

Addressing Social Inflation

definition, drivers
and mitigation strategies

Actuarial Seminar MFF UK,
May 23, 2025
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- We combine **reinsurance broking, capital solutions and strategic advisory** underpinned by insight driven analytics to deliver tailor made solutions for our clients.

One global team

100+

Professionals working in agile teams

3

Strategic hubs: London, Bermuda and the US

Price Forbes Re **Analytics and Advisory**

The capabilities of our team covers several disciplines including **Actuarial, Data Science, Portfolio Management, Cat Modelling and Credit Advisory** services with an average of 10 years' experience gained at both broking and carrier organisations.

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>\$18bn+

Gross Written Premium

12,000+

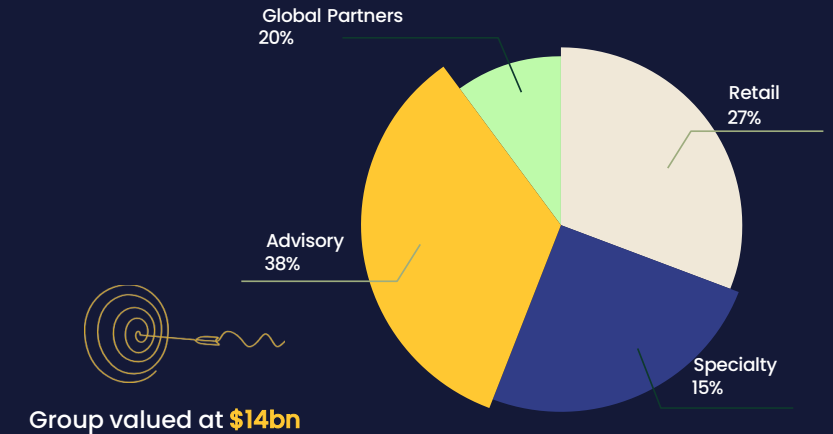
Employees

250+

Offices Worldwide

30+

Territories



The Ardonagh Group

A UK-based, private equity-backed insurance distribution platform offering a wide range of insurance products and services across the UK, Ireland, and international markets. Founded in 2017, it has grown rapidly through acquisitions, including its largest deal with PSC Insurance Group in 2024.

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- Definition and types of inflation
- Drivers of social inflation and managing risks associated with social inflation



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- Implicit approaches: Chain-Ladder, Bornhuetter-Ferguson
- Explicit approaches: Inflation-adjusted Chain-Ladder, PTF model, general



03

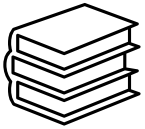
Reinsurance products that address social inflation

- Indexation clause
- Adverse development cover and loss portfolio transfer

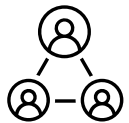


Introduction to Social Inflation

Defining social inflation

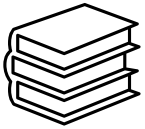


Social inflation refers to the **increasing** severity of liability **claims beyond what can be explained by economic factors**, driven by societal trends.

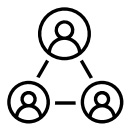


The 'social' part of the term reflects **changing social as well as legal norms and beliefs** regarding who should bear the responsibility for absorbing the risk of injury or loss as well as the entitlement to compensation¹.

Defining social inflation



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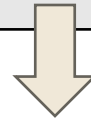
The 'social' part of the term reflects **changing social as well as legal norms and beliefs** regarding who should bear the responsibility for absorbing the risk of injury or loss as well as the entitlement to compensation¹.

As a concept, it dates to at least 1959 when, in an actuarial journal, F.S. Perryman asked whether rising loss costs and claim frequencies in automobile insurance could be attributed **"to reasons which are not economic but social?"**

Perryman, F. S. Discussion of Compulsory Automobile Insurance in Europe, in Proceedings of the Casualty Actuarial Society, vol. 46, 1959

Types of inflation

Economic inflation	Super-imposed inflation	Social inflation
A sustained increase in the price level of goods and services .	Additional inflation specific to certain sectors, beyond general economic inflation , often due to regulatory changes or increased costs.	A part of super-imposed inflation: a rising insurance claim costs due to societal and cultural shifts .
Price levels are measured based on the price of a typical basket of goods and services . ²	Arises from the impact of trends such as: <ul style="list-style-type: none"> • Regulatory changes • Increased litigation costs • Technological advancements • Sector-specific cost increases • Supply chain disruptions 	Arises from the impact of societal trends such as: ² <ul style="list-style-type: none"> • Changing social attitudes • A rising willingness to resolve conflict via the legal system



New regulations can increase compliance and operational costs for specific sectors.

Regulatory changes



Rising legal fees and more frequent lawsuits drive up costs in industries like insurance and healthcare.

Increased litigation costs



The need for advanced technologies, such as cybersecurity, leads to higher operational expenses.

Technological advancements



Certain industries face unique cost pressures, such as higher raw material prices in manufacturing.

Sector-specific cost increases



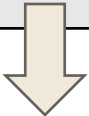
Interruptions in supply chains lead to higher costs for materials and goods.

Supply chain disruptions



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Drivers of Social Inflation

The following information on the drivers³ of social inflation and mitigation strategies⁴ is primarily based on the context of the **United States**.

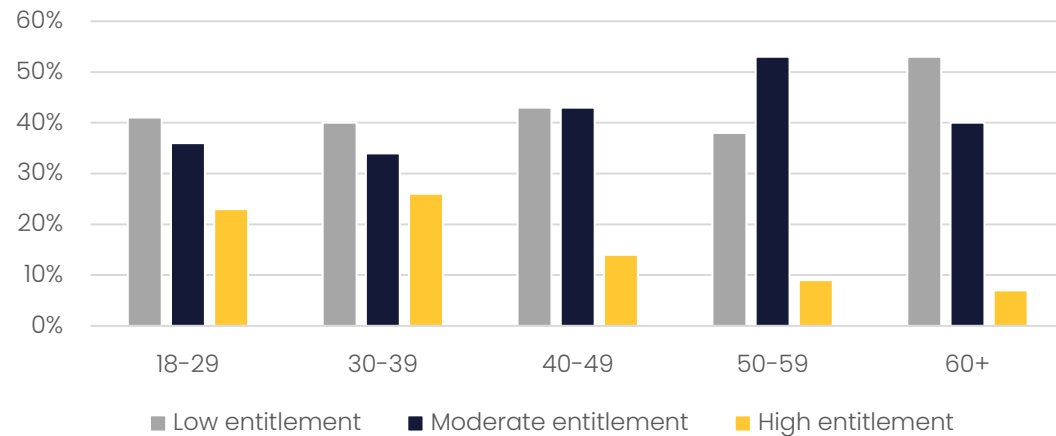
While these insights may not be directly applicable to the Czech Republic, we believe it is **valuable to be aware of these concepts** to better understand potential trends and strategies in managing social inflation risks.

Drivers of social inflation – part 1

Rising public expectations for compensation amounts

Research on entitlement attitudes and jury behaviour provides indirect evidence **that entitlement attitudes** may indeed be **increasing** throughout the general **population**.

Entitlement attitudes among mock jury participants
by age



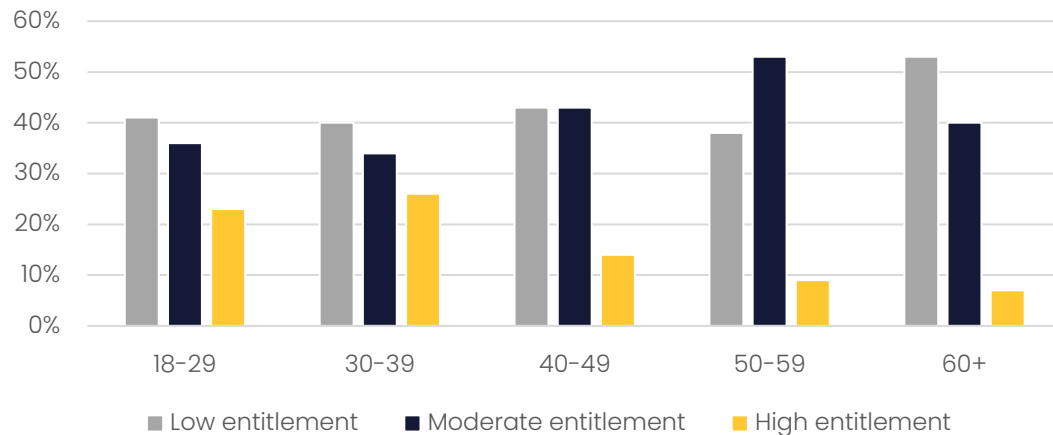
Source: Gary Giewat, Damage Awards: Jurors' Sense of Entitlement as a Predictor, The Jury Expert, May 30, 2011 (accessed April 27, 2020).

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Class action lawsuits

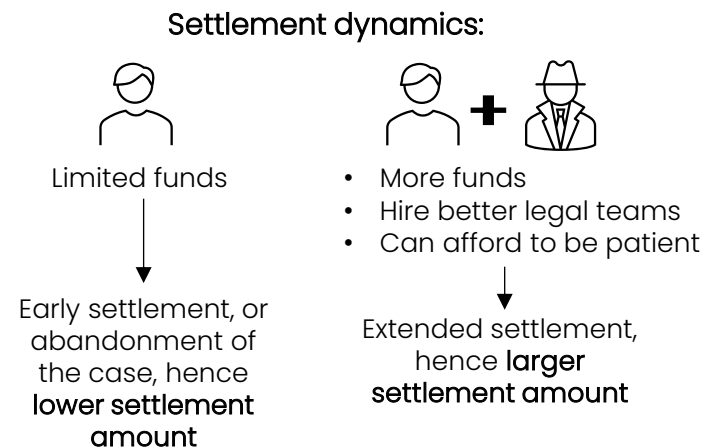
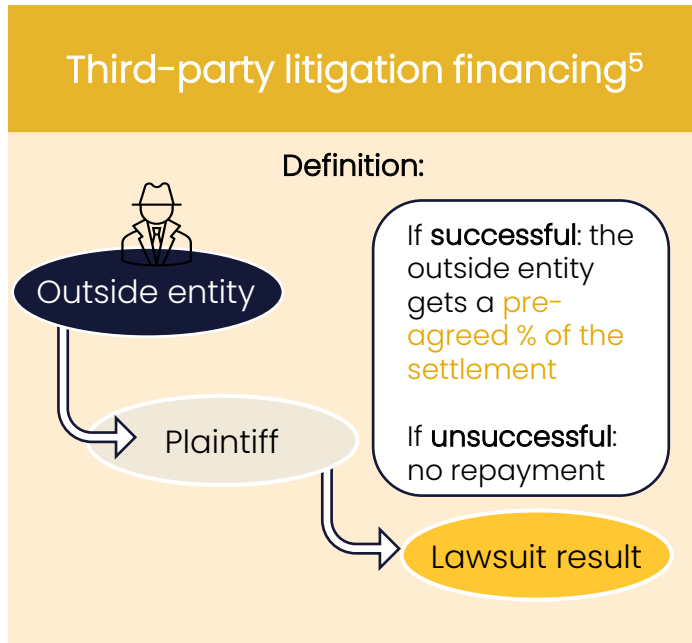
Multiple plaintiffs joining together to seek compensation from a single defendant.

Total U.S. corporate legal spending on class action lawsuits (billion \$)



Source: 2019 Carlton Fields Class Action Survey, Carlton Fields, April 16, 2019, www.classactionsurvey.com (accessed May 26, 2020).

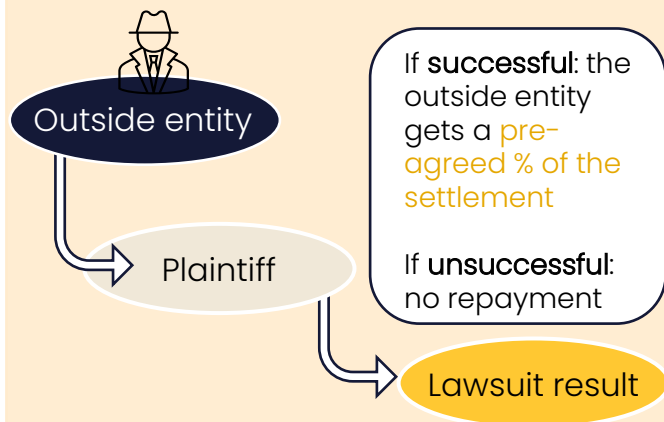
Drivers of social inflation – part 2



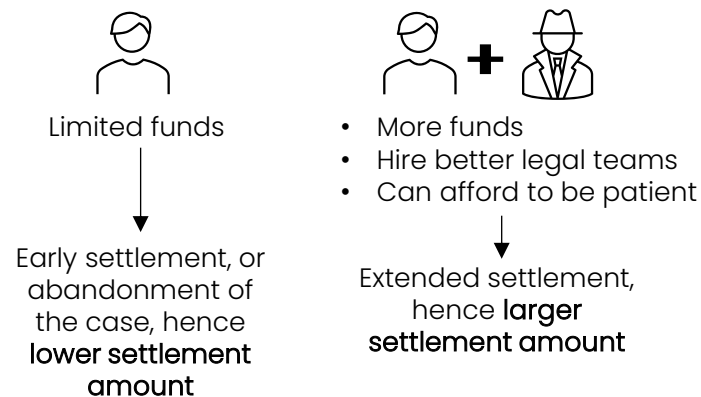
Drivers of social inflation – part 2

Third-party litigation financing⁵

Definition:



Settlement dynamics:



Nuclear verdicts⁶

Definition:

Extremely large jury awards (often exceeding \$10 million)

Include **strong emotional and punitive elements**, far surpassing actual economic losses.

➤ **Perception of deep pockets:** defendants, especially corporations, insurance companies and healthcare providers, are seen as having deep pockets and deserving punishment.

Interesting statistics:

Median

\$21m in 2020 and **\$44m** in 2023

Average

\$89m 2013-2022

Medical malpractice

Number of nuclear verdicts over \$25m: 4 in 2014 and **17** in 2018

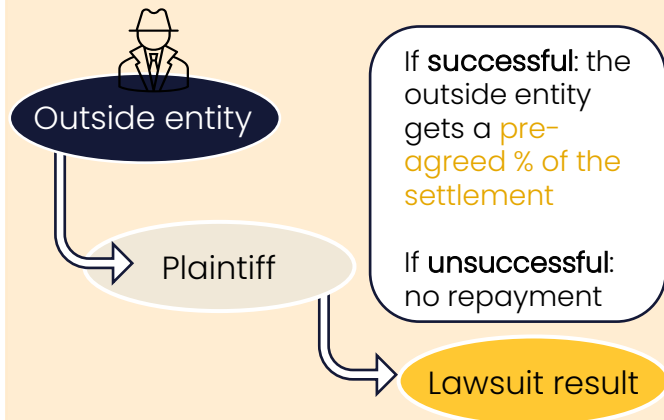
Commercial auto insurance

Against **trucking firms**. Average \$2.3m in 2012 and **\$22.3m** in 2019

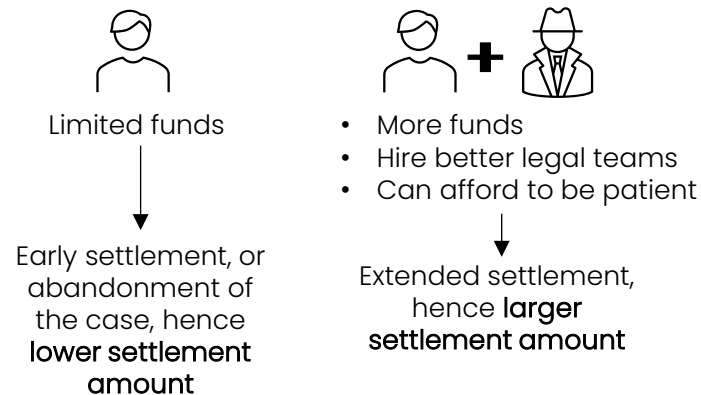
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Tort reforms rollbacks⁷

Tort reforms:

- **Legislative changes to reduce litigation costs** and make the civil justice system more predictable.
- Common reform: cap on non-economic damages (e.g., pain and suffering).

Rollbacks:

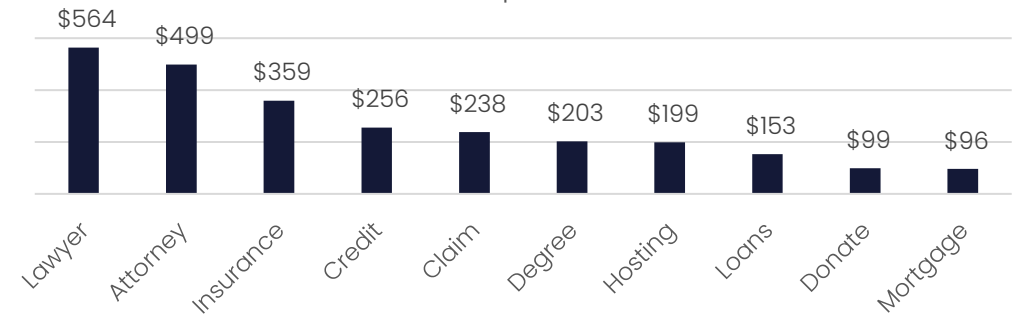
- **Some state supreme courts have overturned these caps**, allowing for higher non-economic damage awards.



Drivers of social inflation – part 3

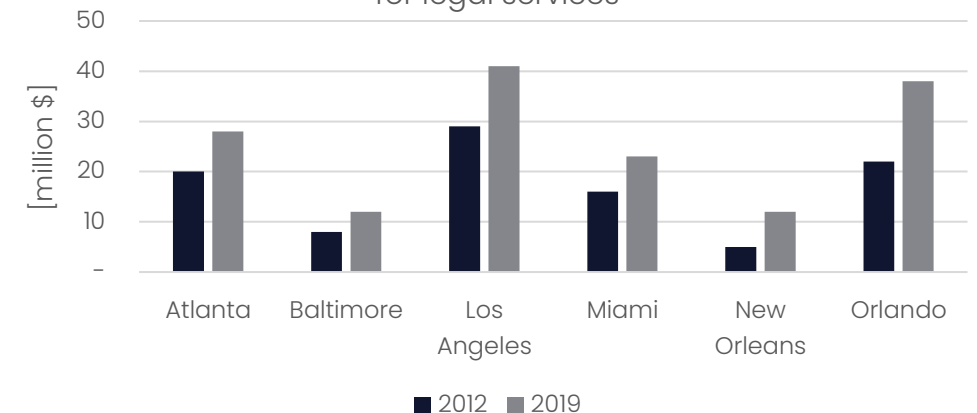


Most expensive Google AdWords keywords, costs per click



Source: Sam Carr, "The Most Expensive AdWords Keywords," PPC Protect, Dec. 30, 2019, ppcprotect.com/mostexpensive-adwords-keyword

Broadcast advertising spending for legal services



Source: Kantar Media

Future social inflation drivers

- The United States is projected to remain the **epicentre** of social inflation due to its **distinct legal and cultural dynamics**.
- However, countries like Australia, Canada, the UK, and **parts of Europe** also show potential liability claims growth driven by **third-party litigation funding and expanded collective redress**.⁸

	US	Australia	UK	Canada	Netherlands	France	Germany	Japan
Claims penetration (ratio of liability claims to a country's GDP)	H	M	H	M	L	M	M	L
Income inequality (is measured using standardized Gini coefficients from the SWIID)	H	M	M	M	L	M	M	M
Third-party litigation funding	H	H	H	M	H	M	M	L
Contingency fees (payment arrangement where a lawyer is only paid a % of the settlement if the case is won)	H	M	M	H	L	L	L	L
Collective redress	H	H	H	H	H	M	M	L
Case law (collection of past court decisions that judges use to help decide new cases with similar facts)	H	H	H	H	L	L	L	L
Jury based (group of citizens decides the outcome of a legal case instead of a judge making the decision alone)	H	L	L	L	L	L	L	L

High risk

Medium risk

Low risk

Source: Swiss Re Institute

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Which industries are most affected?

Social inflation
mostly affects
casualty and
property
insurance

Examples:

Commercial Auto
Insurance⁹

Trucking: high accident rates and the perception of negligence in trucking companies contribute to increased claims costs.



General Liability
Insurance⁹

Product liability: consumer goods and medical devices are frequent targets for class-action lawsuits, attracting TPLF firms and expanding the plaintiff pool through targeted advertising.



Financial and
Professional
Liability Insurance⁹

Intellectual property & trade secrets: litigation is increasingly used to reveal defendants' proprietary information, making companies with valuable intellectual property particularly vulnerable.



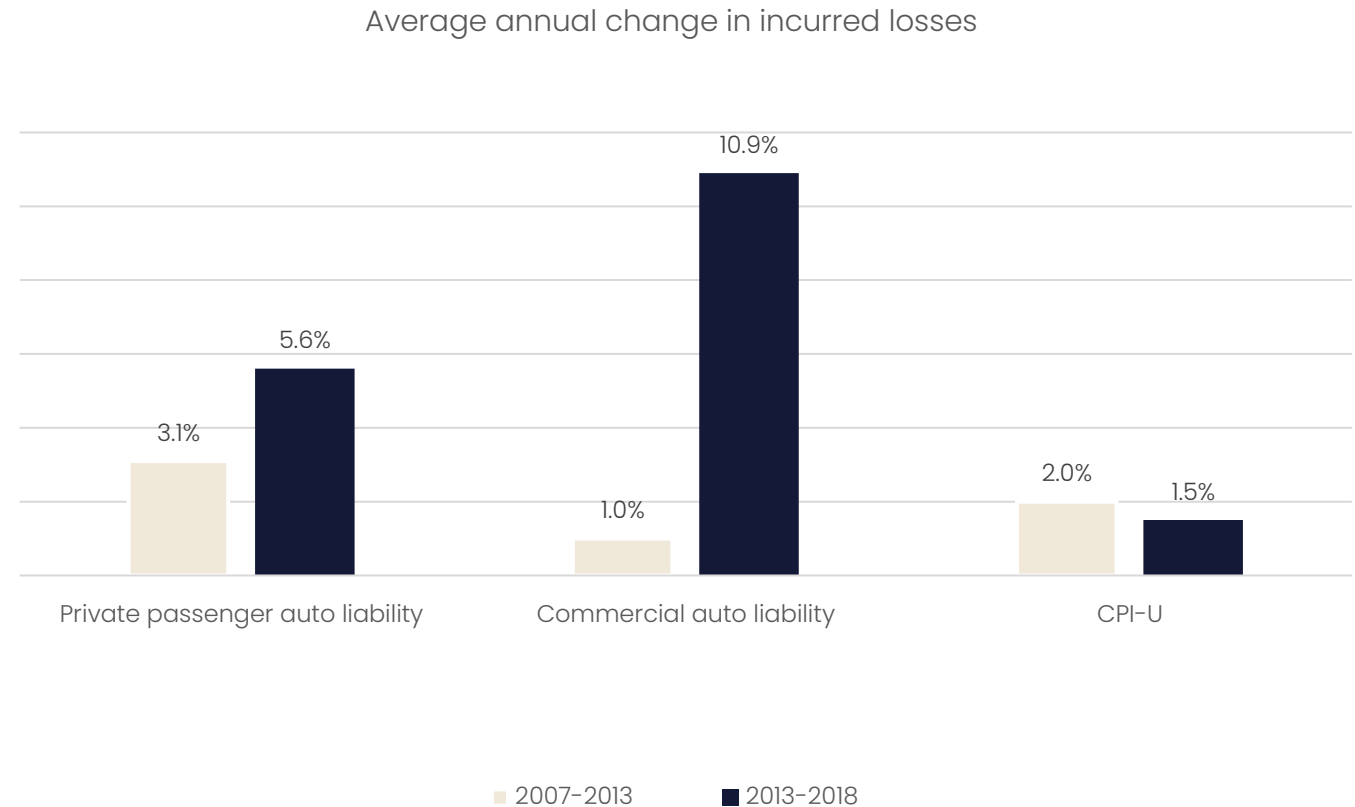
Marine and Energy
Liability¹⁰

Energy offshore liability: notable example of social inflation impacting climate change litigation.



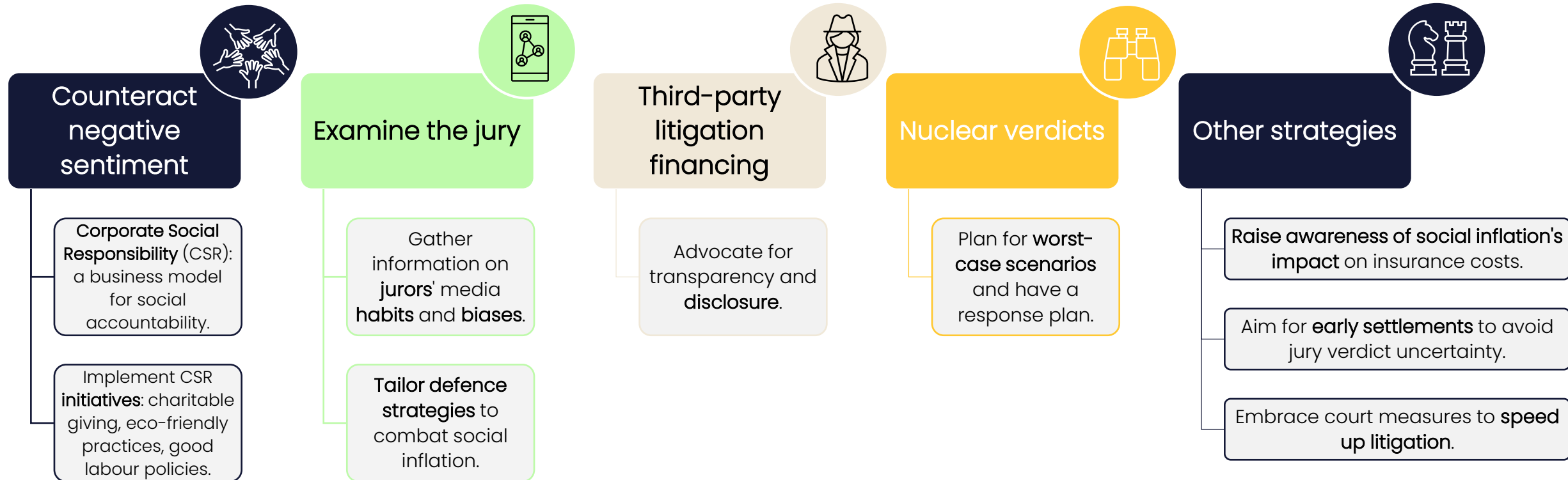
Which industries are most affected?

While social inflation has mainly impacted **commercial liability lines**, it is now beginning to affect **personal liability lines** as well.¹



Source: The Institutes Risk & Insurance Knowledge Group "Social Inflation: Evidence and Impact on Property-Casualty Insurance"

Mitigation of risks coming from the social inflation



Czech experience with social inflation

The New Civil Code: changes in damage compensation

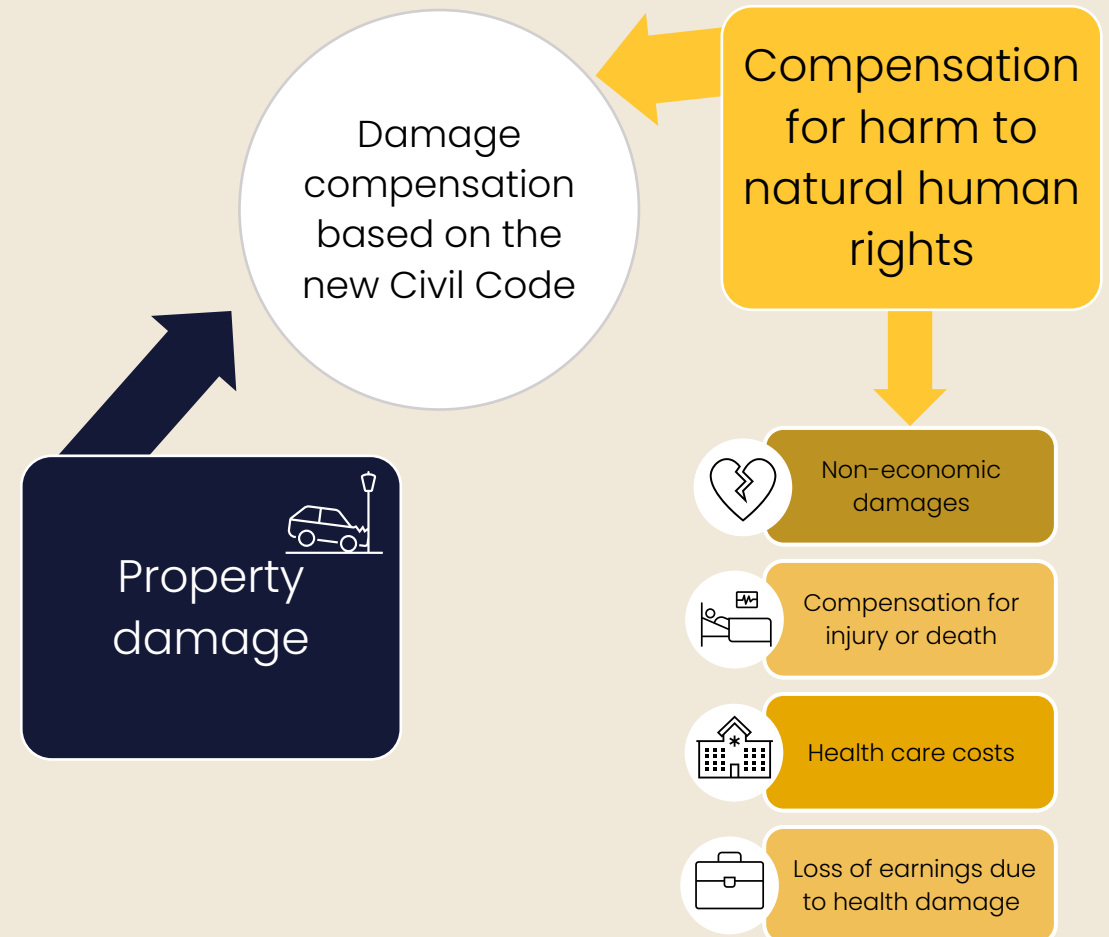
(Nový Občanský Zákoník VII: odpovědnost
za škodu, způsob a rozsah náhrady újmy)

Effective Date: January 1, 2014

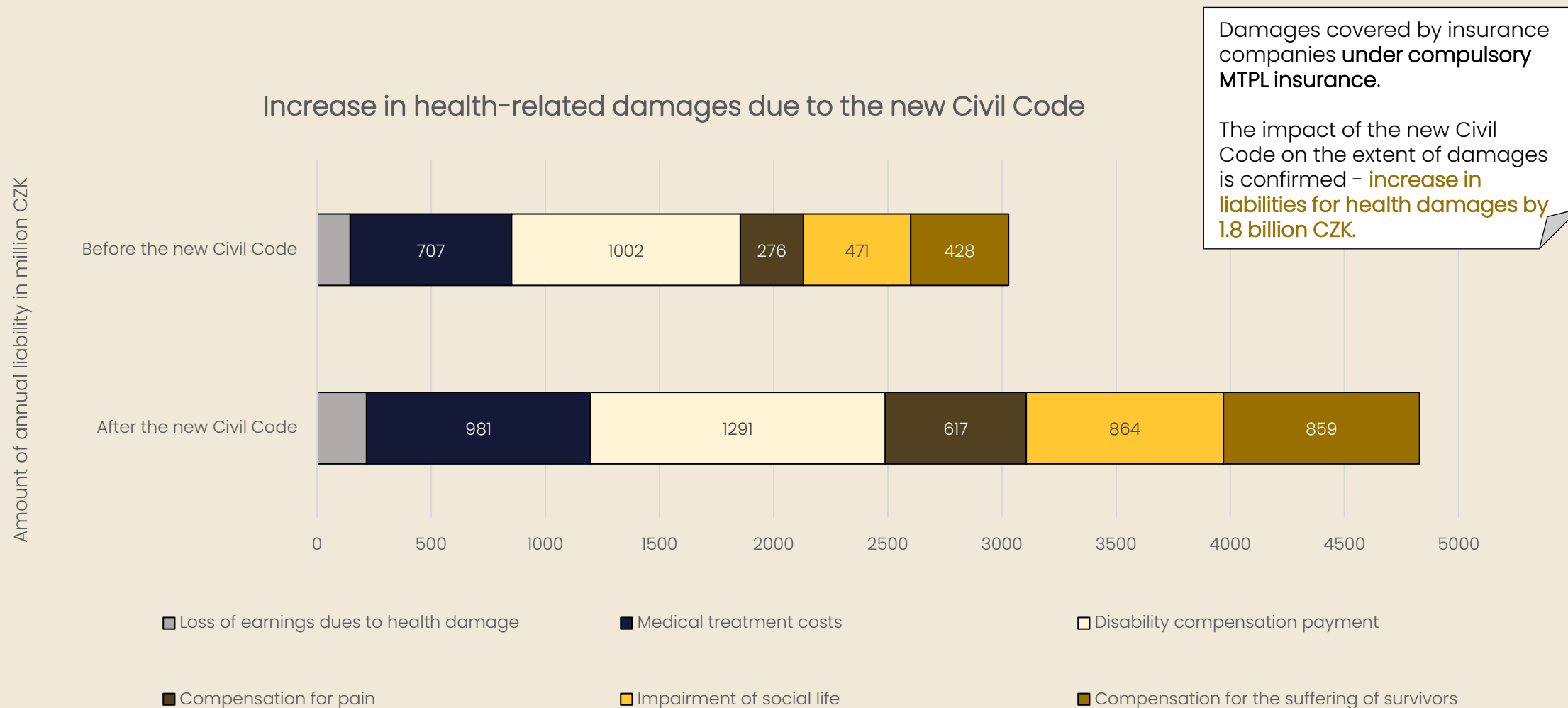
The new Civil Code redefines **damage** to include **both the reduction of assets** and the **increase of debts**. This change ensures that **individuals responsible for causing harm**:¹¹

- Must **compensate the injured party**, either by restoring their original financial state or by paying monetary compensation.
- Are now also **liable for debts incurred by the injured party** as a direct result of their wrongful actions

Previously, courts often did not recognize **debt obligations** caused by a wrongdoer as compensable damage.

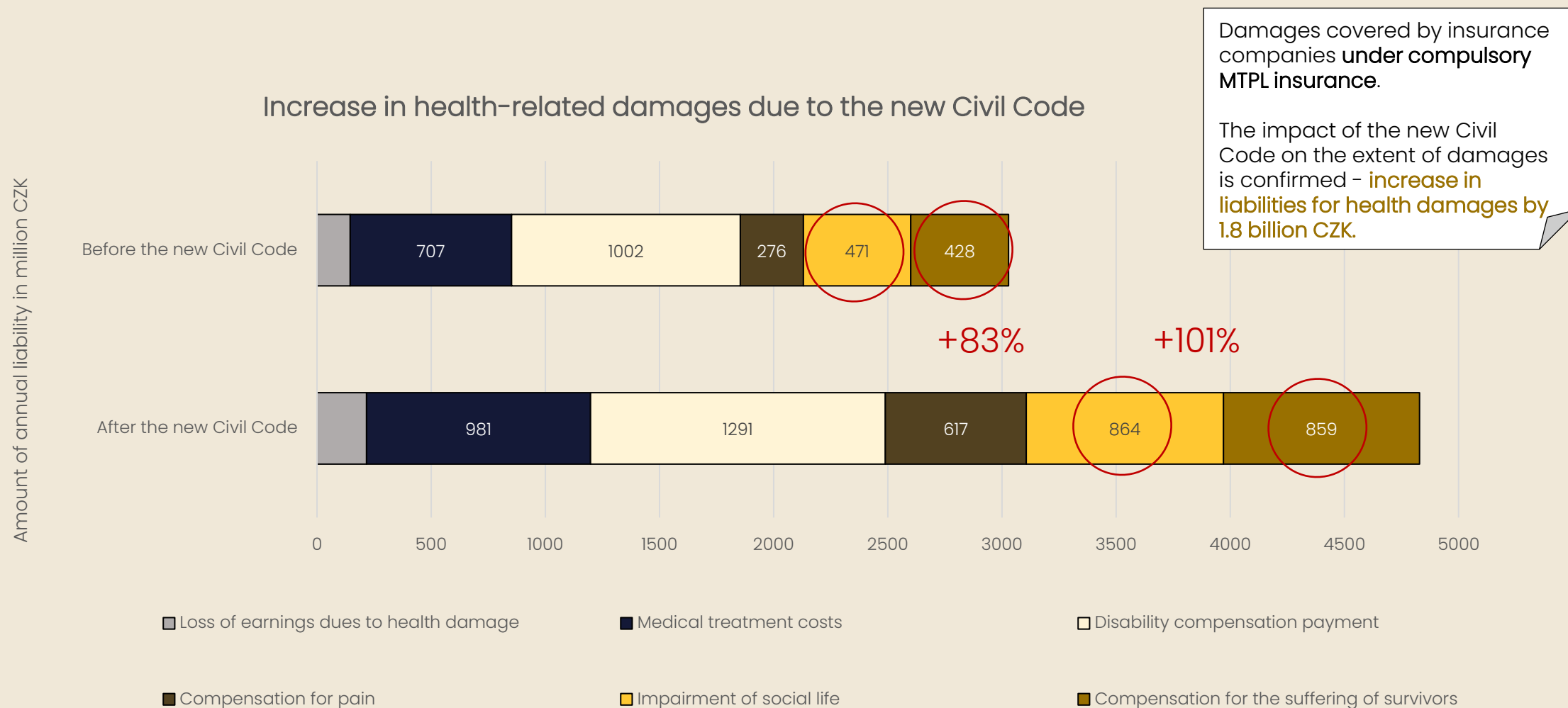


Czech experience with social inflation



Source: ČKP a ČAP Povinné ručení 2017, Tisková konference, Mgr. Jan Matoušek, RNDr. Petr Jedlička, Ph.D., 10. října 2017
https://www.ckp.cz/images/clanky/cz/tiskove_centrum/prezentace_ckp/2017/TK-POV-2017_superfin.pdf

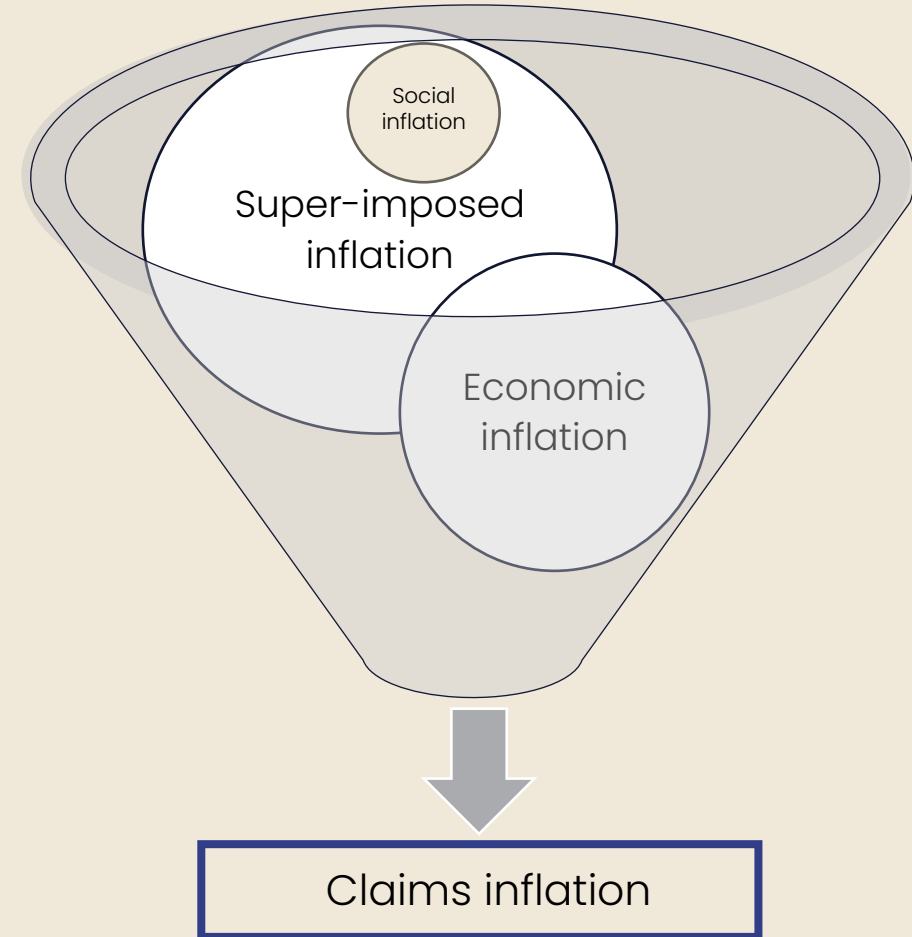
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https://www.ckp.cz/images/clanky/cz/tiskove_centrum/prezentace_ckp/2017/TK-POV-2017_superfin.pdf

Defining claims inflation

While calculating **social inflation** presents **challenges**, our primary focus remains on understanding the **overall impact of total inflation**.



Reserving Methodologies to Account for Claims Inflation

Implicit and explicit allowances for claims inflation in reserving

	Implicit allowances	Explicit allowances
Definition	Implicit allowances for claims inflation are made indirectly , without specifically identifying or quantifying the inflation component.	Explicit allowances for claims inflation involve direct identification and quantification of the inflation component in the reserving process.
Consideration	Suitable when it is assumed that past inflation rates will remain constant and match future inflation rates. ¹³	Suitable when past inflation rates are variable and future inflation is expected to differ . ¹³
Benefits	<ul style="list-style-type: none">• <u>Simplicity</u>: straightforward and easy to implement.• <u>Consistency</u>: ensures uniformity in reserving practices.• <u>Data efficiency</u>: requires less detailed inflation data.	<ul style="list-style-type: none">• <u>Accuracy</u>: provides precise estimates by directly accounting for inflation.• <u>Flexibility</u>: adapts to changing inflation rates and economic conditions.• <u>Transparency</u>: offers clear visibility into the impact of inflation on reserve estimates.
Drawbacks	<ul style="list-style-type: none">• <u>Assumption risk</u>: relies on the continuation of past trends, which may not always be accurate.• <u>Precision limitations</u>: may not adequately account for future inflation changes, leading to potential under- or over-reserving.• <u>Adaptability issues</u>: struggles to adjust to sudden economic shifts.	<ul style="list-style-type: none">• <u>Complexity</u>: more intricate and requires detailed inflation data.• <u>Data intensive</u>: needs comprehensive historical and projected inflation data.• <u>Resource demands</u>: implementation can be resource-intensive, requiring more time and expertise.

Break – 10 minutes

Notation for claims triangle

- ❑ The data in claims triangle are categorised by **accident year** $i \in \{0, \dots, I\}$ and **development year** $j \in \{0, \dots, J\}$
- ❑ We assume the most recent accident year equals the last development year¹⁴ $I = J$
- ❑ **Incremental data** $X_{i,j}$... claim amount in accident year i made in year $i + j$
- ❑ **Cumulative data** $C_{i,j}$ for accident year i after j development years:

$$C_{i,j} = \sum_{k=0}^j X_{i,k}$$

- ❑ **Incremental development triangle**

$$D_I = \{X_{i,j}: i + j \leq I, 0 \leq j \leq J\}$$

- ❑ **Cumulative development triangle**

$$D_I = \{C_{i,j}: i + j \leq I, 0 \leq j \leq J\}$$

Implicit Allowances for Claims Inflation

In this discussion, we will focus on the most used methods: the **Standard Chain Ladder** and **Bornhuetter-Ferguson**.

Standard Chain Ladder method model assumptions

- ❑ Cumulative claims $C_{i,j}$ of different accident years i are independent,¹⁴ i.e.

$\{C_{i,0}, \dots, C_{i,J}\}$ and $\{C_{k,0}, \dots, C_{k,J}\}$ are independent for $k \neq i$.

- ❑ The **exist development factors** $f_0, \dots, f_{J-1} > 0$ such that for all accident years $0 \leq i \leq I$ and all development years $0 \leq j \leq J$ we have

$$E[C_{i,j} | C_{i,0}, \dots, C_{i,j-1}] = f_{j-1} * C_{i,j-1}.$$

Then the expected value of aggregate loss for given accident year $1 \leq i \leq I$, conditional on the history known at the end of year:

$$E[C_{i,J} | D_I] = E[C_{i,J} | C_{i,I-i}] = C_{i,I-i} * f_{I-i} * \dots * f_{J-1}.$$

- ❑ There **exist parameters** $\sigma_0^2, \dots, \sigma_{J-1}^2 > 0$ such that for all accident years $0 \leq i \leq I$ and all development years $0 \leq j \leq J$ we have

$$\text{Var}[C_{i,j} | C_{i,0}, \dots, C_{i,j-1}] = \sigma_{j-1}^2 * C_{i,j-1}.$$

The last assumption is essential for derivation of an expression for the MSE of Chain-Ladder estimate:

$$\text{mse}(\hat{C}_{i,j} - C_{i,j}) = E[(\hat{C}_{i,j} - C_{i,j})^2 | D_I]$$

Standard Chain Ladder method

Paid/incurred cumulative claims triangle

$C_{0,0}$	$C_{0,1}$	$C_{0,2}$	$C_{0,3}$	$C_{0,4}$
$C_{1,0}$	$C_{1,1}$	$C_{1,2}$	$C_{1,3}$	
$C_{2,0}$	$C_{2,1}$	$C_{2,2}$		
$C_{3,0}$	$C_{3,1}$			
$C_{4,0}$				

Calculate estimates \hat{f}_j of development factors

$$\hat{f}_j = \frac{\sum_{i=0}^{I-j-1} C_{i,j+1}}{\sum_{i=0}^{I-j-1} C_{i,j}}$$

$$\hat{C}_{i,j} = C_{i,I-i} * \hat{f}_{I-i} * \cdots * \hat{f}_{j-1}$$

Projected best estimate losses

			$\hat{C}_{1,4} = C_{1,3} * \hat{f}_0 * \hat{f}_1 * \hat{f}_2 * \hat{f}_3$
		$\hat{C}_{2,3} = C_{2,2} * \hat{f}_0 * \hat{f}_1 * \hat{f}_2$	$\hat{C}_{2,4} = C_{2,2} * \hat{f}_0 * \hat{f}_1 * \hat{f}_2 * \hat{f}_3$
	$\hat{C}_{3,2} = C_{3,1} * \hat{f}_0 * \hat{f}_1$	$\hat{C}_{3,3} = C_{3,1} * \hat{f}_0 * \hat{f}_1 * \hat{f}_2$	$\hat{C}_{3,4} = C_{3,1} * \hat{f}_0 * \hat{f}_1 * \hat{f}_2 * \hat{f}_3$
$\hat{C}_{4,1} = C_{4,0} * \hat{f}_0$	$\hat{C}_{4,2} = C_{4,0} * \hat{f}_0 * \hat{f}_1$	$\hat{C}_{4,3} = C_{4,0} * \hat{f}_0 * \hat{f}_1 * \hat{f}_2$	$\hat{C}_{4,4} = C_{4,0} * \hat{f}_0 * \hat{f}_1 * \hat{f}_2 * \hat{f}_3$

Bornhuetter-Ferguson method model assumptions

- ❑ Cumulative claims $C_{i,j}$ of different accident years i are independent,¹⁴ i.e.
 $\{C_{i,0}, \dots, C_{i,J}\}$ and $\{C_{k,0}, \dots, C_{k,J}\}$ are independent for $k \neq i$.
- ❑ The exist parameters $\mu_0, \dots, \mu_I > 0$ and a pattern $\beta_0, \dots, \beta_J > 0$ with $\beta_J = 1$ such that for all accident years $0 \leq i \leq I$ and all development years $0 \leq j \leq J$ we have

$$EC_{i,j} = \beta_j * \mu_i.$$

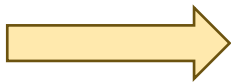
Then the expected value of aggregate loss for given accident year $1 \leq i \leq I$, conditional on the history known at the end of year:

$$E[C_{i,J}|D_I] = C_{i,I-i} + (1 - \beta_{I-i}) * \mu_i.$$

Bornhuetter-Ferguson method

Paid/incurred cumulative claims triangle

$C_{0,0}$	$C_{0,1}$	$C_{0,2}$	$C_{0,3}$	$C_{0,4}$
$C_{1,0}$	$C_{1,1}$	$C_{1,2}$	$C_{1,3}$	
$C_{2,0}$	$C_{2,1}$	$C_{2,2}$		
$C_{3,0}$	$C_{3,1}$			
$C_{4,0}$				



- Estimates of development factors:

$$\hat{f}_j = \frac{\sum_{i=0}^{I-j-1} C_{i,j+1}}{\sum_{i=0}^{I-j-1} C_{i,j}}$$

- Estimates of β_0, \dots, β_J :

$$\hat{\beta}_j = \frac{1}{\prod_{k=j}^{J-1} \hat{f}_k}$$

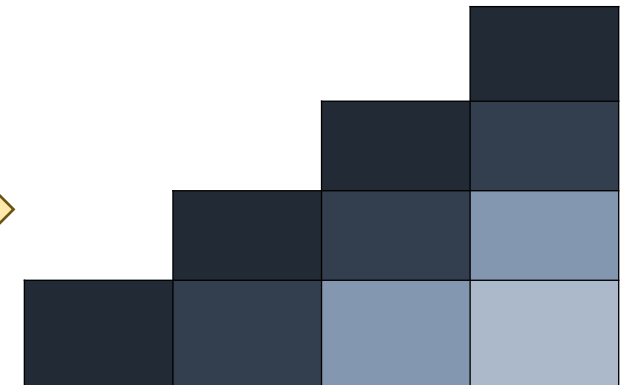
- Estimates of μ_0, \dots, μ_I :

$$\hat{\mu}_i = \text{ELR} * \text{Earned Premium}_i,$$

where ELR is a priori estimate of the loss ratio
(ratio of ultimate losses to earned premiums)

$$\hat{C}_{i,j} = C_{i,j} + (1 - \hat{\beta}_j) * \hat{\mu}_i$$

Projected best estimate losses



Explicit Allowances for Claims Inflation

There are numerous approaches for explicitly accounting for claims inflation. In this discussion, we will focus on the most used method, the **Inflation Adjusted Chain Ladder**, as well as one additional approach that we find particularly interesting – **Probabilistic Trend Family model**.

Inflation-adjusted Chain Ladder method

Incremental claims triangle in nominal terms

$X_{0,0}$	$X_{0,1}$	$X_{0,2}$	$X_{0,3}$	$X_{0,4}$
$X_{1,0}$	$X_{1,1}$	$X_{1,2}$	$X_{1,3}$	
$X_{2,0}$	$X_{2,1}$	$X_{2,2}$		
$X_{3,0}$	$X_{3,1}$			
$X_{4,0}$				

Cumulative claims triangle in
adjusted

$C'_{0,0}$	$C'_{0,1}$	$C'_{0,2}$	$C'_{0,3}$	$C'_{0,4}$
$C'_{1,0}$	$C'_{1,1}$	$C'_{1,2}$	$C'_{1,3}$	
$C'_{2,0}$	$C'_{2,1}$	$C'_{2,2}$		
$C'_{3,0}$	$C'_{3,1}$			
$C'_{4,0}$				

Incremental claims triangle in
adjusted

$X'_{0,0}$	$X'_{0,1}$	$X'_{0,2}$	$X'_{0,3}$	$X'_{0,4}$
$X'_{1,0}$	$X'_{1,1}$	$X'_{1,2}$	$X'_{1,3}$	
$X'_{2,0}$	$X'_{2,1}$	$X'_{2,2}$		
$X'_{3,0}$	$X'_{3,1}$			
$X'_{4,0}$				

Transform to
cumulative claims
triangle

Apply Chain-Ladder method to cumulative claims triangle

Projected cumulative claims

Projected incremental claims

Transform to
incremental claims

Projected incremental claims adjusted for future claims inflation

Adjust claims
for **future**
claims inflation

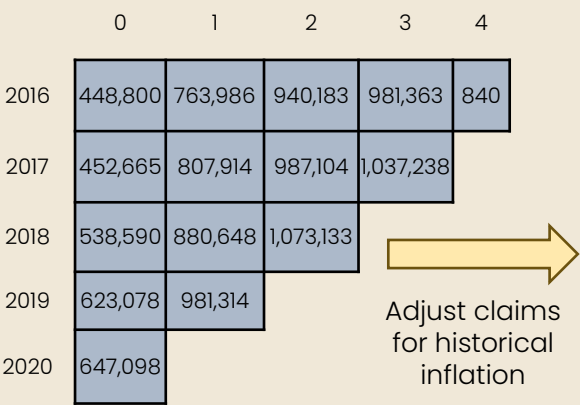
Inflation-adjusted Chain Ladder method

Incremental claims triangle
in nominal terms

	0	1	2	3	4
2016	448,800	763,986	940,183	981,363	840
2017	452,665	807,914	987,104	1,037,238	
2018	538,590	880,648	1,073,133		
2019	623,078	981,314			
2020	647,098				

Inflation-adjusted Chain Ladder method

Incremental claims triangle
in nominal terms




Historical inflation

Year	Inflation rate
2016	-1.5%
2017	1.3%
2018	4.6%
2019	3.8%

Inflation-adjusted Chain Ladder method

Incremental claims triangle
in nominal terms

	0	1	2	3	4
2016	448,800	763,986	940,183	981,363	840
2017	452,665	807,914	987,104	1,037,238	
2018	538,590	880,648	1,073,133		
2019	623,078	981,314			
2020	647,098				



Adjust claims
for historical
inflation

Historical inflation

Year	Inflation rate
2016	-1.5%
2017	1.3%
2018	4.6%
2019	3.8%

Inflation matrix

	0	1	2	3	4
2016	$(1-1.5\%)*(1+1.3%)*(1+4.6%)*(1+3.8\%)$	$(1+1.3%)*(1+4.6%)*(1+3.8\%)$	$(1+4.6%)*(1+3.8\%)$	$1+3.8\%$	1
2017	$(1+1.3%)*(1+4.6%)*(1+3.8\%)$	$(1+4.6%)*(1+3.8\%)$	$1+3.8\%$	1	
2018	$(1+4.6%)*(1+3.8\%)$	$1+3.8\%$	1		
2019	$1+3.8\%$	1			
2020	1				

Inflation-adjusted Chain Ladder method

Incremental claims triangle
in nominal terms

	0	1	2	3	4
2016	448,800	763,986	940,183	981,363	840
2017	452,665	807,914	987,104	1,037,238	
2018	538,590	880,648	1,073,133		
2019	623,078	981,314			
2020	647,098				

Adjust claims for historical inflation

Historical inflation

Year	Inflation rate
2016	-1.5%
2017	1.3%
2018	4.6%
2019	3.8%

Inflation matrix

	0	1	2	3	4
2016	$(1-1.5\%)*(1+1.3%)*(1+4.6%)*(1+3.8\%)$	$(1+1.3%)*(1+4.6%)*(1+3.8\%)$	$(1+4.6%)*(1+3.8\%)$	$1+3.8\%$	1
2017	$(1+1.3%)*(1+4.6%)*(1+3.8\%)$	$(1+4.6%)*(1+3.8\%)$	$1+3.8\%$	1	
2018	$(1+4.6%)*(1+3.8\%)$	$1+3.8\%$	1		
2019	$1+3.8\%$	1			
2020	1				

	0	1	2	3	4
2016	1.08336	1.09986	1.08575	1.03800	1
2017	1.09986	1.08575	1.03800	1	
2018	1.08575	1.03800	1		
2019	1.03800	1			
2020	1				

Inflation-adjusted Chain Ladder method

Incremental claims triangle
in nominal terms

	0	1	2	3	4
2016	448,800	763,986	940,183	981,363	840
2017	452,665	807,914	987,104	1,037,238	
2018	538,590	880,648	1,073,133		
2019	623,078	981,314			
2020	647,098				

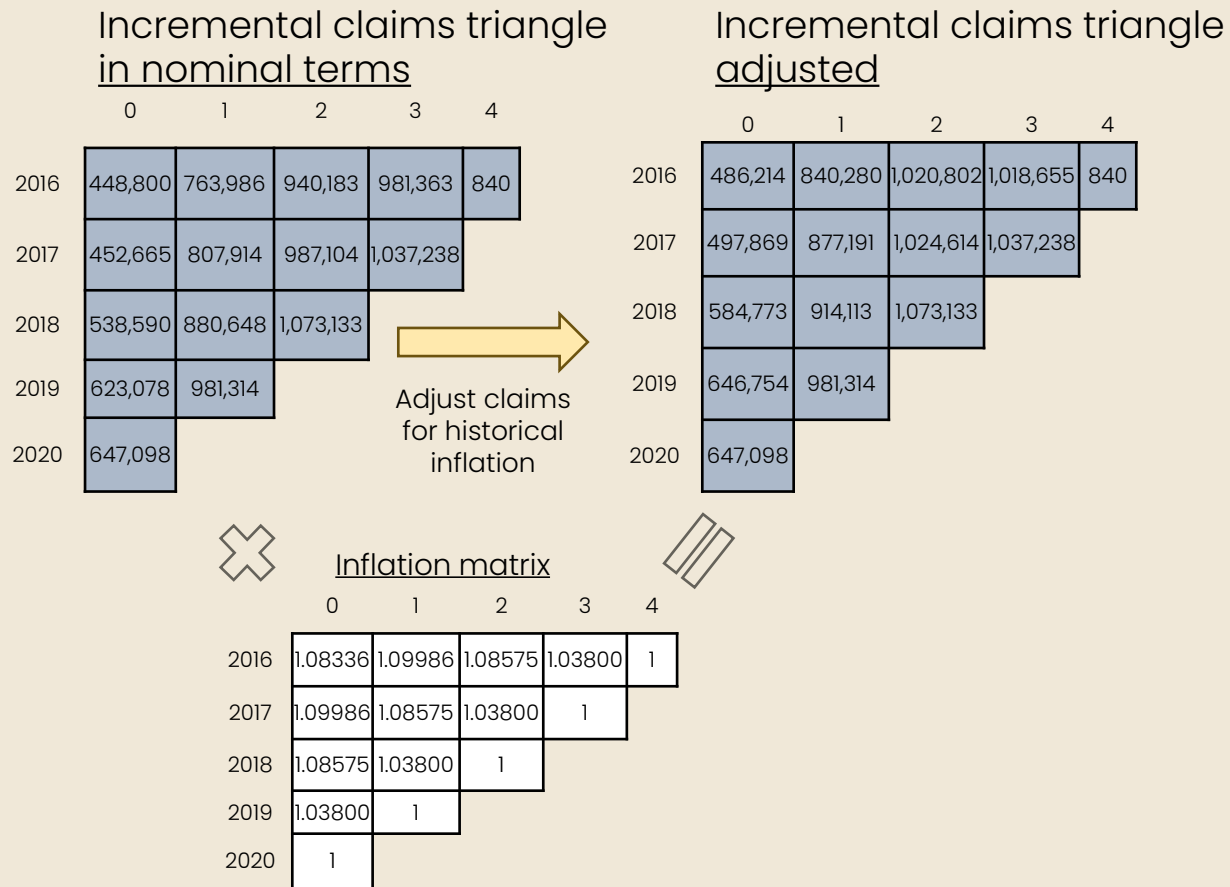
Adjust claims
for historical
inflation



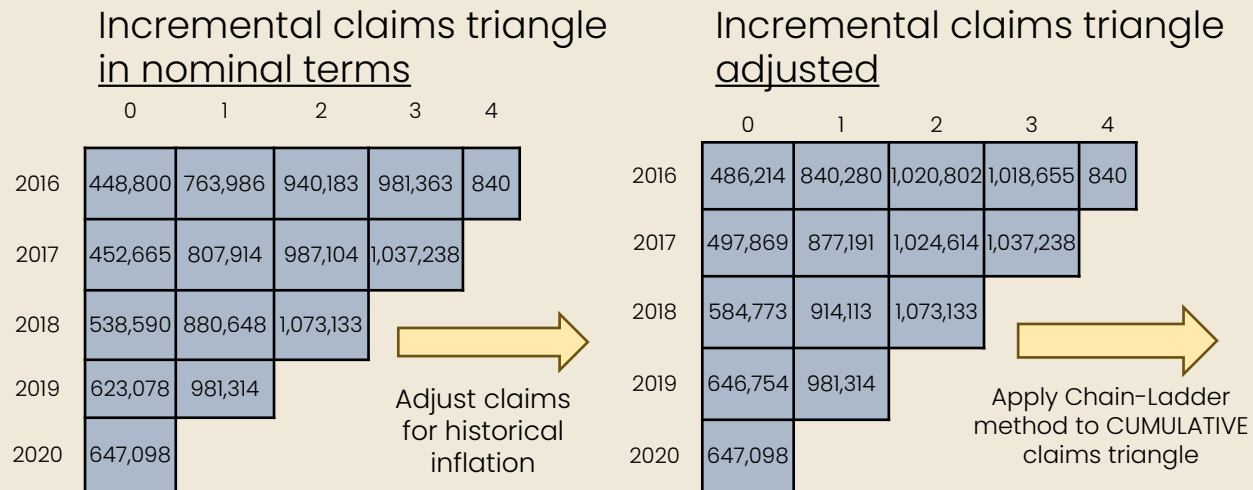
Inflation matrix

	0	1	2	3	4
2016	1.08336	1.09986	1.08575	1.03800	1
2017	1.09986	1.08575	1.03800	1	
2018	1.08575	1.03800	1		
2019	1.03800	1			
2020	1				

Inflation-adjusted Chain Ladder method



Inflation-adjusted Chain Ladder method



Inflation-adjusted Chain Ladder method

Incremental claims triangle
in nominal terms

	0	1	2	3	4
2016	448,800	763,986	940,183	981,363	840
2017	452,665	807,914	987,104	1,037,238	
2018	538,590	880,648	1,073,133		
2019	623,078	981,314			
2020	647,098				

Adjust claims
for historical
inflation

Incremental claims triangle
adjusted

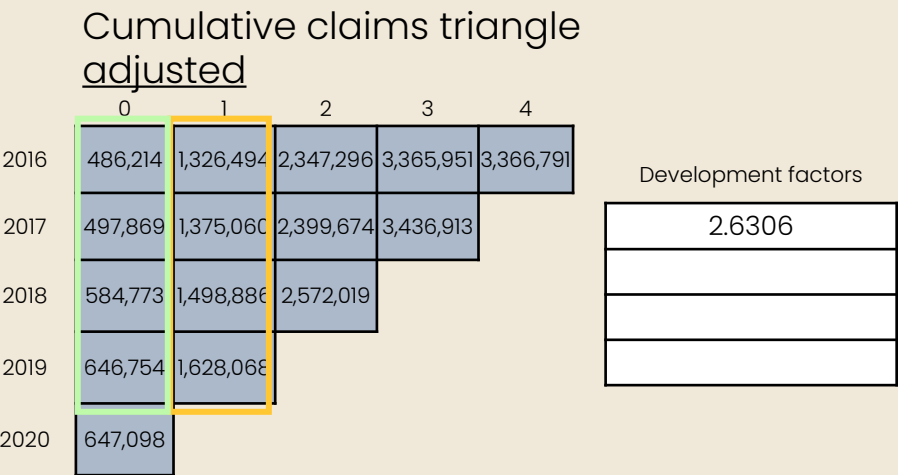
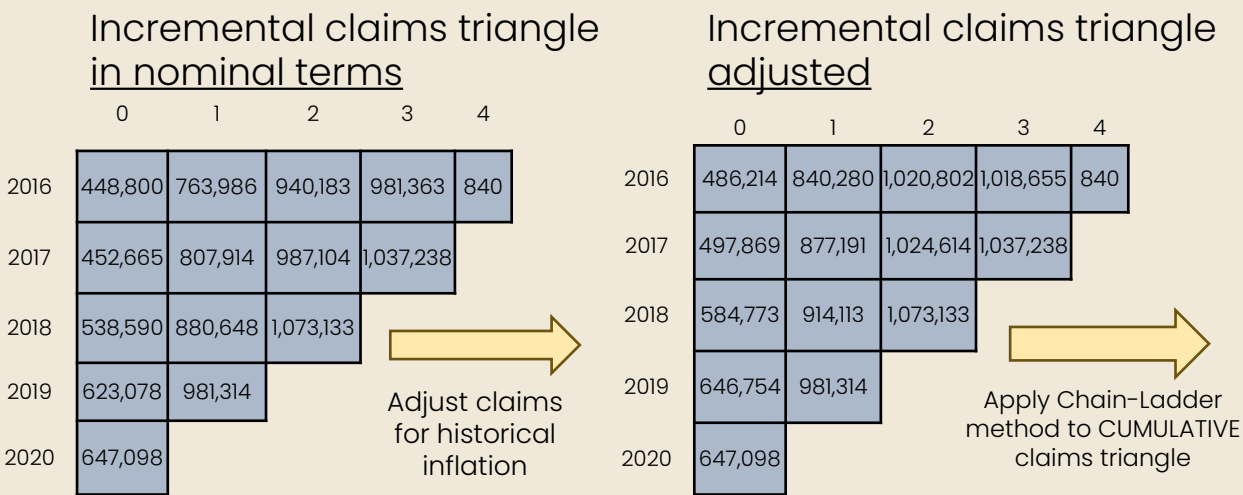
	0	1	2	3	4
2016	486,214	840,280	1,020,802	1,018,655	840
2017	497,869	877,191	1,024,614	1,037,238	
2018	584,773	914,113	1,073,133		
2019	646,754	981,314			
2020	647,098				

Apply Chain-Ladder
method to CUMULATIVE
claims triangle

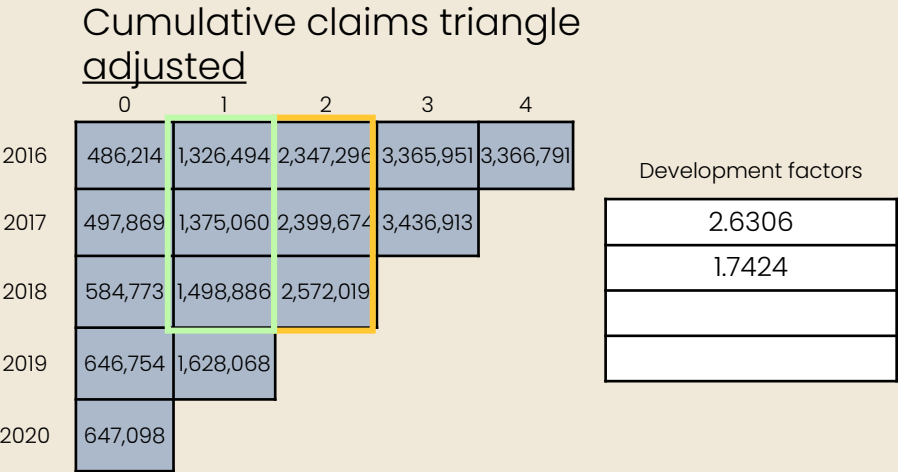
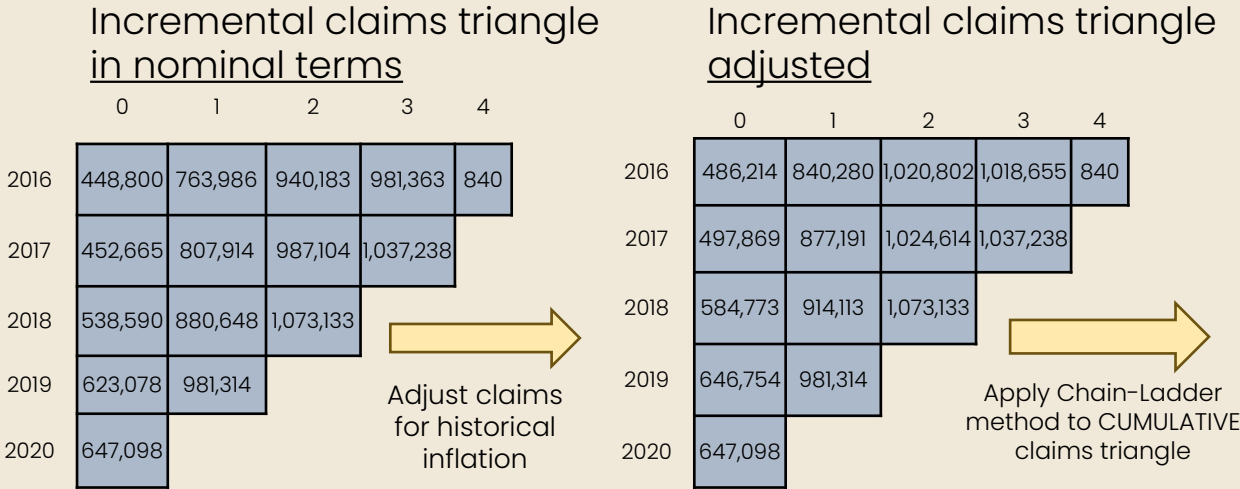
Cumulative claims triangle
adjusted

	0	1	2	3	4
2016	486,214	1,326,494	2,347,296	3,365,951	3,366,791
2017	497,869	1,375,060	2,399,674	3,436,913	
2018	584,773	1,498,886	2,572,019		
2019	646,754	1,628,068			
2020	647,098				

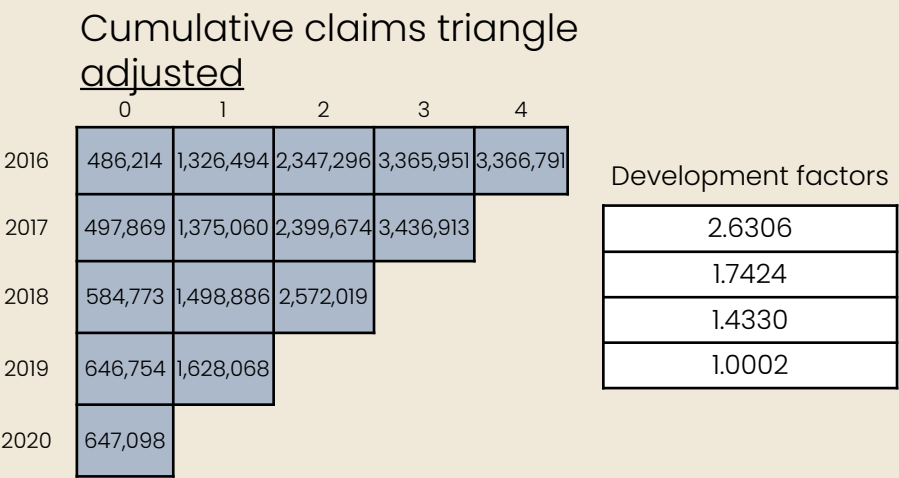
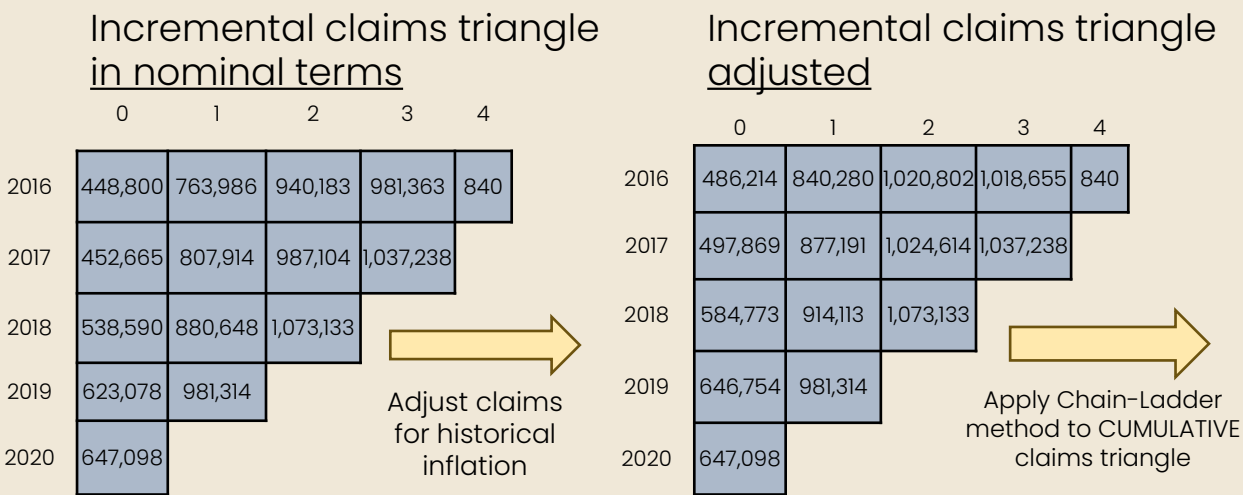
Inflation-adjusted Chain Ladder method



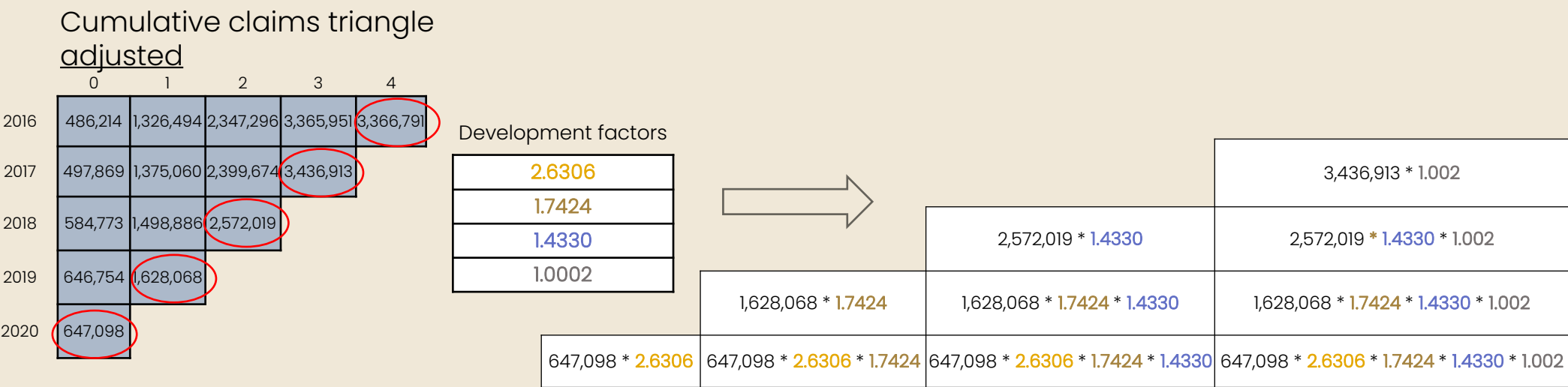
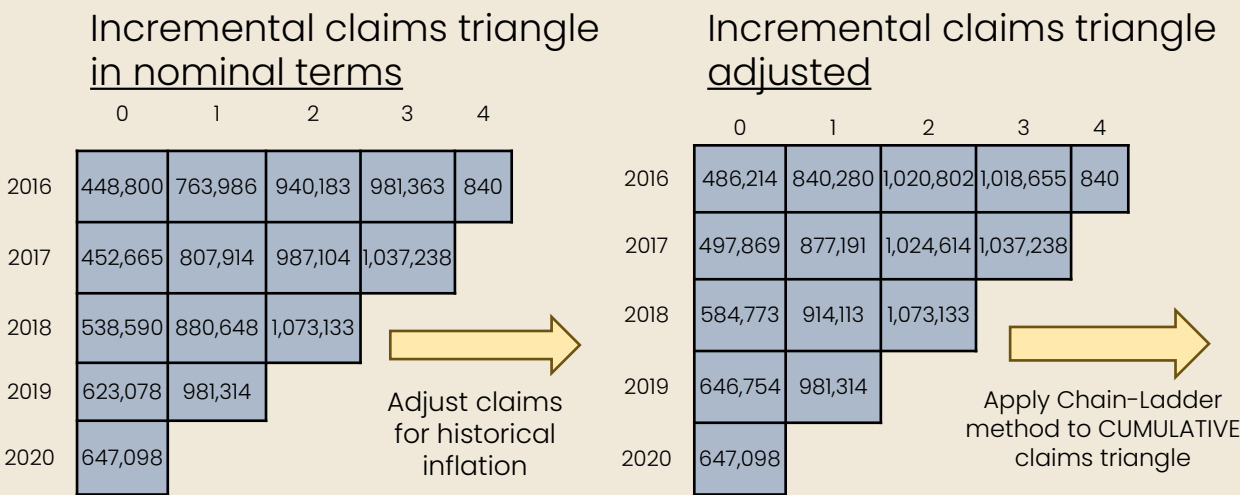
Inflation-adjusted Chain Ladder method



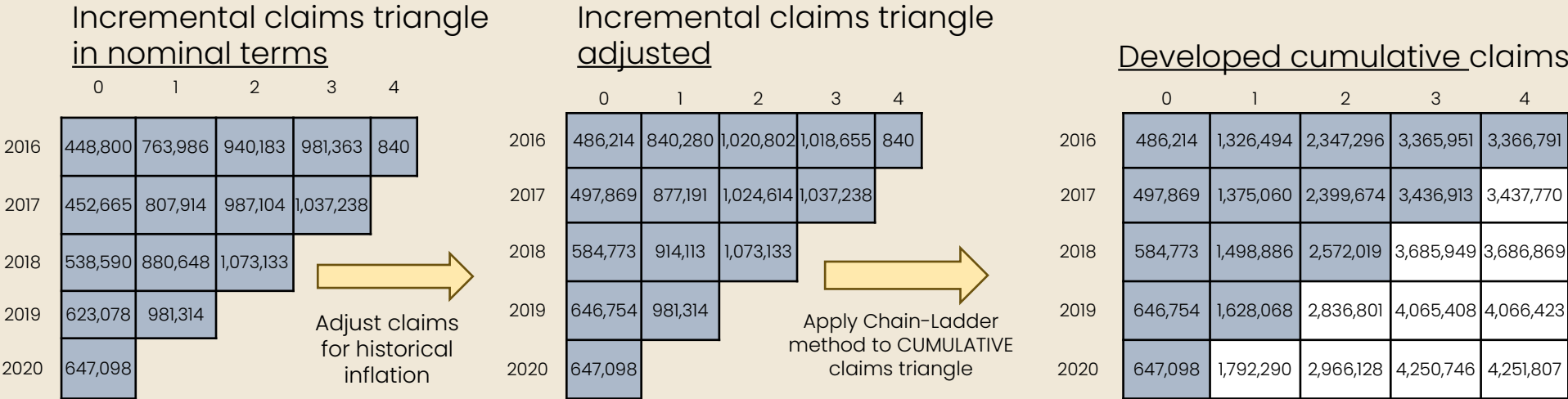
Inflation-adjusted Chain Ladder method



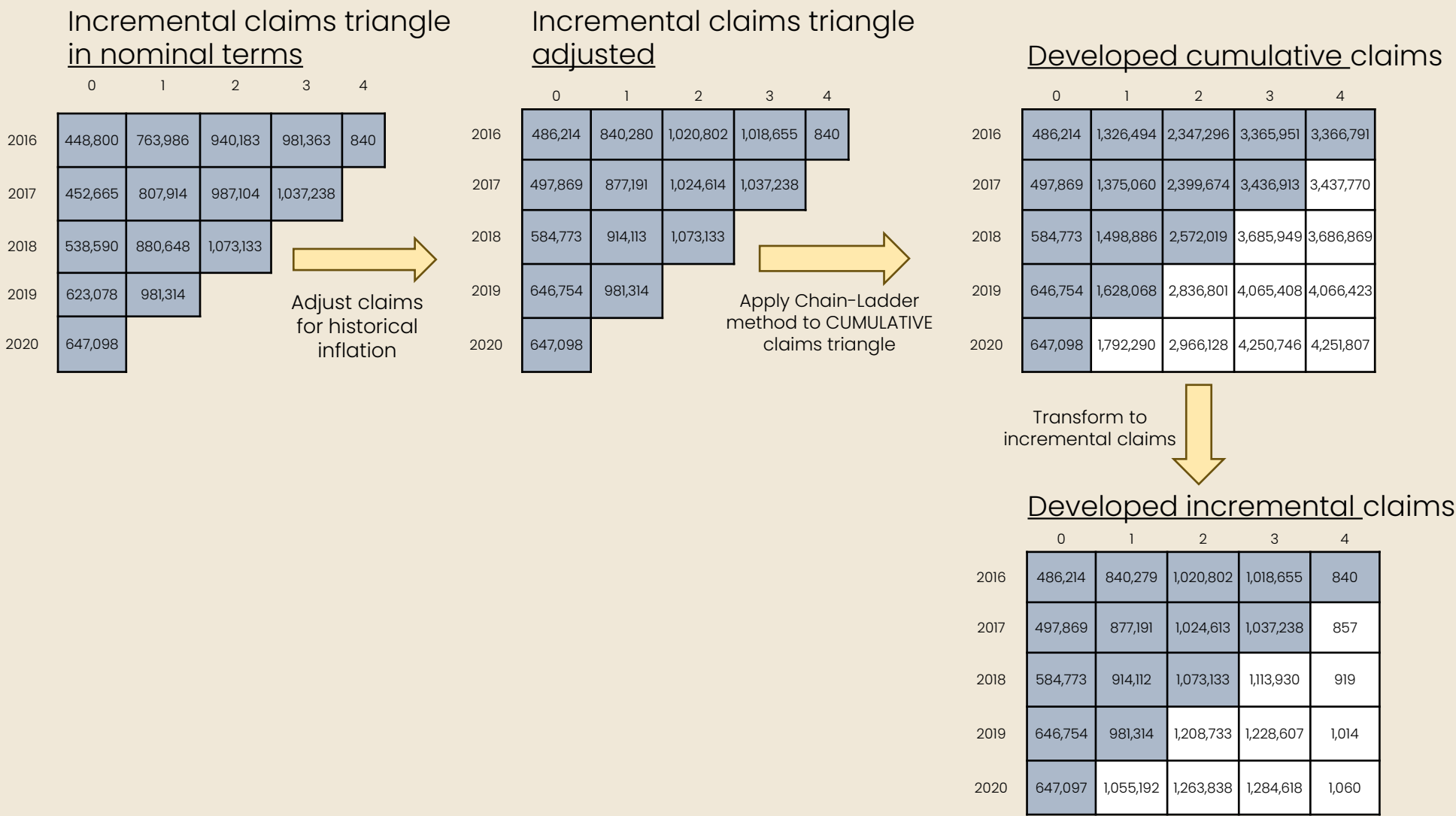
Inflation-adjusted Chain Ladder method



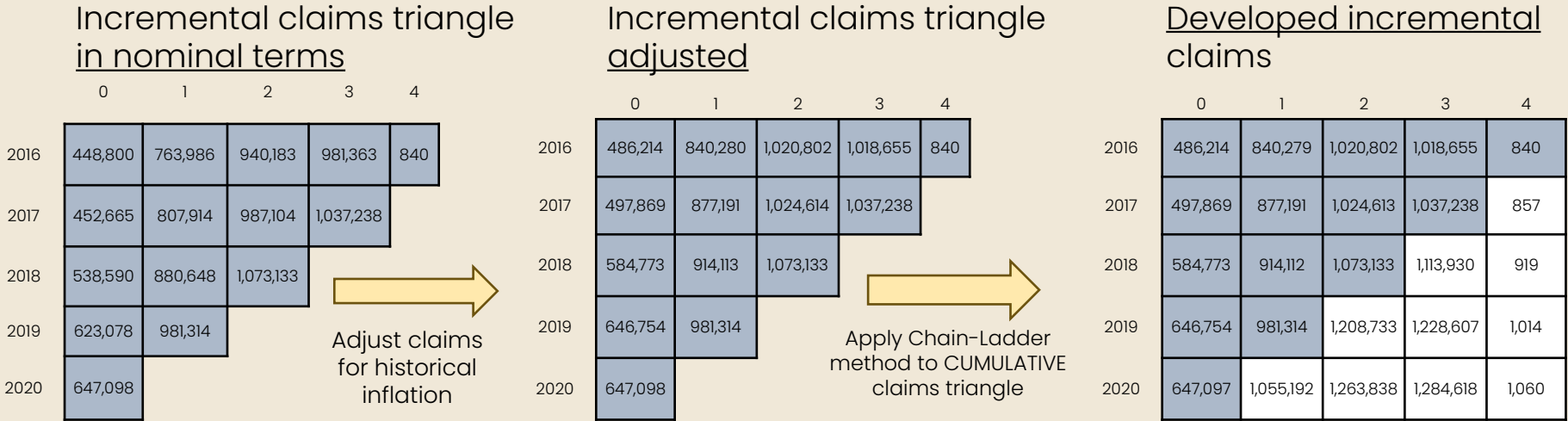
Inflation-adjusted Chain Ladder method



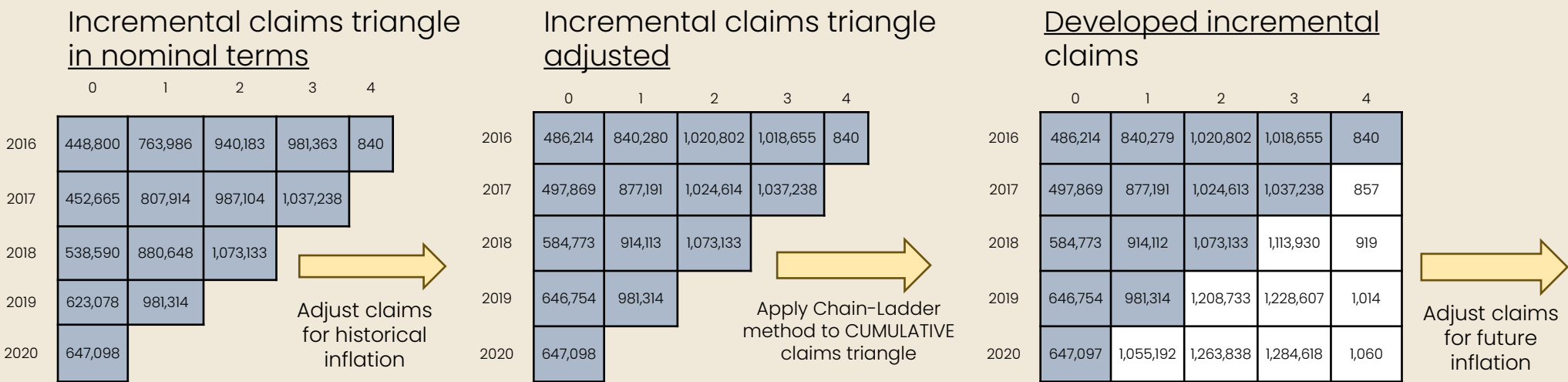
Inflation-adjusted Chain Ladder method



Inflation-adjusted Chain Ladder method



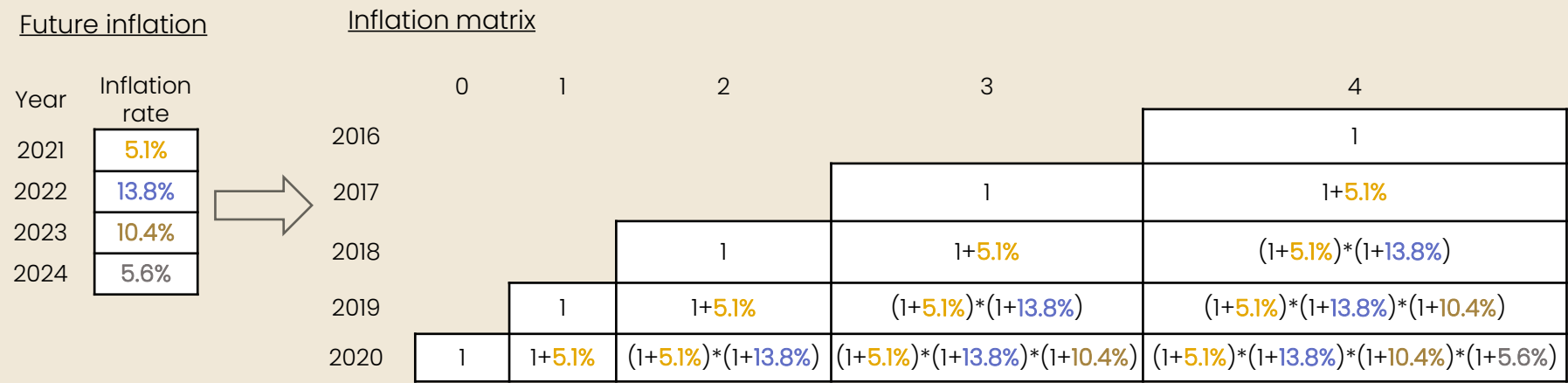
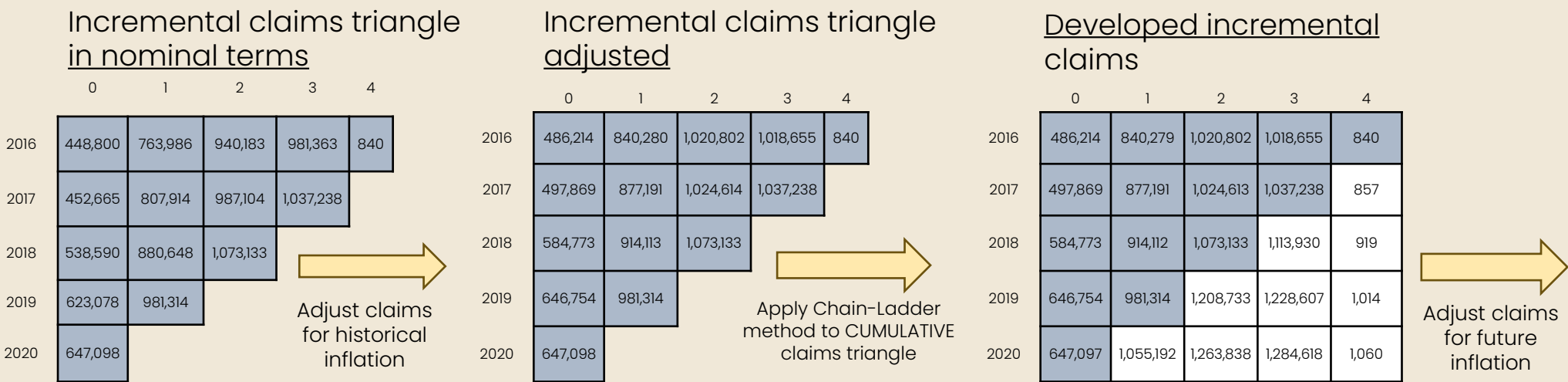
Inflation-adjusted Chain Ladder method



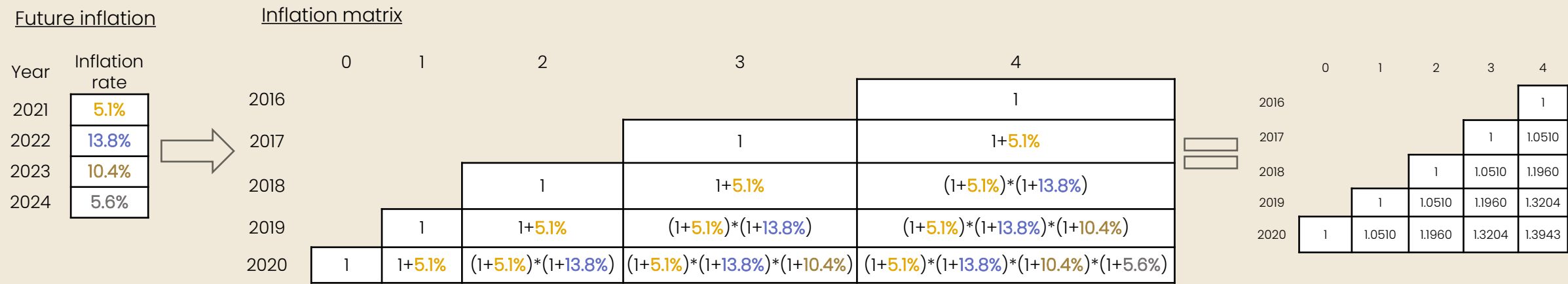
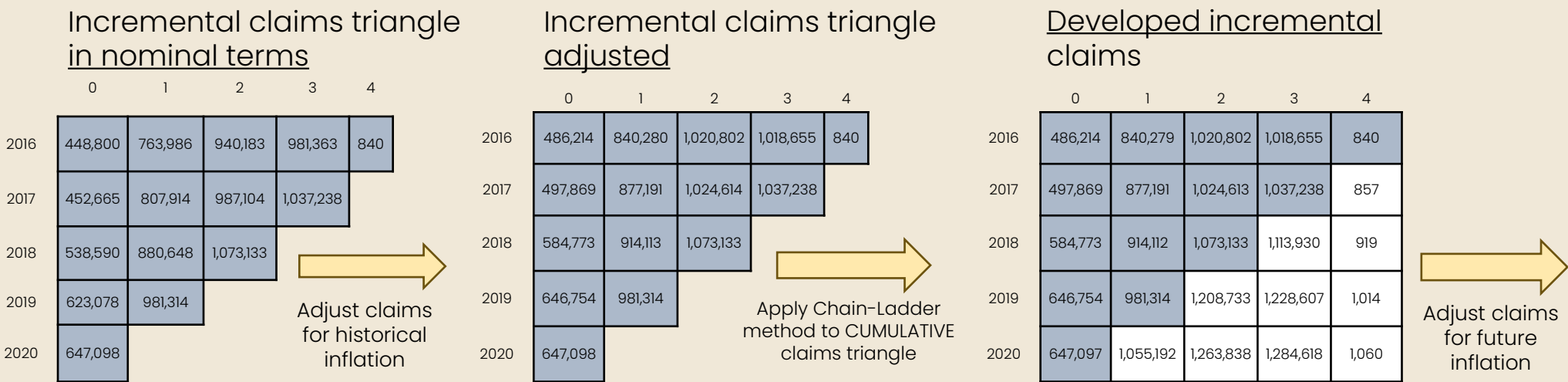
Future inflation

Year	Inflation rate
2021	5.1%
2022	13.8%
2023	10.4%
2024	5.6%

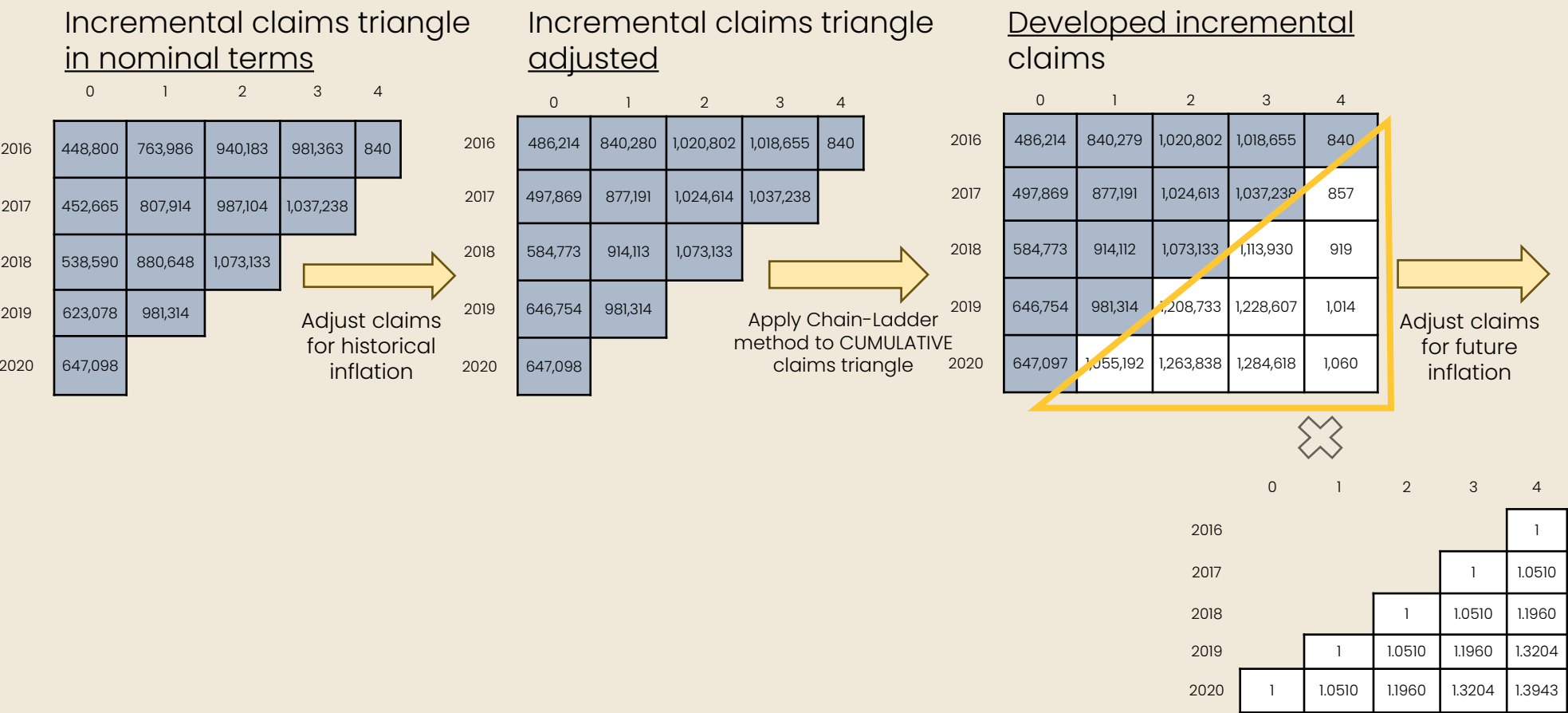
Inflation-adjusted Chain Ladder method



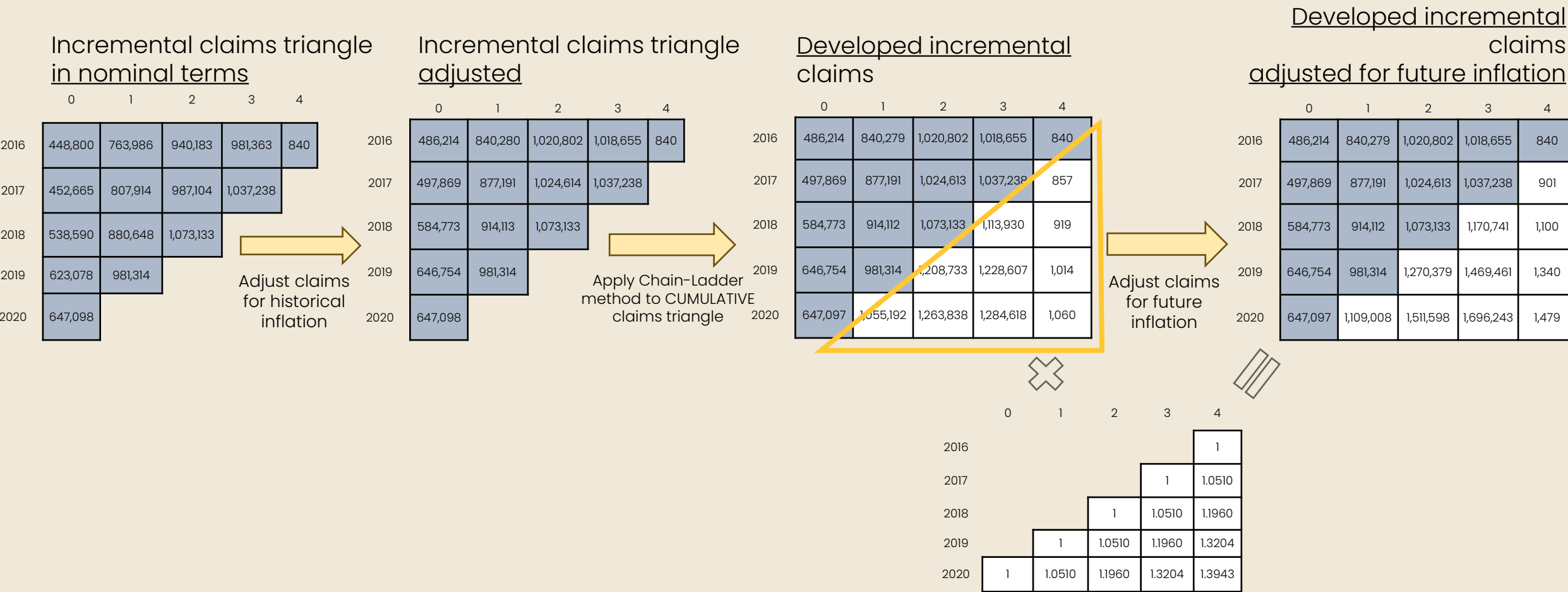
Inflation-adjusted Chain Ladder method



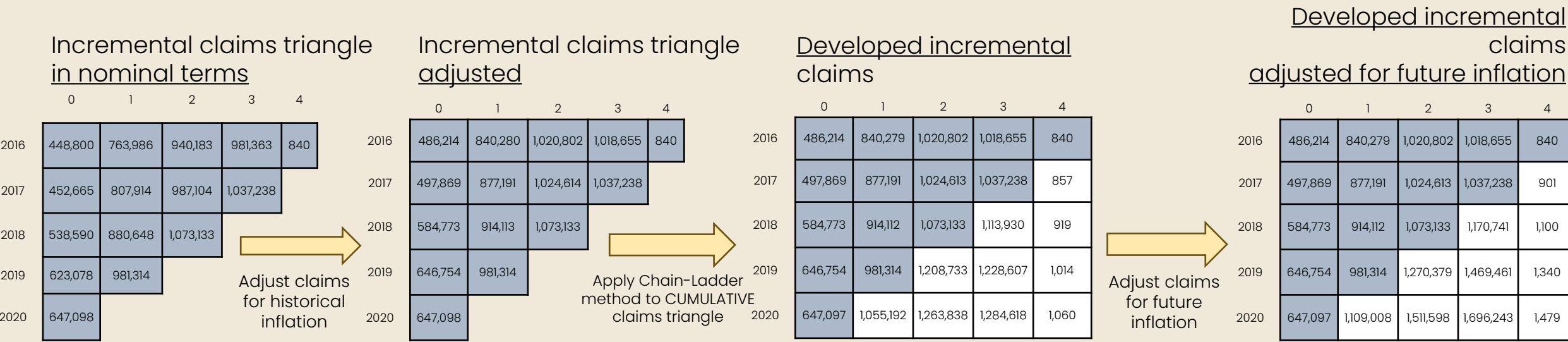
Inflation-adjusted Chain Ladder method



Inflation-adjusted Chain Ladder method



Inflation-adjusted Chain Ladder method



Historical inflation

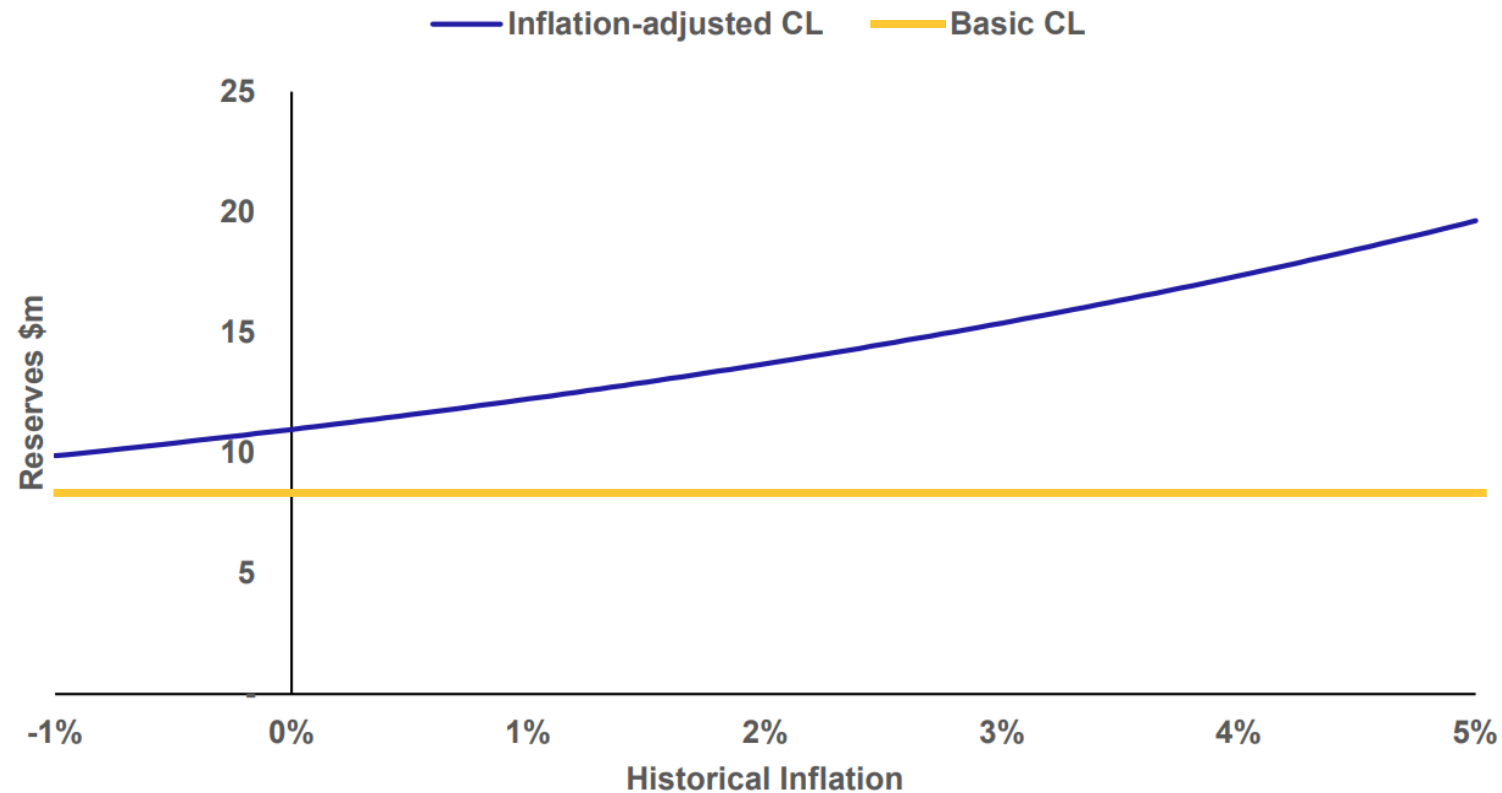
Year	Inflation rate
2016	-1.5%
2017	1.3%
2018	4.6%
2019	3.8%

Future inflation

Year	Inflation rate
2021	5.1%
2022	13.8%
2023	10.4%
2024	5.6%

Inflation-adjusted and standard Chain Ladder method

Reserve projection with basic CL and inflation-adjusted CL for non-marine GL as at 2019 Q3



Source: Martis, Stavros and Stewart, Emma (2020); Claims inflation trends within the Lloyd's and the London Market; GIRO conference

Probabilistic trend family model (PTF)

PTF is a model used in claims reserving that captures trends in accident year, development year, and calendar year directions simultaneously.¹⁵

It is particularly useful for forecasting and risk assessment in environments with significant changes in external factors.

Mathematical model

$$y_{i,j} = \alpha_i + \sum_{k=1}^j \beta_k + \sum_{t=1}^{i+j} \gamma_t + \varepsilon_{i,j}$$

Notation	Meaning
i	Accident year
j	Development year
$t = i + j$	Calendar year
$y_{i,j}$	Natural logarithm of the incremental paid data in accident year i and at development year j
α_i coefficient	Trend for accident year i
β_j coefficient	Trend for development year j
γ_t coefficient	Trend for calendar year t
$\varepsilon_{i,j}$	Zero-mean normally distributed random error with variance σ^2 (can be constant or varying)

Probabilistic trend family model (PTF)

On a **logarithmic scale** the distribution for each **incremental claim** amounts $x_{i,j}$ is **normal**:

$$\log(x_{i,j}) =: y_{i,j} \sim \mathcal{N}(\mu_{i,j}, \sigma^2),$$

where the **means** of the normal distributions are related by the **"trends"** described by the member:

$$\mu_{i,j} = \alpha_i + \sum_{k=1}^j \beta_k + \sum_{t=1}^{i+j} \gamma_t.$$

The **incremental claim** amounts $x_{i,j}$ therefore follow a **lognormal distribution**:

$$x_{i,j} \sim \text{LN}(\tilde{\mu}_{i,j}, \tilde{\sigma}^2),$$

where

$$\tilde{\mu}_{i,j} = e^{\mu_{i,j} + \frac{\sigma^2}{2}},$$

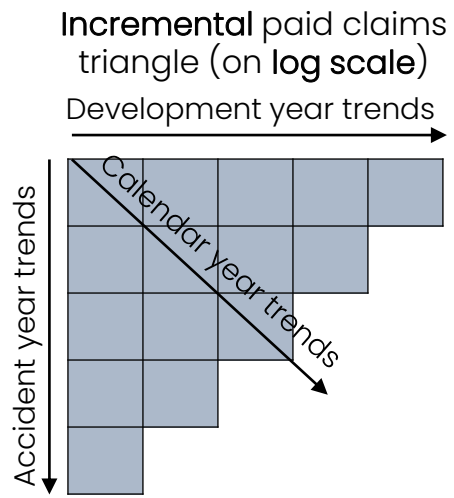
$$\tilde{\sigma}^2 = \tilde{\mu}_{i,j} * (e^{\sigma^2} - 1).$$

Mathematical model

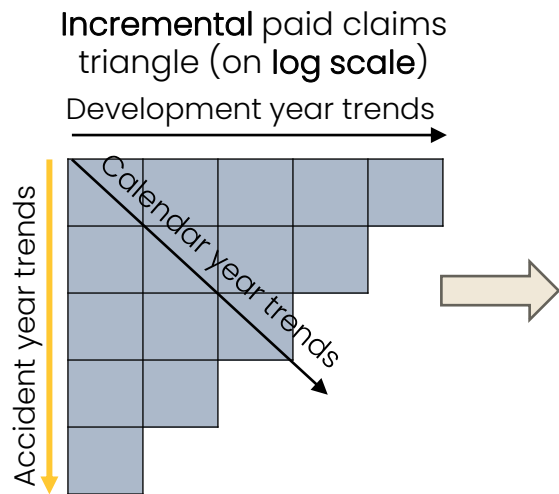
$$y_{i,j} = \alpha_i + \sum_{k=1}^j \beta_k + \sum_{t=1}^{i+j} \gamma_t + \varepsilon_{i,j}$$

Notation	Meaning
i	Accident year
j	Development year
$t = i + j$	Calendar year
$y_{i,j}$	Natural logarithm of the incremental paid data in accident year i and at development year j
α_i coefficient	Trend for accident year i
β_j coefficient	Trend for development year j
γ_t coefficient	Trend for calendar year t
$\varepsilon_{i,j}$	Zero-mean normally distributed random error with variance σ^2 (can be constant or varying)

Accident, development and calendar year trends in PTF model



Accident, development and calendar year trends in PTF model



Accident year (AY) trends:

α_0	α_0	α_0	α_0	α_0
α_1	α_1	α_1	α_1	
α_2	α_2	α_2		
α_3	α_3			
α_4				

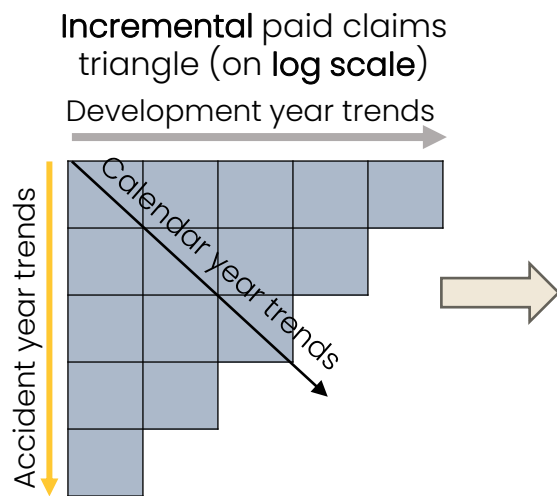
Definition: trends capturing the impact of events or conditions specific to the year in which claims occur.

Examples:¹⁶

- Changes in underwriting practices.
- Economic conditions.
- **Regulatory changes.**

Importance: helps understand how the environment at the time of the accident influences claims development.

Accident, development and calendar year trends in PTF model



Accident year (AY) trends:

α_0	α_0	α_0	α_0	α_0
α_1	α_1	α_1	α_1	
α_2	α_2	α_2		
α_3	α_3			
α_4				

Definition: trends capturing the impact of events or conditions specific to the year in which claims occur.

Examples:¹⁶

- Changes in underwriting practices.
- Economic conditions.
- **Regulatory changes.**

Importance: helps understand how the environment at the time of the accident influences claims development.

Development year (DY) trends:

	β_1	β_2	β_3	β_4
	β_1	β_2	β_3	
	β_1	β_2		
	β_1			

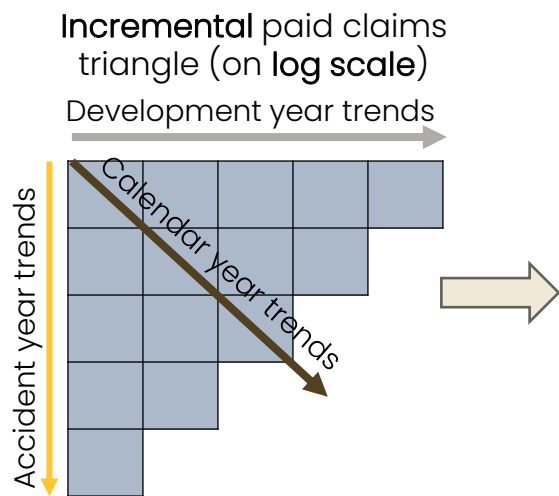
Definition: trends accounting for how claims evolve over time.

Examples:¹⁶

- Time taken for claims to be reported, settled, or paid.
- **Changes in legal processes.**

Importance: models the progression of claims from inception to final settlement.

Accident, development and calendar year trends in PTF model



Accident year (AY) trends:

α_0	α_0	α_0	α_0	α_0
α_1	α_1	α_1	α_1	
α_2	α_2	α_2		
α_3	α_3			
α_4				

Definition: trends capturing the impact of events or conditions specific to the year in which claims occur.

Examples:¹⁶

- Changes in underwriting practices.
- Economic conditions.
- **Regulatory changes.**

Importance: helps understand how the environment at the time of the accident influences claims development.

Development year (DY) trends:

	β_1	β_2	β_3	β_4
	β_1	β_2	β_3	
	β_1	β_2		
	β_1			

Definition: trends accounting for how claims evolve over time.

Examples:¹⁶

- Time taken for claims to be reported, settled, or paid.
- **Changes in legal processes.**

Importance: models the progression of claims from inception to final settlement.

Calendar year (CY) trends:

	γ_1	γ_2	γ_3	γ_4
γ_1	γ_2	γ_3	γ_4	
γ_2	γ_3	γ_4		
γ_3	γ_4			
γ_4				

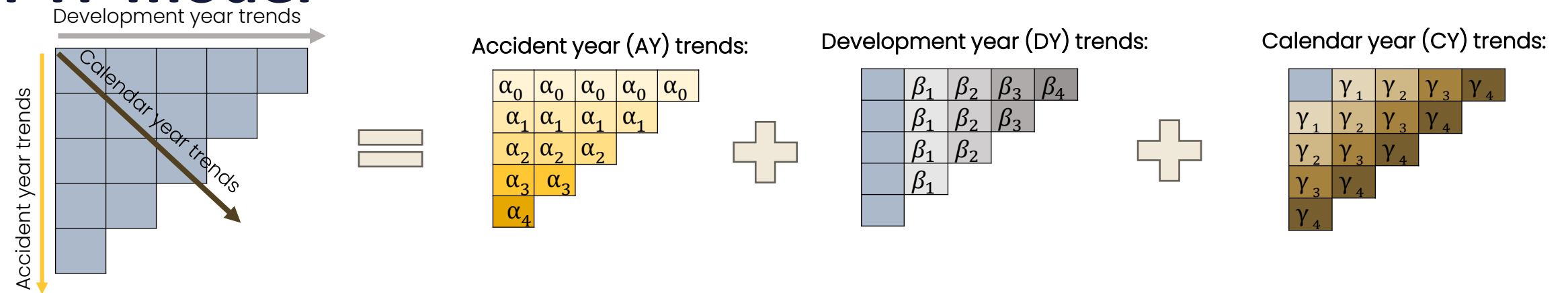
Definition: trends considering external factors affecting claims payments across different calendar years.

Examples:¹⁶

- Economic inflation.
- **Changes in legal environments.**
- **Shifts in social attitudes towards claims.**

Importance: impacts all claims regardless of accident or development year.

Accident, development and calendar year trends in PTF model



α_0	$\alpha_0 + \beta_1 + \gamma_1$	$\alpha_0 + \beta_1 + \beta_2 + \gamma_1 + \gamma_2$	$\alpha_0 + \beta_1 + \beta_2 + \beta_3 + \gamma_1 + \gamma_2 + \gamma_3$	$\alpha_0 + \beta_1 + \beta_2 + \beta_3 + \beta_4 + \gamma_1 + \gamma_2 + \gamma_3 + \gamma_4$
$\alpha_1 + \gamma_1$	$\alpha_1 + \beta_1 + \gamma_1 + \gamma_2$	$\alpha_1 + \beta_1 + \beta_2 + \gamma_1 + \gamma_2 + \gamma_3$	$\alpha_1 + \beta_1 + \beta_2 + \beta_3 + \gamma_1 + \gamma_2 + \gamma_3 + \gamma_4$	
$\alpha_2 + \gamma_1 + \gamma_2$	$\alpha_2 + \beta_1 + \gamma_1 + \gamma_2 + \gamma_3$	$\alpha_2 + \beta_1 + \beta_2 + \gamma_1 + \gamma_2 + \gamma_3 + \gamma_4$		
$\alpha_3 + \gamma_1 + \gamma_2 + \gamma_3$	$\alpha_3 + \beta_1 + \gamma_1 + \gamma_2 + \gamma_3 + \gamma_4$			
$\alpha_4 + \gamma_1 + \gamma_2 + \gamma_3 + \gamma_4$				

Simple example of PTF model

To demonstrate how to estimate parameters in a model from the PTF, we created a **simple example** with clearly **identifiable calendar year trends**

Log-transformed incremental
claims triangle

	0	1	2	3
2018	10	10.2	10.4	11
2019	10.2	10.4	11	
2020	10.4	11		
2021	11			

Simple example of PTF model

Log-transformed incremental claims triangle

	0	1	2	3
2018	10	10.2	10.4	11
2019	10.2	10.4	11	
2020	10.4	11		
2021	11			

Calendar year trends

	0	1	2	3
2018		+0.2		
2019	+0.2			
2020				
2021				

Simple example of PTF model

Log-transformed incremental claims triangle

	0	1	2	3
2018	10	10.2	10.4	11
2019	10.2	10.4	11	
2020	10.4	11		
2021	11			

Calendar year trends

	0	1	2	3
2018		+0.2		
2019	+0.2			
2020				
2021				

	0	1	2	3
2018			+0.2	
2019		+0.2		
2020	+0.2			
2021				

Simple example of PTF model

Log-transformed incremental claims triangle

	0	1	2	3
2018	10	10.2	10.4	11
2019	10.2	10.4	11	
2020	10.4	11		
2021	11			

Calendar year trends

	0	1	2	3
2018		+0.2		
2019	+0.2			
2020				
2021				

	0	1	2	3
2018			+0.2	
2019		+0.2		
2020	+0.2			
2021				

	0	1	2	3
2018				+0.6
2019			+0.6	
2020		+0.6		
2021	+0.6			

Simple example of PTF model

Log-transformed incremental claims triangle

	0	1	2	3
2018	10	10.2	10.4	11
2019	10.2	10.4	11	
2020	10.4	11		
2021	11			

Calendar year trends

	0	1	2	3
2018		+0.2		
2019	+0.2			
2020		+0.2		
2021	+0.2			

	0	1	2	3
2018			+0.2	
2019		+0.2		
2020	+0.2			
2021		+0.2		

	0	1	2	3
2018				+0.6
2019			+0.6	
2020		+0.6		
2021	+0.6			

Log-transformed incremental claims triangle in form of PTF model

	0	1	2	3
2018	$10 = \alpha$	$10.2 = \alpha + \gamma_1$	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$
2019	$10.2 = \alpha + \gamma_1$	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$	
2020	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$		
2021	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$			

Simple example of PTF model

Log-transformed incremental claims triangle

	0	1	2	3
2018	10	10.2	10.4	11
2019	10.2	10.4	11	
2020	10.4	11		
2021	11			

Log-transformed incremental claims triangle in form of PTF model

	0	1	2	3
2018	$10 = \alpha$	$10.2 = \alpha + \gamma_1$	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$
2019	$10.2 = \alpha + \gamma_1$	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$	
2020	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$		
2021	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$			

Linear regression problem

Development year	Response variable	Regression matrix			Regression coefficients
0	10	1	0	0	α
0	10.2	1	1	0	γ_1
0	10.4	1	2	0	γ_2
0	11	1	2	1	
1	10.2	1	1	0	
1	10.4	1	2	0	
1	11	1	2	1	
2	10.4	1	2	0	
2	11	1	2	1	
3	11	1	2	1	

$=: \mathbf{y}$ $=: \mathbf{X}$ $=: \boldsymbol{\beta}$

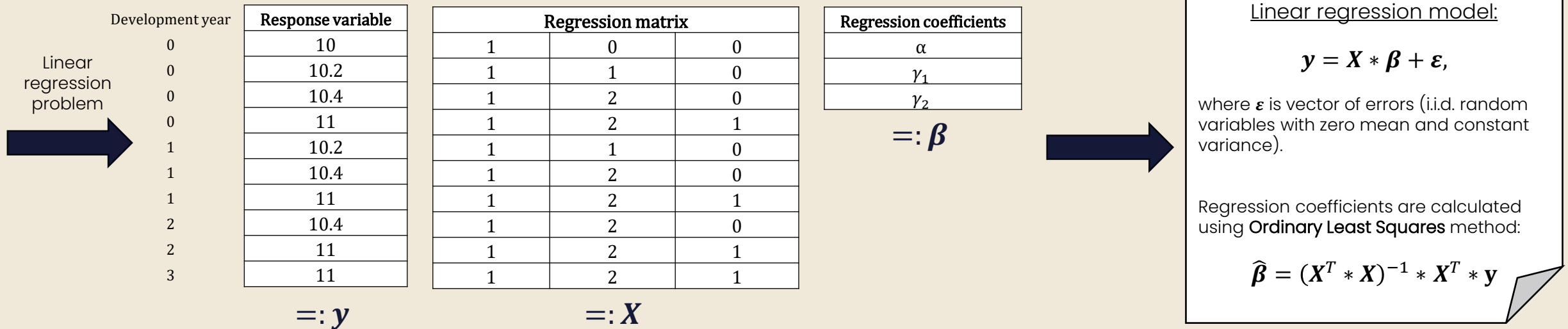
Simple example of PTF model

Log-transformed incremental claims triangle

	0	1	2	3
2018	10	10.2	10.4	11
2019	10.2	10.4	11	
2020	10.4	11		
2021	11			

Log-transformed incremental claims triangle in form of PTF model

	0	1	2	3
2018	$10 = \alpha$	$10.2 = \alpha + \gamma_1$	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$
2019	$10.2 = \alpha + \gamma_1$	$10.4 = \alpha + \gamma_1 + \gamma_1$	$11 = \alpha + \gamma_1 + \gamma_1 + \gamma_2$	
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Linear regression model:

$$\mathbf{y} = \mathbf{X} * \boldsymbol{\beta} + \boldsymbol{\varepsilon},$$

where $\boldsymbol{\varepsilon}$ is vector of errors (i.i.d. random variables with zero mean and constant variance).

Regression coefficients are calculated using **Ordinary Least Squares** method:

$$\hat{\boldsymbol{\beta}} = (\mathbf{X}^T * \mathbf{X})^{-1} * \mathbf{X}^T * \mathbf{y}$$

Linear regression problem



$$\hat{\boldsymbol{\beta}} = (\hat{\alpha}, \hat{\gamma}_1, \hat{\gamma}_2)^T = (10, 0.2, 0.6)^T$$

PTF model: best practices in modelling

Negative increments

Possible ways to handle **negative incremental paid losses**, which **cannot be log transformed**:¹⁷

- Zero out negative values, treating them as 0 after log-transforming the rest.
- Replace the value with $-\log(-q(w, d))$ instead of $\log(q(w, d))$.
- Shift all values to eliminate negatives before taking the logarithm, then shift back after analysis.

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Trends selection

Methods to determine **AY, DY and CY trends coefficients**:¹⁵

- **By inspection:**
 1. Fit the data with a basic model (assuming only one trend in each direction)
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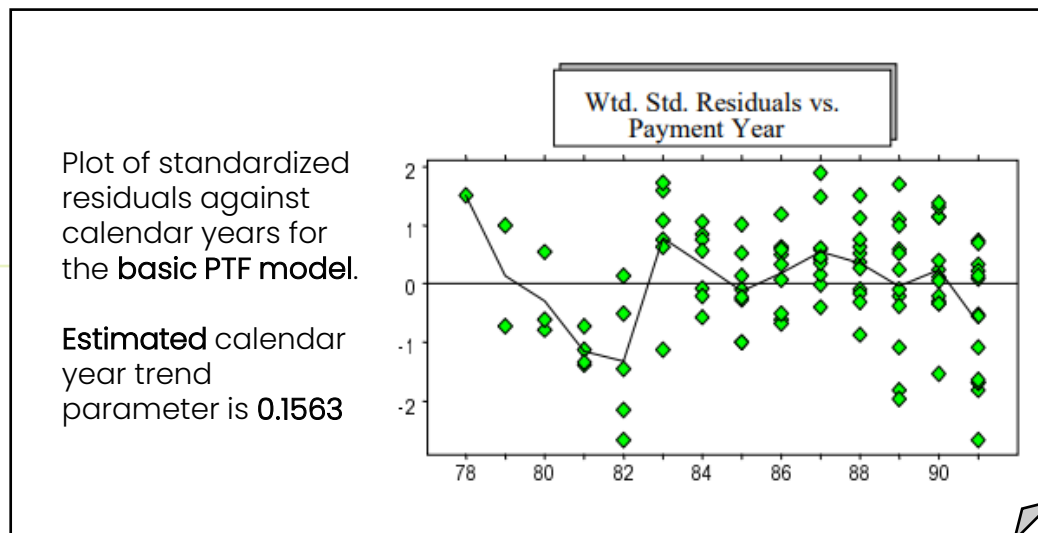
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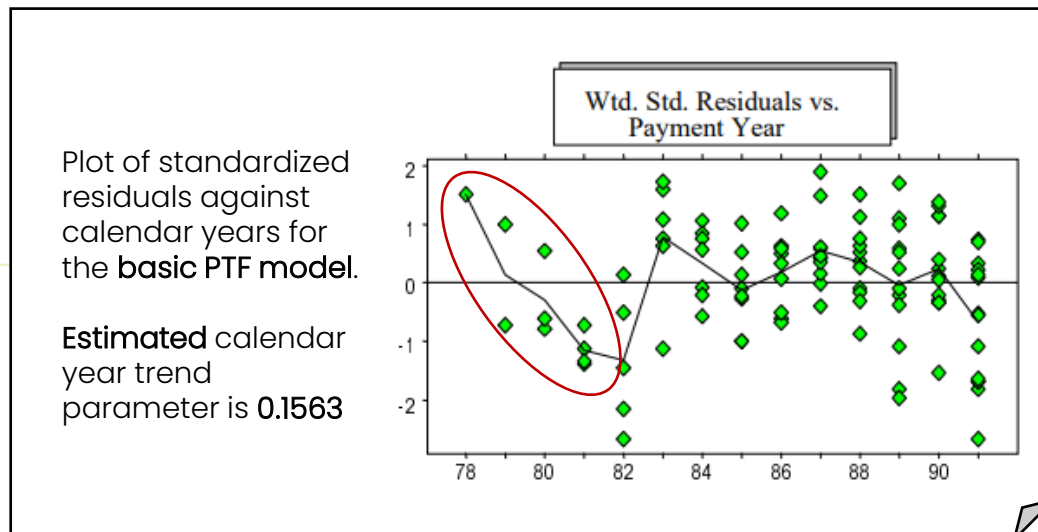
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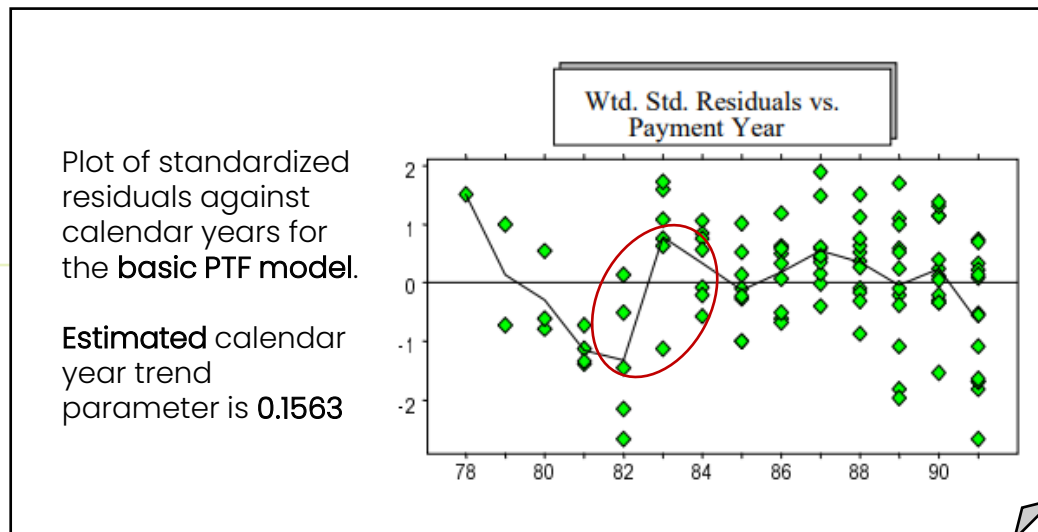
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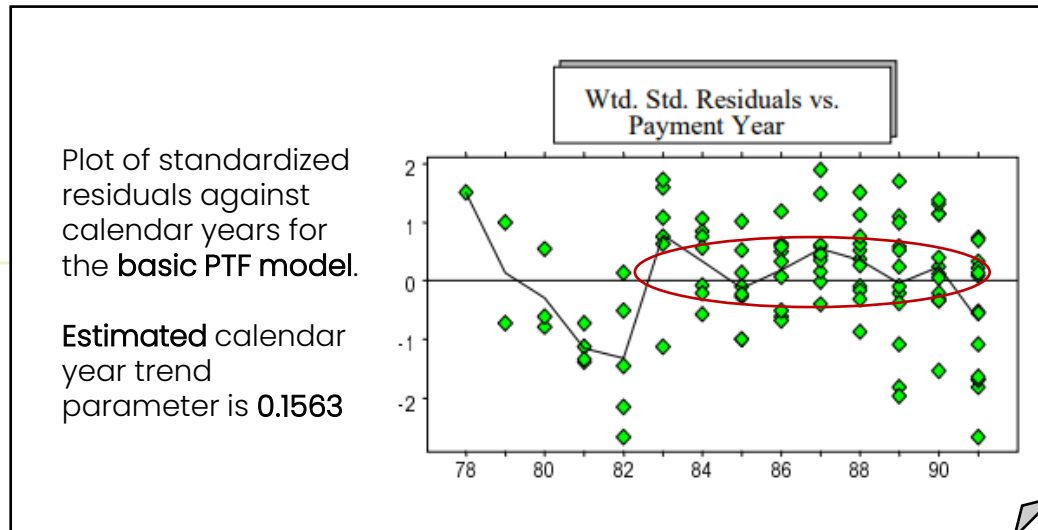
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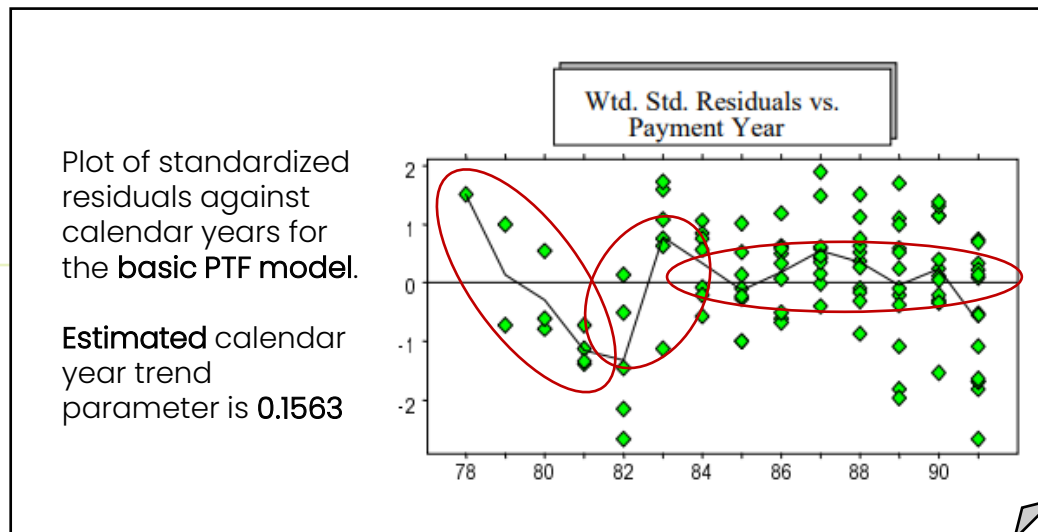
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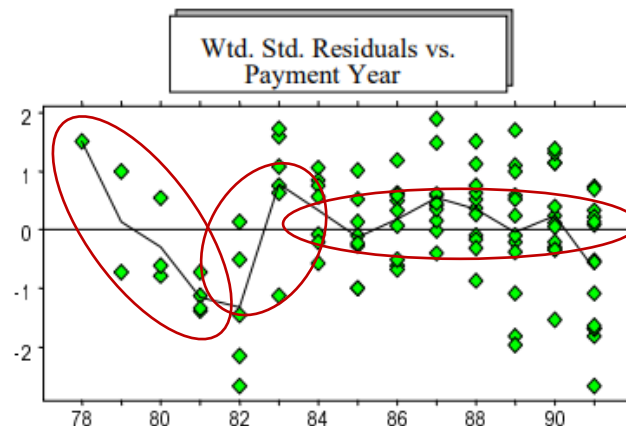
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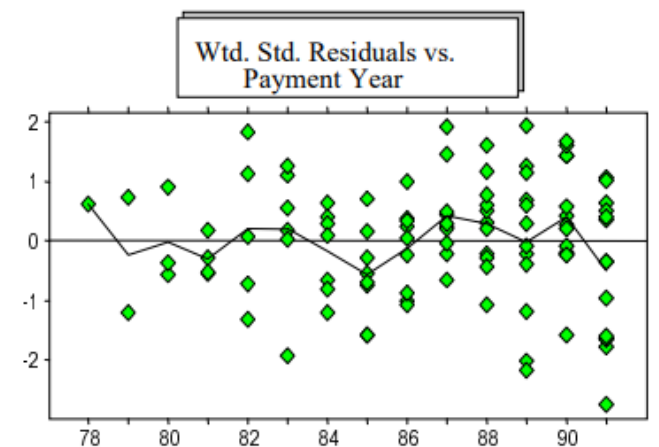
This method can vary based on the analyst's interpretation.

Plot of standardized residuals against calendar years for the **basic PTF model**.

Estimated calendar year trend parameter is **0.1563**



Plot of standardized residuals against calendar years for the PTF model with **three calendar year trends**.



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- **Stepwise selection:** choose the combination of predictors that minimizes the Akaike Information Criterion, balancing goodness of fit with the number of parameters to avoid overfitting (*computationally difficult*).

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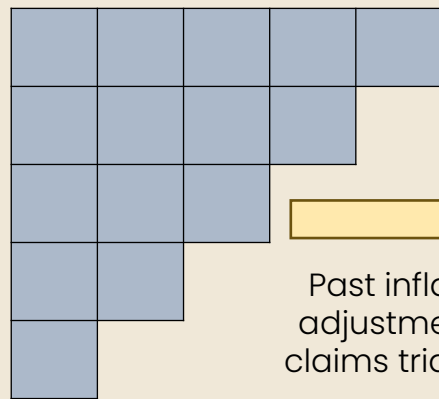
Projecting CY trends

To **estimate ultimate losses**, we need to project **future CY trends**, which may not be observable in current data.

- If **calendar year trend** has been **stable in the more recent years**:¹⁵
 - e.g. if the estimate of calendar year trend in the most recent years is $\hat{\gamma} \pm \text{s.e.}(\hat{\gamma})$, then we assume for the future a **mean trend of $\hat{\gamma}$** with a **standard deviation of trend $\text{s.e.}(\hat{\gamma})$** .
- If **calendar year trend** has been **unstable in the more recent years**:
 - analysing other data types and using any relevant **business knowledge** (though this can be complex as some trends may overlap with development and accident year trends).

Accounting for claims inflation explicitly – general approach

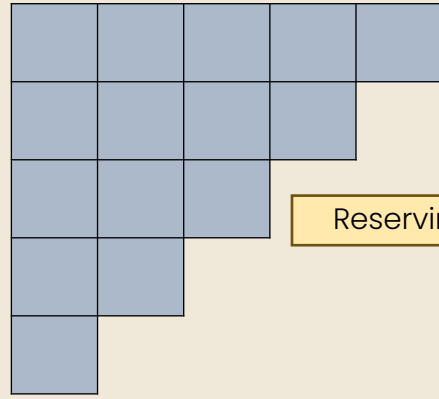
Claims triangle
(paid claims preferably)



Past inflation
adjustment of
claims triangles

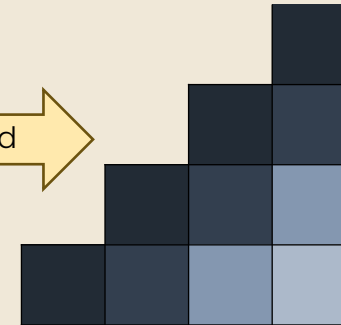
Past inflation can be estimated using mathematical methods or through expert judgement.

Claims triangle in real terms
(at current price levels)

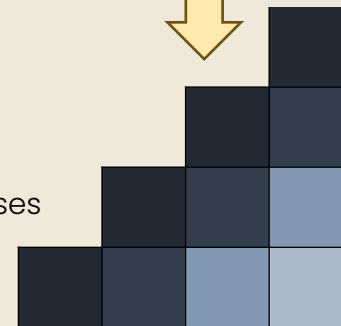


Reserving method

Future best estimate
losses
(at current price levels)



Future best estimate losses
(at future price levels
including explicit future
inflation)



Calibrated expected future
claims Inflation:

Option 1: deterministic

Weighting of the economic inflation drivers must be performed and **adjusted** by other factors (e.g. **social inflation**) to estimate future expected claims inflation per insurance business.

This process uses public information (e.g. **breakdown of overall CPI projections**) in combination with expert judgement.

Option 2: stochastic

Future claims inflation can be modelled as a function of **inflation projections from an Economic Scenario Generator**, allowing it to be stochastic rather than deterministic.

Lloyd's guidance on managing claims inflation

LLOYD'S

Reserving Expectations of Syndicates

"We expect syndicates to explicitly consider economic and excess inflation (including social inflation) in their reserving process **when setting best estimate reserves**. This is particularly important when historical data is unlikely to be representative of the future and traditional reserving techniques do not address this.

Where syndicates are not making an explicit additional allowance in their best estimate reserves for inflation, they must be able to explain why their approach is appropriate and how they have gained sufficient comfort that their reserves are adequate."

LLOYD'S

Best practice

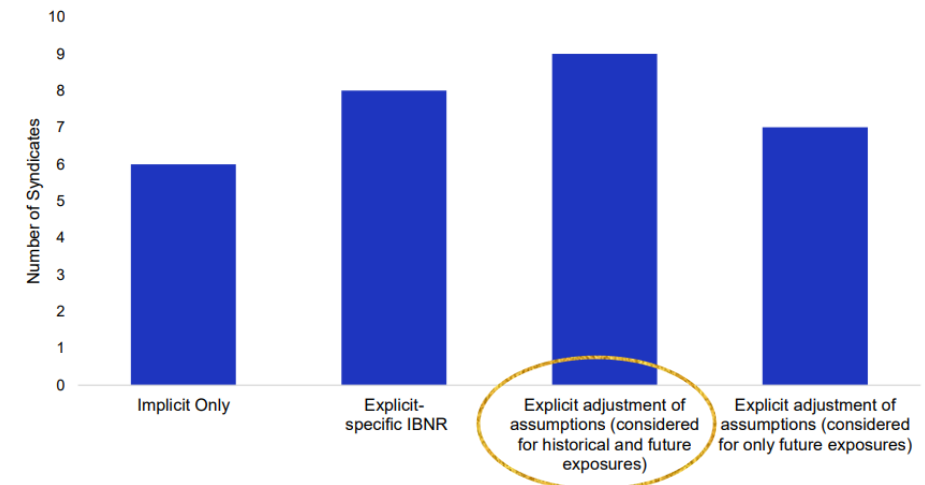
Best estimate reserve inputs

The best approaches have explicit consideration for inflation such as:

Generalised linear models

Inflation-adjusted chain ladder

How is Social Inflation allowed for in Syndicate's reserves?



Reinsurance Products that Address Claims Inflation

The most common risk management measure taken is purchasing reinsurance cover for classes specifically exposed to social inflation or claims inflation in general.

Considerations in Reinsurance Agreements

Reinsurance Contract Exclusions:

Social inflation has led to several **exclusions in reinsurance contracts**.

Some common exclusions include

- nuclear verdicts
- TPLF
- class action lawsuits
- attorney-driven claims inflation.

It is crucial for insurance companies to **carefully review reinsurance agreements** for such exclusions before finalizing any deals.

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Social inflation and different types of reinsurance agreements:

In *proportional reinsurance*, the insurer and reinsurer **share** premiums and losses, **including the effects of social inflation**. This provides the insurer with **proportional protection** against rising claims costs.

Non-proportional reinsurance, however, only covers losses above a set threshold, potentially leaving the insurer **more exposed to inflation-driven increases**:

- To manage this, reinsurers may use **additional tools**—such as **indexation clauses** that adjust limits based on inflation indices—to maintain the real value of coverage.

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Non-proportional reinsurance, however, only covers losses above a set threshold, potentially leaving the insurer **more exposed to inflation-driven increases**:

- To manage this, reinsurers may use **additional tools**—such as **indexation clauses** that adjust limits based on inflation indices—to maintain the real value of coverage.

Reinsurance tools for hedging against social inflation risk:

Adverse development covers (ADC) and **loss portfolio transfers** (LPT) are two possible tools used to hedge against claims inflation.

While non-proportional reinsurance protects against large, unexpected losses,

- ADC provides additional protection by **capping liabilities from worsening claims over time**,
- LPT provides additional protection by **transfers existing and future claim liabilities to the reinsurer**.

Together, they help insurers manage both **immediate high-severity risks and long-term reserve** uncertainties, offering a more comprehensive risk management approach.

Indexation clause

What is it?

- Definition: IC in a reinsurance contract is a contractual provision that **adjusts** the **retention** and **limit** amounts based on a **specified inflation index**.¹⁸
- Purpose: to maintain the real value of the **reinsurance coverage** over time by accounting for inflation.

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How does it work ?

- Redistributes inflation-related increases by adjusting retention and limit amounts based on an inflation index.
- For **XoL treaties**, IC is particularly useful where the **underlying losses take a long time to be paid**.

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Example scenario:

XoL treaty **\$20m xs. \$5m** for motor liability, established in **2010**.

Inflation reached **25% by 2020** (base 1 in 2010 to 1.25 in 2020).

A loss **settled** at **\$15m** in 2020.

Calculation:

Real value of the loss in 2010 terms:

$$\$15\text{m} / 1.25 = \$12\text{m}$$

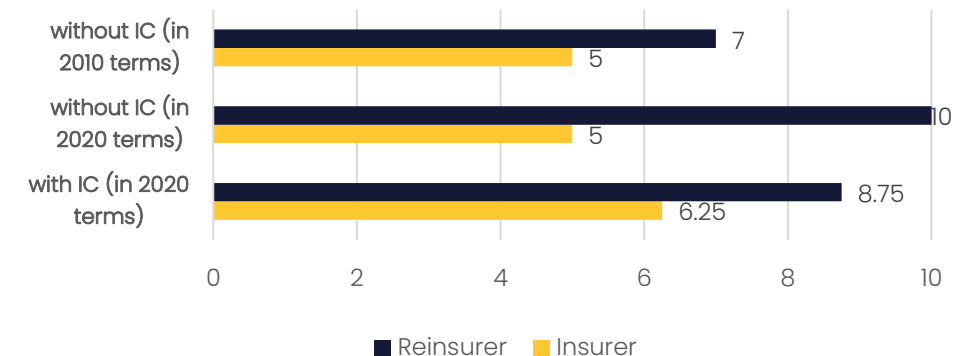
Adjusted retention:

$$\$5\text{m} * 1.25 = \$6.25\text{m}$$

Adjusted limit:

$$\$20\text{m} * 1.25 = \$25\text{m}$$

Distribution of loss



Adverse development cover

What is it?

- Definition: ADC is a type of reinsurance that provides **coverage for losses that exceed the insurer's carried reserves**.¹⁹
- Purpose: ADC helps insurers manage the financial impact of claims inflation by **providing a buffer against unexpected claim developments**.

Adverse development cover

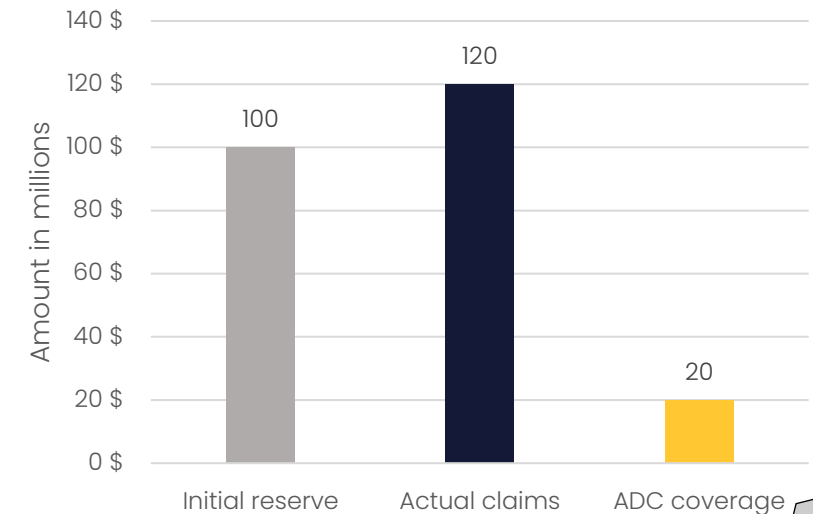
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How does it work?

- ADC activates when **actual claims exceed the insurer's reserved amount, covering the excess**.
- This helps maintain financial stability and manage reserve risk effectively.

Example scenario:



Loss Portfolio Transfer

What is it?

Definition: LPT is a reinsurance agreement where an insurer **transfers existing claim liabilities** and related **reserves to a reinsurer**.²⁰

Purpose: **provides** insurers with **capital relief** by transferring existing claim liabilities to a reinsurer, helping manage claims inflation and stabilizing the insurer's financial position.

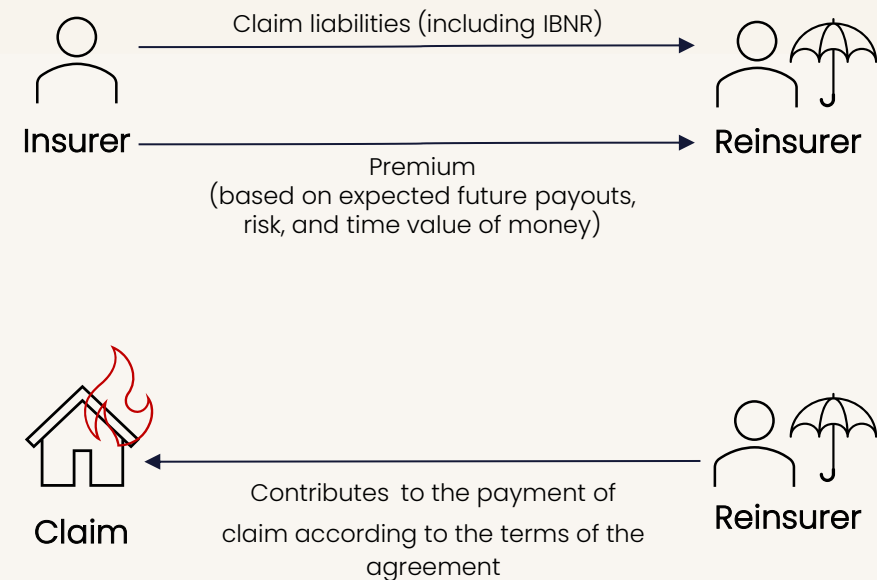
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How does it work?



Thank you for your attention

Contact us:

priceforbesre

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