

Merton-model Approach to Valuing Correlation Products

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Binomial with Merton Model

- Important method for calculating distribution of loan losses:
 - ✓ widely used in banking
 - ✓ used by Basel II regulations to set bank capital requirements
- Linked to distance-to-default analysis

Mixed Binomial: Using Merton's Models as Mixing Distribution

- In Merton model value of risky debt depends on *firm value* and *default risk is correlated because firm values are correlated* (e.g., via common dependence on *market factor*).
- Value of firm i at time T :

$$V_{T,i} = V_i \exp\left(\underbrace{(\mu_i - (1/2)\sigma_{V,i}^2)}_{\text{expected value of } R^C} T \right) + \underbrace{\sigma_{V,i}}_{\text{surprise in } R^C} \sqrt{T} \varepsilon_i \quad \text{where } \varepsilon_i \sim N(0,1)$$

- We will assume that *correlation between firm values* arises because of correlation between surprise in individual firm value (ε_i) and *market factor (m)*

Mixed Binomial: Using Merton's Models as Mixing Distribution

- Suppose correlation between each firm's value and the market factor is the same and equal to $\sqrt{\rho}$.
- This means that we may model correlation between the ε 's as

$$\varepsilon_i = \sqrt{\rho}m + \sqrt{1-\rho}v_i, \quad i = 1, \hat{W}N$$

and

$$\text{corr}(\varepsilon_i, \varepsilon_j) = \rho$$

- Where m and v_i are independent $N(0, 1)$ random variables and ρ is common to all firms
- Notice that if $v_i \sim N(0, 1)$ and $m \sim N(0, 1)$ then $\varepsilon_i \sim N(0, 1)$

Structural Approach, contd.

- From our analysis of *distance-to-default*, we know that under the Merton Model a firm defaults when:

$$\varepsilon_i - R_{D,i} - \left(\mu_i - \frac{1}{2} \sigma_{V,i}^2 \right) T / \sigma_{V,i} \sqrt{T} \text{ where } R_{D,i} = \ln(B_i / V_i)$$

- The *unconditional* probability of default, p , is therefore:

$$p = \text{Prob} \left[\varepsilon_i \leq \frac{R_{D,i} - \left(\mu_i - \frac{1}{2} \sigma_{V,i}^2 \right) T}{\sigma_{V,i} \sqrt{T}} \right] = N \left[\frac{R_{D,i} - \left(\mu_i - \frac{1}{2} \sigma_{V,i}^2 \right) T}{\sigma_{V,i} \sqrt{T}} \right]$$

- In this model we assume that the *default probability*, p , is *constant across firms*

Structural Approach to Correlation – the Idea

- Working out the distribution of portfolio losses directly when the ε 's are correlated is not easy
- But, if we work out the *distribution conditional on the market shock*, m , then we can exploit the fact that the remaining shocks are independent and work out the portfolio loss distribution

Structural Approach, contd.

- The shock to the return, ε_i , is related to the common and idiosyncratic shocks by:

$$\varepsilon_i = \sqrt{\rho}m + \sqrt{1-\rho}v_i$$

- Default occurs when:

$$\varepsilon_i = \sqrt{\rho}m + \sqrt{1-\rho}v_i < \frac{R_{D,i} - (\mu_i - \frac{1}{2}\sigma_{V,i}^2)}{\sigma_{V,i}\sqrt{T}} = N^{-1}(p)$$

or

$$v_i < \frac{N^{-1}(p) - \sqrt{\rho}m}{\sqrt{1-\rho}}$$

The Default Condition

$$v_i < \frac{N^{-1}(p) - \sqrt{\rho}m}{\sqrt{1 - \rho}}$$

- A large value of m means a “good” shock to the market (high asset values)
- The larger the value of m the more negative the idiosyncratic shock, v_i , has to be to trigger default
- The higher the correlation, ρ , between the firm shocks, the larger the impact of m on the critical value of v_i .

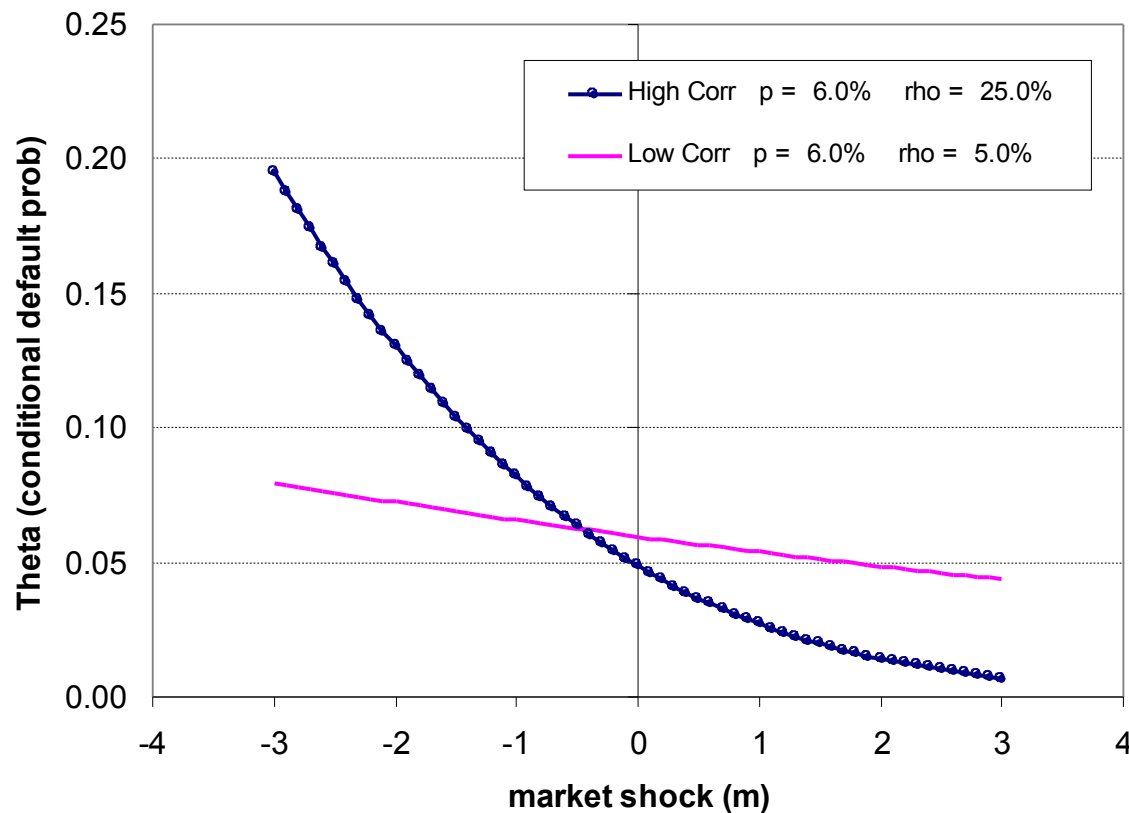
Structural Approach, contd.

- Conditional on the realisation of the common shock, m , the probability of default is therefore:

$$\begin{aligned}\text{Prob}(\text{default} | m) &= \text{Prob} \left(v_i < \frac{N^{-1}(p) - \sqrt{\rho}m}{\sqrt{1-\rho}} \right) \\ &= N \left(\frac{N^{-1}(p) - \sqrt{\rho}m}{\sqrt{1-\rho}} \right) = \theta(m), \text{ say} \\ \text{and therefore} &= \frac{N^{-1}(p) - \sqrt{\rho}m}{\sqrt{1-\rho}} = N^{-1}(\theta)\end{aligned}$$

The relation between m and θ

- For a given market shock, m , θ gives the conditional probability of default on an individual loan



Implications of Conditional Independence

- For a given value of m , as the number of loans in the portfolio $\rightarrow \infty$, the *proportion* of loans in the portfolio that default converges *to the probability θ*
- The probability that the loan-loss proportion, L , is $< \theta$ is therefore:

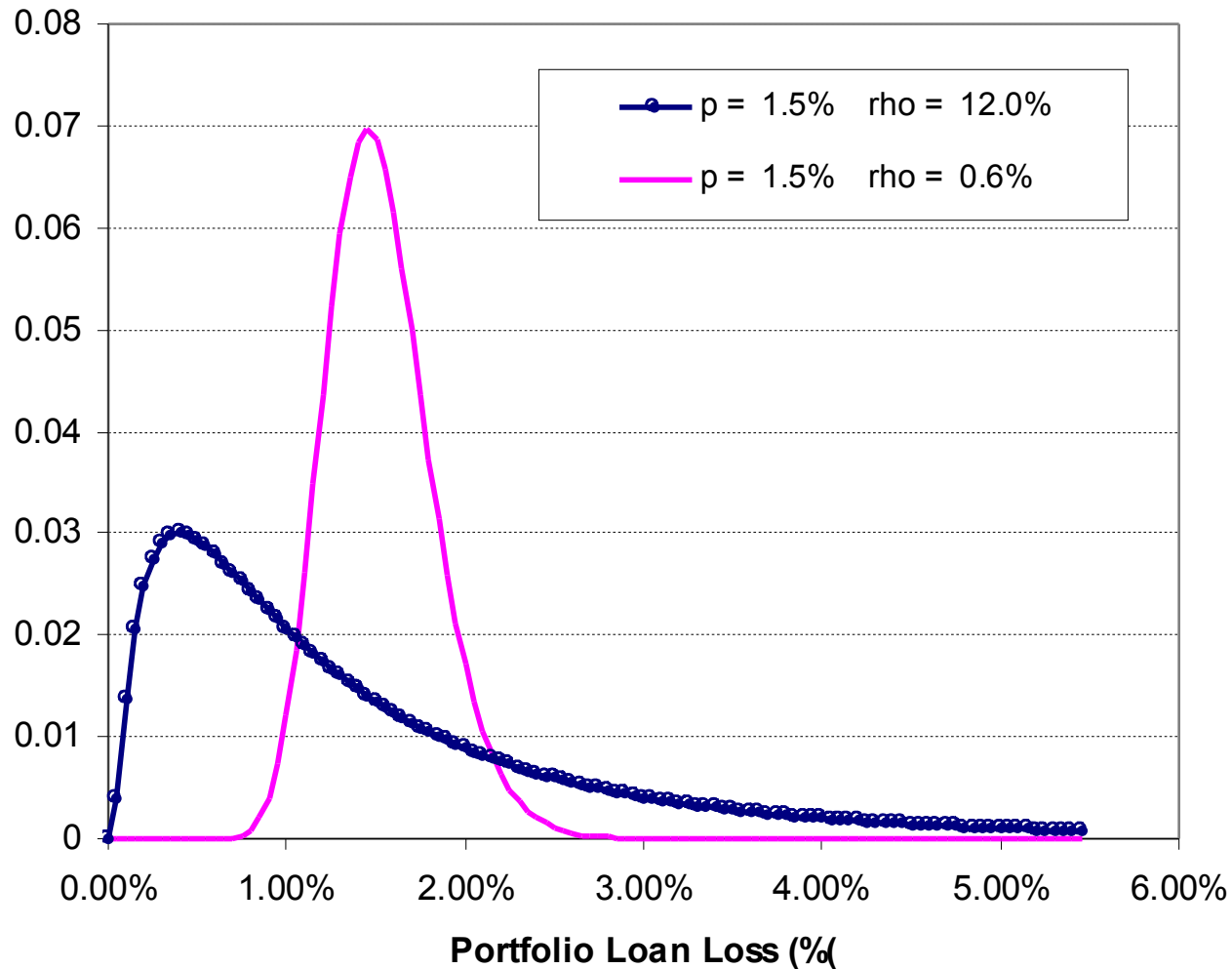
$$\begin{aligned}
 \text{Prob}(L < \theta) &= \text{Prob} \left(\frac{N^{-1}(p) - \sqrt{\rho}m}{\sqrt{1-\rho}} < N^{-1}(\theta) \right) \\
 &= \text{Prob} \left(m \frac{1}{\sqrt{\rho}} \left(N^{-1}(p) - N^{-1}(\theta)\sqrt{1-\rho} \right) < 0 \right) \\
 &= \text{Prob} \left(m \frac{1}{\sqrt{\rho}} \left(\sqrt{1-\rho} N^{-1}(\theta) - N^{-1}(p) \right) > 0 \right) \\
 &= N \left(\frac{1}{\sqrt{\rho}} \left(\sqrt{1-\rho} N^{-1}(\theta) - N^{-1}(p) \right) \right)
 \end{aligned}$$

Loan Loss Distribution – Structural Model

$$\text{Prob}(L = \theta) = N \frac{1}{\sqrt{\rho}} \left(\sqrt{1 - \rho} N^{-1}(\theta) - N^{-1}(p) \right)$$

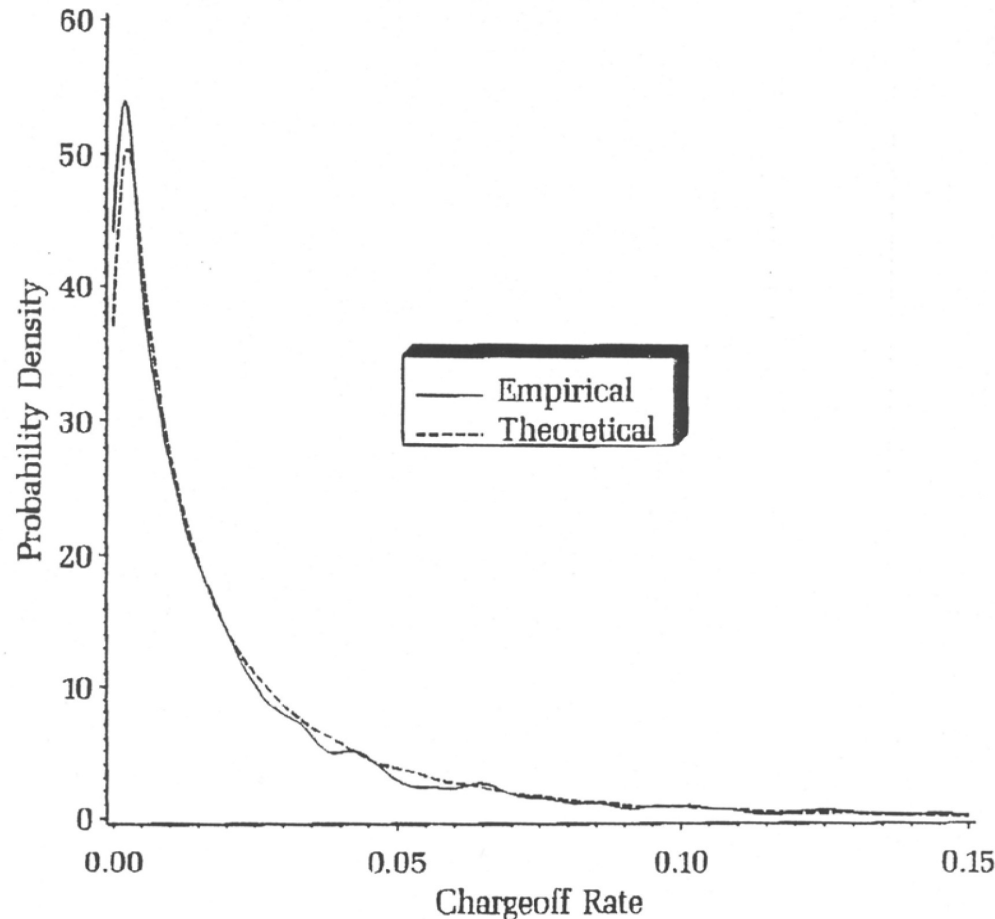
- This result gives the distribution of the *fraction of loans that default* in a *well diversified homogeneous portfolio* where the correlation in default comes from dependence on a *common factor*
- *Homogeneity* means that each loan has:
 - ✓ the *same default probability*, p
 - ✓ (implicitly) the *same loss-given-default*
 - ✓ the *same correlation*, ρ , across different loans
- The distribution has *two parameters*
 - ✓ default probability, p
 - ✓ correlation, ρ

Loan Loss Distribution with $p = 1\%$ and $\rho = 12\%$ and 0.6%



Example of Vasicek formula Applied to Bank Portfolio

Loan Losses as a Fraction of Total Loan Portfolio
(Commercial/Industrial loans, banks < \$300 million assets, 1984 - 1991)



Source: Vasicek