

# Frailty Correlated Default

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Preliminary

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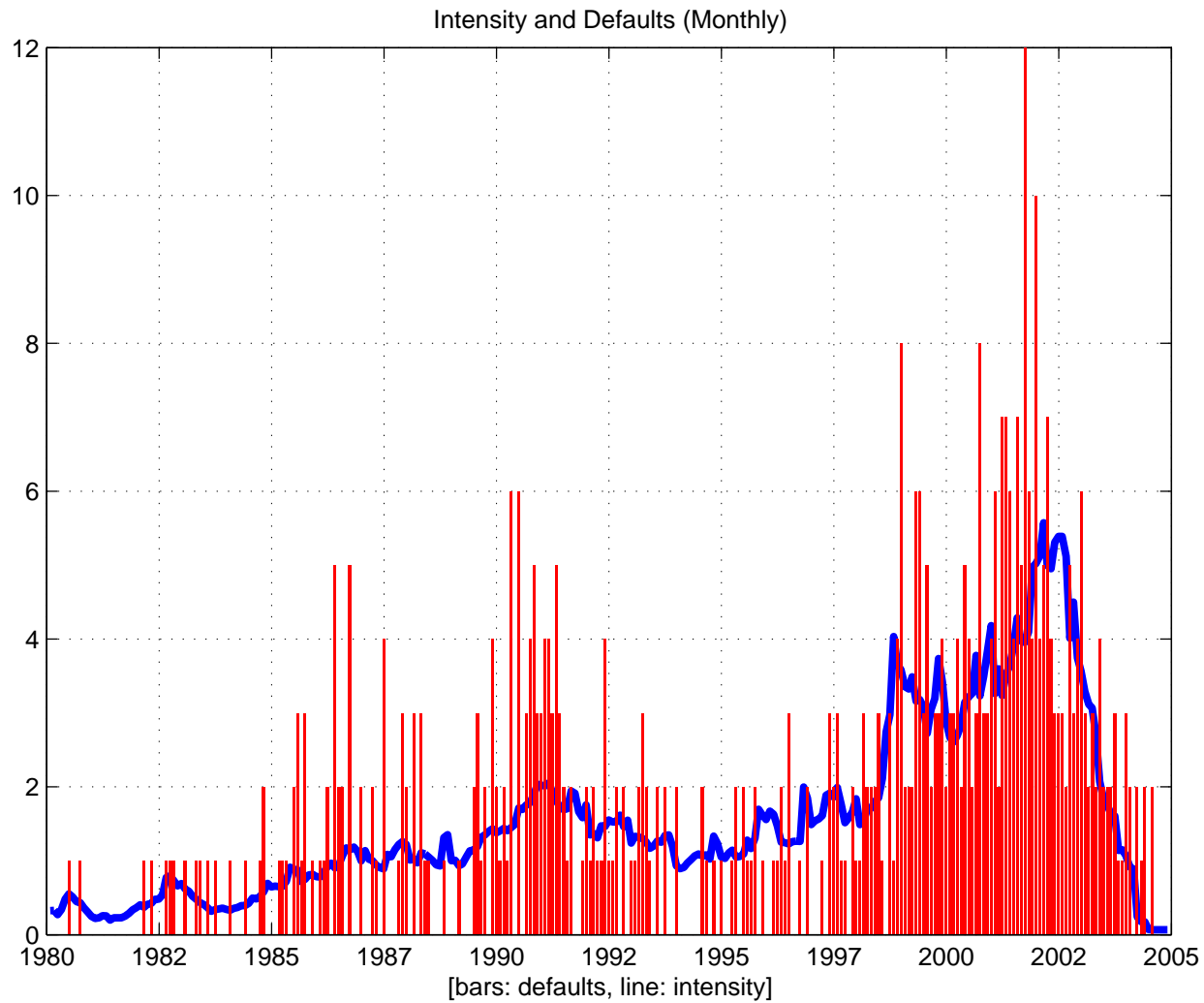


Figure 1: Aggregate default intensities and number of defaults, 1979-2004.

## Default Intensity

- Default intensity: the conditional mean arrival rate of default, in events per year.
- Example: An intensity of 0.2% is a mean arrival rate of 2 defaults per 1000 obligor-years, or roughly a probability of default of 0.2% per year.
- We estimate the parameter vector  $\beta$  for default intensity of the form  $\Lambda(X_t; \beta)$ , where  $X_t$  is a list of covariates, including the firm's leverage, volatility, other firm-specific and macro-economic covariates, and “frailty” factors, both common and firm-specific.

## Dynamic Default Frailty Model

Firm  $i$  default intensity:

$$\lambda_{it} = e^{\alpha + \beta \cdot W_{it} + Y_t + Z_i}$$

where

- $W_{it}$  is a vector of observable covariates.
- $Y$  is a common unobserved Brownian motion.
- and  $Z_i$  is unobserved independent-gamma heterogeneity.

## Matchable U.S. Industrials 1979-2004

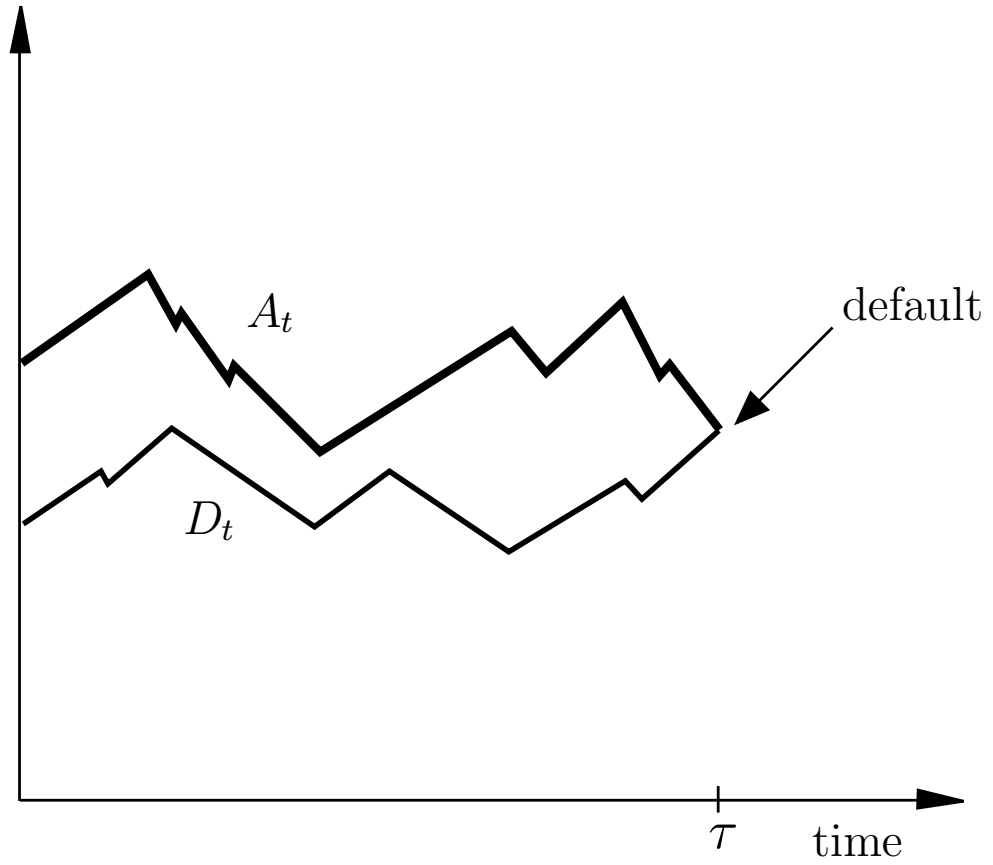
- 402,000 firm-months of data.
- 2,793 firms.
- 496 defaults.
- 1,072 exits by merger or acquisition.

## Some Related Work

- Black and Scholes (1973), Merton (1974), Fischer, Heinkel, Zechner (1989), Leland (1994).
- Beaver (1968), Altman (1968), ..., Shumway (2001), Beaver, McNichols, and Rhie (2004), Duffie, Saita, Wang (2005), Campbell, Hilscher, and Szilagyi (2005), ...
- Schönbucher (2003), Collin-Dufresne, Goldstein, and Hugonnier (2004), Delloy, Fermanian, and Sbai (2005).

# Key covariate: distance to default

Asset and Liability Values



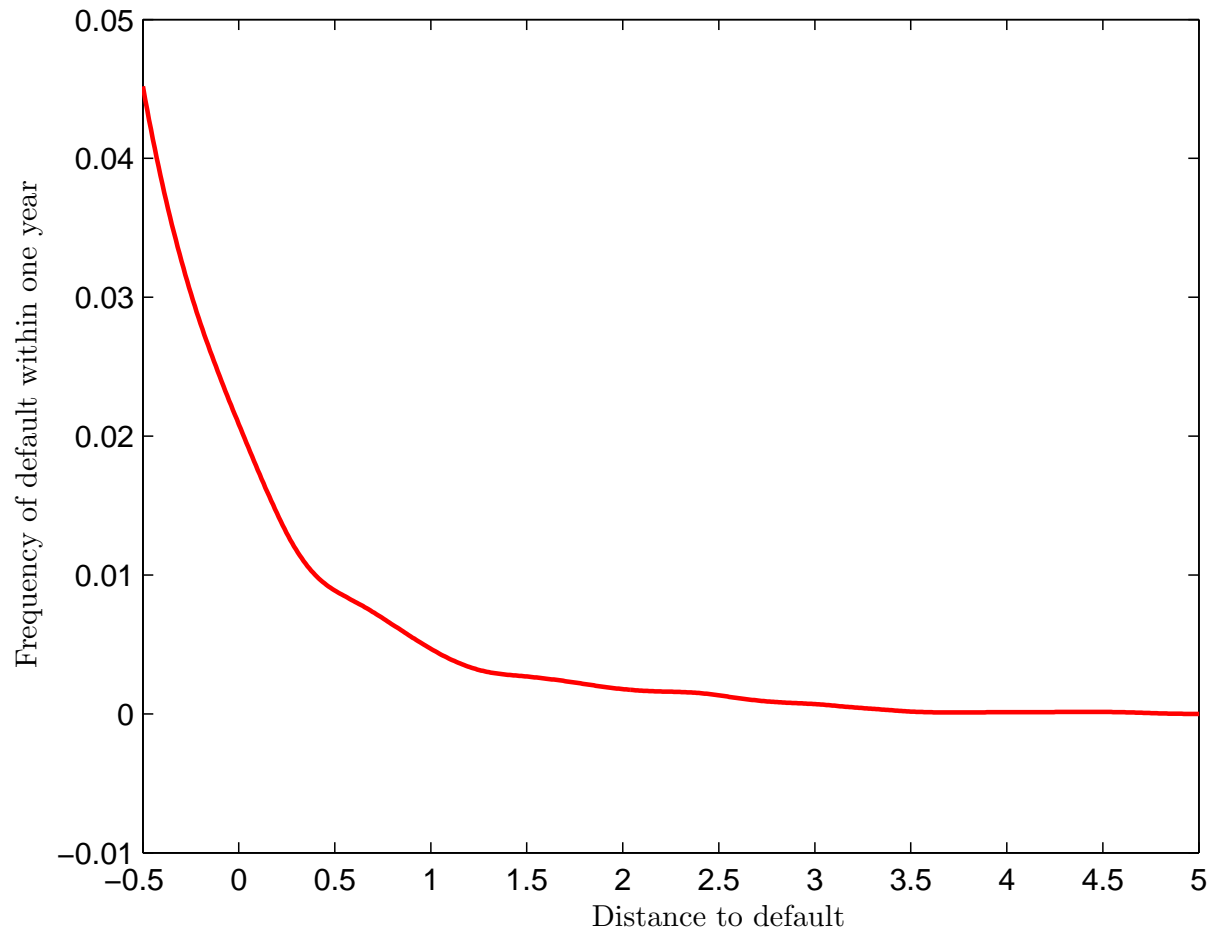


Figure 3: The dependence of empirical default frequency on distance to default.  
Source: Duffie, Saita, Wang, *J. Financial Economics*, 2006.

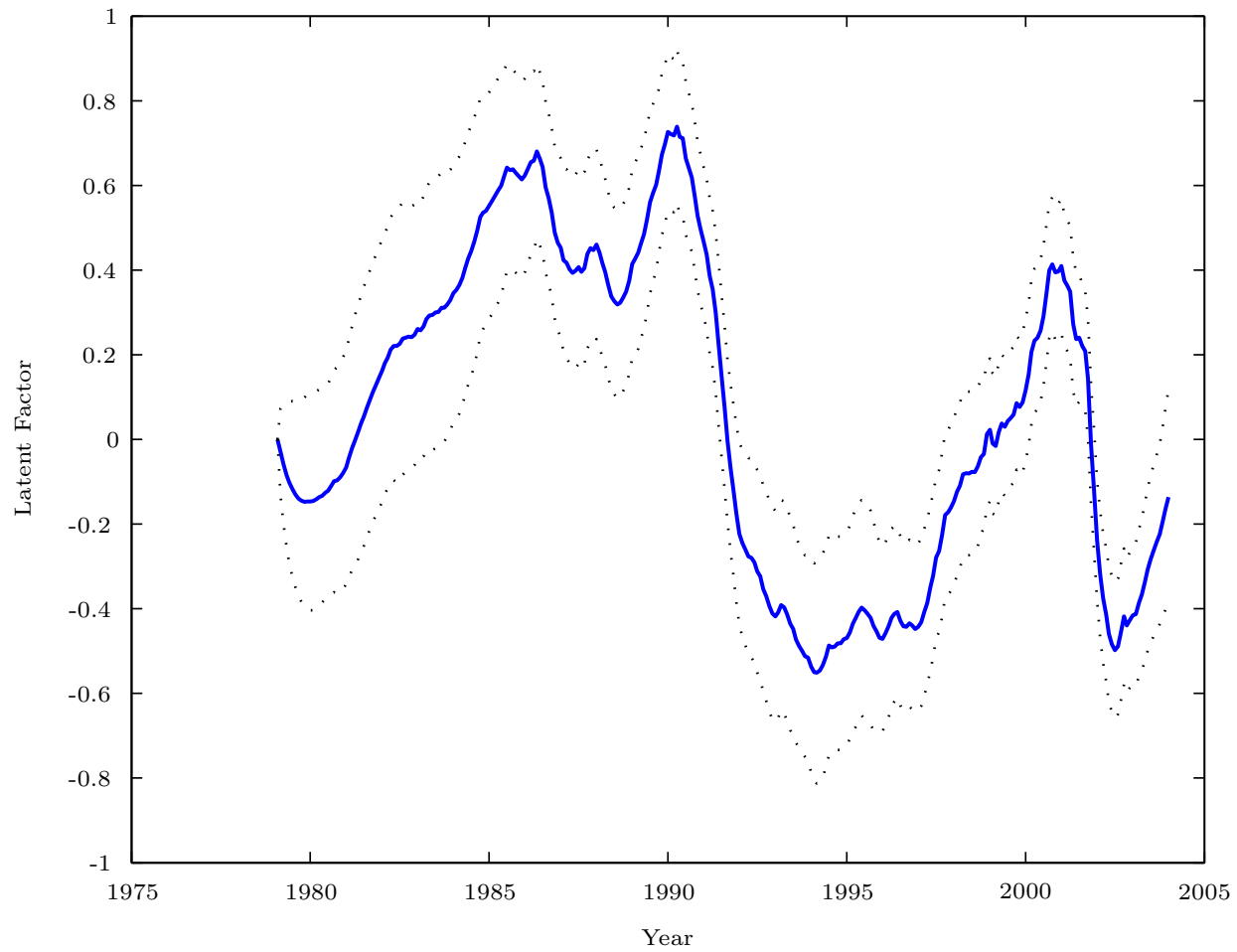


Figure 4: Posterior of the frailty factor, conditioning in 2004.

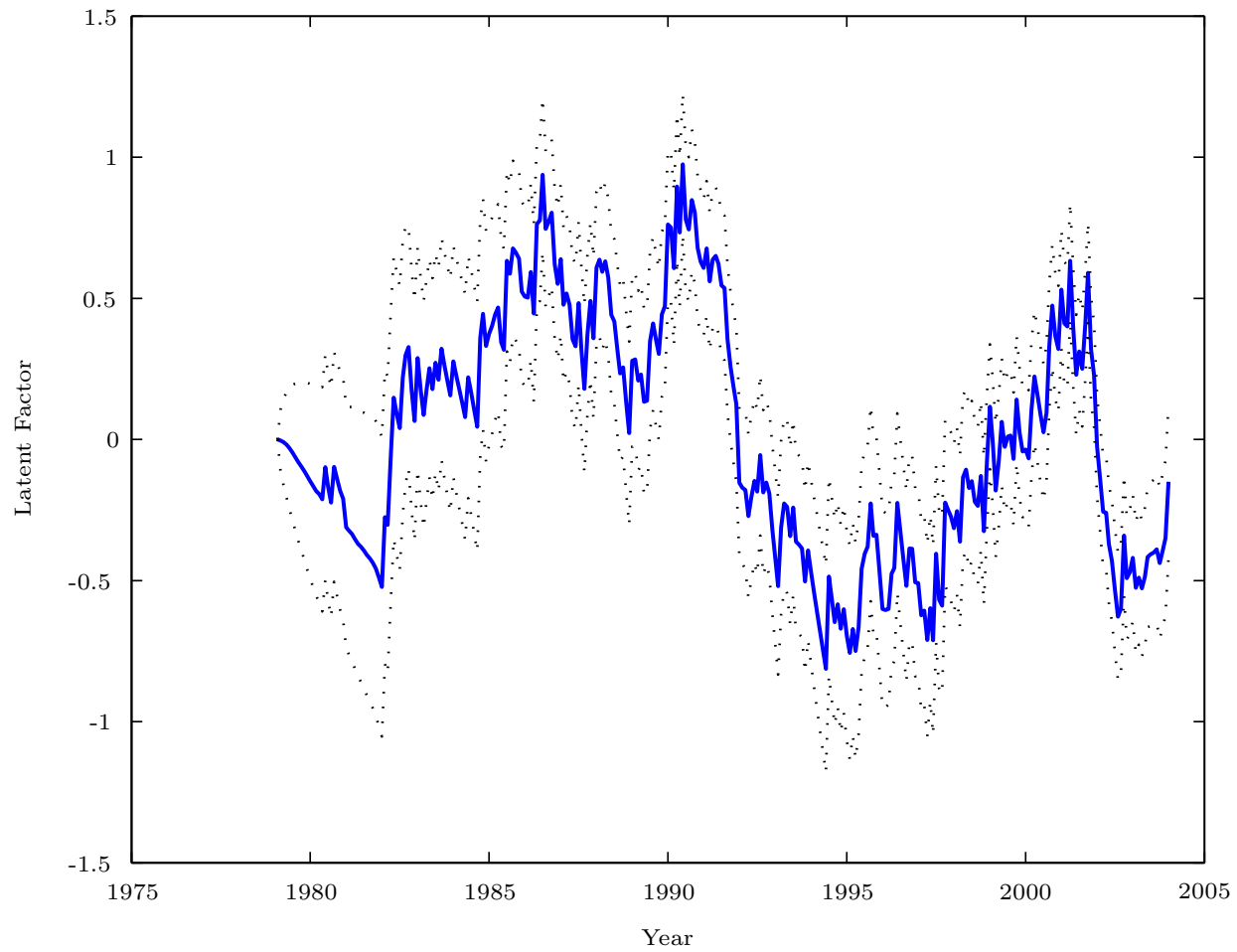


Figure 5: The frailty distribution, conditioning “as you go.”

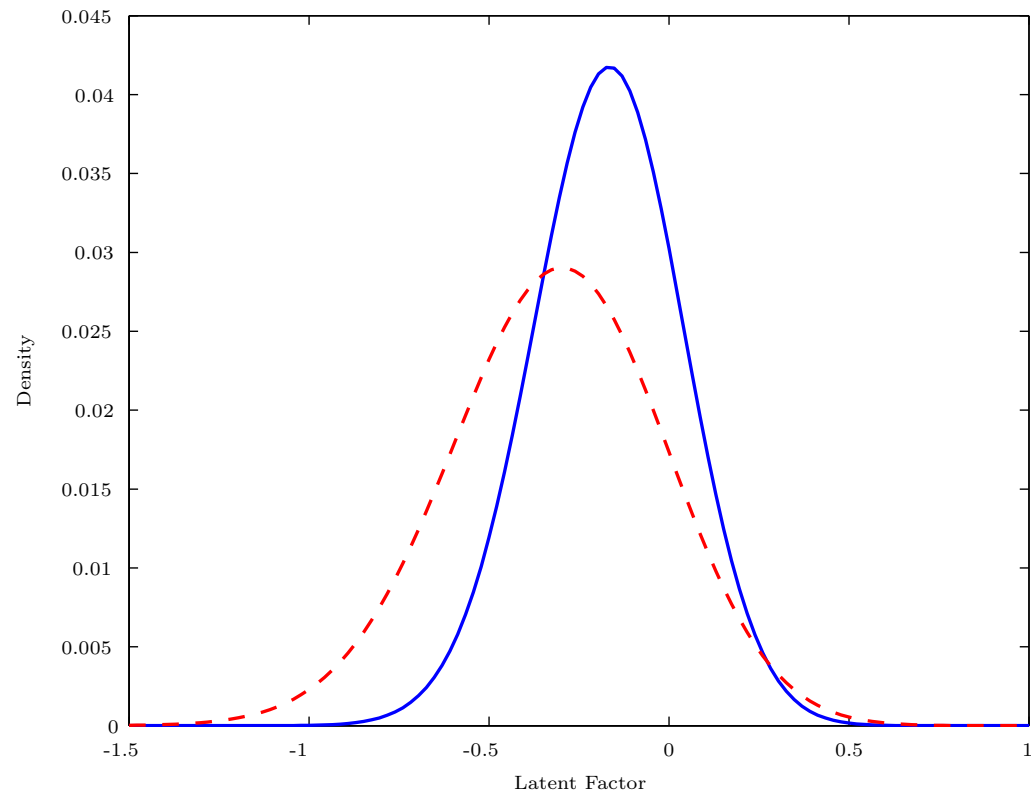
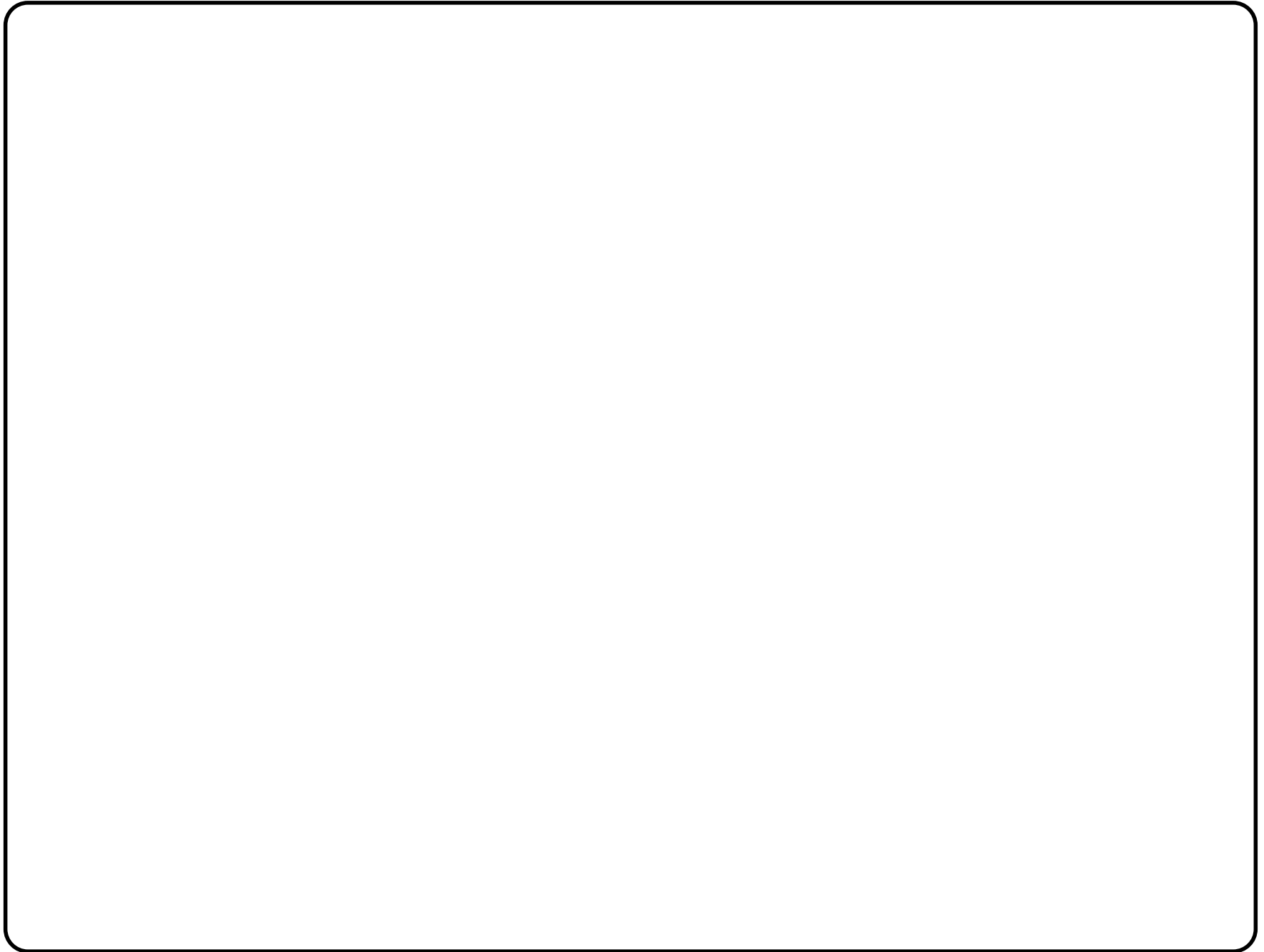


Figure 6: Posterior of the frailty factor in January 1998, conditioning in 2004 (blue), and conditioning in January 1998 (red).



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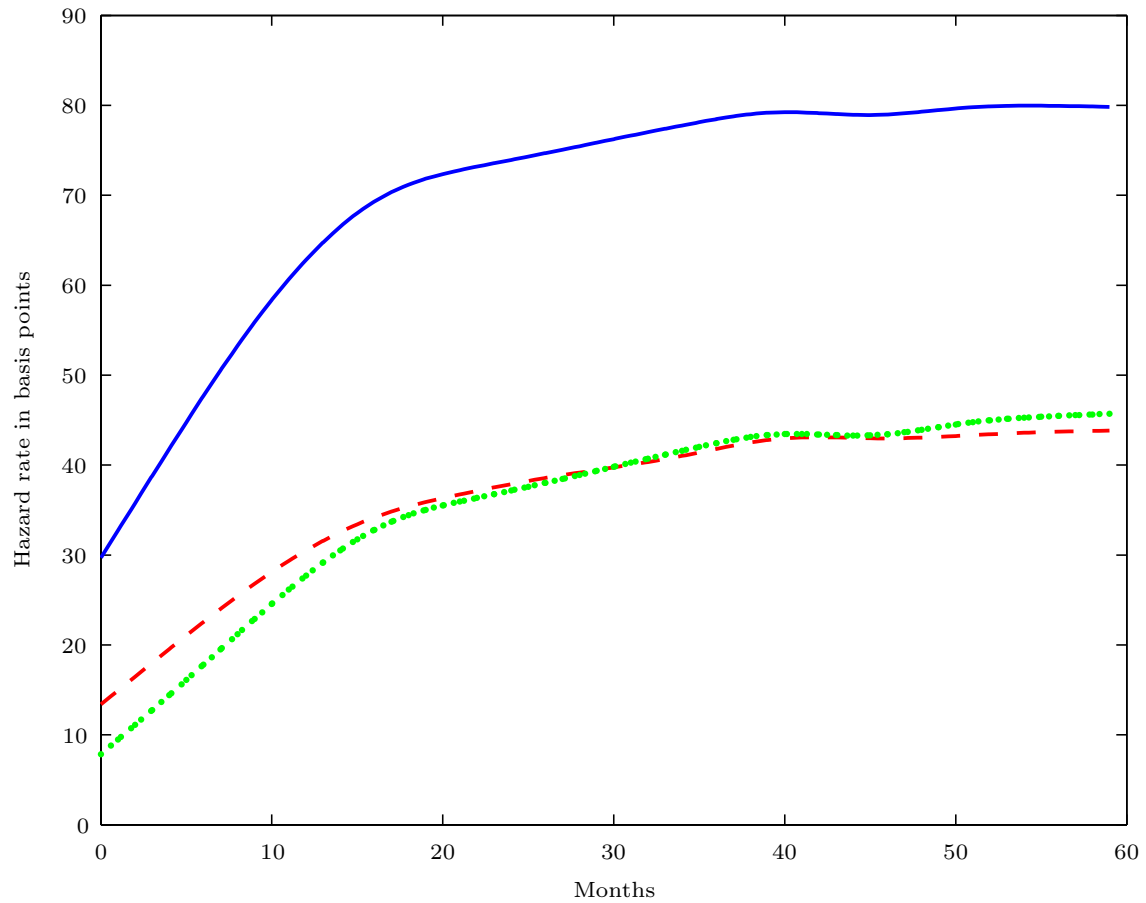


Figure 7: Term structure of hazard rates, Xerox, December 2003: frailty (blue), no frailty (red), unobserved heterogeneity (green).

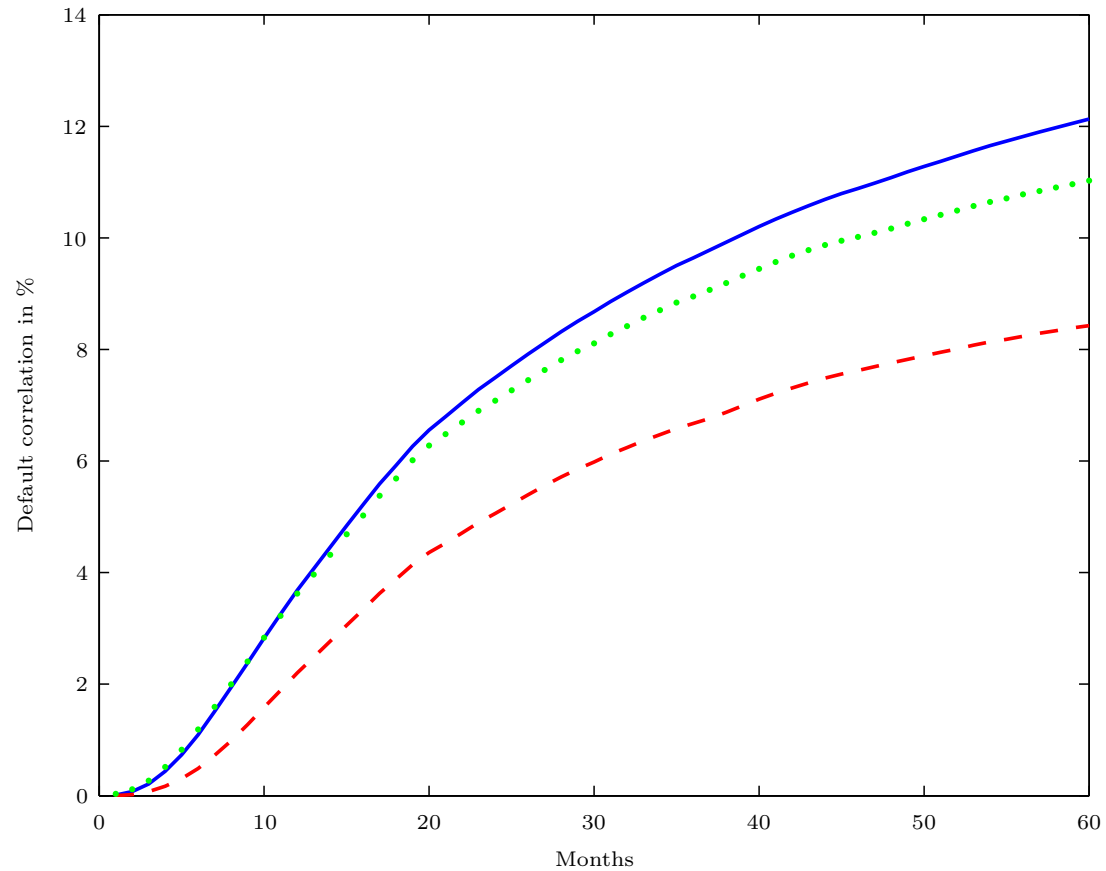


Figure 8: Default correlation of ICO and Xerox with frailty (blue), without frailty (red), and with frailty and unobserved heterogeneity (green).

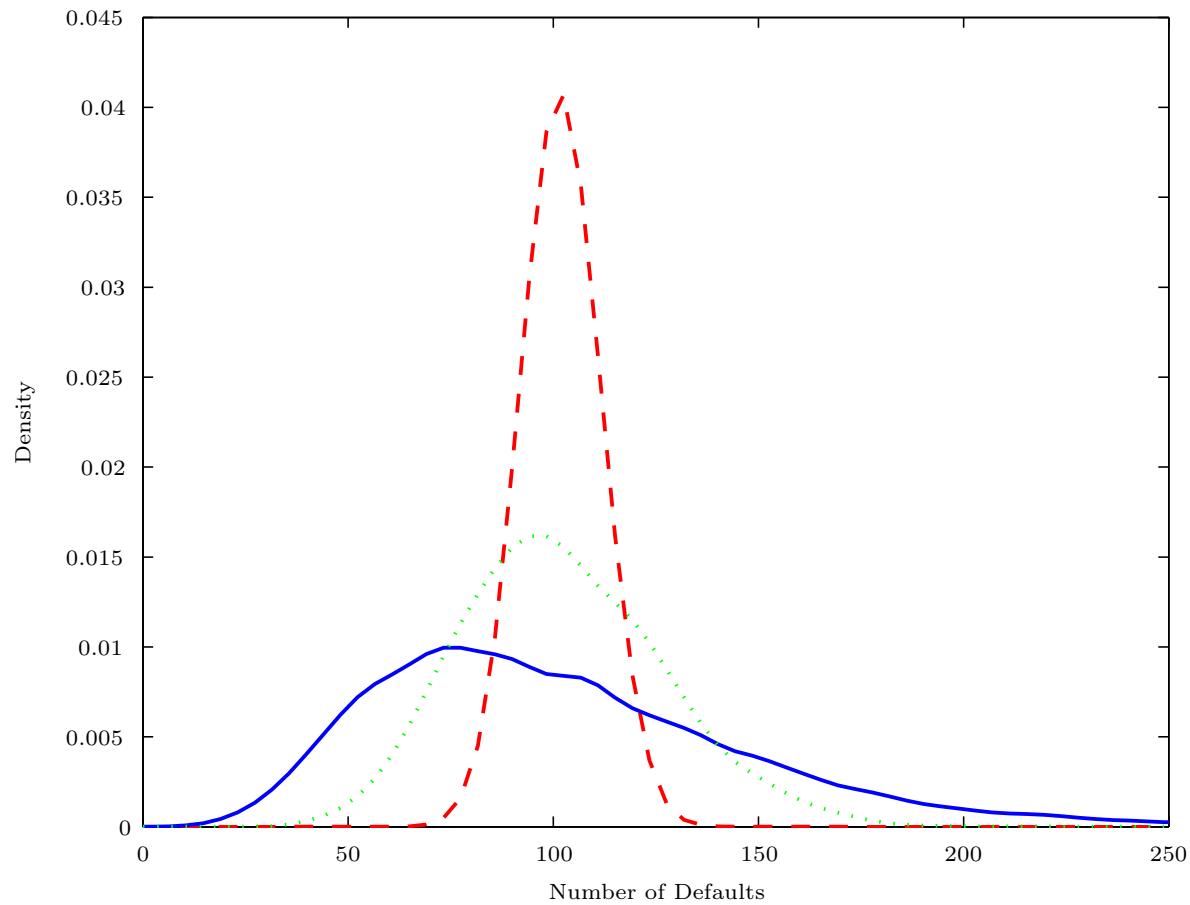


Figure 9: 5-year portfolio defaults: common frailty (blue), independent frailty (red), common start only (green).

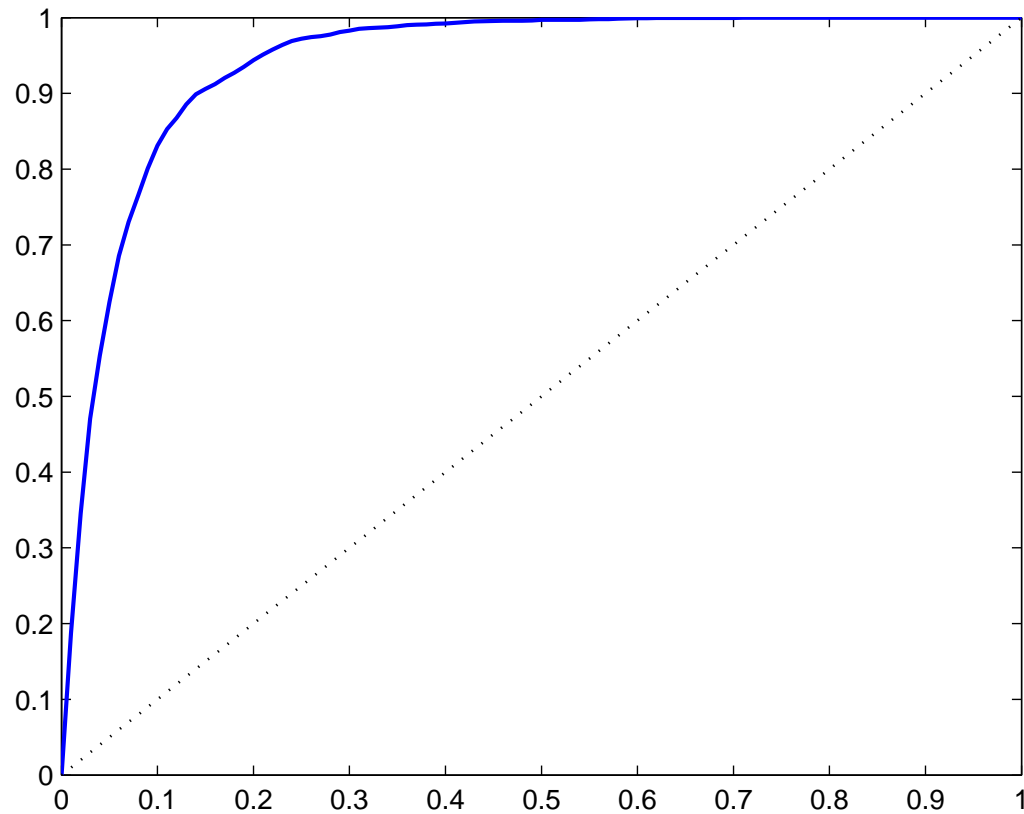


Figure 10: Out-of-sample power curve for 1-year default prediction, January 1993 to December, 2003. Example: The lowest-ranked 10% of firms accounted for 80% of the next year's defaulters. Source: Duffie, Saita, Wang, *J. Financial Economics*, 2006.

## Test of Doubly Stochastic Model

From Das, Duffie, Kapadia, and Saita, *J. Finance* (2006).

- Doubly-stochastic: Defaults are independent conditional on the factors determining default intensities.
- Implication: Re-scaling time, with “new time” passing at rate  $\sum_i \lambda_i(t)$  per unit of calendar time, total default arrivals are Poisson at rate parameter 1.

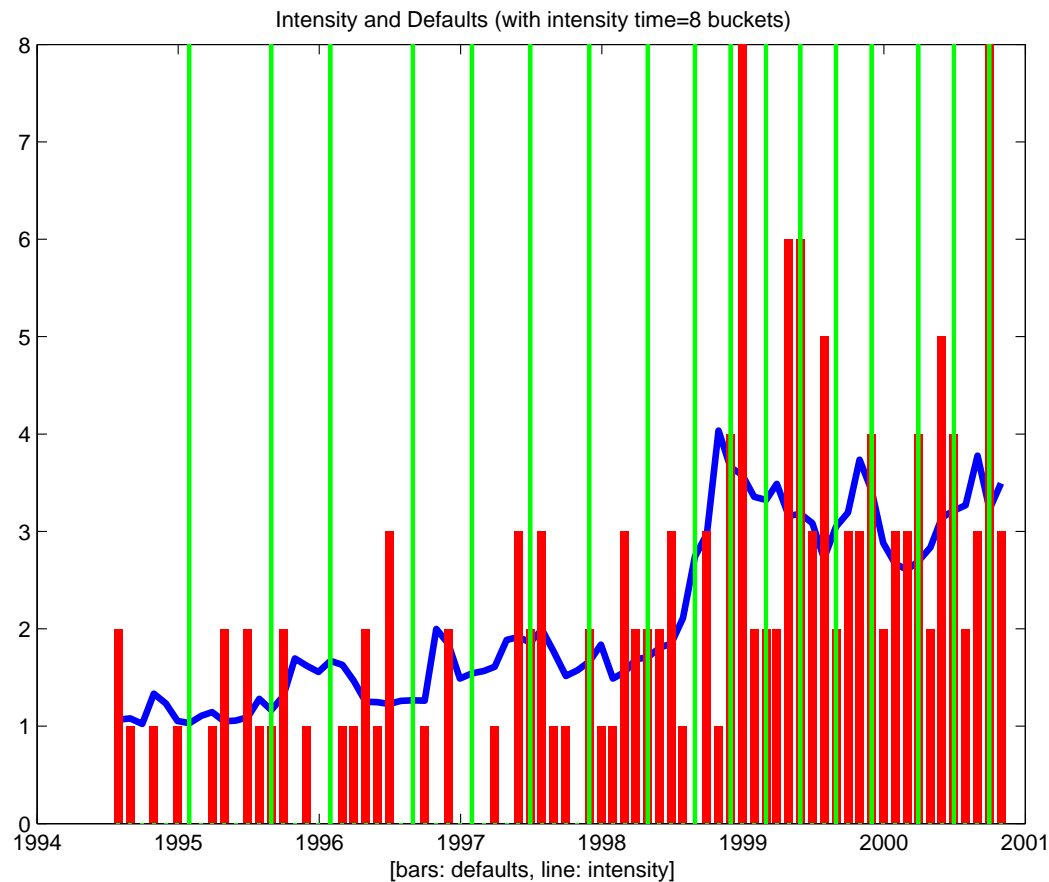


Figure 11: Aggregate intensities and defaults by month, 1996-2001. Bins each include 8 units of total accumulated default intensity. Source: Das, Duffie, Kapadia, and Saita, *J. Finance* (2006).

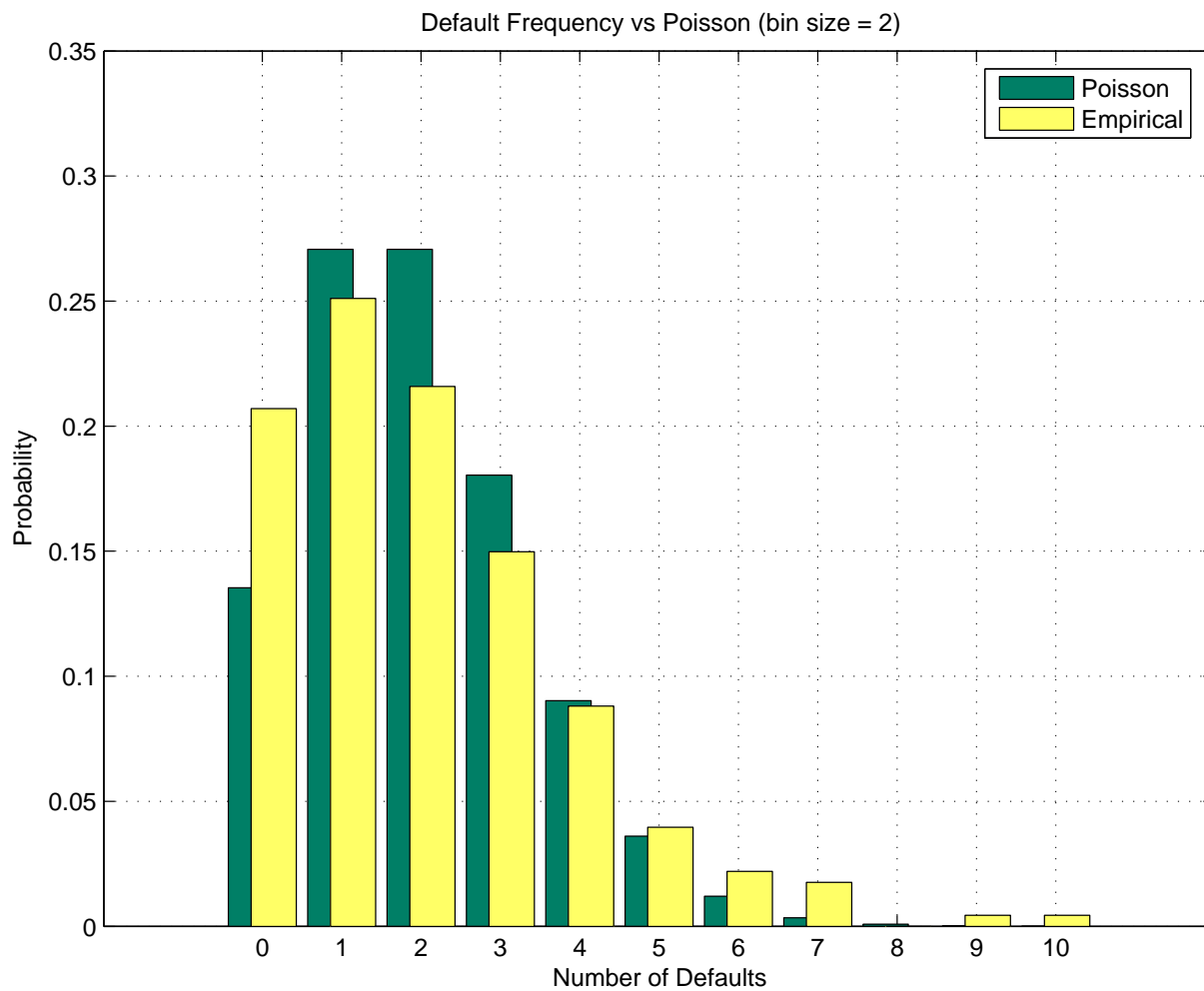


Figure 12: The empirical and Poisson distributions of defaults for bin size 2. Source: Das, Duffie, Kapadia, and Saita, *J. Finance* (2006).

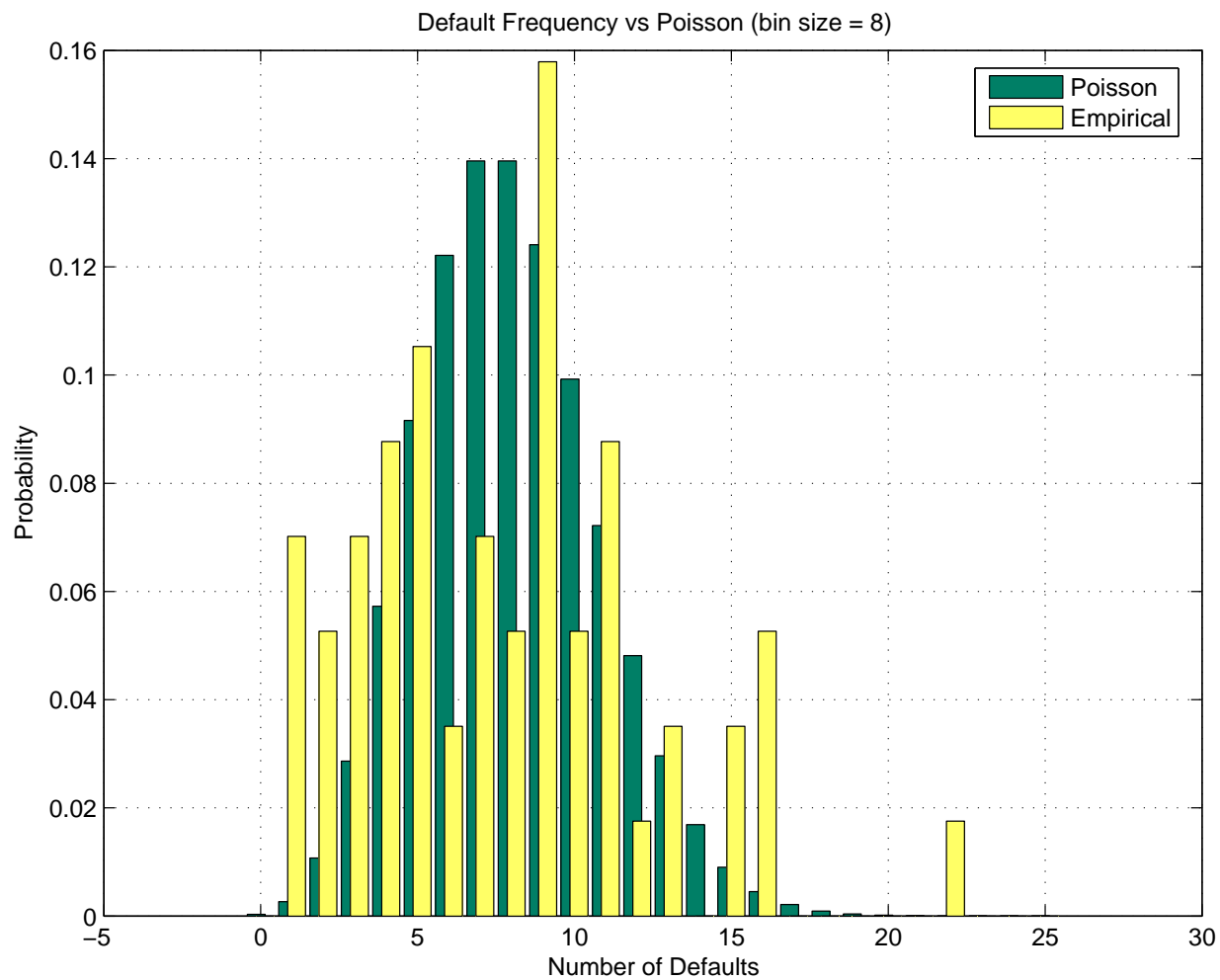


Figure 13: The empirical and Poisson distributions of defaults for bin size 8. Source: Das, Duffie, Kapadia, and Saita, *J. Finance* (2006).

## Joint Likelihood with Frailty

- The joint covariate process  $X = (W, Y)$ , with  $W$  observed,  $Y$  independent and not observed, is a Markov process whose transition distribution is determined by a parameter vector  $\gamma$ .
- By Bayes' Rule, with observable  $X$ , the joint likelihood of  $X$  and the censored default times  $\tau = (\tau_1, \dots, \tau_n)$ , is

$$\mathcal{L}(\tau, X; \beta, \gamma) = \mathcal{L}(\tau | X; \beta) \times \mathcal{L}(X; \gamma).$$

- With unobservable  $Y$ ,

$$\mathcal{L}(\tau, W; \beta, \gamma) = E [\mathcal{L}(\tau | Y, W; \beta) | W] \times \mathcal{L}(W; \gamma),$$

evaluated by Gibbs sampling.

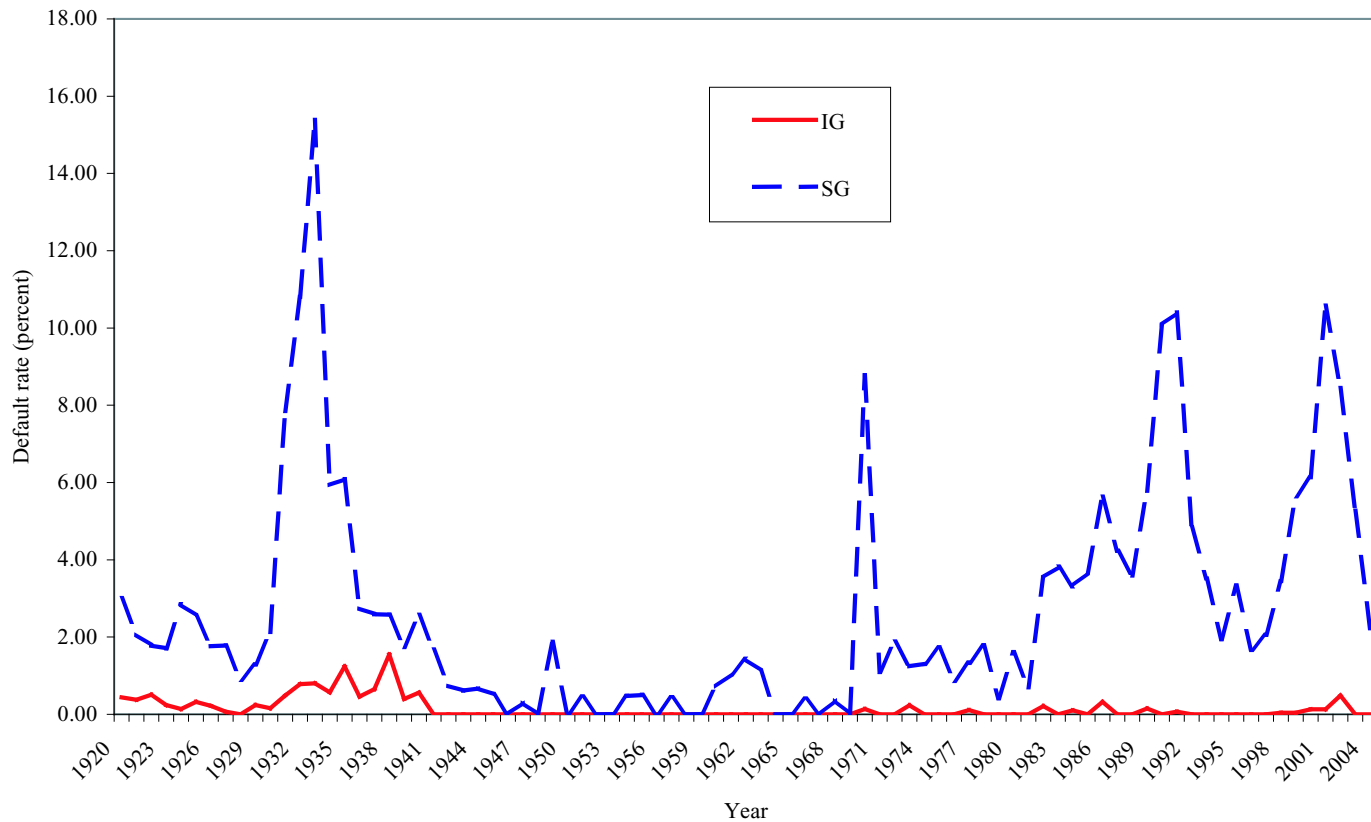


Figure 14: Annual default rates, investment (IG) and speculative (SG). Source: Moodys (2005).

## Default Intensity Estimates

$$\lambda(t) = e^{\beta_0 + \beta_1 X_1(t) + \dots + \beta_n X_n(t)}$$

Maximum Likelihood estimate of  $\beta_i$  (with standard error)

- Distance to default  $-1.13$  (0.036)
- Trailing 1-year stock return  $-0.694$  (0.075)
- 3-month T-Bill rate  $-0.105$  (0.021)
- Trailing 1-year SP500 return  $1.20$  (0.289)

## Conditional Likelihood Function

- For a given firm  $i$ , the  $X$ -conditional likelihood of survival from  $t(i)$  to  $\tau_i$ , censored at  $T(i)$ , is

$$\begin{aligned}\mathcal{L}_i(\tau_i | X; \beta) &= q_{t(i)} q_{t(i)+1} \cdots q_{T(i)-1}, & \text{if } \tau_i > T(i), \\ &= q_{t(i)} q_{t(i)+1} \cdots q_{\tau_i-2} (1 - q_{\tau_i-1}), & \text{if } \tau_i \leq T(i),\end{aligned}$$

where  $q_t = e^{-\Lambda(X_{it}; \beta)}$  is the one-period  $X$ -conditional survival probability.

- The doubly-stochastic assumption implies that the  $X$ -conditional likelihood of  $\tau_1, \dots, \tau_n$ , as censored, is  $\mathcal{L}(\tau | X; \beta) = \prod_i \mathcal{L}_i(\tau_i | X; \beta)$ .