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## **Making Gravity Great Again**

**Will Martin**

*Selected presentation for the International Agricultural Trade Research Consortium's (IATRC's) 2020 Annual Meeting: Economic Implications of COVID-19, December 14-15, 2020, Virtual platform.*

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# Making Gravity Great Again

IATRC Annual Meeting

Will Martin

International Food Policy Research Institute

15 December 2020

# Roadmap


- Importance of the Issue
- Econometric problems
- Approaches to estimation
- Monte Carlo estimates of bias

# Importance of the question

- Gravity model fits data well– widely used for trade, investment, migration etc
- Many recent suggestions to replace traditional trade policy models
- Lack of agreement on the best estimators
  - Wide differences between results from different estimators

# Econometric problems

# Econometric challenges

- Limited-Dependent Variable Bias
  - Large shares of trade, investment, migration data are zeros
    - 40% in aggregate trade flows, higher for disaggregate data
    - Creates bias due to non-zero errors
- Combination of nonlinearity & heteroscedasticity
  - Correlation between errors & explanatory variables  bias
  - Inefficient rather than biased with linear models
    - Estimating in logs gives linearity, but then the zeros truncated
- Heteroscedasticity gives different bias with limited-dependent estimators
  - Whether linear or non-linear
- Some zero observations are not zero, just missing

# Eaton-Kortum Tobit Model

$$y_i^* = f(x_i, \beta) + u_i$$

$$\begin{aligned} y_i &= y_i^* && \text{if } y_i^* > \mathbf{y} \text{ or} \\ y_i &= 0 && \text{if } y_i^* \leq \mathbf{y} \end{aligned}$$



Threshold

- Where  $y$  is a latent variable,  $x$  an explanatory variable,  $\mathbf{y}$  a minimum threshold, and  $u$  an error term, that differs between observations



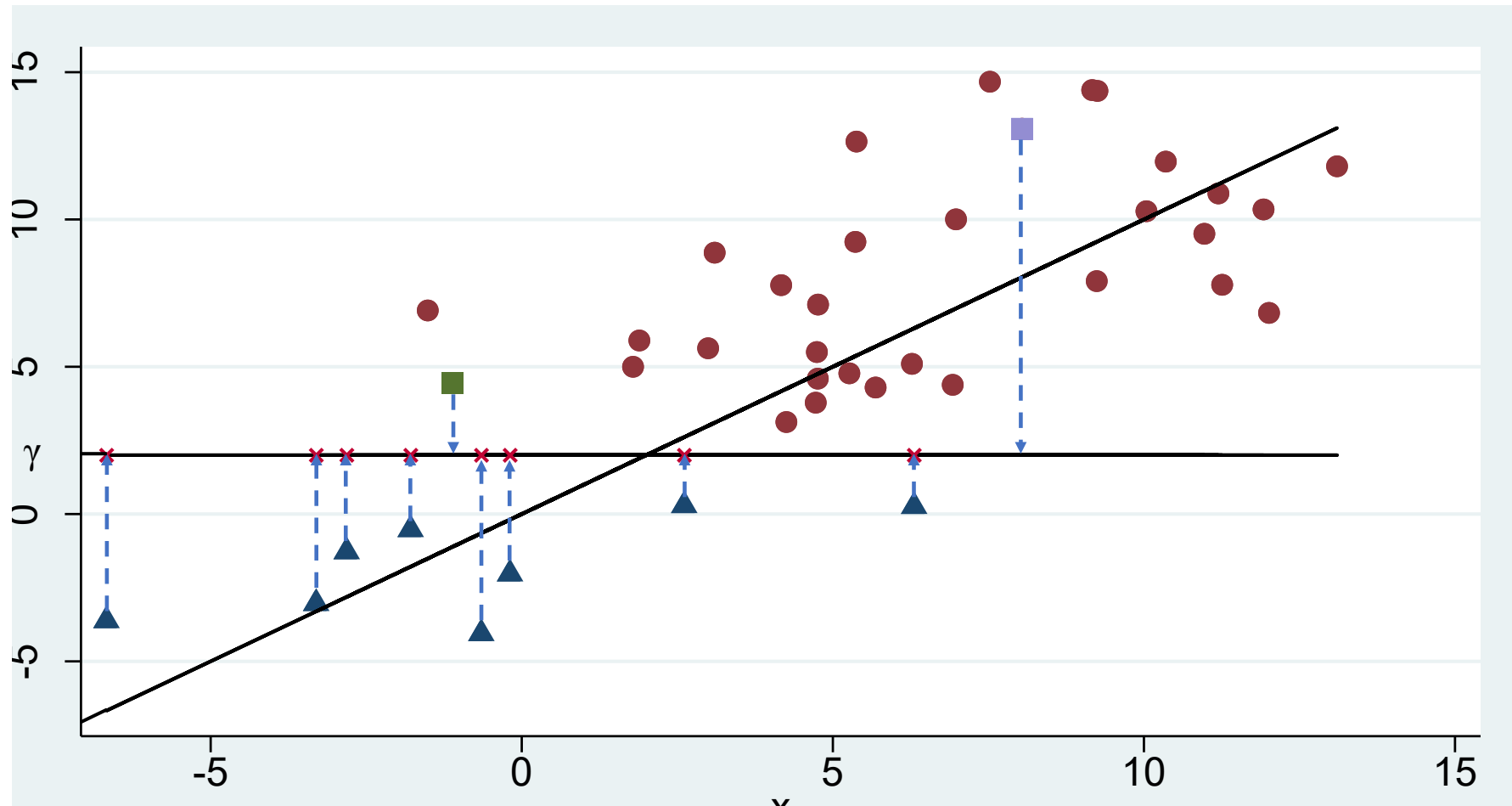
# Likelihood function for this model

- $L(\beta, \sigma) = \prod_{i=1}^{n_0} \Phi\left(\frac{\mathbf{y}-f(x_i, \beta)}{\sigma_i}\right) \prod_{i=n_0+1}^n \frac{1}{\sigma_i} \phi\left(\frac{y-f(x_i, \beta)}{\sigma_i}\right)$
- $\phi$  is the density function for non-limit obsvns
- $\Phi$  is the distribution function for limit observations

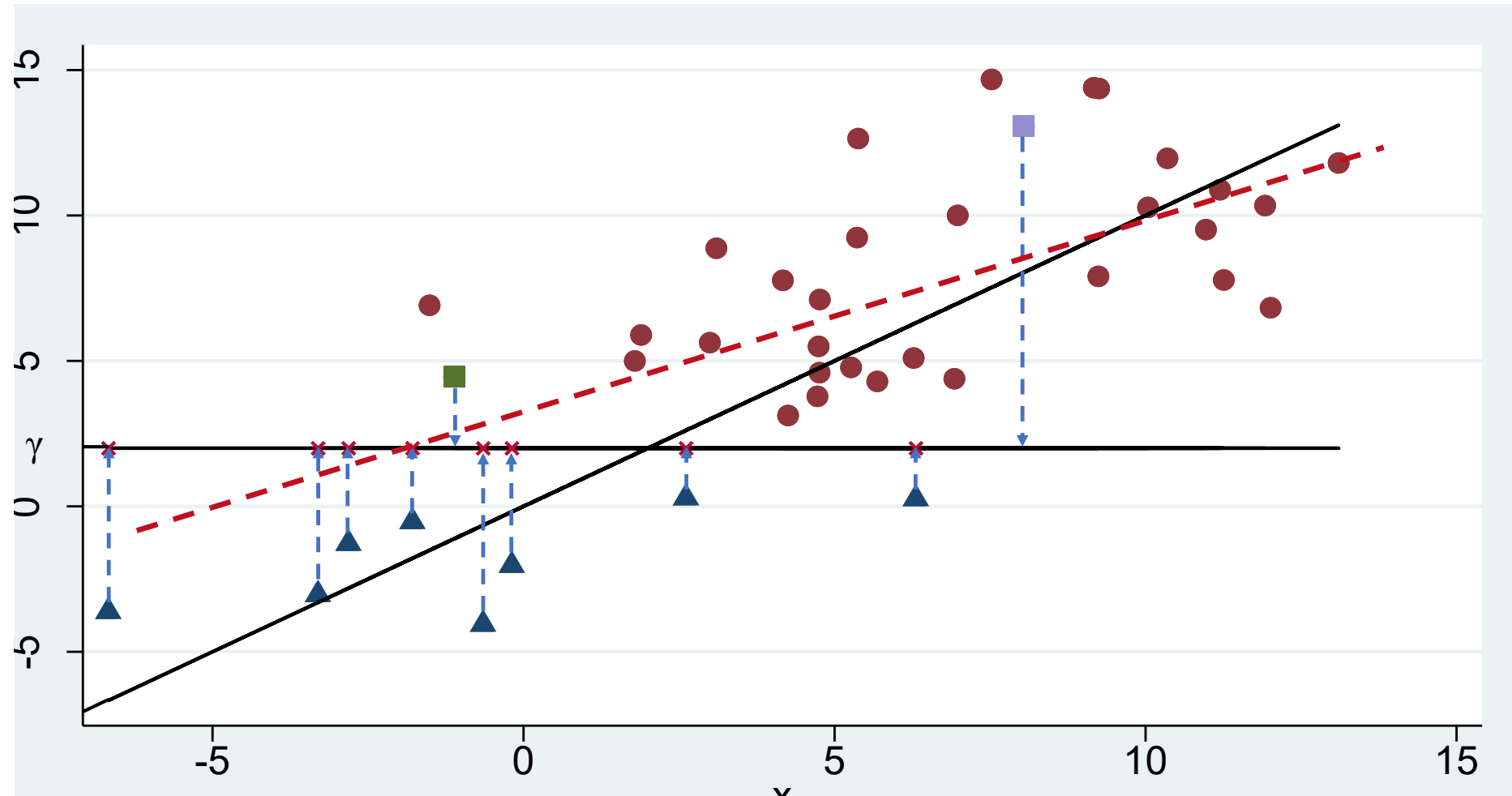
# Eaton-Kortum/Carson-Sun Threshold Model

- Threshold often taken to be zero
  - But the lowest profitable value of trade  $\gamma > 0$
- Carson-Sun examine automobiles
  - Smallest new vehicle purchase the cheapest available car
- Eaton-Kortum propose using the lowest observed import to country
  - Carson-Sun show estimator is super-consistent wrt  $\gamma$
- $\gamma > 0$  eliminates the zero problem in log models

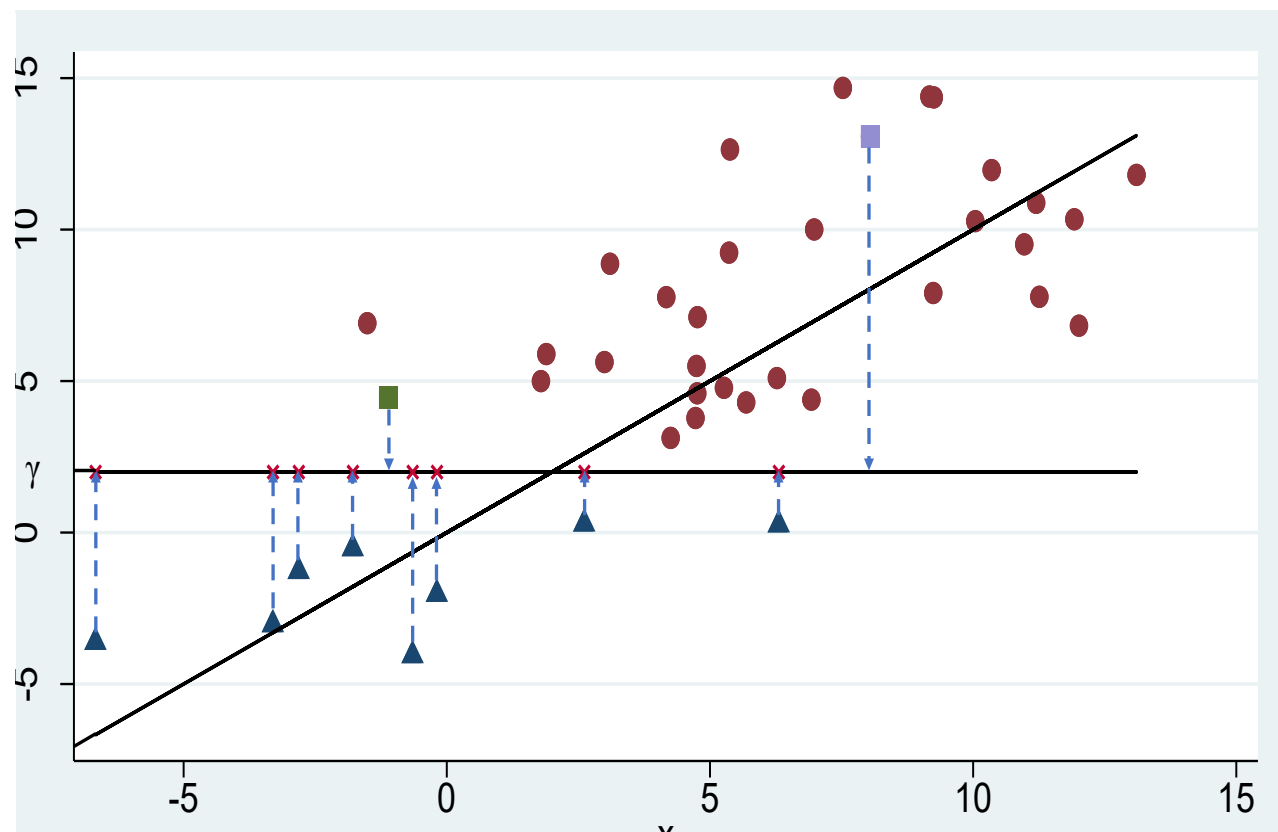
**x** Values represented by Distn fn, ●s by Density



# Limited-Dependent: Downward Bias with OLS



# Missing Data Recorded as Zeros



- Biased up if missing data small
- Biased down if missing data large

# Estimation

# Consider six estimators

- OLS in logs
  - Traditional model– zeros truncated
- PPML
- Generalized Least Squares (for heteroscedasticity)
- Threshold Tobit
- Weighted Tobit
- Eaton-Kortum with heteroscedasticity, EK-H

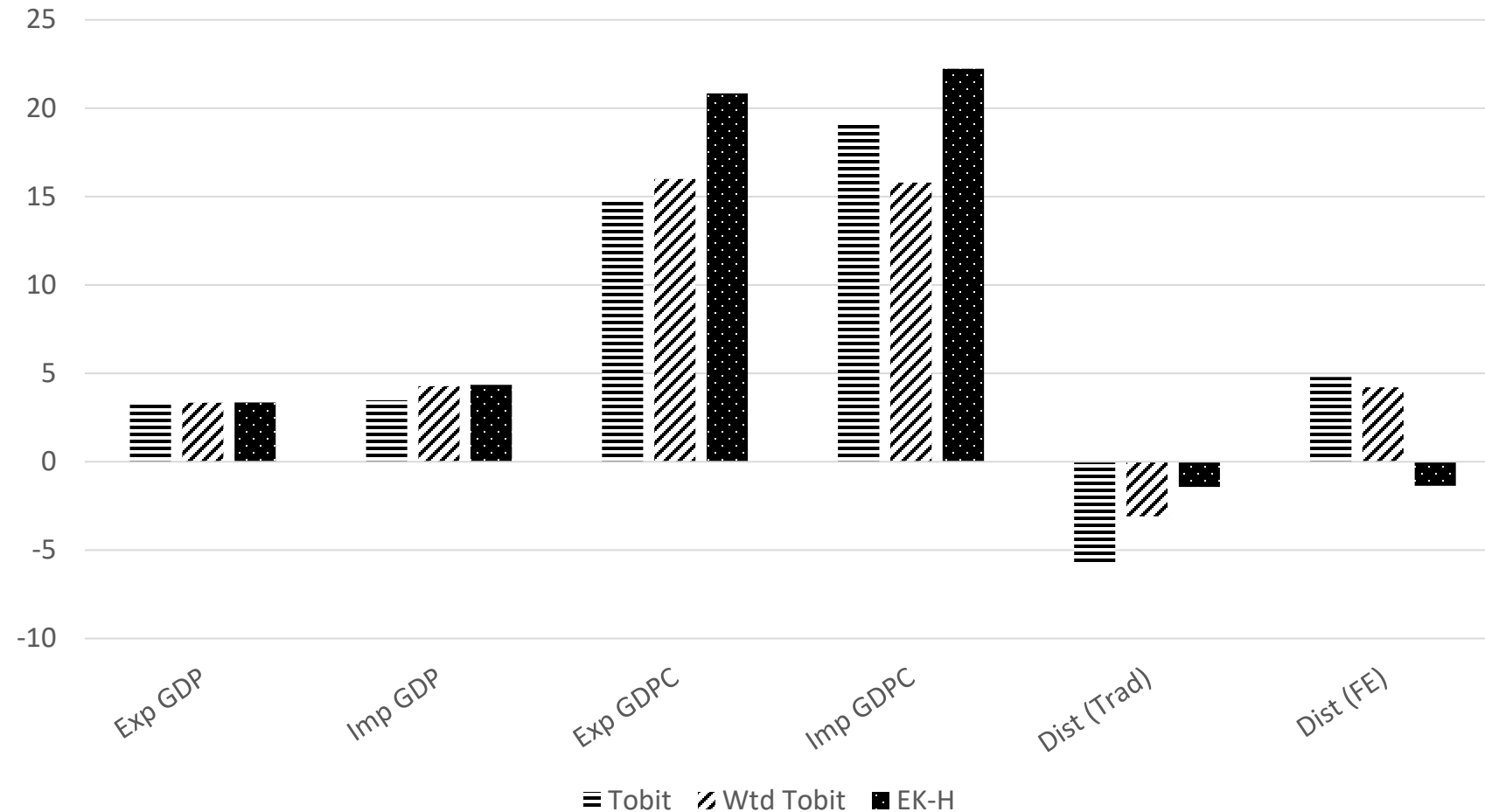
# Dataset

- 136\*136 cross-sectional trade matrix by Feenstra, Lipsey & Bowen
  - Used by Santos Silva & Tenreyro & many other studies
  - 48% zeros
- 46 of 136 economies did not report trade to COMTRADE for 1992
  - 2059 of the potential 2070 trade flows between them were zero
  - Clearly no attempt made to impute these data
  - Safer to exclude them
- Resulting 16290 cross-sectional observations with 41% zeros



# Effect of adding the missing data as zeros

## – upward bias



# Huge Differences in Estimates: Dist. in FE Gravity



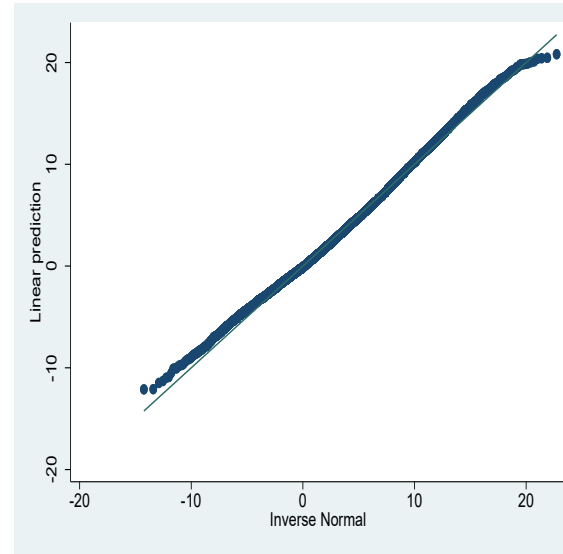
# Monte Carlo Analysis

# Monte Carlo Analysis misleading unless parameters well-chosen. What matters?

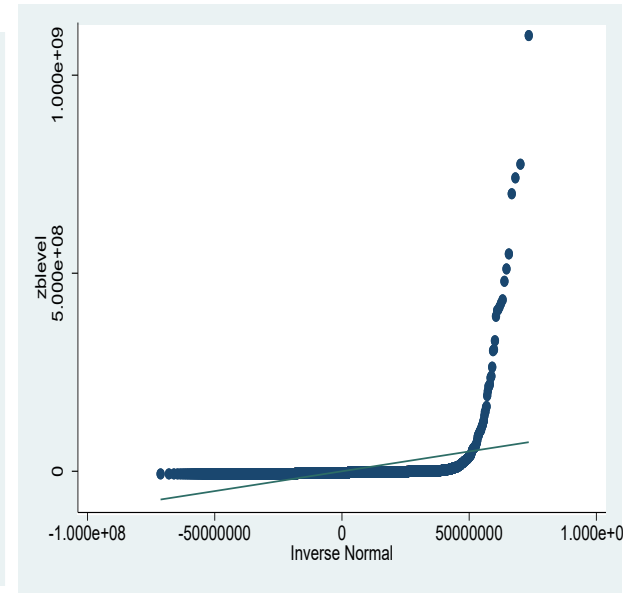
1. The *distribution of predicted* outcomes, including
  - (i) Their distribution & whether normal or log-normal,
  - (ii) The mean & variance of this distribution
2. *Distribution of residuals* around expected trade outcomes, incl:
  - (i) Their distribution,
  - (ii) Their standard deviation at any point,  $\sigma_i$ , and
  - (iii) The pattern of heteroscedasticity
3. The fraction of *observations at the threshold*

# QQ plots 4 Normality

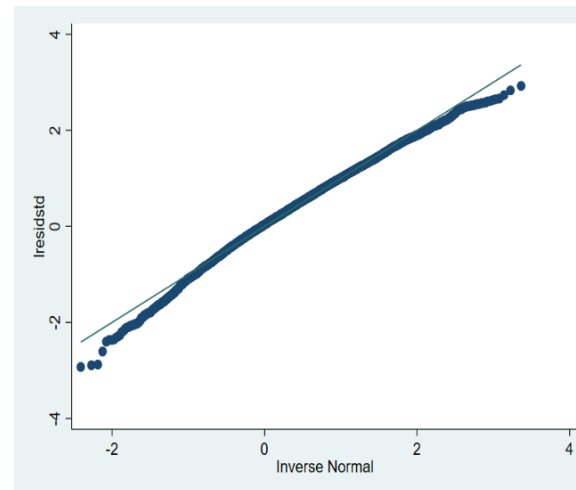
1. Predicted Values, logs



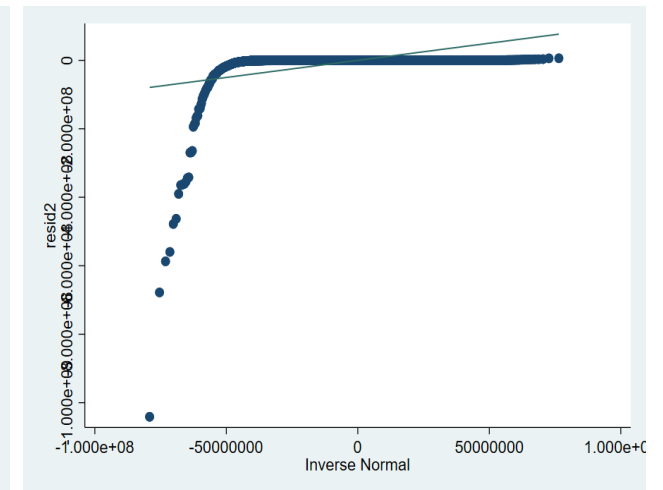
2. Predicted Values, Levels



3. Residuals, logs



4. Residuals, levels



## Parameters from 2 candidate models: PPML & EK-H

- *Test model:*  $y = 1.x + u$
- $x \sim N(4.2, 4.85)$  for EK-H Data Gen Process
- $x \sim N(8.2, 2.8)$  for PPML DGP
- $\sigma_i^2 = e^{(a+by_i)}$  where  $a = 2.7$  &  $b = -0.115$
- Samples of 10,000 repeated 1000 times

# Monte Carlo Results

|                       | EK-H Parameters |           | PPML Parameters |           |
|-----------------------|-----------------|-----------|-----------------|-----------|
|                       | $\hat{\beta}$   | Std Error | $\hat{\beta}$   | Std Error |
| <b>OLS</b>            | 0.62            | 0.01      | 0.52            | 0.01      |
| <b>PPML</b>           | 0.87            | 0.13      | 0.78            | 0.11      |
| <b>Tobit</b>          | 0.89            | 0.01      | 0.90            | 0.01      |
| <b>Weighted Tobit</b> | 0.92            | 0.03      | 0.85            | 0.02      |
| <b>EK-H</b>           | 0.999           | 0.01      | 1.000           | 0.001     |

# Conclusions

- Need good estimates from the gravity model
  - Potentially serious biases from
    - (i) limited dependent variables
    - (ii) heteroscedasticity & nonlinearity, &
    - (iii) missing values mapped to zeros
- Estimates vary enormously between widely-used estimators
  - Need Monte Carlo analyses to distinguish against known parameters
  - Must use distributions in the relevant range
- EK-H estimator allowing for heteroscedasticity & limited-dependent vbls unbiased
  - Traditional OLS very seriously biased
  - PPML biased around 20%, Tobit & Hetero-Wtd Tobit a little better