



# Crop Insurance Pricing using Probability Distributions and Copula Models

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# Overview

Crop Insurance  
Pricing using  
Probability  
Distributions  
and Copula  
Models

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- From 2007 to 2016, the federal crop insurance title had the second-largest outlays in the farm bill after nutrition. The total net cost of the program for crop years 2007-2016 was about \$72 billion. (Rosa, Isabel)



# Importance

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## Why is Crop Insurance Important?

- Financially protects farmers from loss of crop and revenue
- Growing population: more people = more demand for food
- All consumers benefit from a secure agriculture industry



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## Causes of Loss

- Weather: rain, temperature, length of growing season
- Bacteria, viruses, pests
- Implies that insurance amounts may differ between region, and commodity type



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## What is an Indemnity?

- A safeguard against loss
- In terms of US crop insurance: a payment made when crop yields underperform
- USDA RMA: authorizes 15 private companies to provide insurance



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## The Goal

- To predict indemnity amounts for specific farms based on region, and commodity grown



# Data

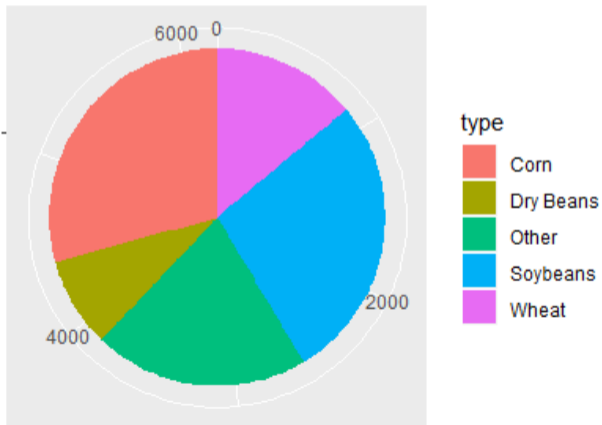
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## Pie Chart for Commodity





# Data

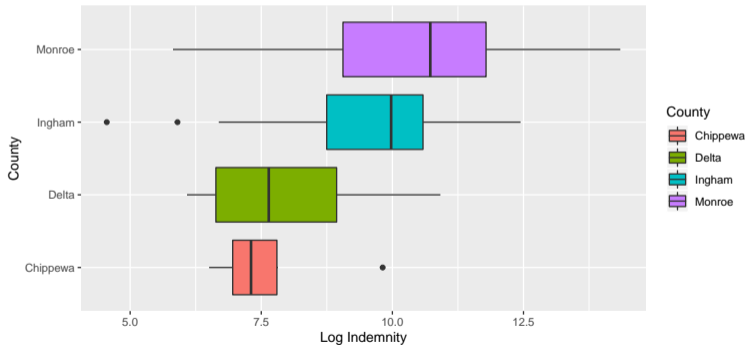
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## Log indemnities by county





# Data

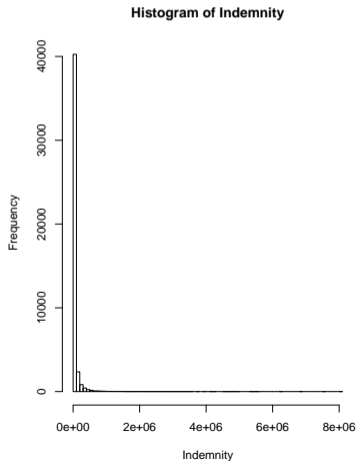
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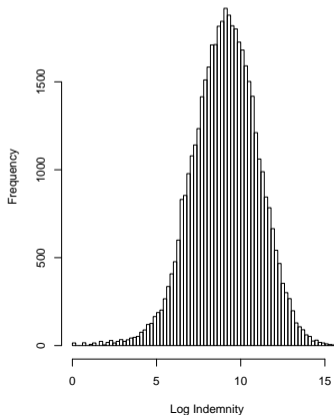
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## Histogram of Indemnity



**Histogram of Log Indemnity**





# Model

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## **What do we want to do?**

Determine the premium



## Gamma distribution

The density for the Gamma distribution is:

$$f(x, \theta, \alpha) = \frac{\left(\frac{x}{\theta}\right)^\alpha}{x\Gamma(\alpha)} e^{(-\frac{x}{\theta})}$$

Statistic indicators

$$E(x) = \alpha\theta \quad \text{Var}(x) = \alpha(\theta)^2$$



## Pareto Distribution

The density for Pareto Distribution:

$$f(x, \theta, \alpha) = \frac{\alpha \theta^\alpha}{(x + \theta)^{\alpha+1}}$$

Statistic indicators

$$E(x) = \frac{\theta}{\alpha - 1} \quad \text{Var}(X) = \frac{\alpha \theta^2}{(\alpha - 1)^2 (\alpha - 2)}$$



## Weibull Distribution

The density for Weibull Distribution:

$$f(x, \theta, \alpha) = \left(\frac{\alpha}{\theta}\right) \left(\frac{x}{\alpha}\right)^{\alpha-1} e^{-\left(\frac{x}{\theta}\right)^\alpha}$$

Statistic indicators

$$E(x) = \theta \Gamma\left(1 + \frac{1}{\alpha}\right) \quad \text{Var}(X) = \theta^2 \left( \Gamma\left(1 + \frac{2}{\alpha}\right) - \left[ \Gamma\left(1 + \frac{1}{\alpha}\right) \right]^2 \right)$$



# Copula Model

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## Sklar's theorem

Sklar's theorem states that an  $m$ -dimensional copula is a function  $C$  from the unit  $m$ -cube  $[0, 1]^m$  to the unit interval  $[0, 1]$  which satisfies the following conditions:

- (1)  $C(1, \dots, 1, a_n, 1, \dots, 1) = a_n$  for every  $n \leq m$  and all  $a_n$  in  $[0, 1]$ ;
- (2)  $C(a_1, \dots, a_m) = 0$  if  $a_n = 0$  for any  $n \leq m$ ;
- (3)  $C$  is  $m$ -increasing



# Practical implications

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## Dependence parameter

Copulas allow researchers to study the dependence between two separate but related issues.

$$F(\mathbf{y}_1, \dots, \mathbf{y}_m) = C(F_1(y_1), \dots, F_m(y_m); \theta)$$

where  $\theta$  is a parameter which measures dependence between the marginals.



# Some Common Bivariate Copulas

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## Product copula

The simplest copula, the product copula, has the form

$$C(\mathbf{u}_1, \mathbf{u}_2) = \mathbf{u}_1 \mathbf{u}_2,$$

where  $\mathbf{u}_1$  and  $\mathbf{u}_2$  take values in the unit interval of the real line.



# Some Common Bivariate Copulas

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## Farlie-Gumbel-Morgenstern copula

The FGM copula takes the form

$$C(u_1, u_2; \theta) = u_1 u_2 (1 + \theta(1 - u_1)(1 - u_2))$$

The FGM copula is a perturbation of the product copula, if the dependence parameter  $\theta$  equals zero, then the FGM collapses to independence.



# Some Common Multivariate Copulas

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## Parameter matrix

Assume the response vector is  $y$ . Then we imagine there is an  $d \times d$  dimensional parameter matrix  $\Sigma$ , such that

$$\sigma_{i,j} = \begin{cases} 1 & \text{if } i = j \\ \rho & \text{if } s_i = s_j \\ 0 & \text{otherwise} \end{cases}$$

where  $s_i$  is the state code for the  $i$ th county, and  $s_j$  is the state code for the  $j$ th county.



# Some Common Multivariate Copulas

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## Gaussian copula

The multivariate Gaussian copula of dimension  $d$  is defined by

$$C(\mathbf{u}) = \Phi_d(\Phi^{-1}(u_1), \dots, \Phi^{-1}(u_d), \Sigma)$$

where  $\Phi^{-1}$  is the inverse cumulative distribution function of the standard normal, and  $\Phi_d$  is the joint cumulative distribution function of a multivariate normal distribution with mean vector zero, and covariance matrix  $\Sigma$

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# Thank you!