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Household's Life Insurance Demand a Multivariate Two Part Model

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## Outline



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## Objective

To understand characteristics of a household that drive life insurance demand with more sophisticated analytical techniques

Data

- 2004 Survey of Consumer Finance
- Build on the work of Lin and Grace (2007) by using covariates that they developed

#### Model features

- Two part Model
  - Frequency model Whether or not to have life insurance
  - Severity model The amount of insurance a household demands given they decide to have life insurance
- Multivariate Model
  - Term life insurance
  - Whole life insurance

#### Important finding







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#### Motivation



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#### • Life insurance demand literature:

- How much life insurance protection a household would seek given their economic and demographic structure (see Goldsmith (1983), Burnett and Palmer (1984) and Lin and Grace (2007))
- Tobit and OLS are widely applied.
- Term and Whole life insurance are substitutes.

#### Two part model

- Analogous to decision making process
- Allow for different explanatory variables for frequency and severity models respectively

#### Multivariate models

- Model two dependent variables simultaneously
- Examine the substitutes or complements effect of term and whole life insurance





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- Survey of Consumer Finances (SCF) data
  - A triennial survey of U.S. families conducted by the Federal Reserve
  - About 4000 household level ("primary economic unit") observations during each survey period
  - A probability sample of the U.S. population
  - Extensive demographic and economic characteristics of the households as well as their behavioral aspects such as the motive to leave a bequest
  - Limitations
    - Life insurance information is aggregate.
    - No information about when the life insurance was purchased.





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- Dependent variable
  - Frequency Part (2150 observations)
    - Term life insurance indicator (65.86%)
    - Whole life insurance indicator (33.40%) \*19.72% have both types of insurance
  - Severity Part (1710 observations—Life insurance purchasers subsample)
    - Face amount of term life insurance (Median \$270,000)
    - Net Amount at Risk (NAR) of whole life insurance (Median \$202,500) \*Positively correlated





4e+05 6e+06 80+08

Whole







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Statistical Models Conclusion The End! We build on the work of Lin and Grace (2007) by using covariates that they developed.

#### • Financial Vulnerability Index (IMPACT)

Measures the adverse financial impact in terms of living standard decline upon the death of one member of the household on the rest







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#### Assets







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#### Assets

Cash and cash equivalents, mutual funds, stocks, bonds, annuities, individual retirement accounts, real estate, and other assets

Debts







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#### Assets

- Debts
- Age







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#### Assets

- Debts
- Age
- Education







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#### Assets

- Debts
- Age
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- Income







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Assets

- Debts
- Age
- Education
- Income
- Bequests (48.8%), Obligations (58.9%), and Inheritance







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Table 1. Summary Statistics									
Variable	Minimum	25th Percentile	Median	75th Percentile	Maximum				
FACETerm	0.8	100	270	1,000	150,000				
NAR	0.66	60.25	202.5	900	45,000				
CASHEQV	0	3	17	98	32,628				
FUND	0	0	0	20	57,500				
STOCK	0	0	0	50	200,000				
BOND	0	0	0	1	100,000				
RETIREMENT	0	0	52	272	35,000				
ANNUITY	0	0	0	0	200,000				
REALESTATE	0	127	350	1,294	194,380				
OTHASSETS	0	15	31	66	97,203				
DEBT	0	13	110	286	121,686				
INHERITANCEExp	0	0	0	0	906,060				
SALARY1	0	29	60	163	80,112				
SALARY2	0	0	13	40	2,700				
IMPACT	0	0.049	0.113	0.340	1265.02				
AGE	21	39.5	47.5	54.5	64				
EDUCATION1	1	12	16	17	17				
EDUCATION2	0	12	15	16	17				

\*All the monetary variables are in thousands.



\* Assets, debts, income and inheritance variables are logarithm transformed and indicator variables for zero values are added for these variables.





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#### Two part model

- $\mathbf{N}_i = (N_{i1}, N_{i2})$ 
  - N<sub>i1</sub> indicator for whether household *i* purchases term life insurance
  - N<sub>i1</sub> indicator for whether household *i* purchases whole life insurance
- $\mathbf{Y}_i = (Y_{i1}, Y_{i2})$ 
  - $Y_{i1}$  the face amount of term life insurance demanded by household i
  - Y<sub>i2</sub> the net amount at risk (NAR) of whole life insurance demanded by household i
- Decompose  $(\mathbf{Y}_i)$  into frequency and severity components

 $f(\mathbf{Y}_i) = f(\mathbf{N}_i) \times f(\mathbf{Y}_i | \mathbf{N}_i).$ 

- Frequency model  $f(\mathbf{N}_i)$ : Bivariate probit regression model
- Severity model  $f(\mathbf{Y}_i|\mathbf{N}_i>0)$ : Generalized linear model with a Gaussian copulas







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#### Two part model

- $\mathbf{N}_i = (N_{i1}, N_{i2})$ 
  - N<sub>i1</sub> indicator for whether household *i* purchases term life insurance
  - $N_{i1}$  indicator for whether household *i* purchases whole life insurance

• 
$$\mathbf{Y}_i = (Y_{i1}, Y_{i2})$$

- $Y_{i1}$  the face amount of term life insurance demanded by household *i*
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- Decompose (Y<sub>i</sub>) into frequency and severity components

 $f(\mathbf{Y}_i) = f(\mathbf{N}_i) \times f(\mathbf{Y}_i | \mathbf{N}_i).$ 

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Bivariate probit regression

- A bivariate probit regression model assumes the joint distribution of the bivariate binary choices is a standard bivariate normal distribution with a correlation coefficient  $\rho$  (see Ashford and Sowden (1970) and Meng and Schmidt (1985)).
- The log-likelihood of the *i*th observation is

$$\begin{split} l_i &= N_{i1}N_{i2}\ln F(\mathbf{x}'_i\beta_1, \mathbf{x}'_i\beta_2; \rho) \\ &+ N_{i1}(1 - N_{i2})\ln[\Phi(\mathbf{x}'_i\beta_1) - F(\mathbf{x}'_i\beta_1, \mathbf{x}'_i\beta_2; \rho)] \\ &+ (1 - N_{i1})N_{i2}\ln[\Phi(\mathbf{x}'_i\beta_2) - F(\mathbf{x}'_i\beta_1, \mathbf{x}'_i\beta_2; \rho)] \\ &+ (1 - N_{i1})(1 - N_{i2})\ln[1 - \Phi(\mathbf{x}'_i\beta_1) - \Phi(\mathbf{x}'_i\beta_2) + F(\mathbf{x}'_i\beta_1, \mathbf{x}'_i\beta_2; \rho)] \end{split}$$

where  $F(\cdot)$  is the cumulative distribution function of the standard bivariate normal distribution with correlation  $\rho$ .





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## Empirical result - Bivariate Probit Regression



Whole Insurance (718) Term Insurance (1416) Parameter Estimate t-ratio Estimate t-ratio 0.6669 0.7241 -0.9387 -0.9923 Intercept Financial Vulnerability Index (IMPACT) 0 1696 2.6724 \*\*\* 0.0558 0.9688 Indicator for IMPACT > 4 -0 4730 -1.9327\* -0 1623 -0.7268 Log (1+ cash and cash equivalent) 0.0304 1 5934 0 0424 2 1641 \*\* Indicator for Izero cash and cash equivalent -0.2411 -1.03590.2903 1.0687 Log (1+stock) -0.0522 -2 5445 \*\* -0.0369 -1 8554 Indicator for zero stock -0 4247 -1.8536 \* -0.4773 -2.1600 \*\* \*\* \*\* Log (1+ bond) -0.0402 -2.4054 -0.0373 -2.3348 -0 4401 -2.6572 \*\*\* -0 5471 -3.5246 \*\*\* Indicator for zero bond 0.0309 1.2265 -0.0437 -1.7953Loa (1+ fund) \*\* Indicator for zero fund 0.3445 1 1329 -0 6971 -2 3807 Log (1+ annuity) -0.0724 -1 8533 0 0229 0.6204 Indicator for zero annuity -0.8718 -1 7882 0.0488 0 1072 Log (1+ retirement) 0.0244 1.0716 -0.0319-1.4329Indicator for zero retirement -0.1217 -0.4814 -0.3881 -1.5228 \*\*\* 2.2573 Log (1+ real estate) -0.2092-5.3364 0.0901 \*\* Indicator for zero real estate -2.5806 -5.6841 \*\*\* 1.7391 \* 0.8182 Log (1+ other assets) 0.0376 1.3837 0.0114 0.4211 0.3720 1.1793 Indicator for zero other assets -0.3394-1.0141Loa (1 + debt)0.0563 2.3066 \*\* 0.0046 0.1822 Indicator for zero debt 0.1954 0.6560 -0.0019-0.0059\*\* Average age of the couple 0.0575 2.2400 0.0035 0.1229 Squared average age of the couple -0.0006 -2.1053\*\* 0.0002 0.6699 0.0577 3,4698 \*\*\* Education level of the resondent -0.0172-0.9852 Education level of the spouse 0.0212 1.3865 0.0141 0.8665 \*\* Log (1+ salary of the respondent) 0.0185 2.2804 0.0040 0.4896 \*\* Log (1+ salary of the spouse) 0.0140 2.3231 0.0148 2.4428 Log (1+ sizable inheritance expected) -0.0234-0.6409-0.0107-0.2944Indicator for zero inheritance expected -0.3234-0.6867-0.1723-0.3676Indicator for the desire to leave a bequest -0.0029-0.04220.1135 1.6806 Indicator for foreseeable major financial obligation 0.0748 1.2013 -0.0005-0.0082Rho -0.2849 -7.6676 \*\*\* Significant at 1% level

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\*\* Significant at 5% level \* Significant at 10% level





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	Term Insurance (1416)			Whole Insurance (718)		
Parameter	Estimate	t-ratio		Estimate	t-ratio	
Intercept	0.6669	0.7241		-0.9387	-0.9923	_
Financial Vulnerability Index	0.1696	2.6724	***	0.0558	0.9688	







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• Financial Vulnerability Index only has impact on the frequency of term life insurance demand.

	Term Insurance (1416)			Whole Insurance (718)		
Parameter	Estimate	t-ratio		Estimate	t-ratio	
Intercept	0.6669	0.7241		-0.9387	-0.9923	_
Financial Vulnerability Index	0.1696	2.6724	***	0.0558	0.9688	

## • In general, the more assets a household has, the less likely that the household demands life insurance.

Log (1+ cash and cash equivalent)	0.0304	1.5934		0.0424	2.1641	**
Indicator for zero cash	-0.2411	-1.0359		0.2903	1.0687	
Log (1+stock)	-0.0522	-2.5445	**	-0.0369	-1.8554	
Indicator for zero stock	-0.4247	-1.8536	*	-0.4773	-2.1600	**
Log (1+ bond)	-0.0402	-2.4054	**	-0.0373	-2.3348	**
Indicator for zero bond	-0.4401	-2.6572	***	-0.5471	-3.5246	***
Log (1+ fund)	0.0309	1.2265		-0.0437	-1.7953	*
Indicator for zero fund	0.3445	1.1329		-0.6971	-2.3807	**
Log (1+ annuity)	-0.0724	-1.8533		0.0229	0.6204	
Indicator for zero annuity	-0.8718	-1.7882		0.0488	0.1072	
Log (1+ retirement)	0.0244	1.0716		-0.0319	-1.4329	
Indicator for zero retirement	-0.1217	-0.4814		-0.3881	-1.5228	
Log (1+ real estate)	-0.2092	-5.3364	***	0.0901	2.2573	**
Indicator for zero real estate	-2.5806	-5.6841	***	0.8182	1.7391	*
Log (1+ other assets)	0.0376	1.3837		0.0114	0.4211	
Indicator for zero other assets	0.3720	1.1793		-0.3394	-1.0141	





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	lerm Insurance (1416)			Whole Insurance (718)		
Parameter	Estimate	t-ratio		Estimate	t-ratio	
Log (1 + debt)	0.0563	2.3066	**	0.0046	0.1822	
Indicator for zero debt	0.1954	0.6560		-0.0019	-0.0059	
Average age of the couple	0.0575	2.2400	**	0.0035	0.1229	
Squared average age of the couple	-0.0006	-2.1053	**	0.0002	0.6699	
Education level of the resondent	0.0577	3.4698	***	-0.0172	-0.9852	
Education level of the spouse	0.0212	1.3865		0.0141	0.8665	
Log (1+ salary of the respondent)	0.0185	2.2804	**	0.0040	0.4896	
Log (1+ salary of the spouse)	0.0140	2.3231	**	0.0148	2.4428	**
Log (1+ sizable inheritance expected)	-0.0234	-0.6409		-0.0107	-0.2944	
Indicator for zero inheritance expected	-0.3234	-0.6867		-0.1723	-0.3676	
Indicator for the desire to leave a bequest	-0.0029	-0.0422		0.1135	1.6806	*
Indicator for foreseeable major financial obligation	0.0748	1.2013		-0.0005	-0.0082	
Rho	-0.2849	-7.6676	***			

#### Finding

The correlation between the likelihood of term life insurance ownership and whole life insurance ownership is significantly negative after controlling for the covariates.







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Rho	-0.2849	-7.6676	***			

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### Severity Model



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$$f(y_i, \theta_i) = \exp(\frac{y_i \theta_i - b(\theta_i)}{\phi_i} + S(y_i, \phi_i))$$

$$E(y_i) = b'(\theta_i), \quad Var(y_i) = \phi_i b''(\theta_i)$$

A link function  $g(\cdot)$  links the covariates  $\mathbf{x}_i$  to the response mean such that  $g(b'(\theta_i)) = \mathbf{x}'_i \boldsymbol{\beta}$ .





## Severity Model



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• Generalized Linear Model (GLM) (see McCullagh and Nelder (1989)) Exponential family

$$f(y_i, \theta_i) = \exp(\frac{y_i \theta_i - b(\theta_i)}{\phi_i} + S(y_i, \phi_i))$$

$$E(y_i) = b'(\theta_i), \quad Var(y_i) = \phi_i b''(\theta_i)$$

A link function  $g(\cdot)$  links the covariates  $\mathbf{x}_i$  to the response mean such that  $g(b'(\theta_i)) = \mathbf{x}'_i \boldsymbol{\beta}$ .

• Copulas (see Frees and Wang (2005))

 $C[F_{i1}(y_{i1}), F_{i2}(y_{i2})] = F_i(y_{i1}, y_{i2})$ 

The log-likelihood of the *i*th household's life insurance demand given they purchase life insurance is

 $l_i = \ln f(y_{i1}, \theta_{i1}) + \ln f(y_{i2}, \theta_{i2}) + \ln c(F_{i1}(y_{i1}), F_{i2}(y_{i2}))$ 





## Severity Model



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• Generalized Linear Model (GLM) (see McCullagh and Nelder (1989)) Exponential family

$$f(y_i, \theta_i) = \exp(\frac{y_i \theta_i - b(\theta_i)}{\phi_i} + S(y_i, \phi_i))$$

$$E(y_i) = b'(\theta_i), \quad Var(y_i) = \phi_i b''(\theta_i)$$

A link function  $g(\cdot)$  links the covariates  $\mathbf{x}_i$  to the response mean such that  $g(b'(\theta_i)) = \mathbf{x}'_i \boldsymbol{\beta}$ .

• Copulas (see Frees and Wang (2005))

 $C[F_{i1}(y_{i1}), F_{i2}(y_{i2})] = F_i(y_{i1}, y_{i2})$ 

The log-likelihood of the *i*th household's life insurance demand given they purchase life insurance is

 $l_i = \ln f(y_{i1}, \theta_{i1}) + \ln f(y_{i2}, \theta_{i2}) + \ln c(F_{i1}(y_{i1}), F_{i2}(y_{i2}))$ 

• Incorporating a parametric distribution function (e.g. a Gamma distribution function with a log link function) and a parametric copula function (e.g. a Gaussian copula) to the above likelihood function, we can get an expression for the log-likelihood of the *i*th observation.





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	Face Value	of Term Insu	rance	NAR of Whole Insurance		
Parameter	Estimate	t-ratio		Estimate	t-ratio	
Intercept	0.6694	0.9030		0.1299	0.1178	
Financial Vulnerability Index (IMPACT)	0.1046	1.7907	*	0.2533	2.7330	***
Indicator for IMPACT ≥ 4	-0.4636	-1.9698	*	-0.8145	-2.3842	**
Log (1+ cash and cash equivalent)	0.1706	8.5447	***	0.0237	0.8551	
Indicator for zero cash and cash equivalent	1.1962	3.8591	***	-1.1153	-2.0780	**
Log (1+stock)	0.0444	2.2057	**	0.0750	2.5311	**
Indicator for zero stock	0.4152	1.8819	*	1.0006	2.9940	***
Log (1+ bond)	0.0635	3.5879	***	0.0737	3.2795	***
Indicator for zero bond	0.4571	2.8738	**	0.6249	2.7952	***
Log (1+ fund)	0.0302	1.2180		0.0557	1.5422	
Indicator for zero fund	0.3965	1.3562		0.9352	2.1561	**
Log (1+ annuity)	0.0161	0.4580		0.0668	1.1762	
Indicator for zero annuity	0.2572	0.6226		0.6278	0.8866	
Log (1+ retirement)	0.0232	1.0801		0.0914	2.8581	***
Indicator for zero retirement	0.1753	0.7126		0.7532	1.9538	*
Log (1+ real estate)	0.2014	5.7790	***	0.3262	5.4281	***
Indicator for zero real estate	2.1948	5.4352	***	3.5057	4.6320	***
Log (1+ other assets)	0.1736	5.9393	***	0.1963	4.9573	***
Indicator for zero other assets	1.8250	5.2204	***	1.2862	2.3854	**
Log (1 + debt)	0.1289	5.2627	***	0.0400	0.9902	
Indicator for zero debt	1.0537	3.3861	***	0.8675	1.6730	*
Average age of the couple	0.0227	2.6742	***	0.0223	1.8322	*
Squared average age of the couple	-0.0005	-5.6999	***	-0.0006	-5.1411	***
Education level of the resondent	0.0458	2.6043	**	0.0057	0.2035	
Education level of the spouse	0.0237	1.3487		0.0560	2.0745	**
Log (1+ salary of the respondent)	0.0174	1.9938	*	0.0122	0.9756	
Log (1+ salary of the spouse)	-0.0244	-3.9509	***	-0.0280	-2.9078	***
Log (1+ sizable inheritance expected)	0.1634	4.5040	***	0.0406	0.6960	
Indicator for zero inheritance expected	1.9633	4.2608	***	0.5633	0.7446	
Indicator for the desire to leave a bequest	0.2058	3.0970	***	0.6351	5.7582	***
Indicator for foreseeable major financial obligation	0.0871	1.3906		0.1625	1.7100	*
Alpha	0.9131	28.4956	***	0.7460	30.6565	***
Rho	0.0990	1.9636	*			

\*\*\* Significant at 1% level

\*\* Significant at 5% level

\* Significant at 10% level





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## • The higher the financial vulnerability index, the more life insurance protection a household seeks for.

	Face Value of Term Insurance			NAR of Whole Insurance		
Parameter	Estimate	<i>t</i> -ratio		Estimate	t-ratio	
Intercept	0.6694	0.9030		0.1299	0.1178	
Financial Vulnerability Index	0.1046	1.7907	*	0.2533	2.7330	***







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• The higher the financial vulnerability index, the more life insurance protection a household seeks for.

	Face Value	of Term Insura	ince NAR of	Whole Insurance
Parameter	Estimate	t-ratio	Estimate	e <i>t</i> -ratio
Intercept	0.6694	0.9030	0.1299	0.1178
Financial Vulnerability Index	0.1046	1.7907 *	0.2533	2.7330 ***

• The more assets a household has, the more life insurance they demand

Log (1+ cash and cash equivalent)	0.1706	8.5447	***	0.0237	0.8551	
Indicator for zero cash	1.1962	3.8591	***	-1.1153	-2.0780	**
Log (1+stock)	0.0444	2.2057	**	0.0750	2.5311	**
Indicator for zero stock	0.4152	1.8819	*	1.0006	2.9940	***
Log (1+ bond)	0.0635	3.5879	***	0.0737	3.2795	***
Indicator for zero bond	0.4571	2.8738	**	0.6249	2.7952	***
Log (1+ fund)	0.0302	1.2180		0.0557	1.5422	
Indicator for zero fund	0.3965	1.3562		0.9352	2.1561	**
Log (1+ annuity)	0.0161	0.4580		0.0668	1.1762	
Indicator for zero annuity	0.2572	0.6226		0.6278	0.8866	
Log (1+ retirement)	0.0232	1.0801		0.0914	2.8581	***
Indicator for zero retirement	0.1753	0.7126		0.7532	1.9538	*
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Log (1+ sizable inheritance expected)	0.1634	4.5040	***	0.0406	0.6960			
Indicator for zero inheritance expected	1.9633	4.2608	***	0.5633	0.7446			
Indicator for the desire to leave a	0.2058	3.0970	***	0.6351	5.7582			
bequest								
Indicator for foreseeable major financial	0.0871	1.3906		0.1625	1.7100			
obligation								
Alpha	0.9131	28.4956	***	0.7460	30.6565			
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#### Finding

The correlation between the amount of term and whole life insurance demand is positive and significant.







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### Conclusion



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#### Contribution

- Improve the understanding of a household's life insurance demand
- Insurance company can develop marketing strategies accordingly
- The demand of term and whole life insurance are substitutes in frequency and complements in severity

#### Further research

The ultimate goal of this study is to project national life insurance demand. Further research will focus on out-of-sample validation and extrapolation to the national population with the proper survey sampling method. We will also explore the demand of life insurance for single person households.





## Conclusion



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#### Thanks



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