 PENNSYLVANIA Compensation Rating Bureau Trusted • Essential • Objective

Presenters:
William Taylor, President \& CEO
Thomas Cleary, Senior Vice President \& COO
Brent Otto, VP \& Chief Actuary Peter Yoon, Director, Actuarial Research

## Today's Agenda

Who is the PCRB?
William Taylor, President \& CEO

The Workers Compensation Classification Code System \& Updates
Thomas Cleary, Senior Vice President \& COO

The "New" Experience Rating Modification Program
Brent Otto, Vice President \& Chief Actuary
Peter Yoon, Director, Actuarial Research

## Who is the PCRB?

It is Pennsylvania's trusted, objective provider of actuarially-based information and research, loss costs, and rating plans essential to a healthy workers' compensation system.

Established in 1915 as is an unincorporated, private, non-profit corporation comprised of all insurance companies licensed to transact workers' compensation insurance in Pennsylvania.

Made up of 100+ staff members, with various divisions, specifically: actuarial, data services, classification, policy review, field operations and administration.

It has over 400-member insurance companies.

No public money is used to fund its operations. The operations are funded primarily by membership fees and assessments.

## Our Vision

To be the trusted, essential, and objective resource for workers' compensation data, research, and information in Pennsylvania and Delaware.

## Our Mission

To provide objective, accurate, and valuable statistical and actuarially based information, marketplace knowledge, research, and rating plans fundamental to a healthy workers' compensation system.

In support of our mission and to achieve success for our stakeholders, we endeavor to:

- Provide ease of doing business by embracing data and technology
- Provide products and services that anticipate and respond to marketplace conditions
- Identify emerging trends and understand marketplace implications
- Conduct innovative research, provide educational services, and engage in outreach
- Deliver knowledge to empower actionable decisions


## Our Values:

We accomplish our mission with a dedicated team and culture that embraces:

Trust. Accountability. Transparency. Innovation. Collaboration. Inclusiveness. Work-life Blend.

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## PCRB's Primary Functions

File loss costs \& related rating values annually with the PA Insurance Department on behalf of its members

Administer a system of classification risk and underwriting rules

Administer an experience and merit rating system

## PCRB Secondary Functions

- Proof of workers' compensation coverage
- Administer a test audit program
- Develop products and services to support the WC system
- Data reporting tools, Supplemental data analysis, Training seminars, etc.
- Work with legislators, government agencies, members, agents, and labor and employer trade groups on workers' compensation issues.


## PA Workers Compensation Marketplace

PCRB's most recent loss cost, effective 4/1/2023, reflected an overall marketplace reduction, year over year, of $-3.3 \%$.

Claims frequency continued to trend downward at -6.2\%.

Claims severity increased about a point to 2.9\% for Indemnity and 3.4\% for Medical.

With a 2021 Combined Ratio of $83 \%$, consumers have benefited from declining ultimate loss costs.

The marketplace remains competitive and heathy for carriers.

To watch: independent contractors; GIG economy; risk management, robotics and AI; medical costs; other.

Recently posted information on the PCRB website highlights many statistics on the PA Workers Compensation marketplace.

## Workers' Compensation Insurance <br> Classification System

## What are "Classifications"?

Method of grouping similar, although not identical, businesses together for Workers Compensation rating and pricing purposes.

Identifies experience attributable to those businesses for ratemaking (Loss Cost Values) and Rating (Experience Modification Factors).


Assigned classification and associated loss cost values go a long way towards determining an employer's premium.

Approximately 335 Direct Assignment Classifications and 296 Temporary Staffing Classifications.

## What is the objective of the classification system?

Assign an employer to the one basic classification which best describes each distinct business enterprise - Field of Business.

Classification assignment is not based on individual job duties, occupation or types of employment.

Business operations not described by an existing classification are assigned to the most analogous classification from the standpoint of process and hazard.

## What are basic classifications?

- The basic classification describes the business of the insured.
- All classifications other than the standard exception classification Codes 822, Telecommuting Clerical Employees, 951, Salesperson - Outside, and 953, Clerical Office Employees.
- For example, employers making products out of sheet metal are assigned to basic class Code 454, Sheet Metal Shop. Employers selling jewelry are assigned to basic class Code 920, Jewelry Store


Recent Updates to the Classification System

## Revised Methodology for Temporary Staffing Classifications

- Temporary staffing classification codes apply to temporary staffing contractors that hire employees and assign those employees to an unrelated business for temporary work assignments varying in duration and are not to be confused with Professional Employer Organizations (PEOs).
- Change eliminated 37 existing temporary staffing classification codes and created 296 new temporary staffing class codes with each direct employment class mapping to one temporary staffing class.
- New temporary staffing classifications are identified by a four-digit number that is 2,000 greater than the associated direct employment classification. i.e., for direct employment classification 323, the corresponding temporary staffing classification is 2323.
- The primary reason for this change is to improve the overall consistency and accuracy of the loss costs between the direct employment class codes and the associated temporary staffing class codes.
- New temporary staffing classifications apply to new or renewal policies effective on or after April 1, 2021.


## Creation of Classification Code 906-Volunteer Ambulance Corps Support Staff and Code 989-Volunteer Fire Company - Support Staff

Act 108 of 2020 defined members of volunteer fire companies/departments and volunteer ambulance corps who do not respond to emergency calls as employees.

Classification study completed of Code 993 , Volunteer Ambulance Corps, and Code 994, Volunteer Fire Company, to determine
percentage of companies whose members are exclusively engaged in responding to emergency calls versus those who perform support duties.

Study resulted in the creation a Code 906, Volunteer Ambulance Corps - Support Staff, to be a companion class for Code 993 and Code 989, Volunteer Fire Company - Support Staff, to be a companion class for Code 993.

Exposure basis for the companion classes is the same as their corresponding class. i.e., Code 989 will use population as the exposure base and Code 906 will use a single loss cost on a per corps basis.

New classifications apply to new and renewal policies effective on or after April 1, 2022.

## Creation of Classification Code 822 Telecommuting Clerical Employees

- New classification for employees performing clerical work from their home via the use of email, telephone, video conferencing, and the Internet.
- Created due to telecommuting becoming increasingly more common across many types of business.
- Administered as a standard exception in addition to Code 951, Salesperson - Outside, and Code 953, Clerical Office Employees.
- Clerical employees who interchange between telecommuting and performing clerical work at their employer's location will be classified to Code 822 when more than $50 \%$ of their time is spent telecommuting, and to Code 953, Office, when $50 \%$ or less of their time is spent telecommuting.
- New classification applies to new or renewal policies effective on or after April 1, 2023.


## Revisions to Pennsylvania Basic Manual Provisions Related to COVID-19

- Applied an expiration date of June 30, 2023, to prior changes made to Basic Manual provisions related to COVID-19 including the COVID-19 Coronavirus Exception, Changing Classifications, Remuneration Payroll Exclusions and Basis of Premium - Payments to Furloughed Employees During Federal, State and/or Local Emergency Orders, Laws or Regulations Issued Due to COVID19 (Coronavirus) Pandemic).
- Removed "Payments to Paid Furloughed Employees Due to COVID-19" from the premium algorithm effective June 30, 2023.
- Sunset the use of statistical code 1212 to report payroll for furloughed employees and have it excluded from premium and experience rating calculations as of June 30, 2023.
- Added language to specify all claims reported with Catastrophe Code 12 with accident dates of December 1, 2019, through June 30, 2023, are excluded from experience rating and merit rating.
- The revisions made to the Basic Manual provisions apply to new and renewal policies effective July 1, 2023.

Experience Rating Plan Update

## Background

- Last major Experience Rating Plan (ERP) change was in 2004.
- Changed from a variable "maximum value" or "split point" to a single maximum value methodology.
- This maximum value is $\$ 42,500$ for all risk sizes.
- The first $\$ 42,500$ of every claim is referred to as "primary losses" and count $100 \%$ in the mod calculation.
- Added a capping rule that restricts mod change to a maximum swing of $+/-25 \%$ compared to the previous mod.


## Purpose of Research

The goal of this research is to enhance plan performance through improved predictive accuracy that incentivize workplace safety.

PCRB conducted a comprehensive review of its methodology, which included examining all the components of the current program.

## Current Experience Rating Plan

Single Maximum Value $=\$ 42,500$

$$
\begin{gathered}
\text { ApxC+ExC×L+E(1.000-C)} \\
E \\
\text { Ap: Actual Primary Losses } \\
\text { E: Expected Losses } \\
\text { C: Credibility } \\
\text { L: Limitation Charge }
\end{gathered}
$$

Eligibility: Premium developed by audited payroll during the experience period is greater than $\$ 10,000$

$$
\text { Includes } 100 \% \text { of Medical Only claims }
$$

Merit Rating plan is used for employers not qualified for the Experience Rating Plan

## Current Plan - Key Findings

- Performance testing showed declines in predictive accuracy since the last major update.
- Variable Maximum Value plan outperform Single Value plan.
- Assigns too little credibility to most risks.
- Inadequately promotes workplace safety.
- Transition between the Merit Rating and ERP plans for small risks can be a large change.


## Current ERP Performance Test

```
2015-2017
```




## Picture of Ideal Plan Performance

An ideal plan results in the same loss ratios for all quintiles after the application of the mod.

Quintile Test: Current Program


Quintile Test: Ideal Result


Our target deviation from 100\% modified loss ratio is $+/-5 \%$.

## How Do We Get There?

There are four crucial components that influence the accuracy and stability of the experience mod, and these components need to be optimized.

- Credibility
- Maximum Values
- Expected Loss Ranges
- L Factor (Limitation Charge)



## Optimization Process

Simultaneously Optimize Credibility, Split Points, Limit Factor, and Expected Loss Ranges

1. Used data for Policy Years 2015, 2016, 2017 and 2018.

- Optimization process built using years 2015-2017 and tested using the 2018 holdout year.

2. Risks grouped into cohorts by risk size with groupings also by expected losses ranges for an experience period.
3. Examine each cohort using an array of split points at a given credibility.

- Compute test statistic to find the maximum dispersion in the manual loss ratio and the minimum dispersion in the modified loss ratio.
- Test Statistic $=\frac{\text { Variance }(\text { Modified Loss Ratio })}{\text { Variance }(\text { Manual Loss Ratio })}$



## Optimal Split Point by Size of Risk - PY 2017

|  | Cohort $=5,000$ |  |  |  |  |  |  |  |  |  | Credibility |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Split Point | 100\% | 95\% | 90\% | 85\% | 80\% | 75\% | 70\% | 65\% | 60\% | 55\% | 50\% | 45\% | 40\% | 35\% | 30\% | 25\% | 20\% | 15\% | 10\% | 5\% |
| 1,000 | 0.533 | 0.551 | 0.569 | 0.588 | 0.607 | 0.627 | 0.648 | 0.669 | 0.691 | 0.713 | 0.735 | 0.759 | 0.783 | 0.809 | 0.834 | 0.860 | 0.886 | 0.913 | 0.945 | 0.971 |
| 2,000 | 0.377 | 0.389 | 0.411 | 0.439 | 0.463 | 0.480 | 0.507 | 0.537 | 0.565 | 0.588 | 0.620 | 0.650 | 0.682 | 0.721 | 0.757 | 0.793 | 0.831 | 0.869 | 0.913 | 0.955 |
| 3,000 | 0.258 | 0.279 | 0.299 | 0.320 | 0.345 | 0.369 | 0.398 | 0.425 | 0.456 | 0.490 | 0.521 | 0.561 | 0.600 | 0.641 | 0.682 | 0.728 | 0.777 | 0.829 | 0.88 | 0.940 |
| 4,000 | 0.190 | 0.206 | 0.225 | 0.245 | 0.267 | 0.291 | 0.317 | 0.348 | 0.379 | 0.413 | 0.448 | 0.487 | 0.529 | 0.576 | 0.625 | 0.677 | 0.733 | 0.793 | 0.857 | 0.926 |
| 5,000 | 0.156 | 0.167 | 0.179 | 0.195 | 0.218 | 0.239 | 0.258 | 0.289 | 0.314 | 0.349 | 0.388 | 0.428 | 0.472 | 0.521 | 0.573 | 0.630 | 0.689 | 0.760 | 0.833 | 0.914 |
| 6,000 | 0.144 | 0.150 | 0.158 | 0.169 | 0.183 | 0.200 | 0.220 | 0.244 | 0.270 | 0.302 | 0.342 | 0.377 | 0.425 | 0.476 | 0.531 | 0.591 | 0.655 | 0.729 | 0.812 | 0.903 |
| 7,000 | 0.156 | 0.156 | 0.160 | 0.166 | 0.176 | 0.189 | 0.206 | 0.226 | 0.251 | 0.279 | 0.319 | 0.351 | 0.394 | 0.443 | 0.505 | 0.566 | 0.631 | 0.703 | 0.796 | 0.892 |
| 8,000 | 0.169 | 0.163 | 0.161 | 0.162 | 0.168 | 0.176 | 0.189 | 0.206 | 0.227 | 0.253 | 0.284 | 0.320 | 0.368 | 0.412 | 0.473 | 0.532 | 0.604 | 0.686 | 0.779 | 0.883 |
| 9,000 | 0.201 | 0.189 | 0.180 | 0.176 | 0.175 | 0.161 | 0.186 | 0.198 | 0.215 | 0.237 | 0.255 | 0.299 | 0.332 | 0.386 | 0.442 | 0.507 | 0.581 | 0.666 | 0.763 | 0.875 |
| 10,000 | 0.202 | 0.185 | 0.173 | 0.164 | 0.161 | 0.161 | 0.166 | 0.176 | 0.190 | 0.210 | 0.237 | 0.270 | 0.310 | 0.357 | 0.414 | 0.479 | 0.556 | 0.645 | 0.747 | 0.865 |
| 11,000 | 0.228 | 0.206 | 0.191 | 0.175 | 0.166 | 0.164 | 0.163 | 0.170 | 0.180 | 0.196 | 0.222 | 0.251 | 0.290 | 0.336 | 0.392 | 0.457 | 0.536 | 0.627 | 0.733 | 0.857 |
| 12,000 | 0.269 | 0.240 | 0.216 | 0.197 | 0.182 | 0.173 | 0.169 | 0.171 | 0.177 | 0.190 | 0.211 | 0.238 | 0.274 | 0.318 | 0.372 | 0.439 | 0.518 | 0.611 | 0.721 | 0.850 |
| 13,000 | 0.307 | 0.273 | 0.243 | 0.215 | 0.198 | 0.184 | 0.173 | 0.173 | 0.174 | 0.184 | 0.202 | 0.226 | 0.259 | 0.303 | 0.356 | 0.422 | 0.500 | 0.596 | 0.709 | 0.842 |
| 14,000 | 0.346 | 0.305 | 0.269 | 0.239 | 0.215 | 0.196 | 0.183 | 0.175 | 0.174 | 0.181 | 0.195 | 0.216 | 0.247 | 0.288 | 0.341 | 0.406 | 0.485 | 0.581 | 0.697 | 0.835 |
| 15,000 | 0.387 | 0.339 | 0.299 | 0.262 | 0.232 | 0.209 | 0.192 | 0.181 | 0.176 | 0.179 | 0.190 | 0.208 | 0.237 | 0.276 | 0.327 | 0.391 | 0.470 | 0.568 | 0.687 | 0.829 |
| 16,000 | 0.430 | 0.376 | 0.328 | 0.287 | 0.252 | 0.224 | 0.202 | 0.187 | 0.178 | 0.178 | 0.186 | 0.201 | 0.227 | 0.264 | 0.314 | 0.378 | 0.457 | 0.555 | 0.676 | 0.823 |
| 17,000 | 0.471 | 0.412 | 0.359 | 0.312 | 0.273 | 0.240 | 0.213 | 0.194 | 0.182 | 0.177 | 0.182 | 0.196 | 0.219 | 0.254 | 0.302 | 0.365 | 0.44 | 0.54 | 0.66 | 0.817 |
| 18,000 | 0.514 | 0.447 | 0.388 | 0.336 | 0.291 | 0.254 | 0.223 | 0.200 | 0.185 | 0.177 | 0.178 | 0.190 | 0.211 | 0.244 | 0.291 | 0.352 | 0.43 | 0.53 | 0.65 | 0.811 |
| 19,000 | 0.570 | 0.495 | 0.429 | 0.370 | 0.320 | 0.278 | 0.242 | 0.215 | 0.195 | 0.183 | 0.181 | 0.189 | 0.207 | 0.238 | 0.282 | 0.34 | 0.42 | 0.52 | 0.64 | 0.805 |
| 20,000 | 0.619 | 0.537 | 0.467 | 0.402 | 0.346 | 0.298 | 0.259 | 0.228 | 0.204 | 0.189 | 0.183 | 0.187 | 0.203 | 0.231 | 0.274 | 0.33 | 0.41 | 0.51 | 0.64 | 0.800 |
| 21,000 | 0.672 | 0.581 | 0.503 | 0.434 | 0.373 | 0.319 | 0.275 | 0.239 | 0.212 | 0.194 | 0.184 | 0.186 | 0.199 | 0.225 | 0.26 | 0.32 | 0.40 | 0.50 | 0.63 | 0.795 |
| 22,000 | 0.723 | 0.627 | 0.543 | 0.466 | 0.399 | 0.342 | 0.293 | . 252 | 0.221 | 0.199 | 0.187 | 0.186 | 0.196 | 0.22 | 0.25 | 0.31 | 0.39 | 0.49 | 0.62 | 0.79 |
| 23,000 | 0.778 | 0.674 | 0.582 | 0.501 | 0.428 | 0.365 | 0.311 | 0.266 | 0.231 | 0.205 | 0.191 | 0.186 | 0.194 | 0.216 | 0.253 | 0.30 | 0.38 | 0.48 | 0.61 | 0.785 |
| 24,000 | 0.844 | 0.734 | 0.635 | 0.544 | 0.466 | 0.396 | 0.338 | 0.288 | 0.249 | 0.219 | 0.200 | 0.192 | 0.196 | 0.215 | 0.250 | 0.303 | 0.37 | 0.478 | 0.610 | 0.781 |
| 25,000 | 0.902 | 0.782 | 0.674 | 0.580 | 0.496 | 0.421 | 0.358 | 0.304 | 0.261 | 0.227 | 0.204 | 0.194 | 0.195 | 0.212 | 0.245 | 0.296 | 0.370 | 0.470 | 0.603 | 0.777 |
| 30,000 | 1.185 | 1.021 | 0.880 | 0.751 | 0.637 | 0.539 | 0.453 | 0.380 | 0.316 | 0.268 | 0.229 | 0.205 | 0.195 | 0.200 | 0.224 | 0.268 | 0.337 | 0.437 | 0.573 | 0.756 |
| 40,000 | 1.786 | 1.529 | 1.307 | 1.105 | 0.934 | 0.781 | 0.651 | 0.536 | 0.439 | 0.356 | 0.291 | 0.240 | 0.208 | 0.193 | 0.199 | 0.230 | 0.288 | 0.384 | 0.525 | 0.723 |
| 50,000 | 2.517 | 2.129 | 1.797 | 1.513 | 1.273 | 1.061 | 0.877 | 0.719 | 0.584 | 0.469 | 0.374 | 0.307 | 0.243 | 0.209 | 0.197 | 0.212 | 0.261 | 0.348 | 0.489 | 0.698 |
| 60,000 | 3.342 | 2.787 | 2.344 | 1.956 | 1.519 | 1.352 | 1.116 | 0.912 | 0.739 | 0.585 | 0.413 | 0.370 | 0.284 | 0.231 | 0.203 | 0.204 | 0.248 | 0.322 | 0.461 | 0.676 |
| 70,000 | 4.054 | 3.325 | 2.736 | 2.255 | 1.863 | 1.530 | 1.249 | 1.012 | 0.811 | 0.645 | 0.504 | 0.386 | 0.289 | 0.233 | 0.199 | 0.193 | 0.221 | 0.293 | 0.434 | 0.657 |
| 80,000 | 5.063 | 4.098 | 3.336 | 2.716 | 2.222 | 1.817 | 1.474 | 1.190 | 0.950 | 0.751 | 0.586 | 0.445 | 0.334 | 0.261 | 0.210 | 0.193 | 0.212 | 0.283 | 0.420 | 0.644 |
| 90,000 | 6.248 | 4.954 | 3.967 | 3.211 | 2.597 | 2.102 | 1.697 | 1.363 | 1.088 | 0.856 | 0.664 | 0.507 | 0.381 | 0.287 | 0.223 | 0.195 | 0.205 | 0.269 | 0.403 | 0.630 |
| 100,000 | 7.564 | 5.895 | 4.656 | 3.713 | 2.979 | 2.395 | 1.924 | 1.539 | 1.222 | 0.959 | 0.742 | 0.565 | 0.423 | 0.312 | 0.237 | 0.198 | 0.200 | 0.259 | 0.390 | 0.619 |
| 150,000 | 17.391 | 12.167 | 8.817 | 6.584 | 5.016 | 3.873 | 3.025 | 2.360 | 1.837 | 1.421 | 1.086 | 0.816 | 0.602 | 0.432 | 0.307 | 0.227 | 0.195 | 0.230 | 0.344 | 0.578 |
| 200,000 | 34.394 | 20.852 | 13.878 | 9.705 | 7.046 | 5.248 | 3.988 | 3.045 | 2.332 | 1.780 | 1.352 | 1.008 | 0.736 | 0.523 | 0.364 | 0.256 | 0.204 | 0.217 | 0.320 | 0.555 |
| 250,000 | 61.160 | 31.783 | 19.196 | 12.735 | 8.871 | 6.411 | 4.764 | 3.575 | 2.711 | 2.049 | 1.538 | 1.143 | 0.830 | 0.586 | 0.404 | 0.276 | 0.210 | 0.212 | 0.308 | 0.542 |
| 300,000 | 97.222 | 43.731 | 24.400 | 15.336 | 10.423 | 7.367 | 5.363 | 3.978 | 2.980 | 2.238 | 1.673 | 1.239 | 0.897 | 0.632 | 0.432 | 0.292 | 0.214 | 0.209 | 0.300 | 0.533 |
| 400,000 | 184.566 | 65.596 | 32.684 | 19.238 | 12.481 | 8.542 | 6.107 | 4.471 | 3.317 | 2.474 | 1.840 | 1.349 | 0.975 | 0.685 | 0.466 | 0.311 | 0.222 | 0.206 | 0.291 | 0.524 |
| 500,000 | 325.787 | 89.996 | 40.059 | 22.355 | 14.035 | 9.529 | 6.700 | 4.820 | 3.546 | 2.627 | 1.943 | 1.419 | 1.026 | 0.719 | 0.487 | 0.323 | 0.226 | 0.198 | 0.286 | 0.518 |

## Optimal Split Point by Size of Risk - PY 2017

|  | Cohort = 40 |  |  |  |  |  |  |  |  |  | Credibility |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Split Point | 100\% | 95\% | 90\% | 85\% | 80\% | 75\% | 70\% | 65\% | 60\% | 55\% | 50\% | 45\% | 40\% | 35\% | 30\% | 25\% | 20\% | 15\% | 10\% | 5\% |
| 1,000 | 0.783 | 0.794 | 0.804 | 0.811 | 0.821 | 0.832 | 0.842 | 0.851 | 0.859 | 0.872 | 0.883 | 0.891 | 0.902 | 0.914 | 0.929 | 0.939 | 0.950 | 0.961 | 0.975 | 0.986 |
| 2,000 | 0.692 | 0.701 | 0.714 | 0.728 | 0.741 | 0.755 | 0.769 | 0.787 | 0.798 | 0.814 | 0.828 | 0.845 | 0.861 | 0.877 | 0.893 | 0.911 | 0.927 | 0.945 | 0.962 | 0.979 |
| 3,000 | 0.625 | 0.639 | 0.652 | 0.673 | 0.688 | 0.702 | 0.718 | 0.734 | 0.756 | 0.774 | 0.787 | 0.807 | 0.828 | 0.847 | 0.868 | 0.886 | 0.908 | 0.931 | 0.954 | 0.976 |
| 4,000 | 0.577 | 0.592 | 0.607 | 0.625 | 0.642 | 0.659 | 0.676 | 0.696 | 0.716 | 0.736 | 0.756 | 0.778 | 0.800 | 0.820 | 0.846 | 0.867 | 0.893 | 0.917 | 0.944 | 0.972 |
| 5,000 | 0.519 | 0.537 | 0.554 | 0.574 | 0.592 | 0.613 | 0.633 | 0.654 | 0.674 | 0.694 | 0.717 | 0.743 | 0.767 | 0.793 | 0.820 | 0.846 | 0.875 | 0.906 | 0.935 | 0.967 |
| 6,000 | 0.481 | 0.499 | 0.517 | 0.536 | 0.556 | 0.576 | 0.598 | 0.620 | 0.642 | 0.666 | 0.691 | 0.716 | 0.743 | 0.771 | 0.800 | 0.830 | 0.861 | 0.894 | 0.928 | 0.963 |
| 7,000 | 0.448 | 0.466 | 0.485 | 0.505 | 0.525 | 0.546 | 0.569 | 0.592 | 0.616 | 0.641 | 0.667 | 0.694 | 0.722 | 0.752 | 0.783 | 0.815 | 0.849 | 0.884 | 0.921 | 0.960 |
| 8,000 | 0.418 | 0.435 | 0.456 | 0.476 | 0.495 | 0.519 | 0.541 | 0.564 | 0.590 | 0.616 | 0.643 | 0.672 | 0.702 | 0.733 | 0.766 | 0.800 | 0.837 | 0.875 | 0.914 | 0.956 |
| 9,000 | 0.391 | 0.412 | 0.429 | 0.450 | 0.472 | 0.493 | 0.517 | 0.542 | 0.568 | 0.594 | 0.623 | 0.653 | 0.684 | 0.717 | 0.751 | 0.787 | 0.826 | 0.866 | 0.908 | 0.953 |
| 10,000 | 0.373 | 0.391 | 0.411 | 0.431 | 0.452 | 0.475 | 0.498 | 0.523 | 0.550 | 0.577 | 0.606 | 0.637 | 0.669 | 0.702 | 0.738 | 0.776 | 0.815 | 0.858 | 0.902 | 0.950 |
| 11,000 | 0.353 | 0.371 | 0.392 | 0.413 | 0.433 | 0.455 | 0.480 | 0.505 | 0.531 | 0.560 | 0.590 | 0.620 | 0.654 | 0.689 | 0.726 | 0.764 | 0.806 | 0.850 | 0.897 | 0.947 |
| 12,000 | 0.351 | 0.369 | 0.388 | 0.408 | 0.429 | 0.451 | 0.475 | 0.499 | 0.526 | 0.554 | 0.584 | 0.615 | 0.648 | 0.681 | 0.720 | 0.760 | 0.802 | 0.848 | 0.896 | 0.946 |
| 13,000 | 0.327 | 0.348 | 0.368 | 0.386 | 0.406 | 0.427 | 0.451 | 0.480 | 0.507 | 0.536 | 0.566 | 0.596 | 0.631 | 0.669 | 0.707 | 0.749 | 0.790 | 0.840 | 0.890 | 0.943 |
| 14,000 | 0.303 | 0.322 | 0.341 | 0.362 | 0.384 | 0.406 | 0.431 | 0.457 | 0.485 | 0.513 | 0.545 | 0.578 | 0.615 | 0.653 | 0.693 | 0.736 | 0.782 | 0.832 | 0.883 | 0.941 |
| 15,000 | 0.292 | 0.310 | 0.329 | 0.351 | 0.372 | 0.395 | 0.420 | 0.447 | 0.474 | 0.504 | 0.536 | 0.570 | 0.603 | 0.643 | 0.685 | 0.728 | 0.774 | 0.824 | 0.879 | 0.937 |
| 16,000 | 0.277 | 0.295 | 0.314 | 0.335 | 0.357 | 0.380 | 0.405 | 0.430 | 0.459 | 0.490 | 0.521 | 0.555 | 0.593 | 0.631 | 0.674 | 0.719 | 0.767 | 0.819 | 0.875 | 0.935 |
| 17,000 | 0.271 | 0.288 | 0.306 | 0.326 | 0.348 | 0.371 | 0.395 | 0.422 | 0.450 | 0.480 | 0.512 | 0.547 | 0.584 | 0.623 | 0.666 | 0.711 | 0.760 | 0.815 | 0.872 | 0.933 |
| 18,000 | 0.270 | 0.285 | 0.303 | 0.322 | 0.343 | 0.366 | 0.390 | 0.416 | 0.443 | 0.473 | 0.505 | 0.539 | 0.577 | 0.616 | 0.659 | 0.706 | 0.755 | 0.809 | 0.867 | 0.931 |
| 19,000 | 0.265 | 0.275 | 0.293 | 0.316 | 0.337 | 0.359 | 0.379 | 0.403 | 0.435 | 0.464 | 0.494 | 0.528 | 0.566 | 0.608 | 0.650 | 0.696 | 0.747 | 0.804 | 0.864 | 0.929 |
| 20,000 | 0.236 | 0.252 | 0.270 | 0.289 | 0.310 | 0.332 | 0.356 | 0.383 | 0.411 | 0.443 | 0.476 | 0.510 | 0.548 | 0.590 | 0.636 | 0.687 | 0.740 | 0.796 | 0.859 | 0.926 |
| 21,000 | 0.223 | 0.239 | 0.256 | 0.275 | 0.296 | 0.318 | 0.342 | 0.368 | 0.397 | 0.428 | 0.461 | 0.497 | 0.536 | 0.578 | 0.624 | 0.674 | 0.729 | 0.787 | 0.853 | 0.924 |
| 22,000 | 0.212 | 0.228 | 0.245 | 0.264 | 0.284 | 0.307 | 0.332 | 0.358 | 0.386 | 0.417 | 0.451 | 0.487 | 0.527 | 0.569 | 0.616 | 0.667 | 0.722 | 0.782 | 0.849 | 0.922 |
| 23,000 | 0.211 | 0.226 | 0.243 | 0.262 | 0.282 | 0.304 | 0.327 | 0.354 | 0.382 | 0.413 | 0.446 | 0.483 | 0.518 | 0.565 | 0.609 | 0.663 | 0.719 | 0.777 | 0.846 | 0.920 |
| 24,000 | 0.204 | 0.219 | 0.235 | 0.253 | 0.272 | 0.294 | 0.318 | 0.344 | 0.372 | 0.402 | 0.436 | 0.472 | 0.512 | 0.555 | 0.603 | 0.654 | 0.711 | 0.775 | 0.842 | 0.917 |
| 25,000 | 0.202 | 0.215 | 0.232 | 0.249 | 0.268 | 0.289 | 0.313 | 0.338 | 0.366 | 0.397 | 0.430 | 0.466 | 0.506 | 0.549 | 0.597 | 0.649 | 0.706 | 0.769 | 0.839 | 0.915 |
| 30,000 | 0.203 | 0.214 | 0.226 | 0.241 | 0.257 | 0.276 | 0.298 | 0.325 | 0.352 | 0.378 | 0.409 | 0.446 | 0.487 | 0.531 | 0.579 | 0.632 | 0.691 | 0.754 | 0.828 | 0.910 |
| 40,000 | 0.286 | 0.281 | 0.280 | 0.282 | 0.287 | 0.294 | 0.305 | 0.319 | 0.337 | 0.358 | 0.385 | 0.414 | 0.449 | 0.488 | 0.536 | 0.588 | 0.647 | 0.719 | 0.796 | 0.891 |
| 50,000 | 0.315 | 0.305 | 0.285 | 0.280 | 0.277 | 0.283 | 0.284 | 0.291 | 0.306 | 0.325 | 0.344 | 0.373 | 0.405 | 0.444 | 0.492 | 0.546 | 0.608 | 0.685 | 0.772 | 0.878 |
| 60,000 | 0.430 | 0.395 | 0.367 | 0.347 | 0.332 | 0.322 | 0.318 | 0.318 | 0.324 | 0.334 | 0.350 | 0.372 | 0.400 | 0.435 | 0.478 | 0.530 | 0.590 | 0.668 | 0.757 | 0.868 |
| 70,000 | 0.605 | 0.543 | 0.493 | 0.452 | 0.417 | 0.394 | 0.378 | 0.368 | 0.365 | 0.364 | 0.374 | 0.386 | 0.407 | 0.440 | 0.478 | 0.526 | 0.582 | 0.659 | 0.748 | 0.862 |
| 80,000 | 0.791 | 0.695 | 0.617 | 0.556 | 0.501 | 0.466 | 0.434 | 0.414 | 0.400 | 0.392 | 0.391 | 0.399 | 0.414 | 0.439 | 0.470 | 0.518 | 0.573 | 0.649 | 0.737 | 0.853 |
| 90,000 | 0.961 | 0.831 | 0.727 | 0.635 | 0.571 | 0.524 | 0.480 | 0.450 | 0.427 | 0.411 | 0.404 | 0.406 | 0.415 | 0.438 | 0.464 | 0.506 | 0.561 | 0.634 | 0.727 | 0.846 |
| 100,000 | 1.221 | 1.067 | 0.929 | 0.804 | 0.714 | 0.645 | 0.586 | 0.542 | 0.505 | 0.476 | 0.461 | 0.452 | 0.453 | 0.468 | 0.487 | 0.521 | 0.571 | 0.639 | 0.724 | 0.844 |
| 150,000 | 2.734 | 2.165 | 1.762 | 1.464 | 1.238 | 1.061 | 0.921 | 0.810 | 0.721 | 0.650 | 0.595 | 0.555 | 0.527 | 0.513 | 0.514 | 0.529 | 0.564 | 0.620 | 0.701 | 0.827 |
| 200,000 | 4.618 | 3.369 | 2.588 | 2.060 | 1.682 | 1.401 | 1.185 | 1.016 | 0.883 | 0.778 | 0.694 | 0.629 | 0.582 | 0.550 | 0.535 | 0.537 | 0.560 | 0.609 | 0.686 | 0.814 |
| 250,000 | 6.923 | 4.618 | 3.367 | 2.583 | 2.053 | 1.673 | 1.391 | 1.175 | 1.006 | 0.874 | 0.769 | 0.686 | 0.624 | 0.580 | 0.554 | 0.547 | 0.562 | 0.604 | 0.677 | 0.806 |
| 300,000 | 9.492 | 5.799 | 4.032 | 2.998 | 2.336 | 1.875 | 1.539 | 1.286 | 1.088 | 0.939 | 0.818 | 0.724 | 0.651 | 0.599 | 0.565 | 0.552 | 0.560 | 0.599 | 0.673 | 0.799 |
| 400,000 | 14.420 | 7.557 | 5.039 | 3.630 | 2.759 | 2.176 | 1.760 | 1.455 | 1.224 | 1.045 | 0.903 | 0.793 | 0.706 | 0.643 | 0.599 | 0.575 | 0.574 | 0.605 | 0.673 | 0.796 |
| 500,000 | 20.889 | 9.307 | 5.836 | 3.958 | 2.956 | 2.312 | 1.858 | 1.549 | 1.285 | 1.101 | 0.946 | 0.825 | 0.731 | 0.658 | 0.610 | 0.581 | 0.576 | 0.602 | 0.668 | 0.791 |

## Optimal Split Point by Size of Risk - PY 2017

|  | Cohort $=1 \mathrm{M}$ |  |  |  |  |  |  |  |  |  | Credibility |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Split Point | 100\% | 95\% | 90\% | 85\% | 80\% | 75\% | 70\% | 65\% | 60\% | 55\% | 50\% | 45\% | 40\% | 35\% | 30\% | 25\% | 20\% | 15\% | 10\% | 5\% |
| 1,000 | 0.882 | 0.888 | 0.894 | 0.900 | 0.905 | 0.910 | 0.917 | 0.923 | 0.928 | 0.934 | 0.940 | 0.945 | 0.953 | 0.957 | 0.964 | 0.970 | 0.973 | 0.982 | 0.989 | 0.993 |
| 2,000 | 0.828 | 0.835 | 0.844 | 0.851 | 0.859 | 0.867 | 0.876 | 0.883 | 0.896 | 0.901 | 0.912 | 0.918 | 0.928 | 0.936 | 0.946 | 0.956 | 0.962 | 0.972 | 0.981 | 0.990 |
| 3,000 | 0.739 | 0.750 | 0.761 | 0.776 | 0.785 | 0.799 | 0.809 | 0.821 | 0.830 | 0.851 | 0.860 | 0.874 | 0.887 | 0.898 | 0.914 | 0.927 | 0.941 | 0.956 | 0.976 | 0.985 |
| 4,000 | 0.705 | 0.717 | 0.740 | 0.753 | 0.765 | 0.778 | 0.791 | 0.804 | 0.809 | 0.828 | 0.838 | 0.855 | 0.869 | 0.885 | 0.904 | 0.919 | 0.931 | 0.949 | 0.965 | 0.983 |
| 5,000 | 0.665 | 0.698 | 0.688 | 0.711 | 0.735 | 0.743 | 0.758 | 0.775 | 0.785 | 0.802 | 0.824 | 0.837 | 0.849 | 0.869 | 0.885 | 0.906 | 0.921 | 0.941 | 0.961 | 0.980 |
| 6,000 | 0.646 | 0.660 | 0.674 | 0.698 | 0.695 | 0.720 | 0.743 | 0.753 | 0.764 | 0.791 | 0.808 | 0.821 | 0.839 | 0.858 | 0.879 | 0.892 | 0.914 | 0.934 | 0.957 | 0.977 |
| 7,000 | 0.619 | 0.634 | 0.649 | 0.666 | 0.681 | 0.697 | 0.714 | 0.732 | 0.749 | 0.767 | 0.786 | 0.805 | 0.825 | 0.848 | 0.868 | 0.882 | 0.908 | 0.930 | 0.954 | 0.975 |
| 8,000 | 0.596 | 0.611 | 0.627 | 0.645 | 0.661 | 0.678 | 0.696 | 0.715 | 0.733 | 0.752 | 0.771 | 0.792 | 0.813 | 0.834 | 0.859 | 0.878 | 0.899 | 0.924 | 0.951 | 0.974 |
| 9,000 | 0.565 | 0.581 | 0.598 | 0.616 | 0.634 | 0.652 | 0.671 | 0.691 | 0.711 | 0.731 | 0.752 | 0.774 | 0.796 | 0.819 | 0.843 | 0.867 | 0.891 | 0.917 | 0.945 | 0.972 |
| 10,000 | 0.544 | 0.561 | 0.579 | 0.597 | 0.616 | 0.635 | 0.654 | 0.674 | 0.695 | 0.717 | 0.739 | 0.762 | 0.785 | 0.809 | 0.834 | 0.859 | 0.886 | 0.913 | 0.941 | 0.970 |
| 11,000 | 0.529 | 0.571 | 0.565 | 0.583 | 0.602 | 0.622 | 0.642 | 0.663 | 0.700 | 0.706 | 0.742 | 0.752 | 0.776 | 0.801 | 0.827 | 0.852 | 0.880 | 0.908 | 0.938 | 0.968 |
| 12,000 | 0.538 | 0.555 | 0.572 | 0.590 | 0.609 | 0.627 | 0.647 | 0.667 | 0.686 | 0.710 | 0.731 | 0.742 | 0.778 | 0.801 | 0.828 | 0.854 | 0.880 | 0.905 | 0.936 | 0.967 |
| 13,000 | 0.496 | 0.514 | 0.532 | 0.552 | 0.571 | 0.592 | 0.612 | 0.634 | 0.657 | 0.680 | 0.704 | 0.729 | 0.756 | 0.782 | 0.809 | 0.839 | 0.875 | 0.905 | 0.932 | 0.965 |
| 14,000 | 0.474 | 0.492 | 0.511 | 0.530 | 0.551 | 0.573 | 0.600 | 0.617 | 0.640 | 0.669 | 0.690 | 0.719 | 0.746 | 0.771 | 0.802 | 0.832 | 0.861 | 0.895 | 0.929 | 0.964 |
| 15,000 | 0.457 | 0.476 | 0.495 | 0.515 | 0.536 | 0.558 | 0.580 | 0.603 | 0.627 | 0.653 | 0.678 | 0.705 | 0.733 | 0.761 | 0.791 | 0.822 | 0.854 | 0.892 | 0.927 | 0.962 |
| 16,000 | 0.444 | 0.462 | 0.483 | 0.502 | 0.524 | 0.546 | 0.569 | 0.592 | 0.617 | 0.642 | 0.668 | 0.699 | 0.724 | 0.754 | 0.785 | 0.817 | 0.851 | 0.885 | 0.922 | 0.961 |
| 17,000 | 0.436 | 0.456 | 0.470 | 0.496 | 0.512 | 0.539 | 0.557 | 0.586 | 0.606 | 0.636 | 0.662 | 0.690 | 0.716 | 0.749 | 0.778 | 0.813 | 0.846 | 0.883 | 0.920 | 0.959 |
| 18,000 | 0.426 | 0.445 | 0.465 | 0.486 | 0.507 | 0.529 | 0.553 | 0.577 | 0.602 | 0.628 | 0.655 | 0.683 | 0.712 | 0.743 | 0.775 | 0.807 | 0.841 | 0.879 | 0.918 | 0.958 |
| 19,000 | 0.413 | 0.433 | 0.453 | 0.474 | 0.495 | 0.518 | 0.541 | 0.566 | 0.591 | 0.618 | 0.645 | 0.674 | 0.706 | 0.736 | 0.769 | 0.803 | 0.839 | 0.876 | 0.915 | 0.957 |
| 20,000 | 0.405 | 0.424 | 0.444 | 0.465 | 0.487 | 0.510 | 0.533 | 0.558 | 0.584 | 0.611 | 0.638 | 0.667 | 0.698 | 0.730 | 0.763 | 0.798 | 0.835 | 0.873 | 0.913 | 0.955 |
| 21,000 | 0.394 | 0.414 | 0.434 | 0.455 | 0.477 | 0.497 | 0.524 | 0.549 | 0.575 | 0.602 | 0.631 | 0.661 | 0.692 | 0.723 | 0.758 | 0.794 | 0.830 | 0.870 | 0.911 | 0.954 |
| 22,000 | 0.380 | 0.400 | 0.420 | 0.442 | 0.464 | 0.488 | 0.512 | 0.537 | 0.564 | 0.592 | 0.620 | 0.653 | 0.684 | 0.717 | 0.750 | 0.787 | 0.825 | 0.866 | 0.909 | 0.953 |
| 23,000 | 0.371 | 0.392 | 0.412 | 0.431 | 0.454 | 0.480 | 0.505 | 0.530 | 0.555 | 0.585 | 0.614 | 0.645 | 0.677 | 0.709 | 0.746 | 0.783 | 0.822 | 0.863 | 0.906 | 0.952 |
| 24,000 | 0.334 | 0.355 | 0.377 | 0.400 | 0.424 | 0.452 | 0.478 | 0.505 | 0.530 | 0.560 | 0.591 | 0.623 | 0.657 | 0.693 | 0.730 | 0.778 | 0.812 | 0.860 | 0.904 | 0.951 |
| 25,000 | 0.326 | 0.347 | 0.368 | 0.392 | 0.416 | 0.442 | 0.467 | 0.494 | 0.523 | 0.554 | 0.584 | 0.617 | 0.651 | 0.687 | 0.725 | 0.765 | 0.808 | 0.852 | 0.899 | 0.950 |
| 30,000 | 0.286 | 0.308 | 0.330 | 0.354 | 0.378 | 0.404 | 0.431 | 0.459 | 0.489 | 0.520 | 0.552 | 0.587 | 0.623 | 0.661 | 0.701 | 0.744 | 0.789 | 0.837 | 0.888 | 0.943 |
| 40,000 | 0.196 | 0.218 | 0.242 | 0.266 | 0.292 | 0.319 | 0.348 | 0.379 | 0.411 | 0.445 | 0.481 | 0.519 | 0.560 | 0.605 | 0.649 | 0.705 | 0.750 | 0.811 | 0.869 | 0.932 |
| 50,000 | 0.157 | 0.177 | 0.199 | 0.222 | 0.247 | 0.274 | 0.303 | 0.327 | 0.365 | 0.400 | 0.437 | 0.472 | 0.515 | 0.564 | 0.613 | 0.663 | 0.721 | 0.782 | 0.852 | 0.922 |
| 60,000 | 0.117 | 0.138 | 0.161 | 0.185 | 0.211 | 0.239 | 0.268 | 0.300 | 0.333 | 0.369 | 0.407 | 0.448 | 0.498 | 0.540 | 0.595 | 0.649 | 0.705 | 0.772 | 0.841 | 0.918 |
| 70,000 | 0.088 | 0.108 | 0.131 | 0.155 | 0.181 | 0.209 | 0.239 | 0.271 | 0.310 | 0.342 | 0.381 | 0.423 | 0.468 | 0.520 | 0.572 | 0.629 | 0.690 | 0.757 | 0.831 | 0.913 |
| 80,000 | 0.072 | 0.092 | 0.113 | 0.137 | 0.162 | 0.190 | 0.219 | 0.251 | 0.285 | 0.321 | 0.361 | 0.403 | 0.449 | 0.498 | 0.552 | 0.611 | 0.675 | 0.745 | 0.822 | 0.908 |
| 90,000 | 0.061 | 0.097 | 0.100 | 0.123 | 0.165 | 0.176 | 0.206 | 0.251 | 0.264 | 0.308 | 0.348 | 0.391 | 0.437 | 0.487 | 0.531 | 0.592 | 0.658 | 0.737 | 0.812 | 0.901 |
| 100,000 | 0.060 | 0.077 | 0.097 | 0.119 | 0.144 | 0.170 | 0.199 | 0.231 | 0.264 | 0.300 | 0.339 | 0.381 | 0.430 | 0.479 | 0.535 | 0.595 | 0.658 | 0.731 | 0.812 | 0.901 |
| 150,000 | 0.021 | 0.022 | 0.033 | 0.049 | 0.071 | 0.096 | 0.124 | 0.156 | 0.191 | 0.229 | 0.279 | 0.315 | 0.371 | 0.425 | 0.468 | 0.540 | 0.614 | 0.698 | 0.784 | 0.885 |
| 200,000 | 0.062 | 0.029 | 0.019 | 0.023 | 0.036 | 0.057 | 0.083 | 0.113 | 0.148 | 0.186 | 0.228 | 0.275 | 0.326 | 0.381 | 0.439 | 0.511 | 0.586 | 0.673 | 0.767 | 0.874 |
| 250,000 | 0.188 | 0.098 | 0.053 | 0.035 | 0.035 | 0.047 | 0.067 | 0.093 | 0.125 | 0.162 | 0.203 | 0.249 | 0.300 | 0.357 | 0.420 | 0.490 | 0.568 | 0.657 | 0.759 | 0.871 |
| 300,000 | 0.337 | 0.187 | 0.106 | 0.068 | 0.054 | 0.057 | 0.070 | 0.093 | 0.122 | 0.156 | 0.196 | 0.241 | 0.291 | 0.348 | 0.411 | 0.481 | 0.560 | 0.650 | 0.752 | 0.867 |
| 400,000 | 0.528 | 0.292 | 0.167 | 0.103 | 0.074 | 0.067 | 0.076 | 0.094 | 0.121 | 0.154 | 0.193 | 0.239 | 0.288 | 0.345 | 0.407 | 0.478 | 0.558 | 0.647 | 0.749 | 0.866 |
| 500,000 | 0.701 | 0.385 | 0.218 | 0.131 | 0.089 | 0.075 | 0.078 | 0.094 | 0.118 | 0.150 | 0.188 | 0.232 | 0.281 | 0.338 | 0.400 | 0.471 | 0.552 | 0.642 | 0.746 | 0.865 |

## ERP Credibility: Current vs. Revised




## Credibilities and Maximum Values

|  | Expected Losses |  | Credibility* | Split Point* | Expected Losses |  | Credibility | Split Point | Expected Losses |  | Credibility | Split Point |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | Low | High |  |  | Low | High |  |  | Low | High |  |  |
|  | - | 5,000 | 0.690 | 10,000 | 382,034 | 405,008 | 0.785 | 73,000 | 1,422,700 | 1,467,472 | 0.881 | 190,000 |
|  | 5,000 | 11,097 | 0.692 | 11,000 | 405,008 | 428,814 | 0.788 | 75,000 | 1,467,472 | 1,513,704 | 0.884 | 194,000 |
|  | 11,097 | 17,683 | 0.694 | 13,000 | 428,814 | 453,416 | 0.791 | 77,000 | 1,513,704 | 1,561,526 | 0.887 | 198,000 |
|  | 17,683 | 23,953 | 0.697 | 15,000 | 453,416 | 478,780 | 0.794 | 80,000 | 1,561,526 | 1,611,076 | 0.890 | 202,000 |
|  | 23,953 | 29,924 | 0.699 | 17,000 | 478,780 | 504,867 | 0.797 | 83,000 | 1,611,076 | 1,662,502 | 0.893 | 206,000 |
|  | 29,924 | 35,614 | 0.701 | 19,000 | 504,867 | 531,643 | 0.800 | 86,000 | 1,662,502 | 1,715,957 | 0.896 | 210,000 |
|  | 35,614 | 41,041 | 0.703 | 21,000 | 531,643 | 559,072 | 0.803 | 89,000 | 1,715,957 | 1,771,606 | 0.899 | 215,000 |
|  | 41,041 | 55,902 | 0.706 | 23,000 | 559,072 | 587,119 | 0.806 | 92,000 | 1,771,606 | 1,829,621 | 0.902 | 220,000 |
|  | 55,902 | 68,958 | 0.711 | 25,000 | 587,119 | 615,751 | 0.809 | 95,000 | 1,829,621 | 1,890,183 | 0.905 | 225,000 |
|  | 68,958 | 80,590 | 0.715 | 27,000 | 615,751 | 644,938 | 0.812 | 98,000 | 1,890,183 | 1,953,479 | 0.908 | 230,000 |
|  | 80,590 | 91,141 | 0.718 | 29,000 | 644,938 | 674,652 | 0.815 | 102,000 | 1,953,479 | 2,019,709 | 0.911 | 235,000 |
|  | 91,141 | 100,920 | 0.722 | 31,000 | 674,652 | 704,871 | 0.818 | 106,000 | 2,019,709 | 2,089,078 | 0.914 | 240,000 |
|  | 100,920 | 110,201 | 0.725 | 33,000 | 704,871 | 735,573 | 0.821 | 110,000 | 2,089,078 | 2,161,801 | 0.917 | 245,000 |
|  | 110,201 | 119,228 | 0.728 | 35,000 | 735,573 | 766,742 | 0.824 | 114,000 | 2,161,801 | 2,238,101 | 0.920 | 250,000 |
|  | 119,228 | 128,218 | 0.731 | 37,000 | 766,742 | 798,366 | 0.827 | 118,000 | 2,238,101 | 2,318,210 | 0.923 | 255,000 |
|  | 128,218 | 137,358 | 0.734 | 39,000 | 798,366 | 830,440 | 0.830 | 122,000 | 2,318,210 | 2,402,367 | 0.926 | 260,000 |
|  | 137,358 | 146,813 | 0.737 | 41,000 | 830,440 | 862,961 | 0.833 | 126,000 | 2,402,367 | 2,490,821 | 0.929 | 265,000 |
|  | 146,813 | 156,724 | 0.740 | 43,000 | 862,961 | 895,933 | 0.836 | 130,000 | 2,490,821 | 2,583,829 | 0.932 | 270,000 |
|  | 156,724 | 167,212 | 0.743 | 45,000 | 895,933 | 929,367 | 0.839 | 134,000 | 2,583,829 | 2,681,655 | 0.935 | 275,000 |
|  | 167,212 | 178,379 | 0.746 | 47,000 | 929,367 | 963,278 | 0.842 | 138,000 | 2,681,655 | 2,784,572 | 0.938 | 280,000 |
|  | 178,379 | 190,306 | 0.749 | 49,000 | 963,278 | 997,690 | 0.845 | 142,000 | 2,784,572 | 2,892,863 | 0.941 | 285,000 |
|  | 190,306 | 203,062 | 0.752 | 51,000 | 997,690 | 1,032,631 | 0.848 | 146,000 | 2,892,863 | 3,006,815 | 0.944 | 290,000 |
|  | 203,062 | 216,698 | 0.755 | 53,000 | 1,032,631 | 1,068,138 | 0.851 | 150,000 | 3,006,815 | 3,126,727 | 0.947 | 295,000 |
|  | 216,698 | 231,254 | 0.758 | 55,000 | 1,068,138 | 1,104,253 | 0.854 | 154,000 | 3,126,727 | 3,252,905 | 0.950 | 300,000 |
|  | 231,254 | 246,756 | 0.761 | 57,000 | 1,104,253 | 1,141,026 | 0.857 | 158,000 | 3,252,905 | 3,385,661 | 0.953 | 300,000 |
|  | 246,756 | 263,220 | 0.764 | 59,000 | 1,141,026 | 1,178,516 | 0.860 | 162,000 | 3,385,661 | 3,525,316 | 0.956 | 300,000 |
|  | 263,220 | 280,654 | 0.767 | 61,000 | 1,178,516 | 1,216,788 | 0.863 | 166,000 | 3,525,316 | 3,672,201 | 0.959 | 300,000 |
|  | 280,654 | 299,053 | 0.770 | 63,000 | 1,216,788 | 1,255,914 | 0.866 | 170,000 | 3,672,201 | 3,826,650 | 0.962 | 300,000 |
|  | 299,053 | 318,410 | 0.773 | 65,000 | 1,255,914 | 1,295,976 | 0.869 | 174,000 | 3,826,650 | 3,989,009 | 0.965 | 300,000 |
|  | 318,410 | 338,707 | 0.776 | 67,000 | 1,295,976 | 1,337,061 | 0.872 | 178,000 | 3,989,009 | 4,159,630 | 0.968 | 300,000 |
|  | 338,707 | 359,924 | 0.779 | 69,000 | 1,337,061 | 1,379,268 | 0.875 | 182,000 | 4,159,630 | 4,338,871 | 0.971 | 300,000 |
| * Preliminary values | 359,924 | 382,034 | 0.782 | 71,000 | 1,379,268 | 1,422,700 | 0.878 | 186,000 | 4,338,871 | Above | 0.974 | 300,000 |

## Updated ERP Performance Test

The following lift charts are produced using the optimized elements.

2015-2018
Deviation within $+/-3 \%$



## Optimization Results Summary

- Smaller risks receive markedly higher credibility.
- Larger risks receive nominally increased credibility.
- Maximum Values for smaller risks are lower.
- Maximum Values for larger risks are higher.
- The efficiency test indicates very good results with a deviation of less than $5 \%$ from unity.



## Policy Count and Premium Distribution

Policy Count


Premium


## Distribution of Mod



## Change in Mod Type



## Change in Modified Premium






* Based on PY 2018 data. Premium based on mods after swing limits are applied.


## Current Capping Rules

1. Final Modification Capped to $+/-25 \%$ of Prior Modification.
2. If the indicated modification is less than unity (1.000) and the capped modification is greater than unity (1.000), then the final modification shall be set equal to unity (1.000).

Example

| Rule | Prior Mod | Indicated Mod | Capped Mod | Final Capped Mod |
| :---: | :---: | :---: | :---: | :---: |
| 2 | 1.563 | 0.827 | 1.172 | 1.000 |

## Leading Option for Revised Capping Method

- Introduction of a Maximum Mod formula for capping, which also includes a $+40 \%$ cap to limit larger annual upward movements. This replaces the current $+/-25 \%$ swing limit approach.
- The current capping rules (+/-25\% swing limits and secondary capping) will also be kept for a two-year period to ensure mod stability as we transition to the new plan.


## Example - Capping

## Max Mod $=1.10+0.0004(E / G)$

| Exp Loss | 5,000 | 10,000 | 25,000 | 50,000 | 250,000 | 500,000 | 1 M |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| Max Mod <br> $(\mathrm{G}=10)$ | 1.30 | 1.50 | 2.10 | 3.10 | 11.10 | 21.10 | 41.10 |

## Capping Examples

| Scenario | Expected Loss | Prior Mod | Indicated | Capped Mod | Max Mod | Final Capped <br> Mod YR 1 | Final Capped <br> Mod YR 2 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Transition | $\$ 10,000$ | 1.02 | 1.60 | $1.28(+25 \%)$ | 1.50 | 1.28 | 1.50 |
| After Transition | $\$ 10,000$ | 1.02 | 1.60 | $1.43(+40 \%)$ | 1.50 | 1.43 | 1.50 |

## Policies Capped

| Expected Loss | Current Plan (+/-25\%) | Updated Plan (+/-25\%) | Updated Plan (Max Mod \& +40\%) |
| :---: | :---: | :---: | :---: |
| Credit | $2 \%$ | $3 \%$ | $0 \%$ |
| Debit | $8 \%$ | $11 \%$ | $8 \%$ |
| Neutral | $1.5 \%$ | $1.6 \%$ | $0 \%$ |
| Total | $\mathbf{1 2 \%}$ | $\mathbf{1 6 \%}$ | $\mathbf{8 \%}$ |

The majority of capped risks are smaller risks that are below $\$ 25,000$.

## Capping Distribution by Expected Loss

|  | EL < \$5,000 |  | \$5,000 < EL < \$10,000 |  | \$10,000 < EL < \$25,000 |  | \$25,000 < EL < \$50,000 |  | \$50,000 < EL |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| Mod Range | Current | Updated | Current | Updated | Current | Updated | Current | Updated | Current | Updated |
| 0-0.6 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 2 | 488 |
| 0.6-0.8 | 0 | 0 | 1 | 0 | 0 | 1,011 | 3 | 4,147 | 3,601 | 4,759 |
| 0.8-0.9 | 12 | 7,695 | 6808 | 16,762 | 15,434 | 14,189 | 6,363 | 2,119 | 3,290 | 1,948 |
| 0.9-1.0 | 9,259 | 499 | 13,558 | 1452 | 2,232 | 1534 | 1,178 | 789 | 2,117 | 1,790 |
| 1.0-1.1 | 1 | 201 | 229 | 557 | 845 | 748 | 610 | 539 | 1,708 | 1,588 |
| 1.1-1.2 | 0 | 114 | 155 | 311 | 545 | 496 | 422 | 862 | 1,315 | 1,079 |
| 1.2-1.5 | 0 | 155 | 245 | 540 | 975 | 2049 | 1,363 | 1323 | 1,677 | 1,803 |
| 1.5-2.0 | 1 | 127 | 147 | 1267 | 1287 | 1325 | 433 | 563 | 610 | 771 |
| 2.0-3.0 | 0 | 417 | 354 |  | 349 | 314 | 72 | 101 | 90 | 176 |
| >3.0 | 0 | 65 | 34 | 83 | 40 | 41 | 6 | (7) | 6 | 14 |

NCCI Max Mod $=1.10+0.0004(E / G)$

| G | 10 | 10 | 10 | 10 | 10 | 10 | 10 |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| E (Exp Loss) | 0 | 5,000 | 10,000 | 25,000 | 50,000 | 250,000 | 500,000 |  |
| Max Mod | 1.10 | 1.30 | 1.50 | 2.10 | 3.10 | 11.10 | 21.10 |  |
| Min Mod | 0.85 | 0.85 | 0.84 | 0.79 | 0.76 | 0.59 | 0.50 | $P C / R \mid B$ <br> PENNSYLVANIA |

[^0]
## Mod and Premium Change



- The percentage change is calculated from the 2017 capped mod to the 2018 indicated mod, based on the new plan's mod calculation method.


## Capping and Eligibility Summary

## Capping

Necessary to maintain a form of capping, particularly for smaller risks, to promote mod stability.

Most states use a Maximum Mod approach to limit the upside volatility primarily on smaller risks.

The Maximum Mod helps prevent the mod from becoming excessively high due to one or more large losses that may not accurately reflect overall loss experience.

The Maximum Mod approach primarily benefits small risks while swing limits provides stability for all risks.

## Eligibility

Analyzed to address the transition between the Merit Plan and ERP plan for smaller risks.

Results show that the use of a Maximum Mod formula and potentially a lower eligibility amount improve the transition between the two plans.

Lowering the eligibility to $\$ 5,000$ results $11 \%$ more risks being experience rated.

## Appendix

## Recommendations

|  | Current | Updated |
| :---: | :---: | :---: |
| Plan | Single Split Point | Variable Split Point |
| Eligibility | $\$ 10,000$ | $\$ 5,000$ |
| Credibility | $0.283-0.938$ | $0.690-0.974$ |
| Expected Loss Range | $10,706-5,806,852$ | $5,000-4,338,871$ |
| Split Points | Single (1): $\$ 42,500$ | Variable (88): $\$ 10,000-\$ 300,000$ |
| Med-Only Claims | $100 \%$ | $100 \%$ |
| Capping \% | $+-25 \%$ | Max Mod and 40\% swing limit <br> (2-year Transition) |
| Secondary Capping | Capping Rule \#4 | Eliminate (After Transition Period) |

## Mod Point Change (MaxMod and $+40 \%$ Swing Limit)




## Policies Capped (Max Mod and $+40 \%$ Swing Limit)



Note: "Capped Mod" is based on using Max Mod and $+40 \%$ swing limit capping and "Transition Mod" uses both Max Mod and $+/-25 \%$ swing limit capping used during a transition period.


[^0]:    * Based on PY 2018 data

