

# Estimation of transition probabilities of credit ratings for several companies

Gan Chew Peng and Pooi Ah Hin

Citation: AIP Conference Proceedings **1782**, 050007 (2016); doi: 10.1063/1.4966097 View online: http://dx.doi.org/10.1063/1.4966097 View Table of Contents: http://scitation.aip.org/content/aip/proceeding/aipcp/1782?ver=pdfcov Published by the AIP Publishing

Articles you may be interested in Estimation of transition probabilities of credit ratings AIP Conf. Proc. **1691**, 050005 (2015); 10.1063/1.4937087

Calculating rate constants and committor probabilities for transition networks by graph transformation J. Chem. Phys. **130**, 204111 (2009); 10.1063/1.3133782

Theoretical Study of Several Oscillator Strengths and Lifetimes of Germanium, Thallium and Bismuth. Measures of Some Relative Transition Probabilities AIP Conf. Proc. **1058**, 152 (2008); 10.1063/1.3026428

Accurate estimation of the density of states from Monte Carlo transition probability data J. Chem. Phys. **125**, 144905 (2006); 10.1063/1.2358345

Experimental Stark widths, shifts, and transition probabilities of several ArII lines AIP Conf. Proc. **386**, 151 (1997); 10.1063/1.51832

# Estimation of Transition Probabilities of Credit Ratings for Several Companies

Gan Chew Peng<sup>1, a)</sup> and Pooi Ah Hin<sup>2, b)</sup>

<sup>1,2</sup>Sunway University Business School, Centre for Actuarial Studies, Applied Finance and Statistics, No. 5, Jalan Universiti, Bandar Sunway, 47500, Subang Jaya, Selangor.

<sup>a)</sup> Corresponding author: chewpengg@sunway.edu.my <sup>b)</sup> ahhinp@sunway.edu.my

**Abstract.** This paper attempts to estimate the transition probabilities of credit ratings for a number of companies whose ratings have a dependence structure. Binary codes are used to represent the index of a company together with its ratings in the present and next quarters. We initially fit the data on the vector of binary codes with a multivariate power-normal distribution. We next compute the multivariate conditional distribution for the binary codes of rating in the next quarter when the index of the company and binary codes of the company in the present quarter are given. From the conditional distribution, we compute the transition probabilities of the company's credit ratings in two consecutive quarters. The resulting transition probabilities tally fairly well with the maximum likelihood estimates for the time-independent transition probabilities.

# **INTRODUCTION**

Credit risk is the risk of loss due to the probability that an obligor is unable or unwilling to pay its credit. In the New Basel Capital Accord (Basel II), the matrix of transition probabilities between rating classes plays an essential role in the supervision of banking activities.

Multivariate Markov chain model for credit ratings has been used to estimate credit transition matrices ([1] - [3]). The method based on multivariate Markov chain was applied to the rating data of 15 industries in Taiwan to estimate the probability of the *j*-th ( $1 \le j \le 15$ ) company's credit rating in the next quarter as a function of the ratings of all the companies in the current quarter [4]. Multivariate power-normal mixture distribution was used to fit the rating data of 10 companies in the electronic industries in Taiwan in order to estimate the transition probability of the future credit rating of a given company in the next quarter as a function of all the 10 companies' credit ratings in the current quarter [5].

In the situation in which the number of companies under investigation is fairly large but the number of quarters spanning the dataset is fairly small, we may not be able to estimate the transition probability of the future credit rating of a given company in the next quarter as a function of all the companies' credit ratings in the current quarter. In this paper, we show that in such situation, we are able to estimate the transition probability of the future credit rating of the given company as a function of only the given company's rating in the present quarter. The procedure involves the use of the following binary codes.

The index of the *i*-th company in a group of N companies may be represented by the N - 1 binary codes  $0 \cdots 010 \cdots 0$  in which the value "1" takes the *i*-th position for  $1 \le i \le N - 1$ , while the N-th company may be represented by N - 1 zeros. In the case when the number of possible credit ratings is M, the rating j of a company may be represented by the M - 1 binary codes  $0 \cdots 010 \cdots 0$  in which the value "1" takes the *j*-th position for  $1 \le j \le M - 1$ , while the rating M may be represented by M - 1 zeros.

The 4th International Conference on Quantitative Sciences and Its Applications (ICOQSIA 2016) AIP Conf. Proc. 1782, 050007-1–050007-5; doi: 10.1063/1.4966097 Published by AIP Publishing. 978-0-7354-1444-0/\$30.00

#### 050007-1

Consider a vector  $\mathbf{y}$  of [N-1+2(M-1)] components consisting of the codes for the index of a company together with those for the company's ratings in the present and next quarters. Initially we fit the data for  $\mathbf{y}$  with a multivariate power-normal distribution. When the rating of a given company in the present quarter is given, we use the fitted multivariate power-normal distribution to find a conditional distribution for the codes of the company's rating in the next quarter. From the conditional distribution, we compute the transition probability of the company's rating in the next quarter. The resulting transition probabilities are found to agree fairly well with those obtained by the method of maximum likelihood.

The layout of the paper is as follows. In Section 2, we describe the computation of the transition probabilities of credit ratings of M companies. Section 3 presents the numerical results for the transition probabilities obtained from the data on the quarterly credit ratings of 10 companies in 15 years taken from the database of the Taiwan Economic Journal (TEJ). Section 4 concludes the paper.

# COMPUTATION OF TRANSITION PROBABILITIES USING MULTIVARIATE POWER-NORMAL DISTRIBUTION

Consider the following vector **y** consisting of *k* correlated random variables:

$$\mathbf{y} = \boldsymbol{\mu} + \mathbf{H}\boldsymbol{\varepsilon} \tag{1}$$

where **H** is an orthogonal matrix,  $\mu = E(\mathbf{y})$ , and  $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_k$  are uncorrelated random variables. Furthermore, assume that  $\varepsilon_i$  can be expressed as

$$\varepsilon_{i} = \sigma_{i} [\widetilde{\varepsilon}_{i} - E(\widetilde{\varepsilon}_{i})] / \{ \operatorname{var}(\widetilde{\varepsilon}_{i}) \}^{\frac{1}{2}}$$
(2)

where  $\sigma_i > 0$  is a constant, and  $\tilde{\varepsilon}_i$  has a power-normal distribution [6] with parameters  $\lambda_i^+$  and  $\lambda_i^-$  [7].

From the multivariate power-normal distribution for the vector  $\mathbf{y}$  of [N-1+2(M-1)] binary codes, we generate a large number  $N_s$  of the values of  $\mathbf{y}$ . The components of  $\mathbf{y}$  may be divided into three groups of which group 1 consists of the initial N-1 components, group 2 consists of the next M-1 components and group 3 consists of the last M-1 components. Let  $N_1 = N-1$  and  $N_2 = N_3 = M-1$ . The distance of the  $N_j$  components  $y_1^{(j)} y_2^{(j)} \cdots y_{N_j}^{(j)}$ 

in Group *j* of **y** from the vector of  $N_j$  components  $(0, \dots, 0, 1, 0, \dots, 0)$  with the only component of "1" taking the *i*-th position is given by

$$D_{i}^{(j)} = \left\{\sum_{\substack{i=1\\i\neq i}\\i\neq i}^{N_{j}} y_{i}^{(j)2} + \left(y_{i}^{(j)} - 1\right)^{2}\right\}^{1/2}$$
(3)

Let  $D_0^{(j)} = \sum_{i=1}^{N_j} y_{i'}^{(j)2}$ . Let  $i^{(j)*}$  be the value of *i* such that among  $D_0^{(j)}, D_1^{(j)}, D_2^{(j)}, \dots$ , and  $D_{N_j}^{(j)}, D_{i'^{(j)*}}^{(j)}$  has the minimum value. Next let

$$i^{(j)} = \begin{cases} i^{(j)*}, 1 \le i^{(j)*} \le N_j \\ N_j + 1, \quad i^{(j)*} = 0 \end{cases}$$
(4)

Suppose there are  $n_{i,r_1,r_2}$  generated values of **y** of which  $i^{(1)} = i$  and  $i^{(2)} = r_1$  and  $i^{(3)} = r_2$  and there are  $n_{i,r_1}$  generated values of **y** of which  $i^{(1)} = i$  and  $i^{(2)} = r_1$ . Then for company  $i (1 \le i \le N)$ , the probability of the transition from rating  $r_1$  to  $r_2$  is estimated by

$$P_{i,r_1,r_2} = \frac{n_{i,r_1,r_2}}{n_{i,r_1}}, \quad 1 \le r_1, r_2 \le M .$$
(5)

#### 050007-2

### NUMERICAL RESULTS

Consider the quarterly credit ratings of 10 companies in the electronic sector in 15 years (from 1999 to 2013) taken from the database of the Taiwan Economic Journal. The credit rating ranges from 1 (lowest default risk) to 9 (highest default risk), and 10 represents the default state. Thus, the values of N and M (see Section 1) are both equal to 10 and the vector y (see Section 1) has 27 components. A multivariate power-normal distribution is first fitted to the 590 observed values of y. Some of the estimated transition probabilities based on the multivariate power-normal distribution for the 10 companies are presented in Table 1 along with those estimated by using the maximum likelihood procedure. Let  $M_{ij}$  be the number of times when the rating of company i is j in the current quarter. When  $M_{ij}$  is large, the maximum likelihood procedure would be able to give fairly good estimate of the transition probability of the rating of company i in the next quarter given that the rating of company i in the current quarter is j. Table 1 shows that the transition probabilities estimated by the two procedures are fairly comparable when the corresponding  $M_{ij}$ is not too small.

**TABLE 1.** Transition probability estimated by maximum likelihood procedure, and transition probability (in parentheses) based on multivariate power-normal distribution  $(N_s = 100, 000)$ .

	(a)	1	-		
Rating This Quarte	er (j)	Rating 1 7	Next Qu 8	arter	$M_{1j}$
7	0.9	0773 C	0.0227	0 (0.0200)	44
8	,	0	0.8	0.2 (0.0671)	5
9	Ċ	0,1 (0) (0) (0) (0) (0) (0) (0) (0) (0) (0)	0 (0)	(0.0071) 0.9 (0.794)	10
		Compan	y 2		
Rating This Quarte	er ( <i>i</i> )	<b>Rating Next Quarter</b>			
Training This Quality		8	9	10	$M_{2j}$
8			.1667 .0556)	$\begin{pmatrix} 0\\ (0) \end{pmatrix}$	6
9			.8667	0.0667	15
			.8223)	(0.1293)	15
10			.0263	0.9737 (0.8542)	38
	(c)			(0.00.12)	
		Rating I		arter 🕠	
<b>Rating This Quarter</b>		3	(ene Qui	arter <sub>M<sub>3</sub></sub>	j
3		0.9 (0.8044)	-	$\frac{.1}{2235}$ 10	)
4		0 (0)		l 49 946) 49	)
	(d)	Compan	y 4		
g This Quarter (j)		Rating I	-		
S mis Quarter ()	7	8			10
7	l (0.8737)	0 (0)		-	0 (0)
0	0.5	0.5	·	· · ·	0
8	(0.1056)	(0.5466)		)) (0	(0)

9	0			.9091	0	11	
)	(0	) (0.0		.8237)	(0)	11	
10	0				.9545	22	
	(0	) (	(0) (0.	05697) (0	.8158)		
		()	-				
			mpany 5				
Rating This Quarter (j)		Ra	ting Next Q	uarter		$M_{5i}$	
	5	6	7	8	9	51	
	1	0	0	0	0		
5	(0.9061)	(0)	(0)	(0)	(0)	20	
6	0.25	0.75	0	0	0		
0	(0.05)	(0.6333	) (0)	(0)	(0)	4	
7	0	0.5	0.5	0	0	2	
1	(0)	(0.0035	/ .		(0)	2	
8	0	0	0.0833			12	
0	(0)	(0)	(0.0578	· · ·			
9	0	0	0	0.0476		//	
	(0)	(0)	(0)	(0.0244	) (0.80	1) -1	
		(f) Co	mpany 6				
			ting Next Q	uarter		_	
Rating This Quarter (j)		6	7	8	$M_{6j}$		
(		1	0	0	1.4	-	
6		(0.7672)	(0)	(0)	14		
7		0.0526	0.8421	0.1053	19		
1		(0.0173)	(0.8348)		) 19		
8		0	0.0769	0.9236	26		
0		(0)	(0.0754)	(0.7148)	20	_	
		$(\alpha)$					
			mpany 7 ting Next Q	uarter		-	
Rating This Q	uarter (j)	7	8	9	$M_{7j}$		
7			0	0	22	-	
7		(0.8859)	) (0)	(0)	32		
8		0.5	0.5	0	2		
0	8		(0.5302)	· · · ·	2		
9	9		0.04	0.96	25		
			(0.0151)	) (0.8358)		_	
		(h) Co	mpany 8				
		Da	ting Next Q	uarter			
Rating Thi	Rating This Quarter (j) Rating Next Quarter $M_{8j}$						
	8	0	1	0	1		
	8		(0.0829)	(0)	1		
9		0 (0)	0.9444	0.0556	18		
	10		(0.8353)	(0.1136)	10		
			0.025	0.975	40		
		(0)	(0.072)	(0.8523)			
		(i) Co	mpany 9				
			ting Next Q	uarter			
Rating This Quarte	r (j)	5	6	7	8	$M_{9j}$	
5		1	0	0	0	26	
5	(0.9	308)	(0)	(0)	(0)	20	

6	0.05882	0.8824	0.0588	0	17
	(0.0426)	(0.7593)	(0.037)	(0)	
7	0	0.1111	0.7778	0.1111	9
	(0)	(0.0192)	(0.8219)	(0.0159)	
8	0	0	0.1429	0.8571	7
	(0)	(0)	(0.0584)	(0.539)	/
	(j)	Company 1	0		
Rating This	Quarter (i)	Rating Next Quarter		$M_{10j}$	
Kating This	ating This Quarter ()		10	10 j	
	9		0.1429	7	
7		(0.8029)	(0.1421)	/	
10		0.0100	0 0000		
1	n	0.0192	0.9808	52	

## **CONCLUSION**

The numerical results for the transition probabilities in Section 3 show that the fit given by the multivariate powernormal distribution is fairly satisfactory. A possible reason for the satisfactory fit is that we have used all the companies' rating data to estimate the transition probability matrix of a given individual company. A limitation of the computed transition probabilities is that they are assumed to be time-independent. To obtain time-dependent transition probabilities, we may first fit the data of m (m < 59) consecutive quarters with a multivariate power-normal distribution and obtain the transition probabilities for the ratings of the *i*-th company in the immediate future quarter for  $1 \le i \le 10$ . In this way we can perform an out-of-sample prediction of the future ratings.

# REFERENCES

- 1. M. Kijima, K. Komoribayashi, and E. Suzuki, Journal of Risk, 4, pp. 1-32 (2002).
- 2. T-K Siu, W-K Ching, S. E. Fung, and M. K. Ng, Quantitative Finance, 5(6), pp. 543-56 (2005).
- 3. S-L Lu, Applied Economics Letters, 16(11), pp. 1143-8 (2009).
- 4. S-L Lu, Advances in Management and Applied Economics, 2(4), pp. 243 (2012).
- C. P. Gan and A. H. Pooi. "Estimation of Transition Probabilities of Credit Ratings" in the 2<sup>nd</sup> Innovation and Analytics Conference & Exhibition 2015, AIP Conference Proceedings 1691, pp. 050005-1-7.
- 6. I. K. Yeo and R. A. Johnson, Biometrika, 87(4), pp. 954-959 (2000).
- 7. A. H. Pooi, Applied Mathematical Sciences, 6(115), pp. 5735-5748 (2012).