

# Equity-Based Insurance Guarantees Conference

Nov. 6-7, 2017

Baltimore, MD

## Behavioral Analytics for Annuities

Timothy Paris

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# 2017 Equity-Based Insurance Guarantees Conference

**Session 2B – Behavioral Analytics for Annuities**

**November 6, 2017**

**1:30-3:00pm**

**Timothy Paris, FSA, MAAA**

**Ruark Consulting LLC**



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Source: LIMRA	Variable Annuities	Fixed Indexed Annuities
Gross Sales (p.a.)	~\$150 billion	~\$100 billion
Net Sales (p.a.)	~\$0 billion	?
% Qualified	65%	55%
% Guaranteed Living Benefit	77%	68%

# Overview of VA Industry Experience

# VA Industry Data

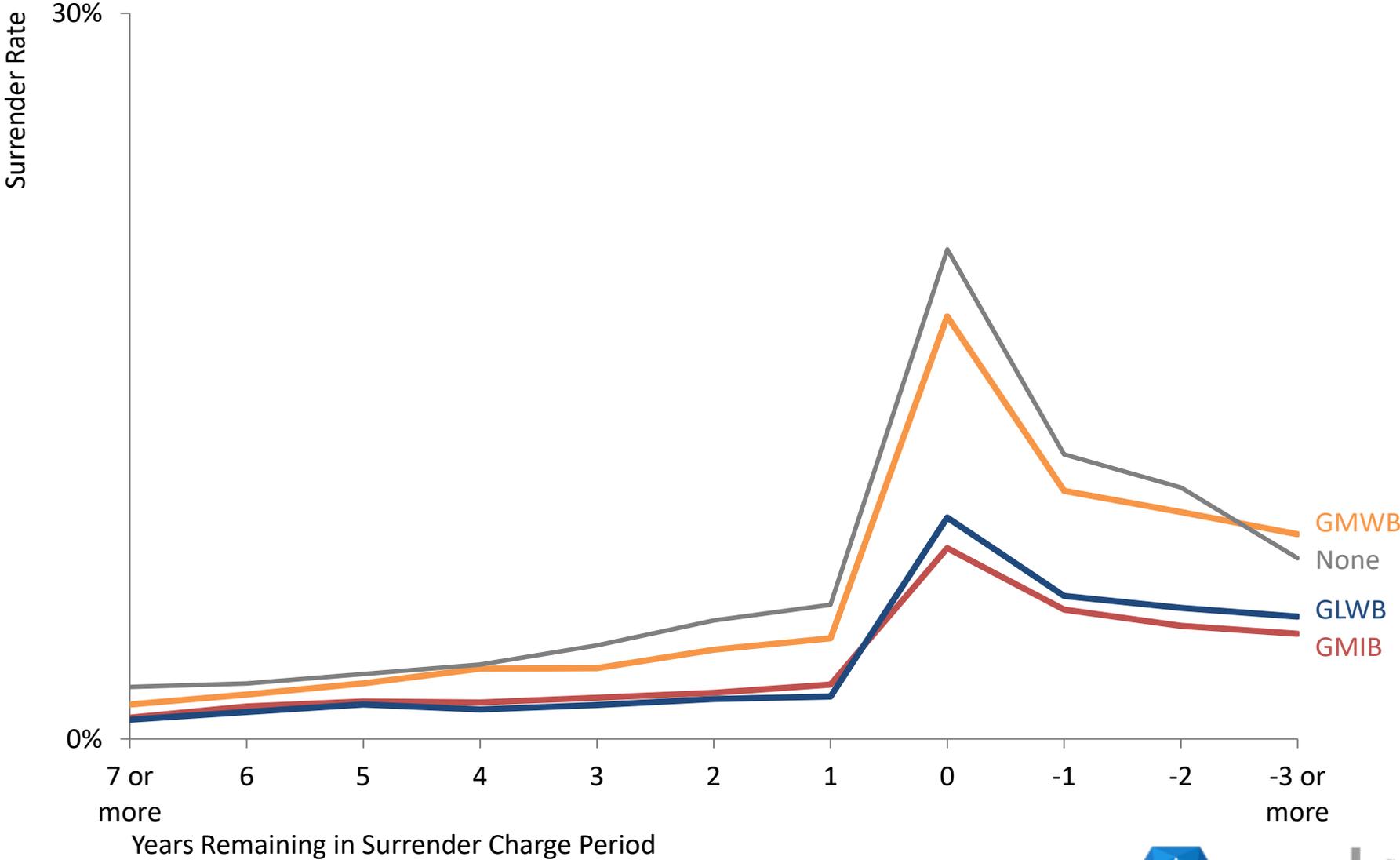
22 participating companies

2008 to present

68 million contract years of exposure

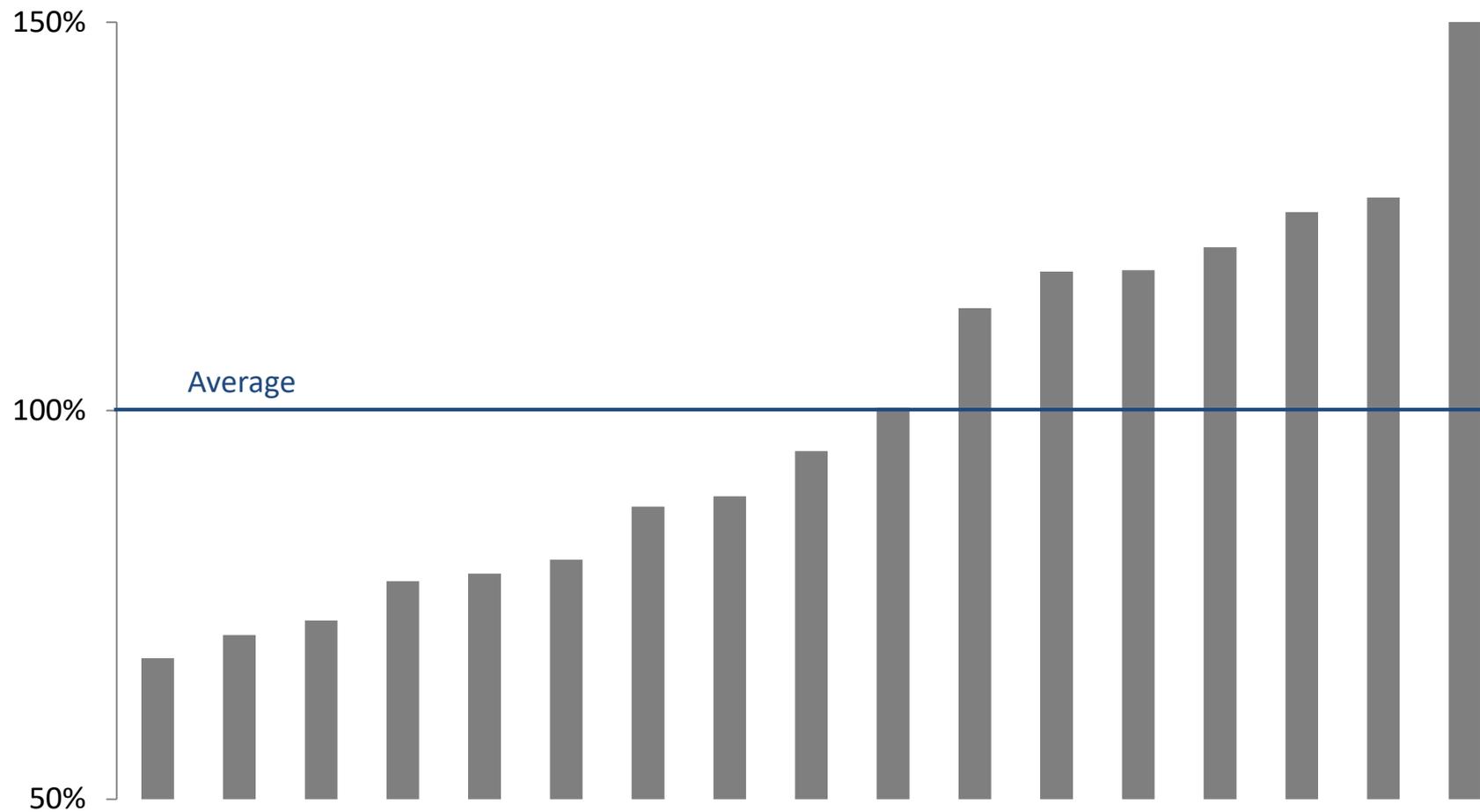
+22% from last year

# Surrenders vary by living benefit type

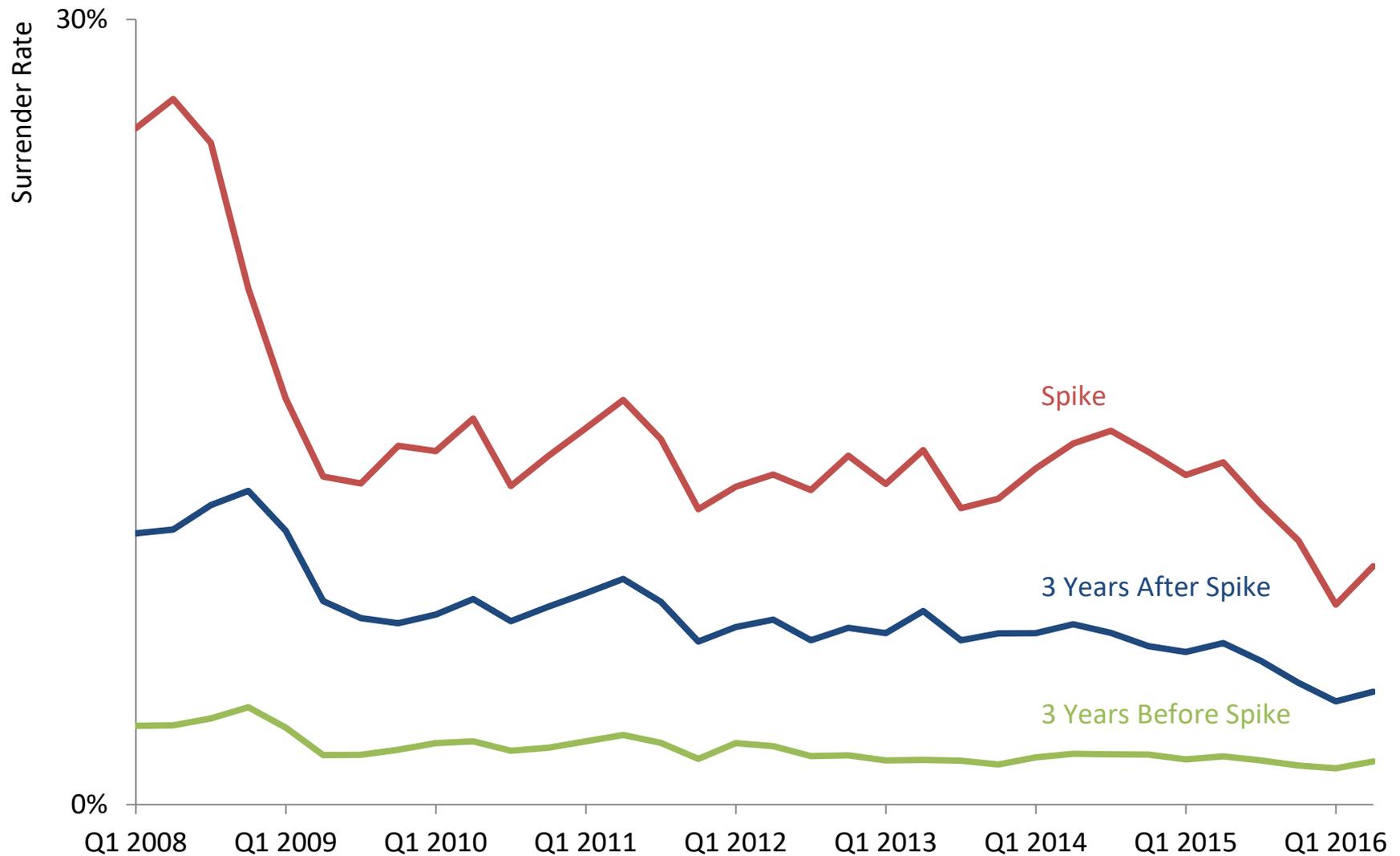


# Experience varies by company, but why?

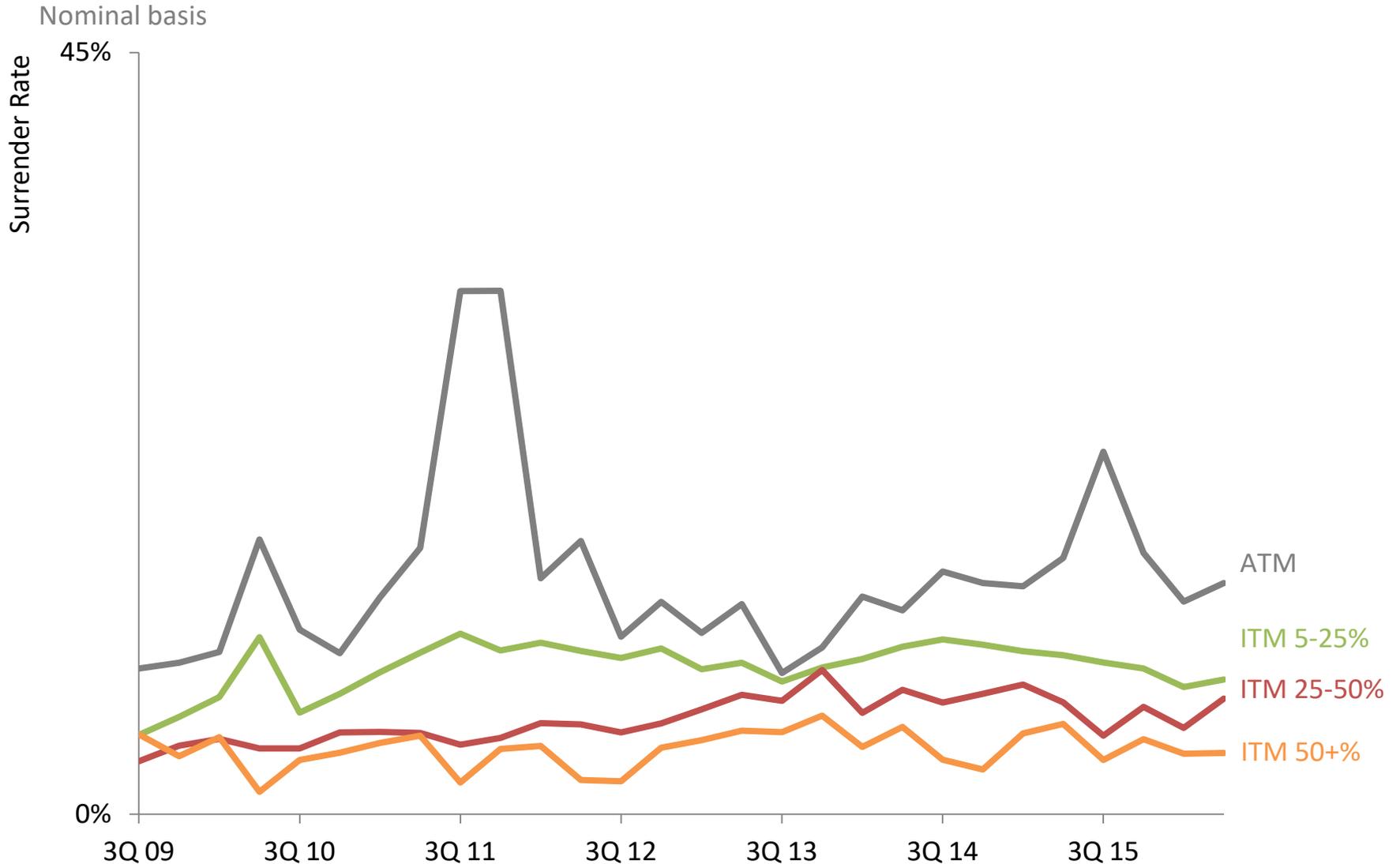
GLWB, Normalized by Years Remaining in Surrender Charge Period



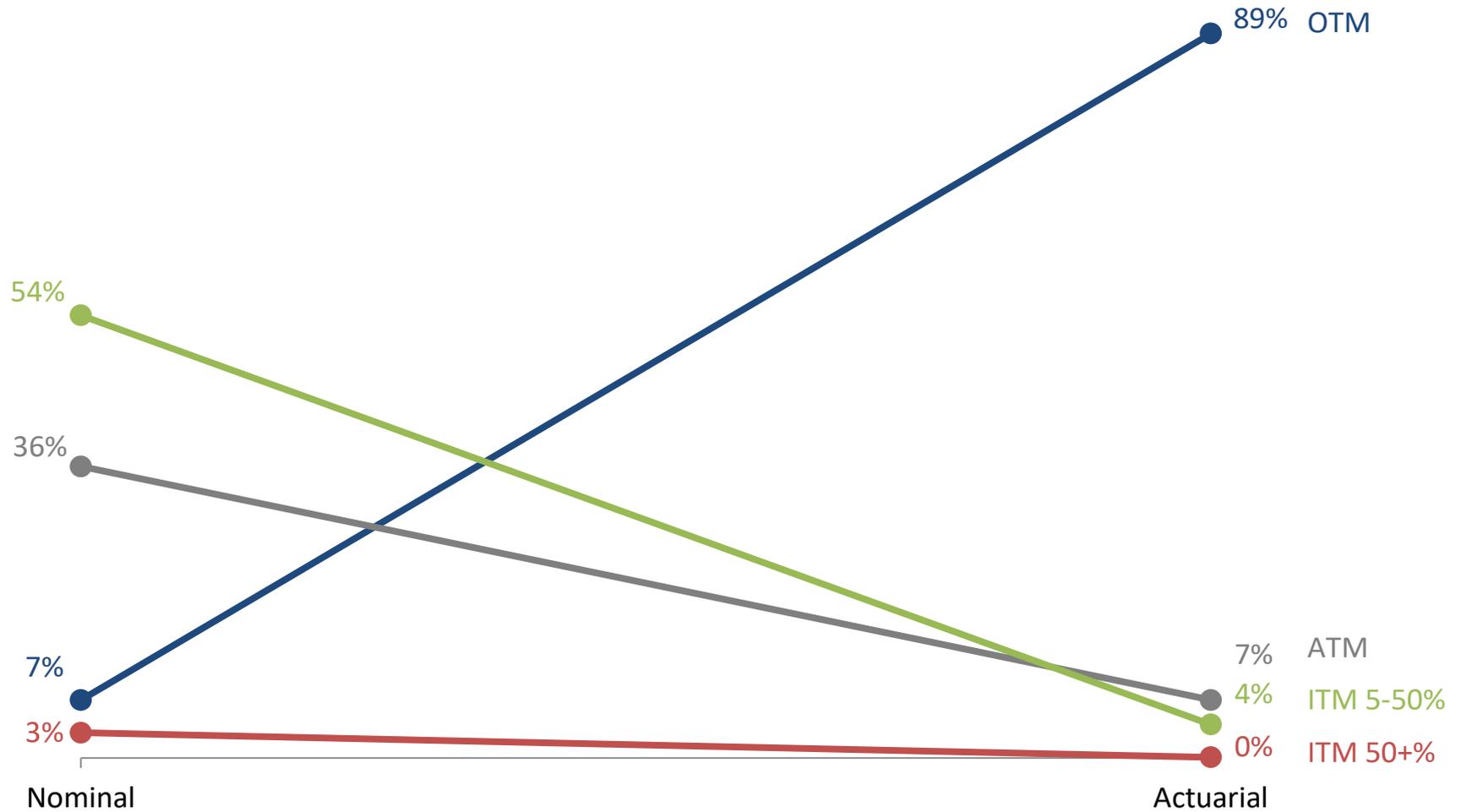
# Surrenders have decreased since the crisis



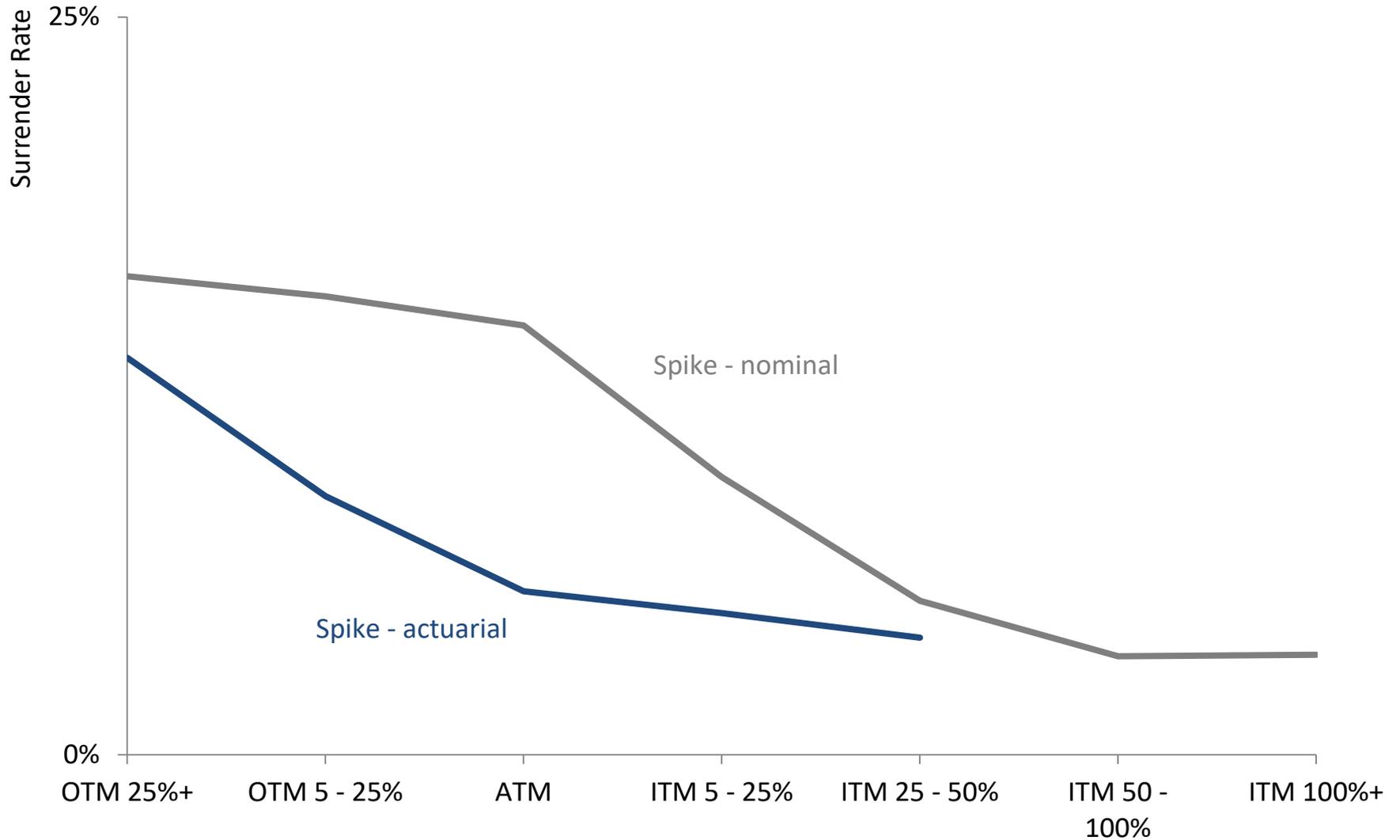
# However, a different trend for GLWB “spike”



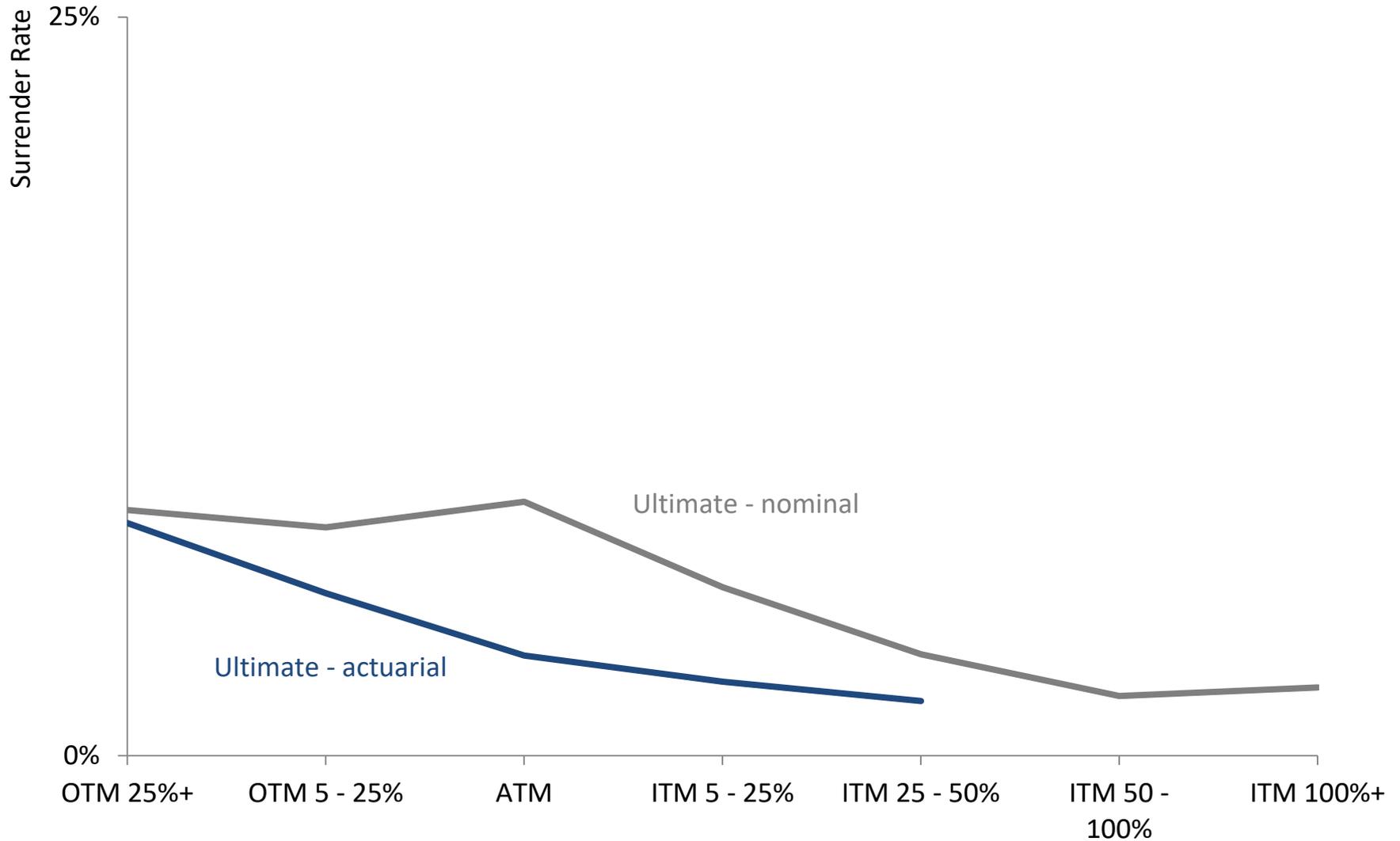
# Most GLWBs are actuarially out-of-the-money



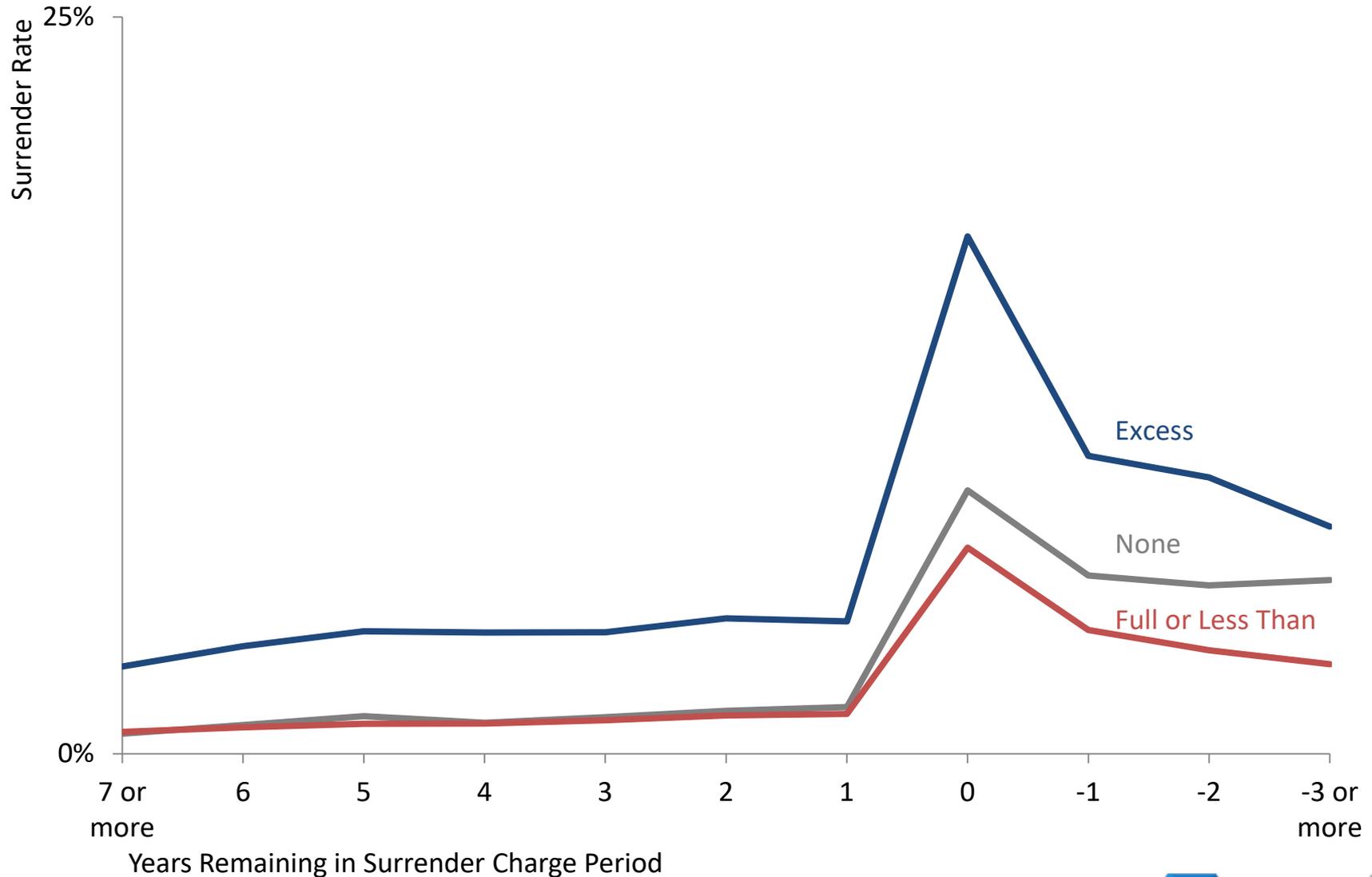
# GLWB moneyness basis matters



# GLWB moneyness basis matters

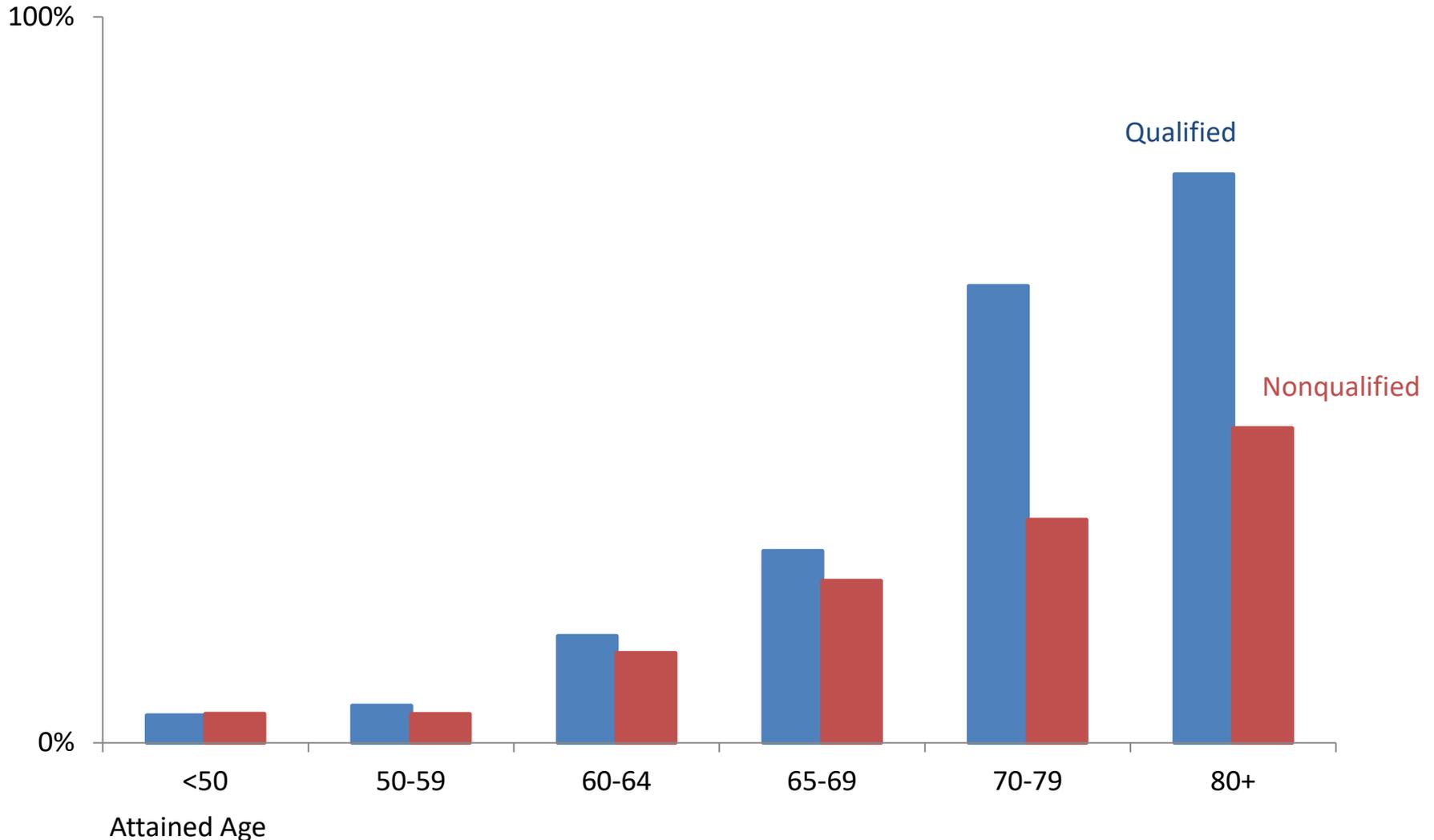


# GLWB income utilization affects surrenders



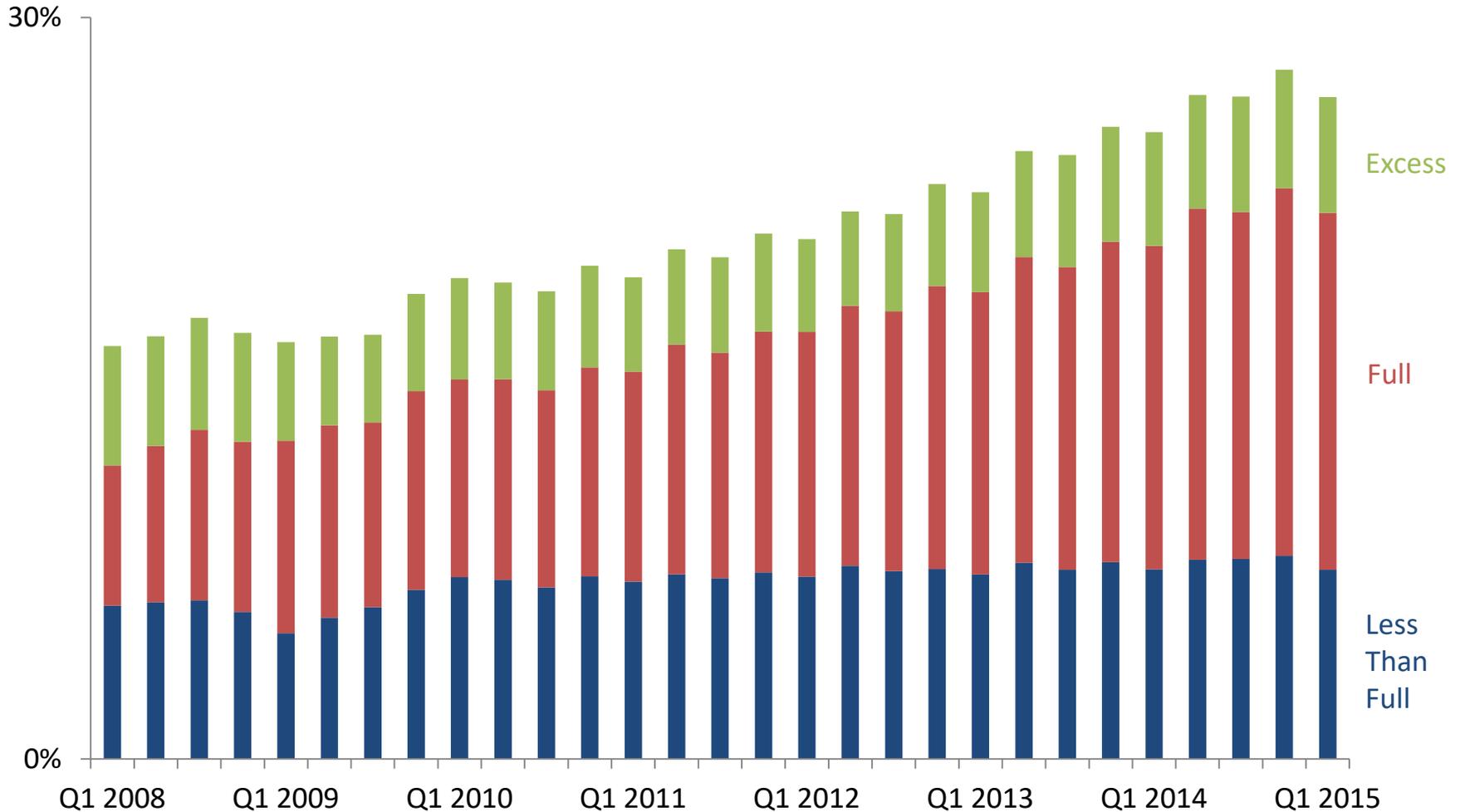
# Income utilization varies by age and tax status

GLWB Partial Withdrawal Frequency



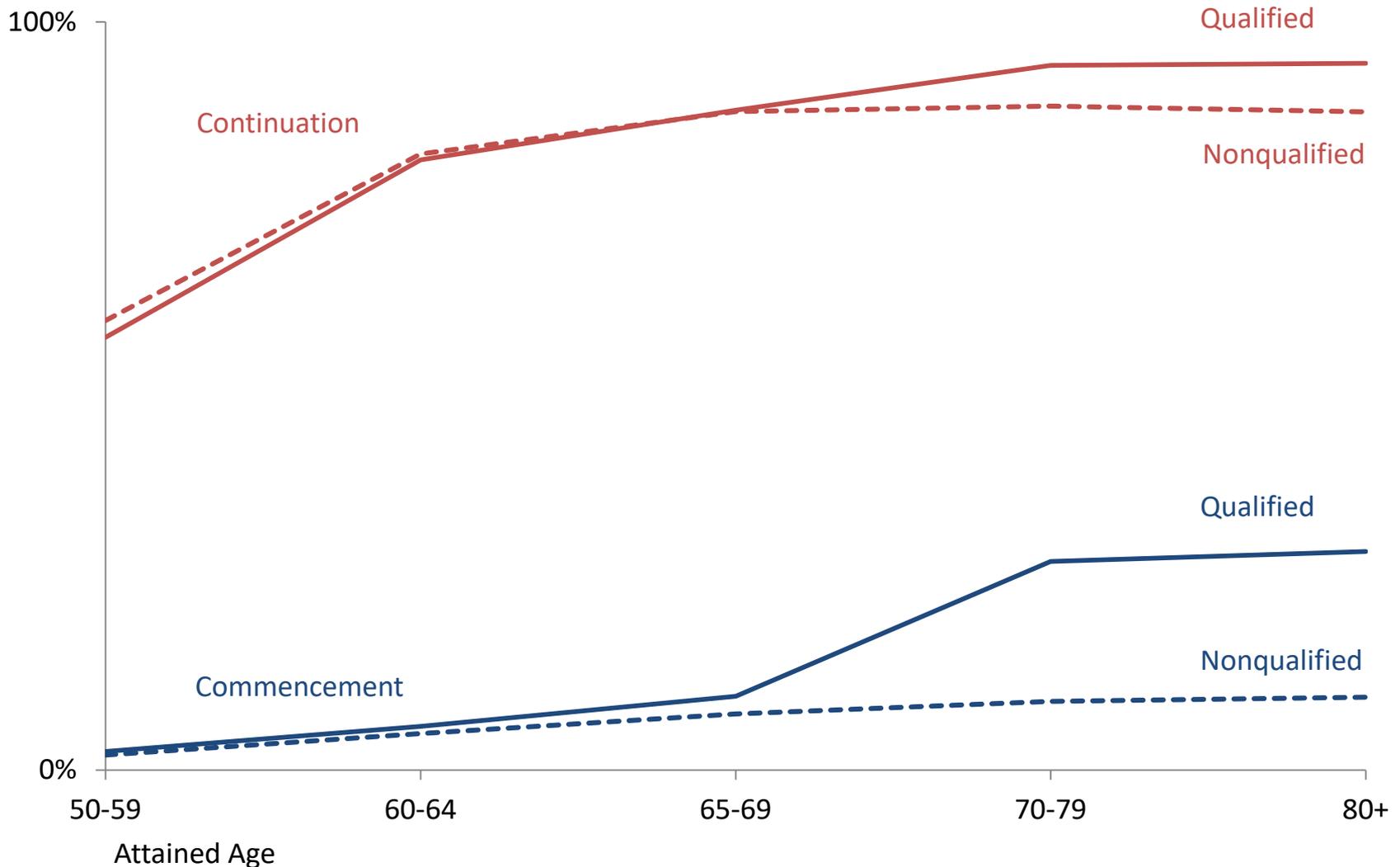
# Income utilization efficiency has increased

GLWB Partial Withdrawal Frequency and Amounts



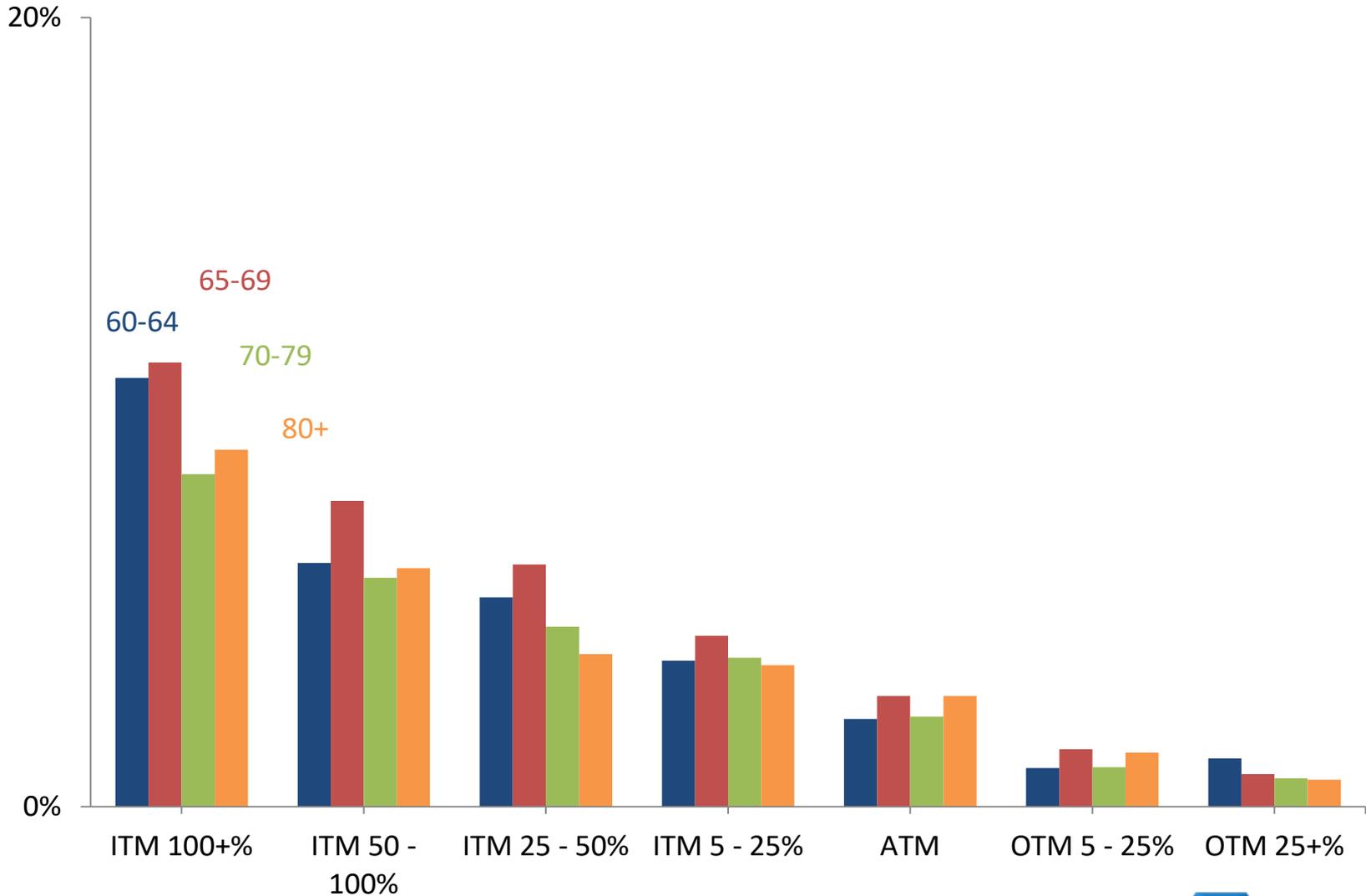
# Income commencement is the key question

GLWB Partial Withdrawal Frequency

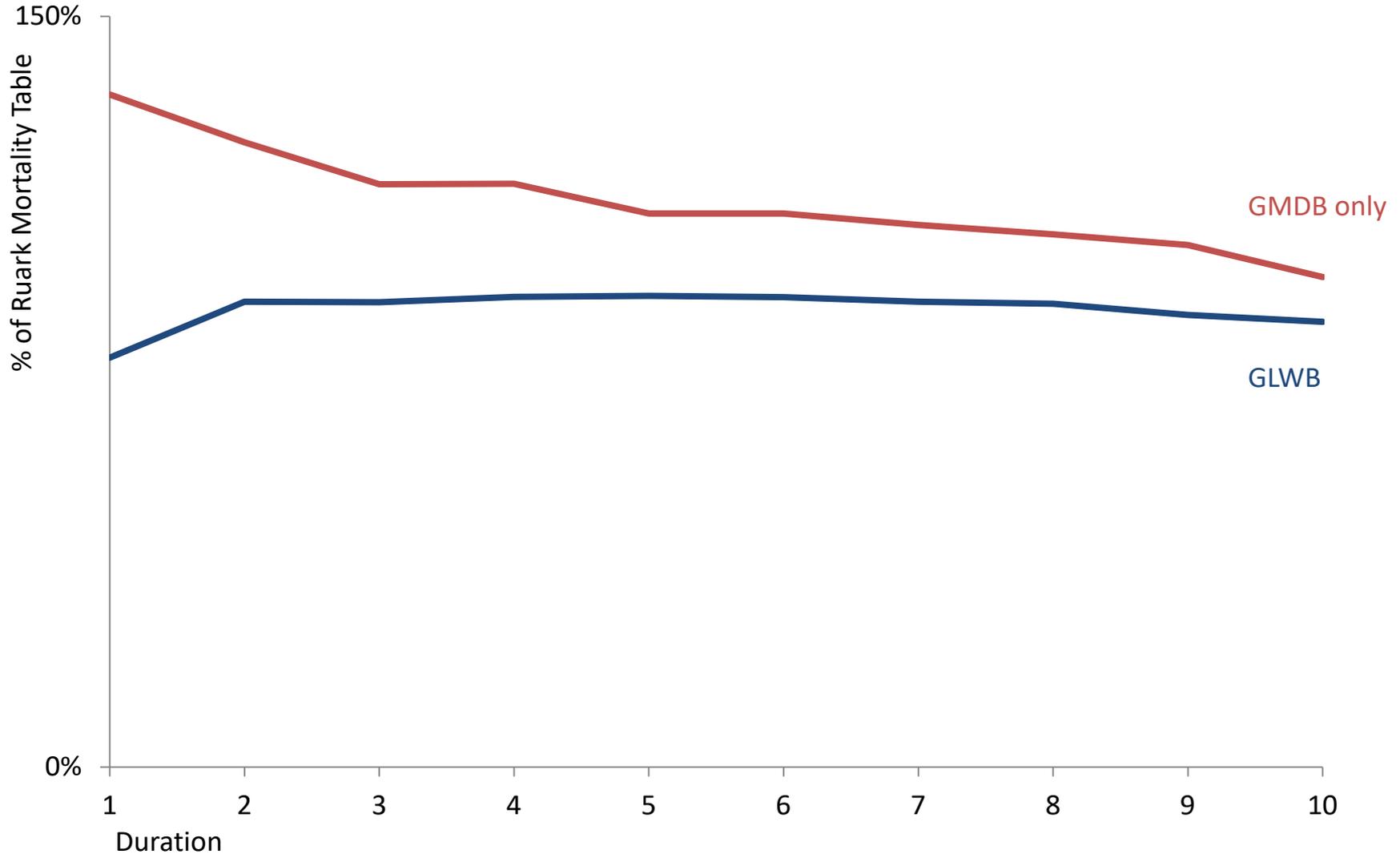


# GMIB annuitizations are low

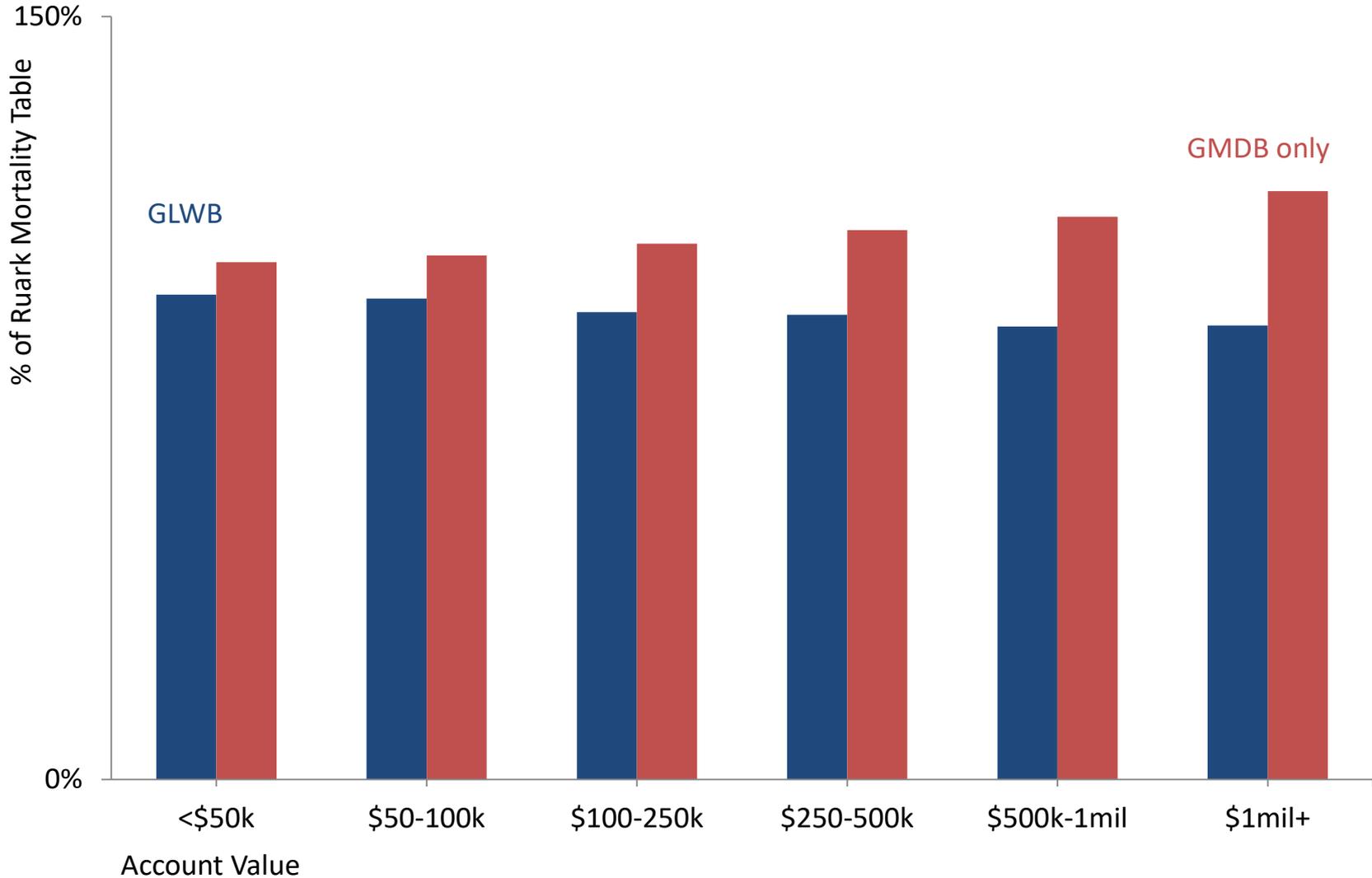
Actuarial basis



# Guarantees can affect mortality too



# Mortality effects are amplified by policy size



# Overview of FIA Industry Experience

# FIA Industry Data

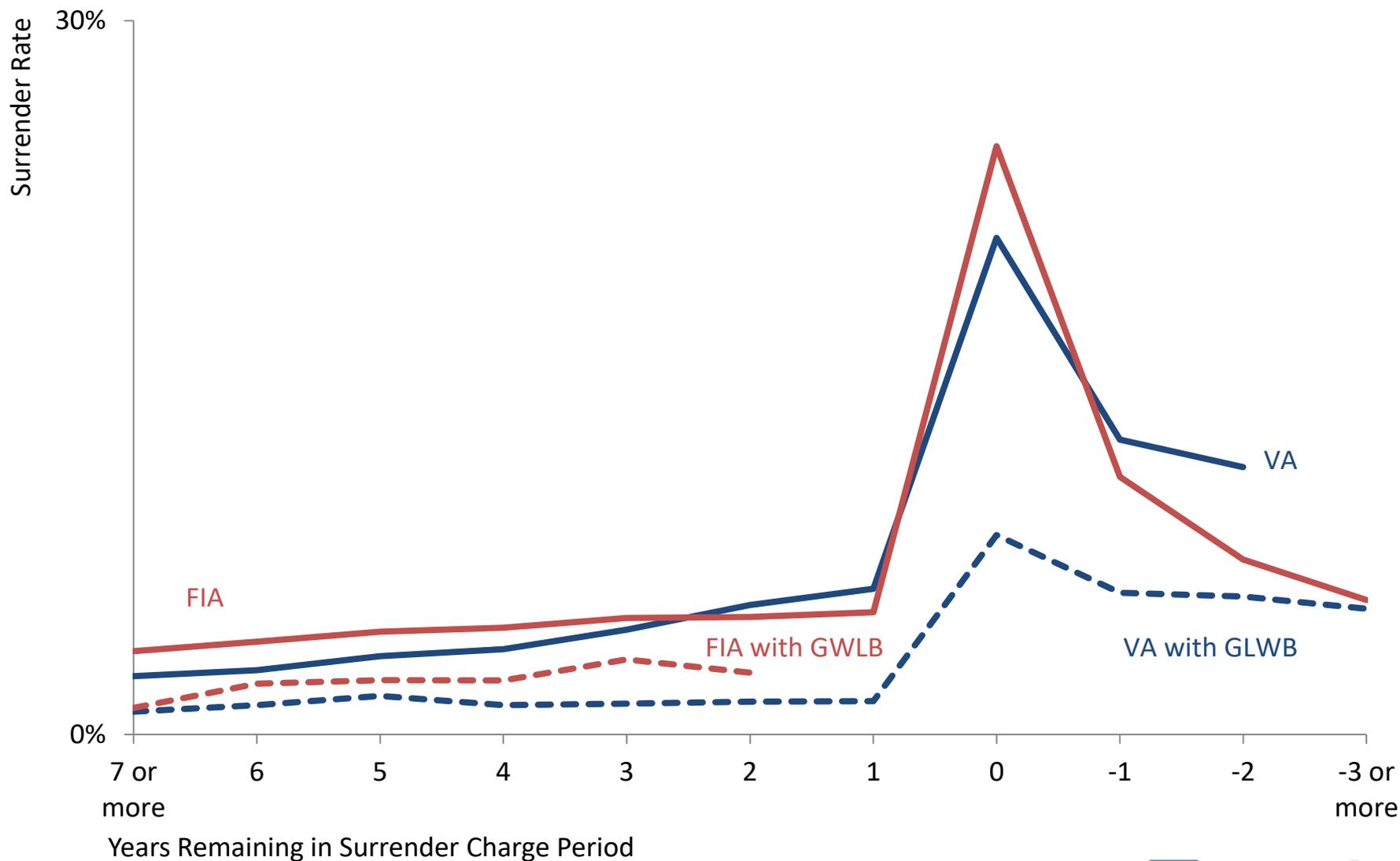
12 participating companies

2007 to present

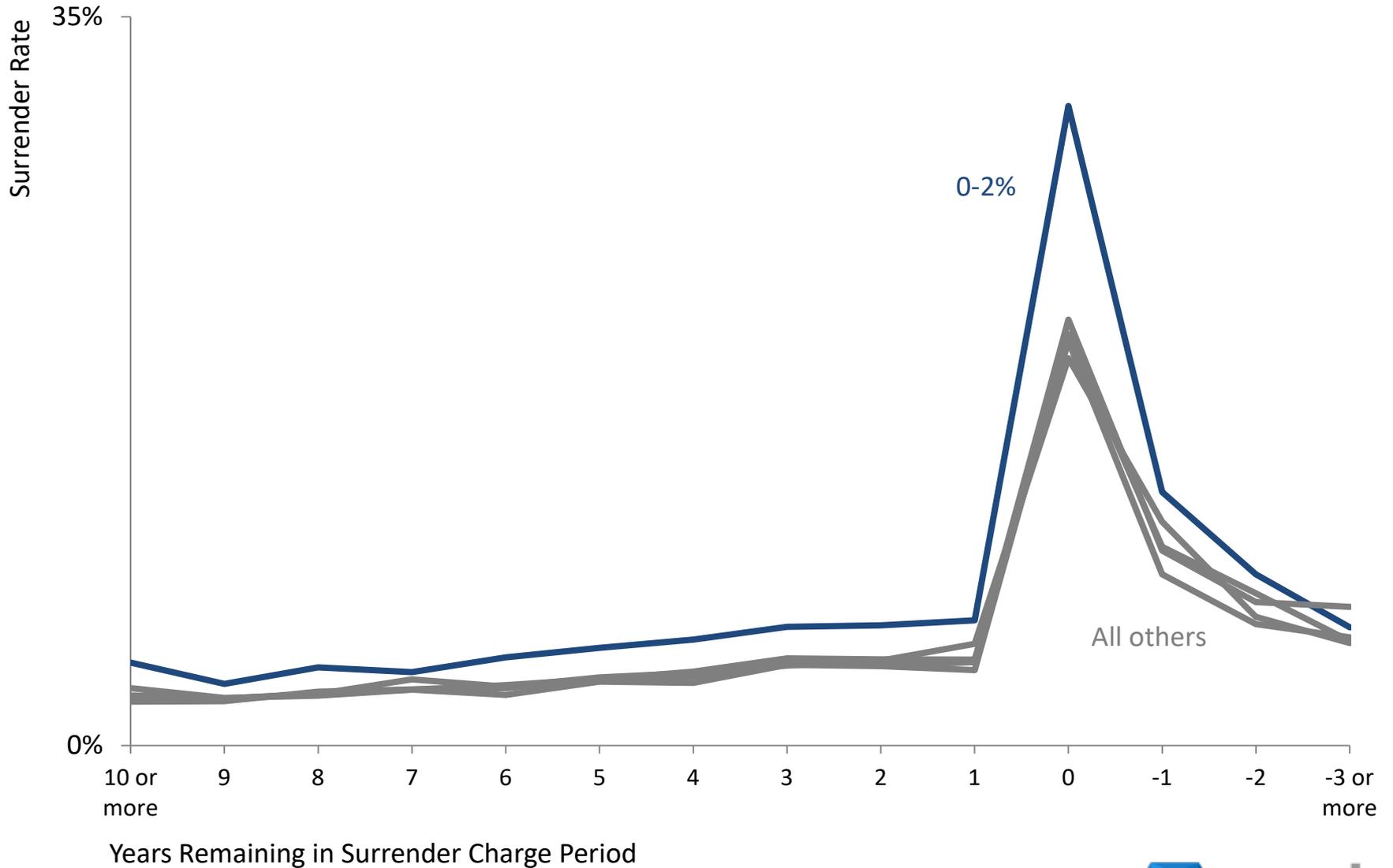
13 million contract years of exposure

+30% from last year

# VA and FIA surrenders are lower with GLWB



# FIA surrenders vary based on interest credited



# Behavioral Analytics Framework



# Model Development

Start with maximum data set (industry)

Extract relevant subset for a company

Develop a model on this basis

Do likewise using only company's data

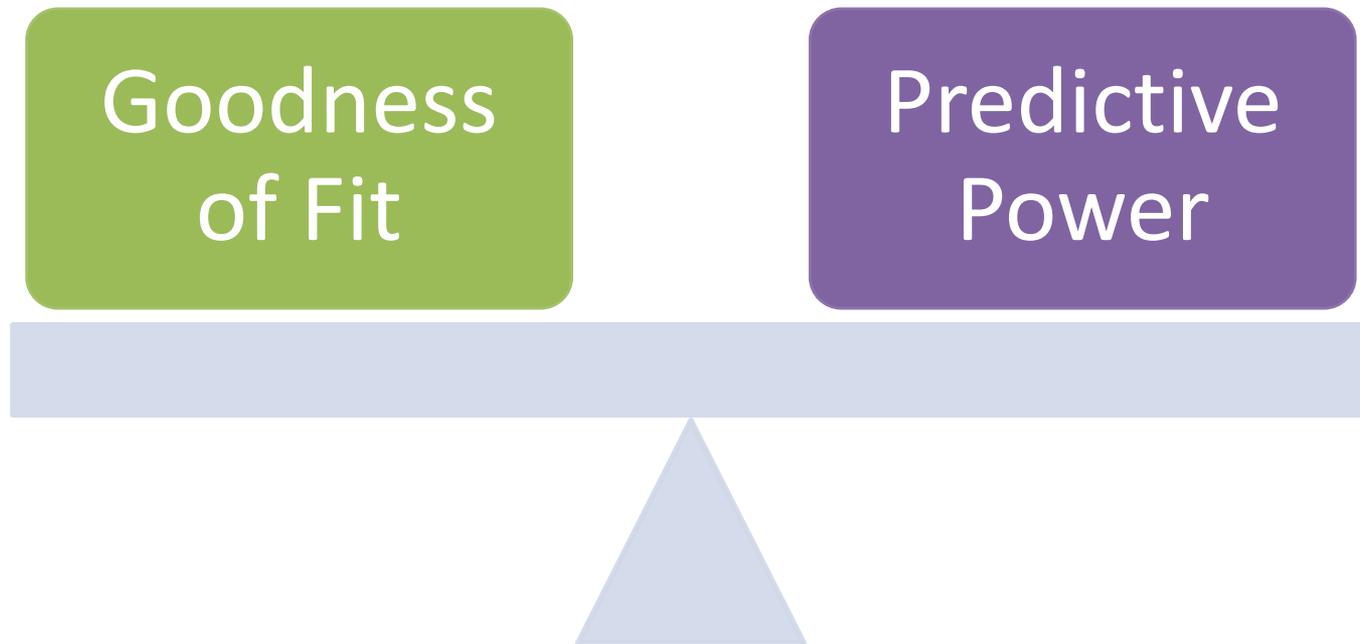
Customize model to reflect both, so that most important factors are included, with stable coefficients, balancing goodness-of-fit and predictive power

You can go far with Generalized Linear Models (GLM)

# Logistic Regression Model

$$\ln\left(\frac{\mu}{1-\mu}\right) = \beta_0 + \sum \beta_i x_i$$

“Log of odds” is a linear function of key factors  
Binary values, such as surrenders or deaths



# Bayesian Information Criterion

Rewards goodness-of-fit to historical data, but penalizes for additional factors used in your model

One of many metrics to help guide your model selection process

# Actual-to-Expected Ratios

“Predictive Power” in the new vernacular

Develop E using train data, compare to A from test data

Out-of-sample, out-of-time, and k-fold cross-validations

Examine in aggregate, by cohorts, and over time

Look at range of outcomes and tails

# Expert Judgment is Vital

Business context, sensibility, materiality, parsimony

Let the data speak

More data usually beats more complex models

Build simple models for complex data, and complex models for simple data

# Sample Models

## VA Surrenders

Yrs Remaining in  
Surr Chg Period

LB Type and PW  
History

Moneyiness

Contract Size

Interactions

# Using industry data

For each factor coefficient, standard error terms  $\left(\frac{\sigma}{\mu}\right)$  are typically very small  $\sim 1/300$  to  $1/100$ .

Then testing predictive power using 5-fold cross-validation, average A/E errors are also very small  $\sim 1/700$ .

# Using company-only data

In some cases, company-only data is insufficient to even identify the key factors observed in the industry data, or it demonstrates factor coefficient estimates that are not sensible.

Even if they do, the coefficient standard error terms  $\left(\frac{\sigma}{\mu}\right)$  can be 20x larger.

Similarly, the average cross-validation A/E errors can be 10x larger.

# Combining industry and company-only data

A customized combination of industry and company-only data can produce a vastly superior model with much better fit and predictive power.

Such a model should identify and quantify the effects of each additional factor in the presence of the others, and the interactions between them.

Confidence increases with additional data.

# Integration Across Behaviors

Factor		Coefficient
Intercept		-2.0
	7	-4.0
	6	...
Years	...	...
Remaining	1	-1.0
in SurrChg	0	Intercept
Period	-1	-0.5
	-2	...
	-3	-0.8
LB Type	GMDB only	Intercept
and PW	Less than Full	...
History	Full	-1.2
	Excess	...
	...	...
	OTM	...
	ATM	Intercept
Moneyiness	ITM Band1	...
	ITM Band2	...
	ITM Band3	...
	ITM Band4	-1.0
Size	log (AV)	-0.2

Very important to model behavior on integrated basis

## VA GLWB / GMIB Income Utilization

Attained Age

Tax Status

Historical  
Income  
Utilization

Contract Size

Interactions

# The power of more data

As above, but for GWLB / GMIB income utilization, need to address complexities of frequency and severity relative to guarantee amounts.

	Average A/E Error		
	Less Than	Full	Greater Than
Industry	0.20%	0.30%	0.50%
Company-only	2.80%	1.50%	3.60%
Customized	2.00%	0.70%	3.20%

Customized model using industry data can reduce error by half where it matters most, for Full income utilization.

# Discussion