

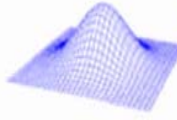
# **An Empirical Comparison of Default Risk Forecasts from Alternative Credit Rating Philosophies**

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2nd International Conference on Credit Risk

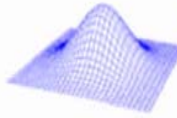
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Montreal



# Agenda

- **The Problem**
- **Through-the-Cycle vs. Point-in-Time Forecasts**
- **Backtesting**
- **Implications and Discussion**



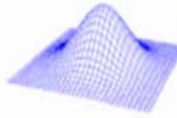
# The Problem

## Regarding Ratings

- What is a Credit Rating?
- How to “link” Ratings and (forecasts for) Probabilities of Default?
- Which kind of PD forecasts to take for regulatory capital?
- Potential consequences w.r.t.
  - accurateness of risk capital
  - procyclicality

## The paper seeks to

- analyse and compare TTC and PIT Rating Philosophies w.r.t. their implied forecasts of PDs
- show an application of how to “backtest” these forecasts
- discuss consequences for banks



# TTC vs. PIT Forecasts

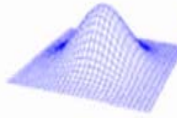
- Starting point: Standard & Poors Default Data from 1982 up to 2000-> **“Through-the-Cycle Rating”**
- Estimate their implied PD and asset correlation by the latent random effects probit model (see e.g. Gordy/Heitfield, 2000, Hamerle/Liebig/Rösch, 2002):

$$R_{it} = b F_t + \sqrt{1-b^2} U_{it}$$

$$\lambda(f_t) = P(R_{it} < \beta_0 | f_t) = P\left( U_{it} < \frac{\beta_0 - b f_t}{\sqrt{1-b^2}} \right) = \Phi\left( \frac{\beta_0 - b f_t}{\sqrt{1-b^2}} \right)$$

p-values in parentheses

Grade	$b$	$\beta_0$
BB	0.229 (0.003)	-2.290 (<0.001)
B	0.210 (<0.001)	-1.628 (<0.001)
CCC	0.256 (0.004)	-0.809 (<0.001)



# TTC vs. PIT Forecasts

- **(Mimicking) Point-in-Time-Rating:** “Add” current information to the rating (e.g. macroeconomic condition, “z”-factors)
- Estimate implied PD and asset correlation by the extended latent random effects probit model (see Hamerle/Liebig/Rösch, 2002, Hamerle/Liebig/Rösch, 2003):

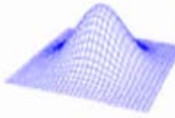
$$\lambda(z_t, f_t) = P\left( U_{it} < \frac{\beta_0 + \beta' z_t - b f_t}{\sqrt{1-b^2}} \right) = \Phi\left( \frac{\beta_0 + \beta' z_t - b f_t}{\sqrt{1-b^2}} \right)$$

p-values in parentheses

Grade	<i>b</i>	$\beta_0$	$\beta_1$	$\beta_2$	$\beta_3$	$B_4$
BB	0.085 (0.333)	-2.791 (<0.001)	0.676 (<0.001) FEDR_1			
B	0.093 (0.010)	-2.104 (<0.001)	3.137 (0.034) DR_B_1	1.453 (<0.001) DSER_2	-0.387 (0.009) DIND_1	7.449 (0.022) RAT4chg_1
CCC	0.048 (0.815)	0.294 (0.264)	-1.798 (<0.001) UNEM_1			



# Backtesting

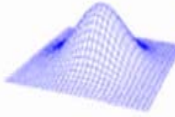


- Altogether we obtain 5 years of out-of-time loss distribution forecasts for each rating and subsequent default realizations
- With these sets the “Berkowitz” backtest is conducted (see Berkowitz, JBES, 2001, Frerichs/Löffler, JoR, 2003):

$$\hat{g}(d_{T+1}) = \begin{cases} \binom{N_{T+1}}{d_{T+1}} \cdot \int_{-\infty}^{+\infty} \left[ \hat{\lambda}(f_{T+1})^{d_{T+1}} \cdot [1 - \hat{\lambda}(f_{T+1})]^{(N_{T+1} - d_{T+1})} \right] \varphi(f_{T+1}) df_{T+1} & d_{T+1} = 0, 1, 2, \dots, N_{T+1} \\ 0 & \text{otherwise} \end{cases}$$

$$x_{T+1} = \int_{-\infty}^{q_{T+1}} \hat{g}(u) du \quad \text{is i.i.d. uniform if the forecasted density is correct}$$

# Backtesting



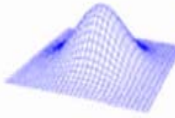
$z_{T+1} = \Phi^{-1}(x_{T+1}) = \Phi^{-1}\left(\int_{-\infty}^{q_{T+1}} \hat{g}(u) du\right)$  is i.i.d. standard normal if the forecasted density is correct

The Likelihood-Ratio is

$$LR = -2 \cdot (L(0,1) - L(\hat{\mu}, \hat{\sigma})) =$$
$$= -2 \cdot \left( \left( -\frac{L}{2} \log 2\pi - \sum_{l=1}^L \frac{z_{T+l}^2}{2} \right) - \left( -\frac{L}{2} \log 2\pi - \frac{L}{2} \log \hat{\sigma}^2 - \sum_{l=1}^L \frac{(z_{T+l} - \hat{\mu})^2}{2\hat{\sigma}^2} \right) \right)$$

$$\hat{\mu} = \frac{1}{L} \sum_{l=1}^L z_{T+l} \quad \hat{\sigma}^2 = \frac{1}{L} \sum_{l=1}^L (z_{T+l} - \hat{\mu})^2$$

which is asymptotically Chi-squared (2) distributed if the forecasted distributions are correct

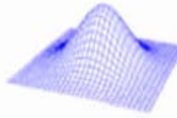


## Testresults

Grade	LR „Naive“	LR Through the Cycle	LR Point in Time
BB	5.94 (0.05)	5.48 (0.06)	1.29 (0.52)
B	6.42 (0.04)	3.24 (0.20)	0.30 (0.86)
CCC	2.30 (0.32)	0.29 (0.87)	4.22 (0.12)

p-values in parentheses

- Power of the test is checked via Monte-Carlo simulations

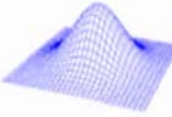


# Implications and Discussion

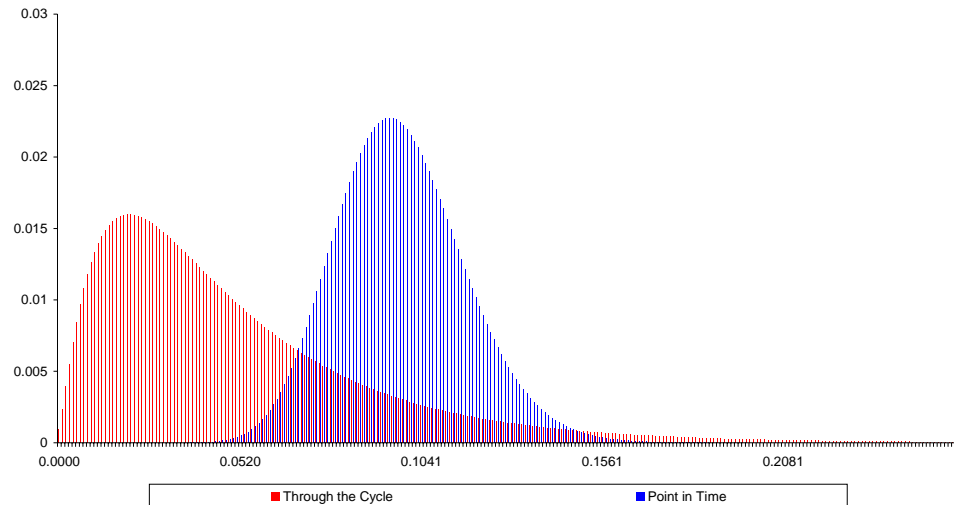
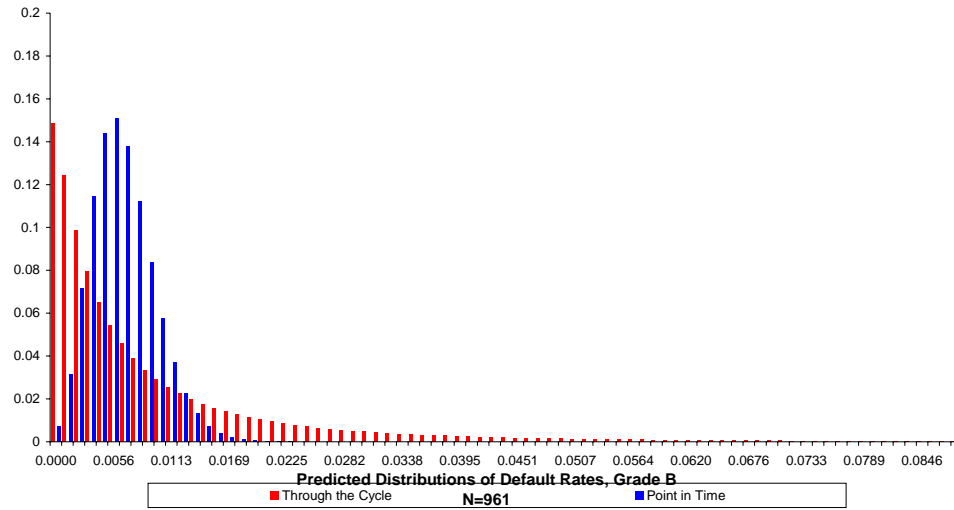
## The Rating Puzzle

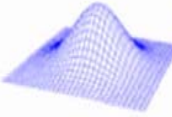
- Rating, PD, and **correlation** are strongly interweaved
- Through the Cycle rating states similar “risk” over a **multi-year** horizon for borrowers within the same grade (e.g. BB)
- These borrowers may well exhibit very different **one-year** probabilities of default (in the cross-section and over time)
- Hence, one-year default rates of a grade over time are generated by **different default probabilities** and thus are noisy
- Consequently, if historical default rates from TTC are used as forecasts, their noise should be taken into account (by higher correlation)
- PIT PD forecasts are less noisy -> lower correlation!

# Implications and Discussion



Predicted Distributions of Default Rates, Grade BB  
N=887





## Implications for Banks and Regulators

- PD forecasts from TTC Ratings are relatively stable over time
- But they are noisy forecasts for true one-year PDs and thus rather unreliable regarding true economic risk!
- TTC Rating PD forecasts may be compatible with (high) asset correlations in the Basel Accord
- PIT PD forecasts fluctuate over the cycle
- But they are more precise estimates for one-year PDs and imply lower asset correlations!
- Procyclicality due to PIT is coherent with economic risk
- If Procyclicality is unwanted: Regulatory guidelines could take this into account: Change the outputs (capital charge rules) instead of the inputs (PD forecasts)