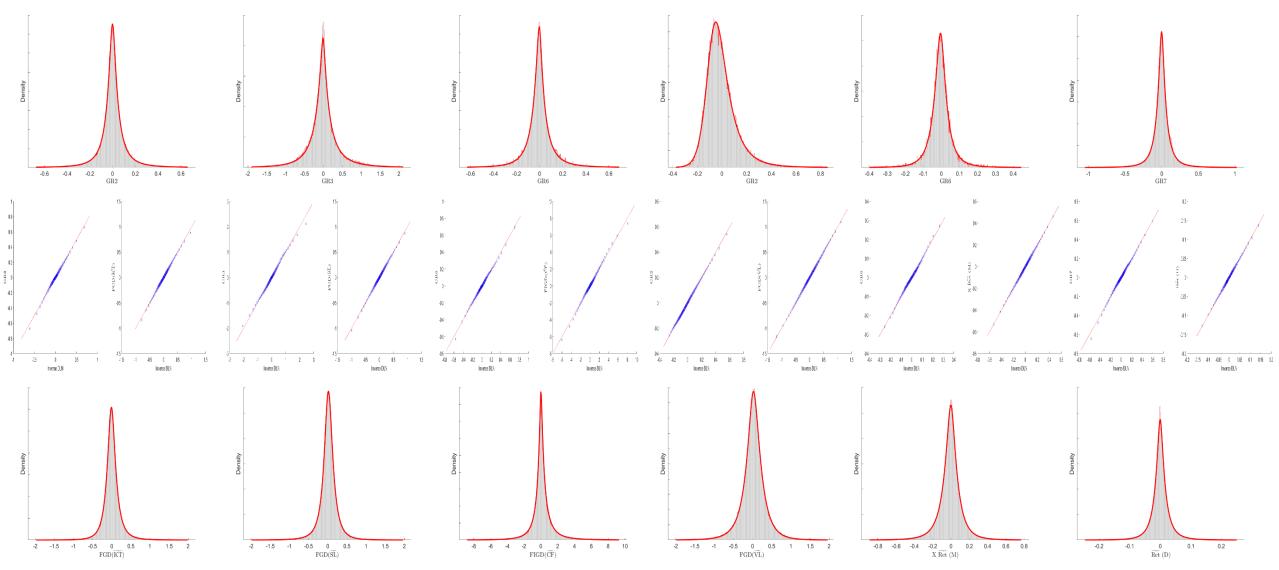
Growth and Differences of Log-Normals

Robert Parham



What do these all have in common?



Grey: data dist., red line: fitted diff-log-N, middle: q-q plots; Why, and how does it help us write models? (prod func)

Normal

What's so normal about **The Normal**?

$$Y = \lim_{M \to \infty} \frac{1}{M} \sum_{i=1}^{M} X_i \sim \mathcal{N}$$

$$\mathcal{N} + \mathcal{N} \sim \mathcal{N}$$

$$\mathcal{N} - \mathcal{N} \sim \mathcal{N}$$

The "brown" of nature (maximum entropy dist.) e.g. AR(I); Closed under addition and subtraction

Log-Normal

What if we **multiply** rather than add?

$$Y = \lim_{M \to \infty} \left(\prod_{i=1}^{M} X_i \right)^{\frac{1}{M}} \sim \log - \mathcal{N}$$

$$\log - \mathcal{N} \times \log - \mathcal{N} \sim \log - \mathcal{N}$$

$$\log -N + \log -N \sim \log -N$$

$$\log - \mathbb{N} \div \log - \mathbb{N} \sim \log - \mathbb{N}$$

$$\log - N - \log - N \sim ?$$

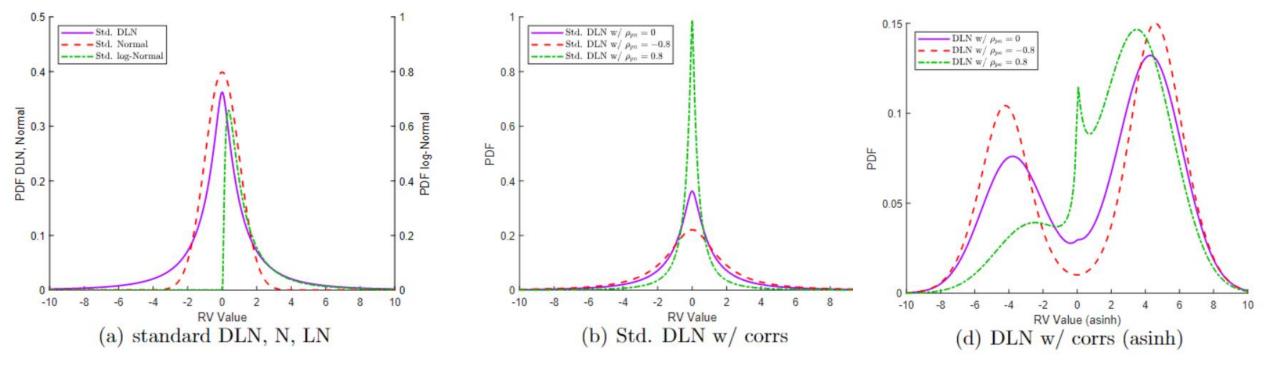
A word from our sponsor*

- "There are only two ways to make money: increase sales and decrease costs."
 - -- Fred DeLuca (founder of Subway)



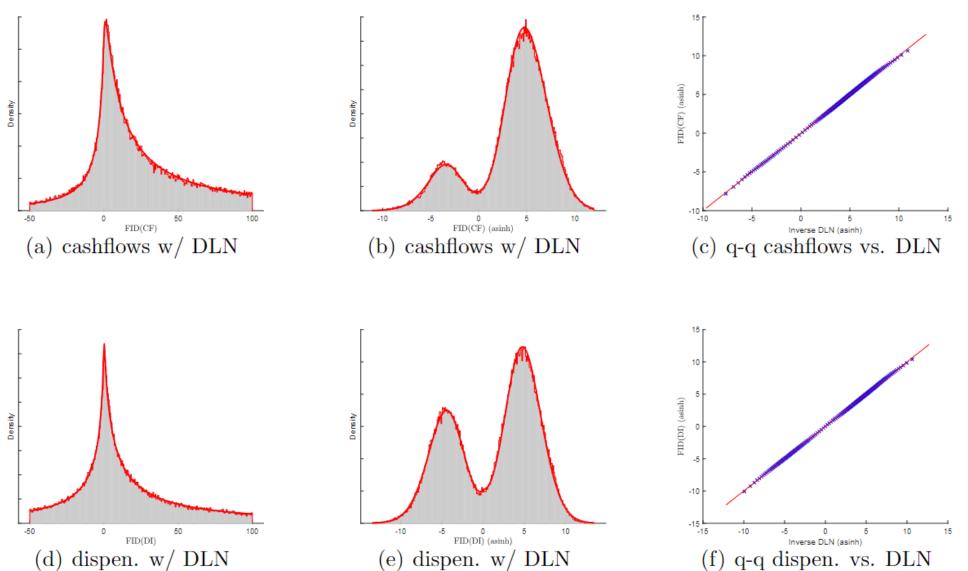
Difference of Log-Normals

$$W = Y_p - Y_n = \exp(X_p) - \exp(X_n)$$
 with $\mathbf{X} = (X_p, X_n)^T \sim \mathcal{N}(\boldsymbol{\mu}, \boldsymbol{\Sigma})$



Every marginal benefit has a marginal cost and they're both multiplicative

Example: Firm cashflows



The fundamental sources and uses equation of the firm: SALES - EXPENSES = INCOME = DISPENSATIONS + INVESTMENT; "double log"

Production function

$$Y_{LL} = \prod_{i=1}^{M} X_i^{\beta_i} = \exp\left(\boldsymbol{x} \cdot \boldsymbol{\beta}\right)$$

$$Y_{DLL} = Y_p - Y_n = \exp(\boldsymbol{x} \cdot \boldsymbol{\beta}_p) - \exp(\boldsymbol{x} \cdot \boldsymbol{\beta}_n)$$

$$Y_{DLL} = 2 \cdot \exp(\lambda) \cdot \sinh(\tau)$$

$$\lambda = \boldsymbol{x} \cdot \frac{\beta_p + \beta_n}{2} = \log\left(\sqrt{Y_p \cdot Y_n}\right) = \log(\sqrt{\text{Sales} \cdot \text{Expenses}})$$

$$\tau = \boldsymbol{x} \cdot \frac{\beta_p - \beta_n}{2} = \log\left(\sqrt{Y_p/Y_n}\right) = \log(\sqrt{\text{Sales/Expenses}})$$

Production function cool stuff

$$Y_p = \exp\left(\lambda + \tau\right)$$

$$Y_n = \exp(\lambda - \tau)$$

$$\frac{dY_t/dt}{|Y_t|} = \operatorname{sgn}(\tau_t) \cdot \left[\frac{d\lambda_t}{dt} + \frac{d\tau_t}{dt} \cdot \frac{1}{\tanh(\tau_t)} \right] \approx \operatorname{sgn}(\tau_t) \cdot \left[(\lambda_{t+1} - \lambda_t) + \frac{\tau_{t+1} - \tau_t}{\tau_t} \right]$$

$$\frac{\partial Y_{DLL}}{\partial X_i} / \frac{Y_{DLL}}{X_i} = \frac{\beta_{p,i} \cdot Y_p - \beta_{n,i} \cdot Y_n}{Y_{DLL}} = \frac{\beta_{p,i} \cdot Y_p - \beta_{n,i} \cdot Y_n}{Y_p - Y_n}$$

Summary

Gibrat: size is approx. log-Normal

Here: growth is approx. difference-of-log-Normals

Predicted by a sensible production function and confirmed in data Example prod. func. for firms, cities, and populations in paper

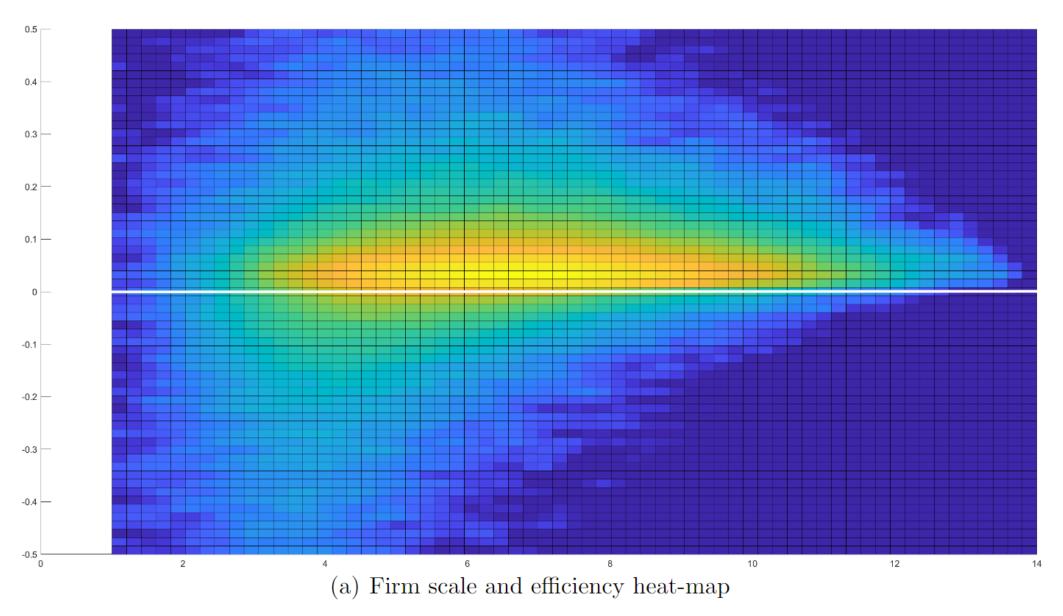
Riddle: A firm has \$100 in sales and \$90 in expenses.

What is **income growth** if both sales and expenses increase by 10%?

What if sales increase by 10% but expenses decrease by 10%?

What if sales decrease by 10% and expenses increase by 10%?

Sunrise



Lam-Tau heatmap for all yearly firm obs 1970-2019