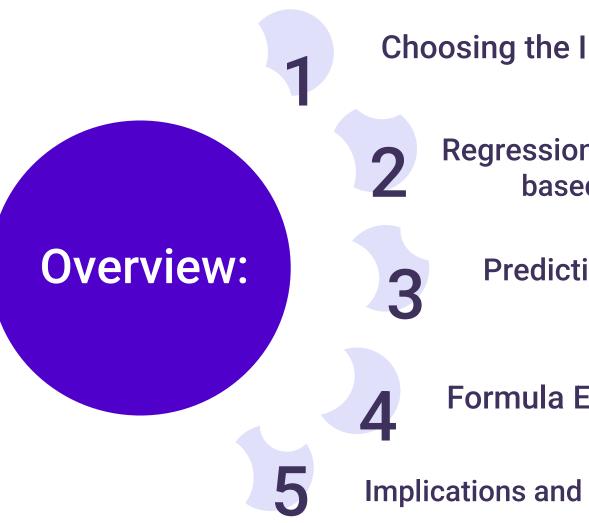
2023 BAS Case Competition

Team #19: Jru George, Shadrick Thompson, Ismael Martinez, Patricia Lansang



Choosing the Influential Factors

Regression and Calculating Formula based on Historical Data

Predicting and Modeling Future Lapse Rates

Formula Enhancement for GLB

Implications and Recommendations





Choosing the Influential Factors

Policy Years

Market Value Adjustments (MVA)

Mortality rates

Crediting Rates

5yr Treasury Rates

General Account Portfolio Yields

Statutory reserves

10yr Treasury Rates

Surrender Charges

Choosing the Influential Factors



2 Regression 2 Models

Logistic Regression

Linear Regression

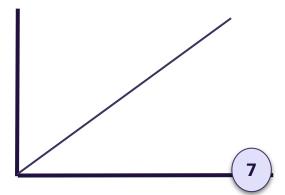
Logistic Regression Model:

- Typically used when the response variable is discrete
- Would require individual client data and their response (lapse or no lapse)
- Creates a model that indicates the probability of a client lapsing

Linear Regression Model:

VS.

- Typically used when the response variable is continuous
- Assumes that the relationship is linear
- Allows predictions on the rate of lapses



Methodology/Regression Model

- Inputs: 3 data sets (Issue Years: 1981, 1990, 1993)
 - 4 Input variables * 86 Inputs/data set
 - Outlier Years Removed
- Method: Least-Squares Linear Regression
 - Solves for each Independent Variable and 1 Constant
 - <u>Dummy Variable</u> activates on the Policy Year the Surrender Charge Fee ends
- How: Excel Plug-In (Analysis ToolPak)
 - Calculates Regression Statistics.
 - Fast Calculation allows for optimization

Residual Output

Table 1: Residual Output											
Issue-Year-1981				Issue-Year-1990				Issue-Year-1993			
Observation	Pred. Full Lapse	Residuals	Res. $\%$	Observation	Pred. Full Lapse	Residuals	Res. $\%$	Observation	Pred. Full Lapse	Residuals	Res. $\%$
1	0.1705	-0.0173	-1.73%	31	0.0940	-0.0088	-0.88%	59	0.0717	0.0018	0.18%
2	0.1636	0.0100	1.00%	32	0.1141	0.0128	1.28%	60	0.0854	-0.0076	-0.76%
3	0.1656	0.0193	1.93%	33	0.1154	0.0052	0.52%	61	0.0927	0.0248	2.48%
4	0.1471	0.0117	1.17%	34	0.0915	0.0043	0.43%	62	0.0682	-0.0230	-2.30%
5	0.1562	0.0308	3.08%	35	0.1341	-0.0121	-1.21%	63	0.1267	0.0122	1.22%
6	0.1562	-0.0009	-0.09%	36	0.1564	0.0222	2.22%	64	0.0904	0.0075	0.75%
7	0.1646	-0.0021	-0.21%	37	0.1419	-0.0177	-1.77%	65	0.1162	-0.0422	-4.22%
8	0.1695	-0.0065	-0.65%	38	0.1339	0.0201	2.01%	66	0.1309	0.0101	1.01%
9	0.1604	-0.0060	-0.60%	39	0.1374	0.0238	2.38%	67	0.1225	0.0106	1.06%
10	0.1495	-0.0178	-1.78%	40	0.1396	-0.0241	-2.41%	68	0.1286	0.0095	0.95%
11	0.1446	-0.0104	-1.04%	41	0.1451	0.0374	3.74%	69	0.1361	0.0023	0.23%
12	0.1539	-0.0169	-1.69%	42	0.1475	-0.0077	-0.77%	70	0.1422	-0.0273	-2.73%
13	0.1557	-0.0159	-1.59%	43	0.1414	-0.0056	-0.56%	71	0.1489	-0.0199	-1.99%
14	0.1530	-0.0069	-0.69%	44	0.1427	0.0318	3.18%	72	0.1459	-0.0047	-0.47%
15	0.1526	-0.0040	-0.40%	45	0.1456	0.0163	1.63%	73	0.1533	0.0071	0.71%
16	0.1381	-0.0221	-2.21%	46	0.1337	0.0101	1.01%	74	0.1357	-0.0126	-1.26%
17	0.1376	-0.0154	-1.54%	47	0.1292	-0.0128	-1.28%	75	0.1421	-0.0110	-1.10%
18	0.1480	-0.0083	-0.83%	48	0.1348	-0.0025	-0.25%	76	0.1275	0.0037	0.37%
19	0.1381	-0.0223	-2.23%	49	0.1453	-0.0326	-3.26%	77	0.1367	0.0347	3.47%
20	0.1370	-0.0199	-1.99%	50	0.1455	0.0126	1.26%	78	0.1308	0.0335	3.35%
21	0.1324	-0.0033	-0.33%	51	0.1405	0.0100	1.00%	79	0.1396	0.0013	0.13%
22	0.1318	-0.0192	-1.92%	52	0.1411	0.0059	0.59%	80	0.1347	-0.0016	-0.16%
23	0.1447	-0.0154	-1.54%	53	0.1288	0.0137	1.37%	81	0.1272	0.0023	0.23%
24	0.1256	-0.0150	-1.50%	54	0.1312	0.0067	0.67%	82	0.1471	0.0356	3.56%
25	0.1292	-0.0040	-0.40%	55	0.1442	-0.0135	-1.35%	83	0.1279	0.0201	2.01%
26	0.1348	-0.0232	-2.32%	56	0.1394	0.0144	1.44%	84	0.1180	-0.0129	-1.29%
27	0.1304	-0.0033	-0.33%	57	0.1234	0.0391	3.91%	85	0.1228	0.0022	0.22%
28	0.1200	0.0044	0.44%	58	0.1410	0.0026	0.26%	86	0.1465	0.0281	2.81%
29	0.1377	-0.0190	-1.90%								
30	0.1479	-0.0170	-1.70%								

Regression Statistics

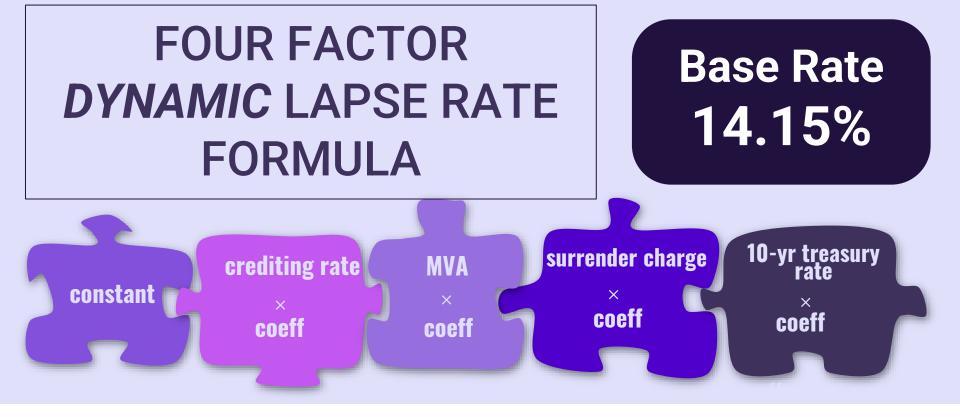
Table 2: Regression	Statistics
Multiple R	0.7444
R Square	0.5542
Adjusted R Square	0.5322
Standard Error	0.0179
Observations	86

R-Square Value: 0.5542

Low **P-Value** Indicates our Variables are significant in explaining Lapse Rate

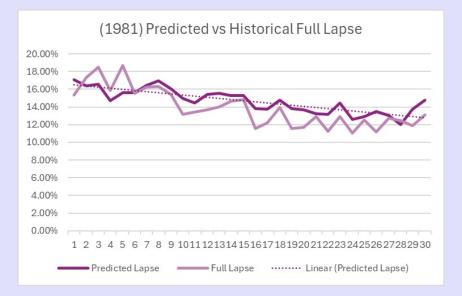
 Table 3: Regression Coefficients and P-values

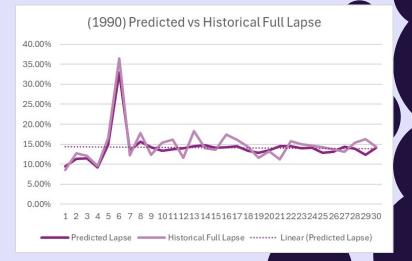
Variable	Coefficient	P-value
Intercept	0.1415	0.0000
MVA	-0.3125	0.0003
Crediting Rate	-0.9074	0.0023
10-Yr Treasury rate	0.6238	0.0000
Surrender charge	-0.7778	0.0000



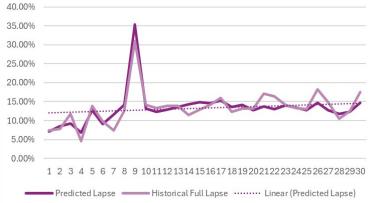
Lapse Rate = $\begin{cases} 0.142 + (-0.907 \cdot \text{C.R.}) + (-0.313 \cdot \text{MVA}) + (0.778 \cdot \text{S.C.}) + (0.624 \cdot 10\text{YR}), & \text{Policy Years} \neq 9\\ 0.142 + (-0.907 \cdot \text{C.R.}) + (-0.313 \cdot \text{MVA}) + (0.778 \cdot \text{S.C.}) + (0.624 \cdot 10\text{YR}) + 0.21, & \text{Policy Year} = 9 \end{cases}$

Accuracy Of Model





(1993) Predicted vs Historical Full Lapse



Accuracy Of Model - Adjusting for Std. Error

1993 Predicted Lapse Adjusted for Std. Error 35.00% 20.00% 10.00% 5.00% 0.00% Pred. 1993 + Std. Error Pred. + Std. Error 1993

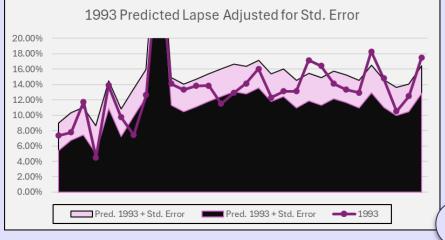
Zooming in on the graph

- Spikes in Historical Data
- Our Model attempts to predict these Spikes

Accounting for a Standard Error Margin (±1.79%)

Predictive Model shows **23/30** Historical Lapse Rates fall within Standard Error Margin

- Historical Lapse Rates taken from (1993)





Predicting Future Lapse Rates

New Product Specifications

Product Name	Pepper Back Fixed Annuity	
Target Launch Date	March 1975	
Surrender Charge Period	8 years	
Surrender Charge Schedule	9%, 8%, 7%, 6%, 5%, 4%, 3%, 2%	
MVA interest Rate Period	15 years	
Premium	Single	
Guaranteed Minimum Crediting Rate	0.70%	
First available Annuitization Date	15 years after issue	
Free Partial Withdrawal	15%	

Product Name	Pepper Back Fixed Annuity
Target Launch Date	January 2023
Surrender Charge Period	8 years
Surrender Charge Schedule	12%, 10%, 10%, 7%, 5%, 4%, 3%, 2%
MVA interest Rate Period	15 years
Premium	Single
Guaranteed Minimum Crediting Rate	1.50%
First available Annuitization Date	15 years after issue
Free Partial Withdrawal	10%





Projected Future Lapse Rates

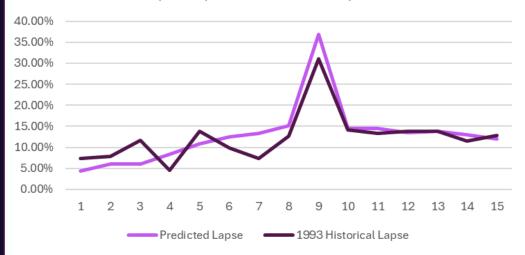
Table 5: Policy Data and Predicted Lapse						
Policy Year	MVA	Crediting Rate	10-Yr Treasury rate	Surrender charge	Predicted Lapse	
1	0.10%	4.50%	5.87%	12.00%	(4.37%)	
2	0.20%	4.50%	6.00%	10.00%	5.97%	
3	-0.45%	4.50%	5.80%	10.00%	6.05%	
4	-2.35%	4.70%	5.00%	7.00%	8.30%	
5	-3.04%	4.80%	6.35%	5.00%	10.82%	
6	-7.96%	4.80%	5.26%	4.00%	12.46%	
7	-8.15%	5.00%	5.65%	3.00%	13.35%	
8	-10.63%	5.00%	6.03%	2.00%	15.14%	
9	-9.98%	5.00%	5.02%	0.00%	36.87%	
10	-7.08%	5.20%	4.61%	0.00%	14.53%	
11	-8.03%	5.20%	4.01%	0.00%	14.45%	
12	-5.39%	5.50%	4.27%	0.00%	13.51%	
13	-6.34%	5.50%	4.29%	0.00%	13.82%	
14	-2.47%	5.50%	4.80%	0.00%	12.93%	
15	0.30%	5.50%	4.63%	0.00%	(11.96%)	

Trends in the Data

Three Trends observed in Lapse Rates

- 1. SLOW INCREASE (1-8yrs)
- 1. SPIKE IN LAPSE (9yr)
- SLIGHT DECREASE LAPSE (>9yr)

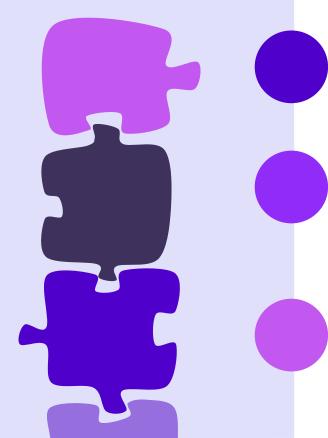
Future Predicted Lapse compared to (1993) Historical Full Lapse



Formula 4 Enhancement for GLB



Considering an Additional Factor



The Additional Factor:

Age of Policyholder

Effect on Lapsing

Increase in age seems to correlate to lower full lapse rates

- Health Concerns
- Outliving Savings

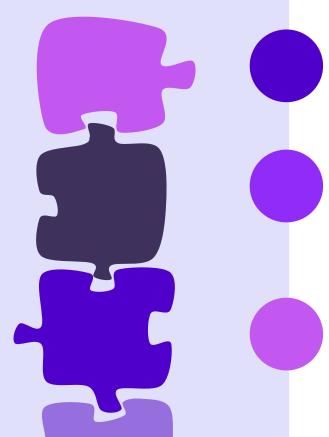
Guaranteed Living INCOME Benefit

Example

21 Year Old vs 55 Year Old

- Who benefits most from GLB?
- Who is more likely to Lapse?

Considering an Additional Factor



The Additional Factor:

Income Level of Policyholder

Effect on Lapsing

Higher income policyholders are less inclined to liquidate assets, such as an annuity, for emergency funds.

Annuities may offer a taxed-deferred growth:

- Attractive to high income policyholders
- Leading to Lower lapse rates.

Example

Wealthy CEO vs Retail Employee

- Who benefits most from GLB?
- Who is more likely to Lapse?

Implications / 5 Recommendations

Implications and Recommendations

Remove Outliers 04 05 from Dataset Compare Against **Other Regression** Use More Models Independent Variables in Calculations 03 01 Increase Size Consider of Dataset Competitive Products 02 22

