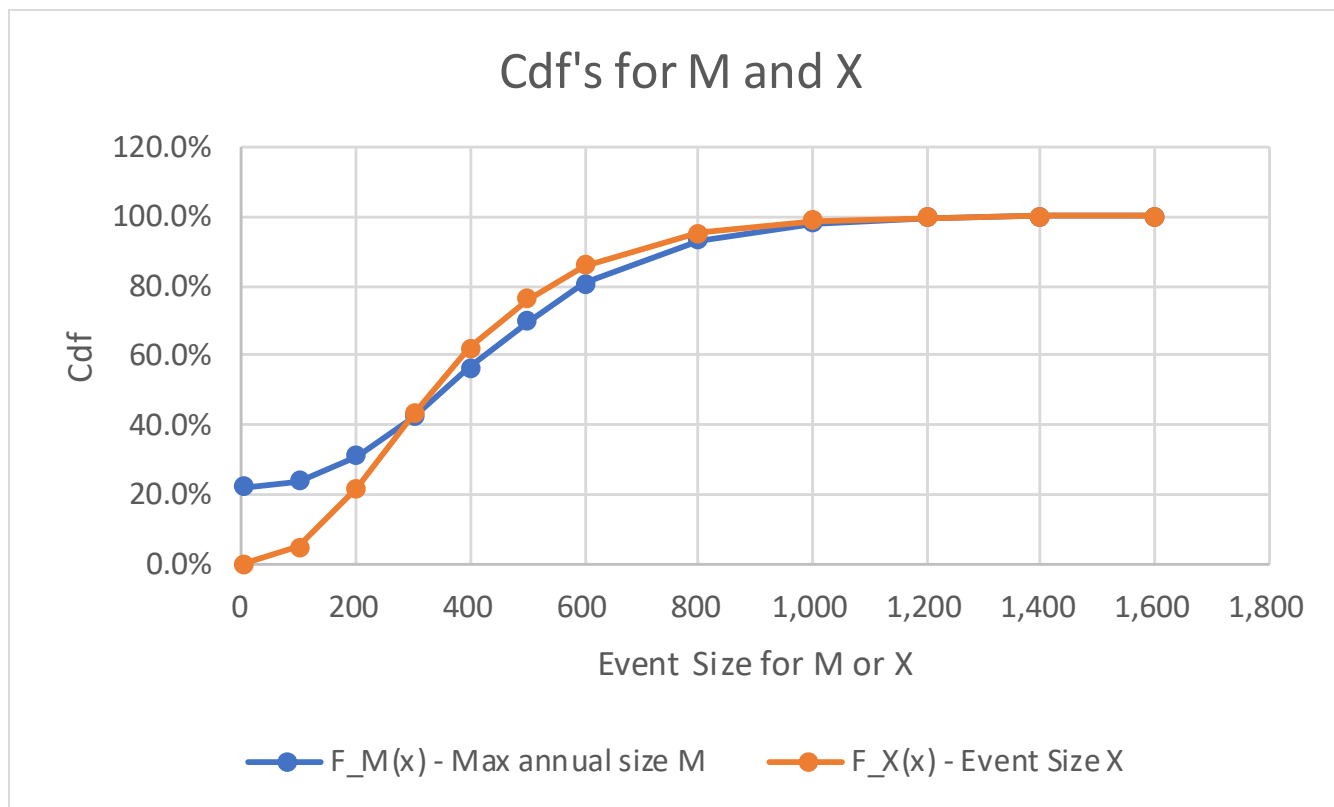
Comparison of Return Periods for Maximum Annual Loss M and Event Size X



Return Period for the OEP curve is $r=1/O(x)=1/(1-F_M(x))$

Return Period for the Event Size curve is $r=1/(1-F_X(x))$

Comparison of Cdf's for Maximum Annual Loss M and Event Size X



Recall that
 $F_M(x) = \text{PGF}(F_X(x))$

Simulation of Cat Losses from RMS-style ELT



Draw the Number of Events N from a Poisson Distribution with mean = $\lambda = \sum Rate_i$

| Event ID | Rate | Mean | Sdi | Sdc | Exposure |
|----------|------|------|-----|-----|----------|
| 1 | .10 | 500 | 500 | 500 | 10,000 |
| 2 | .10 | 300 | 400 | 800 | 5,000 |
| 3 | .50 | 200 | 300 | 400 | 4,000 |

$$\lambda = (.1 + .1 + .5) = .7$$
$$N \sim \text{Poisson}(\lambda)$$

For each event, draw a random row R from the ELT in proportion to the rates.

| Event ID | Rate | Mean | Sdi | Sdc | Exposure |
|----------|------|------|-----|-----|----------|
| 1 | .10 | 500 | 500 | 500 | 10,000 |
| 2 | .10 | 300 | 400 | 800 | 5,000 |
| 3 | .50 | 200 | 300 | 400 | 4,000 |

$$U \sim \text{Uniform}(0,1)$$
$$R = \min(r: U \leq \sum_{i=1}^r \text{Rate}_i / \lambda)$$

The size of the event is drawn from a Beta distribution with parameters computed from the ELT

$$a_R = \left(\frac{\text{Mean}_R}{\text{Sdi}_R + \text{Sdc}_R} \right)^2 \left(1 - \frac{\text{Mean}_R}{\text{Exposure}_R} \right) - \frac{\text{Mean}_R}{\text{Exposure}_R}$$

$$b_R = a_R \left(\frac{\text{Exposure}_R}{\text{Mean}_R} - 1 \right)$$

$$X \sim (\text{Exposure}_R)(\text{Beta}(a_R, b_R))$$

Excel formula:
 $X = \text{BETA.INV}(\text{RAND}(), a_R, b_R) * \text{Exposure}_R$

When the ELT has sub-lines there are additional steps.

| | | Personal Lines | | | | Commercial Lines | | | |
|----------|------|----------------|-----|-----|----------|------------------|-----|-----|----------|
| Event ID | Rate | Mean | Sdi | Stc | Exposure | Mean | Sdi | Sdc | Exposure |
| 1 | 0.1 | 300 | 400 | 300 | 3000 | 200 | 300 | 200 | 1000 |
| 2 | 0.1 | 100 | 371 | 267 | 1000 | 200 | 150 | 533 | 4000 |
| 3 | 0.5 | 100 | 224 | 200 | 2000 | 100 | 200 | 200 | 2000 |

- ▶ Aggregate the two sublines and apply the simulation recipe
- ▶ Allocate the simulated losses to the sublines

Aggregate the sub-lines

| Event ID | Rate | Mean | Sdi | Stc | Exposure |
|----------|------|---------|----------------------|---------|-----------|
| 1 | 0.1 | 300+200 | $\sqrt{400^2+300^2}$ | 300+200 | 3000+1000 |
| 2 | 0.1 | 100+200 | $\sqrt{371^2+150^2}$ | 267+533 | 1000+4000 |
| 3 | 0.5 | 100+100 | $\sqrt{224^2+200^2}$ | 200+200 | 2000+2000 |

$$\text{Mean}_R = \sum_k \text{Mean}_{R,k}$$

$$\text{Exposure}_R = \sum_k \text{Exposure}_{R,k}$$

$$\text{Sdi}_R = \sqrt{\sum_k \text{Sdi}_{R,k}^2}$$

$$\text{Sdc}_R = \sum_k \text{Sdc}_{R,k}$$

Once the sub-line table is aggregated the event size X can be simulated using the earlier recipe.

k indexes the sublines.

Allocate the simulated losses in proportion to the subline means.

$$X_k = X \frac{\text{Mean}_{R,k}}{\text{Mean}_R}$$

Model Blending



ELT/YELT Blending

| Trial | Model Uniform | Model Selected | Event Count | Loss |
|-------|---------------|----------------|-------------|------|
| 1 | 0.599 | AIR | 1 | 100 |
| 2 | 0.041 | RMS | 0 | - |
| 3 | 0.401 | RMS | 2 | 168 |
| 3 | - | - | - | 268 |
| 4 | 0.5 | AIR | 1 | 100 |

- ▶ It produces a blended set of results that can be used to model dependencies with other portfolio results under certain constraints

OEP Blending

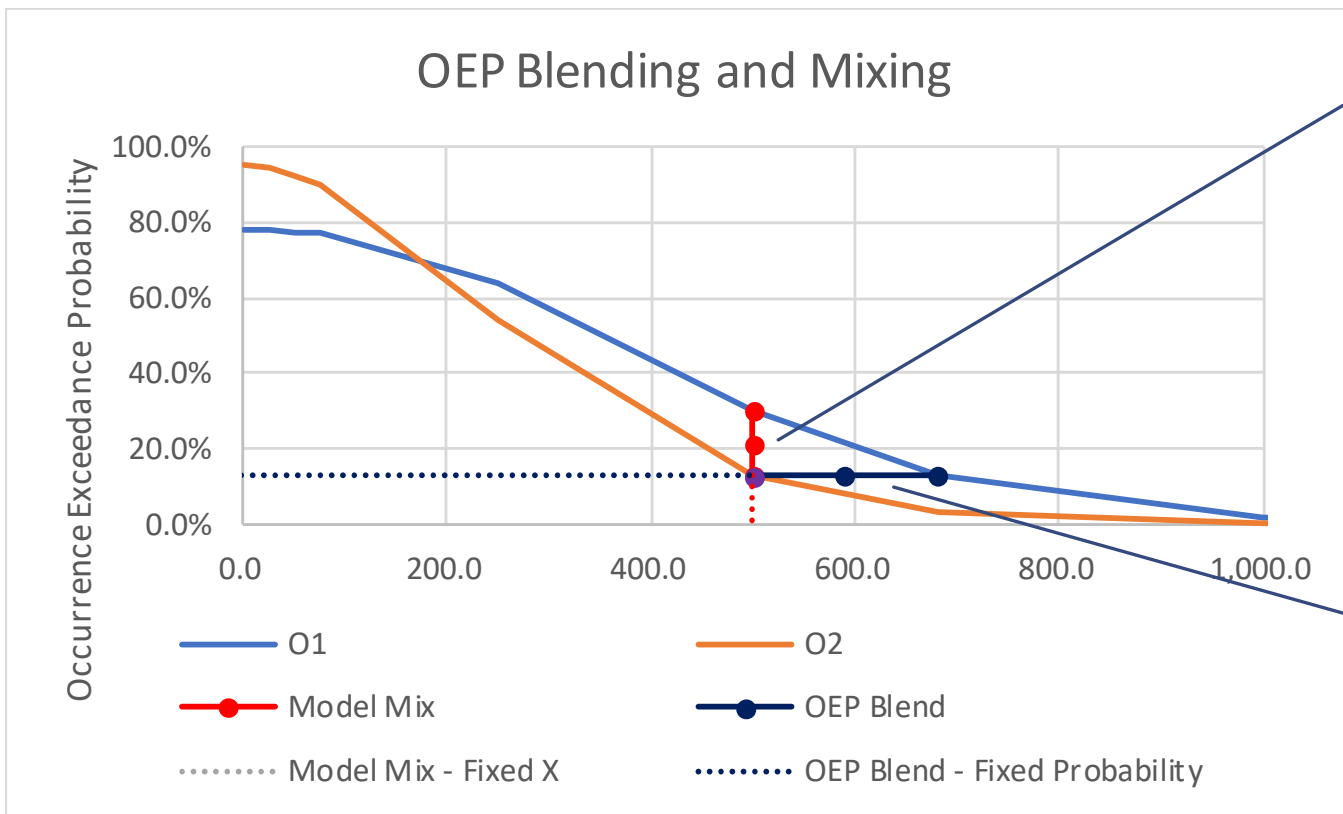
| Return Period | AIR PML | RMS PML | 50/50 PML |
|---------------|---------|---------|-----------|
| 1.33 | 0 | 0 | 0 |
| 2 | 50 | 204 | 127 |
| 4 | 100 | 268 | 186 |
| ∞ | 500 | 272 | 384 |

- ▶ It is intuitive and it has become a common practice to present the blended results at a high level

ELT/YELT Blending vs. OEP Blending

- ELT/YELT blending is essentially a blend of probabilities
- OEP blending is essentially a blend of losses
- The two approaches produce different results and the differences vary by return periods
- The choice of which approach to use depends on business context and application of blended results

Model Mixing weights probabilities, OEP Blending weights PML's or Event Sizes



Model Mix - randomly draw from different models or weight the probabilities for fixed event sizes(PML's)

OEP Blend - weight the event sizes (PML's) for fixed probabilities

When is the AEP like the OEP?



AEP vs OEP

$$Z = X_1 + \dots + X_N$$

$$F_Z(x) = \sum_n P_N(n) F_X^{(n)}(x)$$

$$A(x) = 1 - F_Z(x)$$

$$M = \max(X_1, \dots, X_N)$$

$$F_M(x) = \sum_n P_N(n) (F_X(x))^n$$

$$O(x) = 1 - F_M(x)$$

$$A(x) - O(x) = \sum_{n=2}^{\infty} P_N(n) (F_X^{(n)}(x) - (F_X(x))^n)$$

“The next step is magic in actuarial science.”

$$\mathcal{F}(F_Z(x)) = \text{PGF}(\mathcal{F}(F_X(x)))$$

$$F_M(x) = \text{PGF}(F_X(x))$$

Mildenhall, Stephen J., The Report of the Research Working Party on Correlations and Dependencies Among All Risk Sources (Part 1): Correlation and Aggregate Loss Distributions With An Emphasis On The Iman-Conover Method, pg.136. <https://www.casact.org/pubs/forum/06wforum/06w107.pdf>