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# **An Empirical Study on Optimal Reinsurance about crop insurance in China**

**—Based on data from Inner Mongolia, Jilin and Liaoning**

Yunbo Wang

AnHua Agricultural Insurance  
Company, Beijing, China.

# Outline

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- Motivation
- Model and data
- Results
- Conclusion

# Motivation

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- Agricultural Insurance
  - Adverse weather, flood, draught, hail etc.
  - Diseases of livestock.
  - Resembles catastrophe insurance.
  - Reinsurance is a major risk transfer instrument.
- Agricultural Insurance in China
  - The world's second-largest agricultural insurance market.
  - Crop insurance accounts for over 90%.
  - Government subsidy since 2007.

# Motivation

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- Reinsurance of crop insurance during 2008-2012
  - Cumulated original premiums: 62.68 billion RMB
  - Cumulated ceded premiums: 9.50 billion RMB.
  - Cumulated losses recovered from reinsurers: 4.57 billion RMB.
  - Cumulated expenses recovered from reinsurers :2.33 billion RMB.
  - Cumulated net ceded profit:2.13 billion RMB.
  - Ceded profit accounts for 3.4% of original premiums.
  - Ceded profit accounts for 22.42% of ceded premiums.

# Motivation

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- Considering the high cost of reinsurance for crop insurance:
  - How to determine the optimal reinsurance arrangement?
  - And what is the appropriate cost for it?

# Model and data

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- Empirical model: 
$$\begin{cases} \min_f \rho(\widehat{x}, f) \\ s. t. 0 \leq f_i \leq x_i \\ \widehat{\pi}(f) \leq \pi \end{cases}$$
  - $x = (x_1, x_2, \dots, x_n)$ , n-dimension samples, can be collected directly, or generated randomly.
  - for loss  $x_i$ , the insurer cedes  $f_i$  to a reinsurer.
- Empirical models can be transformed into Second Order Conic Programming problems (Weng, 2009)
- CVX MATLAB toolbox (Grant et al., 2013) can be used to solve this problem

# Model and data

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- Advantages of empirical model (Weng,2009)
  - Simple, intuitive, practical.
  - It exploits directly the observed data.
  - Applies to a number of premium principles.
- Premium principles
  - Expectation principle: safety loading is unrelated to the variation of the risk:

$$\pi(f) = (1 + \beta)E[f(X)]$$

- Standard deviation principle: safety loading is positively related to the variation of the risk:

$$\pi(f) = E[f(X)] + \beta\sqrt{Var[f(X)]}$$

# Model and data

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- Risk measures
  - $VaR_X(\alpha) = \inf\{x: \Pr(X > x) \leq \alpha\}$
  - $CTE_X(\alpha) = E[X|X > VaR_X(\alpha)]$
  - CTE is a better risk measure compared to VaR (Cai et al., 2008).
- In this study
  - To minimize the risk measure under the reinsurance premium budget constraint.



# Model and data

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- Other assumptions
  - Risk tolerance level  $\alpha$ : 2%, 5%, 10%, 20%.
  - Reinsurance premiums budget  $\pi$ : 3%, **3.75%**, 4%, 5% of total premiums.
  - Safety loading coefficient  $\beta$ : 20%, 30%, 40%, 50%.
  - Samples size: 1000.

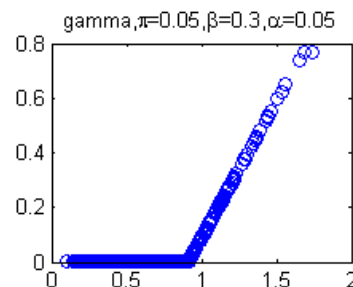
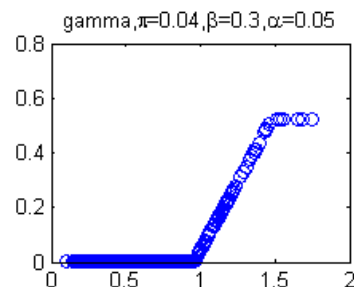
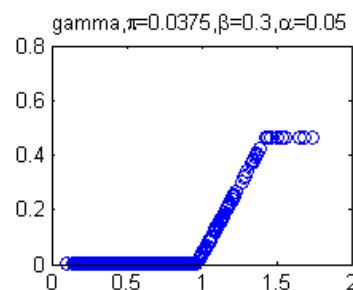
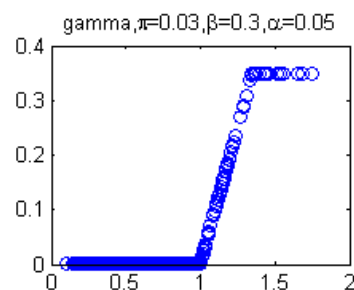
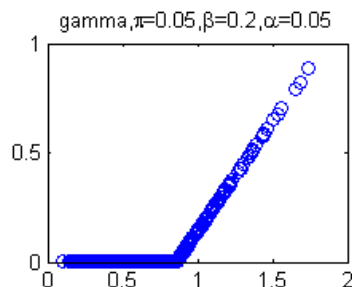
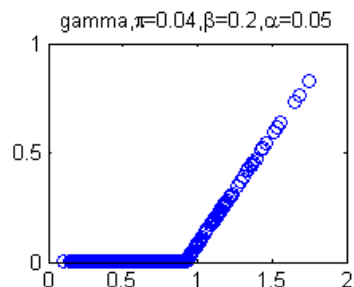
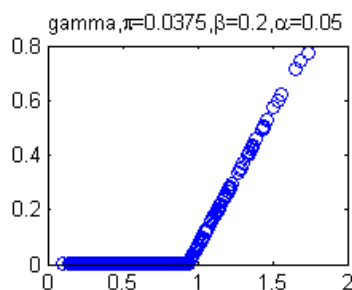
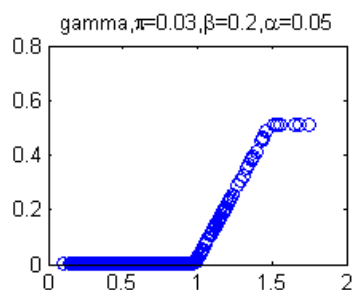
# Model and data

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- Loss ratio distribution(AHCRES)
  - Inner Mongolia: Gamma Distribution,  $a = 5.74987$  and  $b = 0.105108$ .
  - Liaoning: Gamma Distribution,  $a = 4.1405$  and  $b = 0.1796$ .
  - Jilin: Generalized Extreme Value Distribution,  $a = 0.105086$ ,  $b = 0.173391$  and  $c = 0.53431$ .
  - Gamma Distribution:  $f(x) = \frac{1}{b^a \Gamma(a)} x^{a-1} e^{-\frac{x}{b}}$
  - Generalized Extreme Value Distribution:  
$$f(x) = \left(\frac{1}{b}\right) \exp\left(-\left(1 + a\frac{(x-c)}{b}\right)^{-\frac{1}{a}}\right) \left(1 + a\frac{(x-c)}{b}\right)^{-1-\frac{1}{a}}$$

# Results- Inner Mongolia

Fixed  $\alpha$ , varied  $\pi$  and  $\beta$  – scatter plots of  $\{(x_i, f_i)\}$



- When  $\pi$  is small, limited stop loss or its variants are optimal, with the increase of  $\pi$ , the optimal form may become stop loss without a limit.
- With the increase of  $\beta$ , the optimal form may become limited stop loss from stop loss.
- Small  $\pi$  and large  $\beta$  play similar roles in the process of optimal reinsurance design.
- The reality is, fairly large  $\beta$  and small  $\pi$  (strict reinsurance premiums budget).

# Robustness test- Inner Mongolia

- For  $\{(x_i, f_i)\}$ , fit  $f(x) = \min\{c(x - d)_+, m\}$ , we replicate the random samples 500 times independently to obtain 500 independent estimates of  $\hat{c}$  and  $\hat{d}$  using  $\varepsilon = 0.001$ .
- For one simulation, if  $|f(x_i) - f_i| \leq \varepsilon$  for  $i=1,2,\dots,1000$ , one admission is obtained, admissibility=times of admission/500.

$\beta = 0.3$ , samples size=1000,  $\varepsilon = 0.001$

$\alpha$	$\pi$	admissibility	$\bar{c}$	$\bar{d}$	$\overline{d + m}$
5%	3.00%	100.00%	1.00	88.69%	107.81%
	3.75%	100.00%	1.00	85.03%	107.54%
	4.00%	100.00%	1.00	84.16%	107.93%
10%	3.00%	100.00%	1.00	81.67%	94.42%
	3.75%	100.00%	1.00	79.41%	94.94%
	4.00%	100.00%	1.00	78.80%	95.28%
20%	3.00%	100.00%	1.00	72.14%	80.42%
	3.75%	100.00%	1.00	70.11%	80.10%
	4.00%	100.00%	1.00	69.80%	80.49%

# Results- Inner Mongolia

## Optimal reinsurance arrangement of Inner Mongolia under different constraints

$\beta$		30%		40%		50%	
$\alpha$	$\pi$	$d$	$d + m$	$d$	$d + m$	$d$	$d + m$
5%	3.00%	89%	108%	91%	108%	93%	109%
	3.75%	85%	108%	88%	109%	90%	108%
	4.00%	84%	108%	87%	109%	90%	109%
10%	3.00%	82%	94%	83%	94%	84%	95%
	3.75%	79%	95%	81%	95%	83%	96%
	4.00%	79%	95%	80%	95%	82%	96%
20%	3.00%	72%	80%	73%	80%	73%	80%
	3.75%	70%	80%	71%	81%	72%	81%
	4.00%	70%	80%	71%	81%	72%	81%

# Conclusion

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- When the primary insurer's loss function, the reinsurance premium calculation principle, risk measure  $\rho$  are given,  $\alpha, \beta, \pi$  all affect the optimal reinsurance design.
- When strict constraint on reinsurance premiums budget  $\pi$  are implemented (which is often the reality), Limited Stop Loss Reinsurance is optimal.
- Reinsurance premium calculation principle and safety loading coefficient of reinsurers play important roles in the optimal reinsurance decision-making process.

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Thank you for your attention