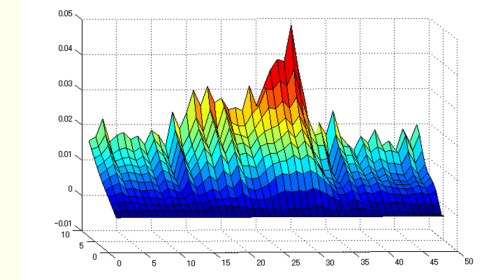


# Measuring Credit Risk of Individual CBs and Deriving Term Structure of Default Probabilities

-Market approach



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**(GSB, Meiji University)**  
**Zhu Wang (ZW System)**

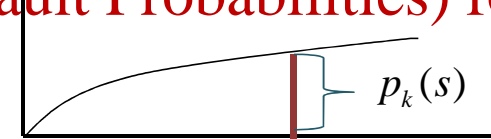
## Abstract

Via Kariya's cross-sectional CB model (2012) with GB model in Kariya, et al (2012), for individual CBs we define a **measure of CRPS (Credit Risk Price Spread)** for the  $k$ -th CB by

$$y_k^{(i)} = V_k - \hat{P}_k^{(i)}, \quad \hat{P}_k^{(i)} = \sum_{j=1}^{M(k)} C_k(s_{kj}) \bar{D}_k^{(i)}(s_{kj})$$

and derive the **TSDPs (Term Structure of Default Probabilities)** for the  $k$ -th firm;

$$p_k(s) = P(\tau_k \leq s)$$



- Market Prices are formed by investors with a future perspective on credit risks and take into account **bond attributes of GBs and CBs**
- **Credit-homogeneous groups** via R&I rating, cluster analysis and **FIR (fixed interval rating)** via S-CRPS are analyzed and compared where industry category is combined.
- **TSDPs** of these groups and some individual firms are derived. For TEPCO and Mitsubishi Corp, time series changes of TSDPs are described.

# Basic information on credit risk

## Forward-Looking vs Backward-Looking

**Backward-Looking**: model using past time series data over a period, *generated under different regimes*

- interest rates: statistical or econometric model
- credit data: past defaults and non-defaults, intensity model, classification, transition, logit-probit model

**Forward-Looking**: model using current (cross-sectional) market price data that reflects investors' future views, projection and perspectives on economic and financial movements of firms, given the past time series information

- interest rate: current GB prices, swap rates
- credit: current CB prices, CDS, stock prices,

## Information content of prices at $n$ of CBs

CB prices reflect

- **Investors' evaluations, views and perspectives** on the term structure of credit risk in the CBs over maturities, given past and current information
- The evaluation includes their considerations on the business portfolio structure of each firm.
- Hence prices at  $n$  of many CBs implicitly carry the investor's view on **the TSDP** for each specific features of firms, provided the CB market is efficient.
- **In short, the information contains investors' forward-looking evaluation on the TSDP of CBs for existing firms that have different portfolio of business lines.**

## GB prices and Attribute Effects

- **GB prices are formed on investors' different motives**
  - Does GB prices reflects **bond-attribute effects** ? :  
investors' coupon preference and maturity preference
  - **Motives** : **Investment (Buy & Hold) or Trading or both**  
on their future perspectives,
  - **Institutional Investors (life insurance, pension etc)** :  
ALM, Duration  $\Leftrightarrow$  coupon & maturity preferences
  - **The existence of these attribute effects denies the empirical validity of no-arbitrage theory in math finance (logic of trading motives)**
- **Testing Hypothesis of No Attribute Effect**

# Yield Approach vs Price Approach

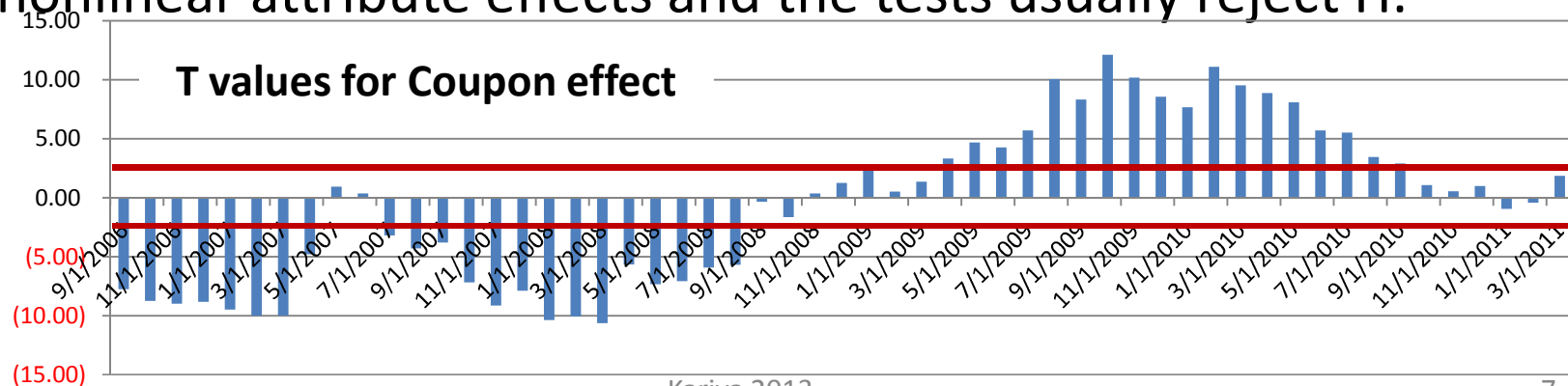
- Attribute effects are directly included in each prices
- Hypothesis of No attribute effect is tested either by price model or by yield model
- Traditionally it is often tested and rejected in a yield model (YTM, par rate) by assuming the linearity between yield and attributes
- But, this approach fails to specify a model that takes them into account in order to derive yields that get rid of the attribute effects
- Our bond model by price approach aims to price individual GBs .

# Yield Approach

- GB price is a convex function of YTM  $\Rightarrow$  YTM is a function of coupon, maturity and price, and it is convex in price
- A yield curve derived from all the GB prices is often regarded as risk-free or zero mean curve and the spreads between the mean yields and yields of individual bonds are used to test the hypothesis of no attribute effect by such a model

$$\text{Spread}(M) = \alpha + \sum_{i=1}^9 \beta_i MT_i + \gamma \text{Coupon}$$

- However, the mean curves themselves are distorted by the nonlinear attribute effects and the tests usually reject H.



# Testing Hypothesis of **$H_0$ : No Attribute Effect**

## H: No Attribute Effect

- 1) what is GB-equivalent CB price?
- 2) Investors' behaviors in the GB market
- 3) Maturity preference and coupon preference



# Attribute-dependent GB pricing model

Currently at  $t=0$ , future CFs are generated at

$$s_{g1} < s_{g2} < \dots < s_{gM(g)} \quad (g = 1, \dots, G)$$

$C_g(s)$  ; CF function = 0 unless  $s = s_{gj}$        $s_{aM(a)} = \max_g s_{gM(g)}$

$D_g(s)$  ; At-dependent stochastic DF       $0 < s \leq s_{aM(a)}$

$$P_g = \sum_{j=1}^{M(g)} C_g(s_{gj}) D_g(s_{gj}) \quad (g = 1, \dots, G)$$

$P_g$  realization  $\longleftrightarrow \{D_g(s) : 0 \leq s \leq s_{gM(g)}\}$

$$D_g(s) = \bar{D}_g(s) + \Delta_g(s)$$

$$P_g = \sum_{m=1}^{M(g)} C_g(s_{gm}) \bar{D}_g(s_{gm}) + \eta_g \quad \eta_g = \bar{C}'_g \bar{\Delta}_g$$

$$\tilde{y} = X \tilde{\beta} + \tilde{\eta} \quad Cov(\tilde{\eta}) = (Cov(P_g, P_h)) \equiv \sigma^2 \Phi(\rho, \xi)$$

## Attribute-Dependent Mean Discount Function

$(w_1, w_2, w_3)$        $z_{g1}=1, \quad z_{g2} : \text{maturity}, \quad z_{g3} : \text{coupon}$

M0 : (1,0,0); basic model with no attributes

M1 : (1,1,0): M0 + maturity effect,

M2 : (1,0,1); M0 + coupon effect

M3 :(1,1,1); M0 +maturity effect + coupon effect

$$\begin{aligned} \bar{D}_g(s) = & 1 + (\delta_{11}w_1z_{g1} + \delta_{12}w_2z_{g2} + \delta_{13}w_3z_{g3})s \\ & + \dots + (\delta_{p1}w_1z_{g1} + \delta_{p2}w_2z_{g2} + \delta_{p3}w_3z_{g3})s^p \end{aligned}$$

$$\bar{D}_g(s_{gj}) = E[D_g(s_{gj})] = E\left[\exp\left(-\int_0^{s_{gj}} f_{gs} ds\right)\right]$$

$$R_s = -\frac{1}{s} \log \bar{D}(s) \quad \text{Term structure of interest rates}$$

## Covariance Structure and GLS Estimation

$$[\text{Cov}(P_g, P_h)] = [\text{Cov}(\eta_g, \eta_h)] = \sigma^2 [\lambda_{gh} \varphi_{gh}] = \sigma^2 \Phi(\rho, \xi)$$

Attribute-dependent cov naturally induces duration effect

$$\varphi_{gh} = \sum_{j=1}^{M(g)} \sum_{m=1}^{M(m)} C_g(s_{gj}) C_h(s_{hm})$$

$$\lambda_{gh} = \begin{cases} \sigma^2 & (g = h) \\ \sigma^2 \rho \exp(-\xi |s_{gM(g)} - s_{hM(h)}|) & (g \neq h) \end{cases}$$

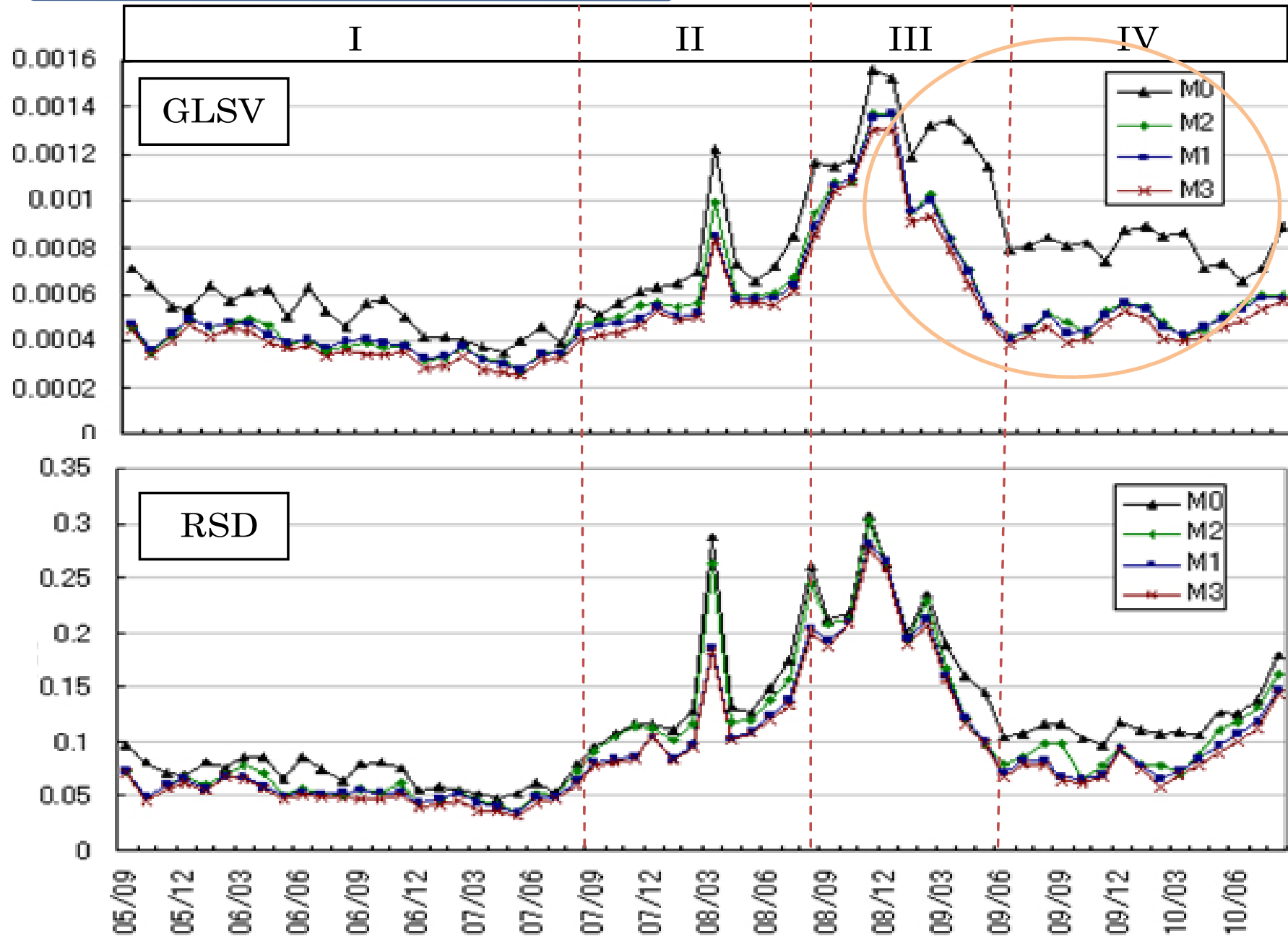
$$\underline{\tilde{y}} = X \underline{\tilde{\beta}} + \underline{\tilde{\eta}} \quad \text{GLS Estimation to minimize}$$

$$\psi(\underline{\tilde{\beta}}, \rho) = [\underline{\tilde{y}} - X \underline{\tilde{\beta}}]' [\Phi(\rho, \xi)]^{-1} [\underline{\tilde{y}} - X \underline{\tilde{\beta}}]$$

Kariya&Kurata(2004) Generalized Least Squares Wiley

# Performances of 4 models:M0-M3

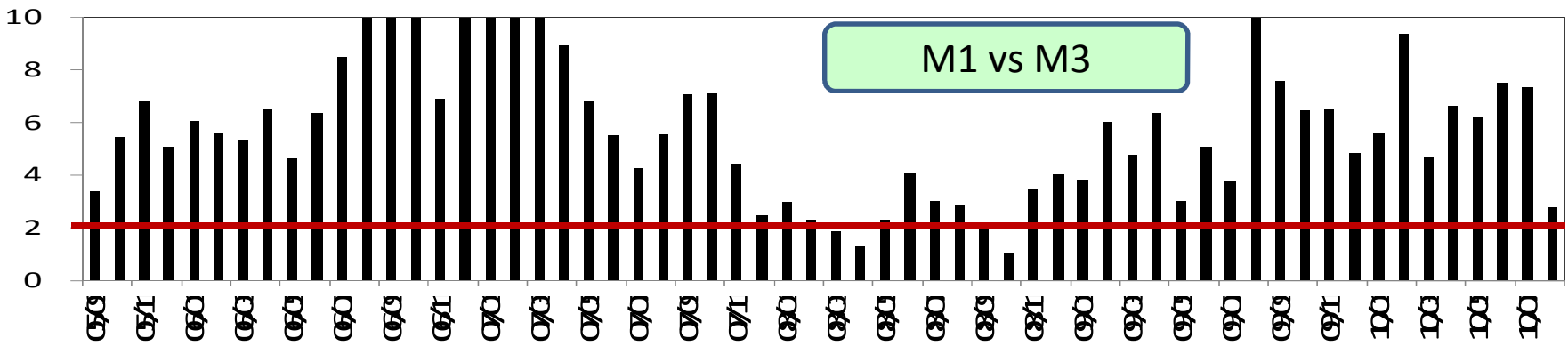
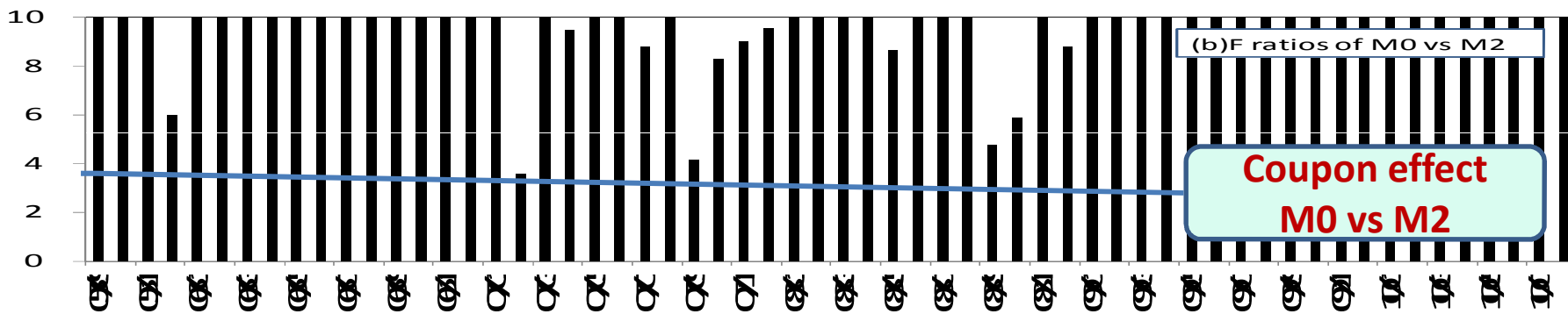
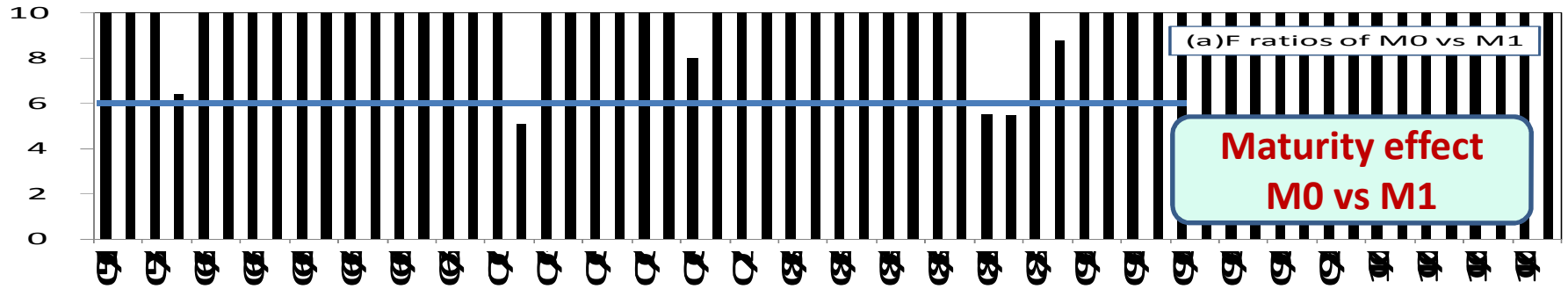
Monthly analysis 2005.09-2010.08



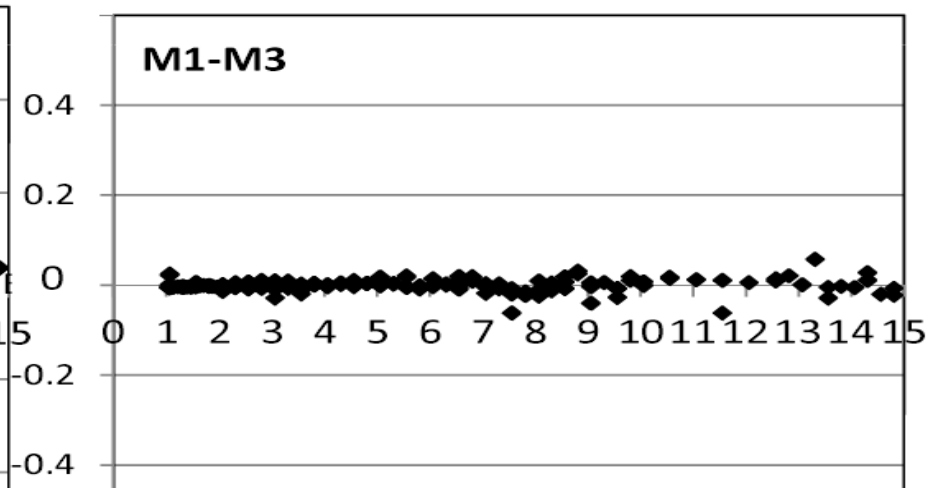
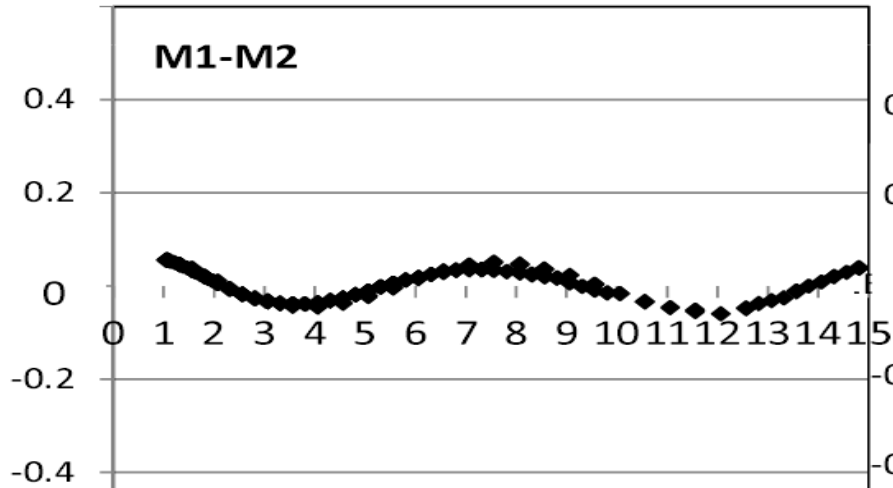
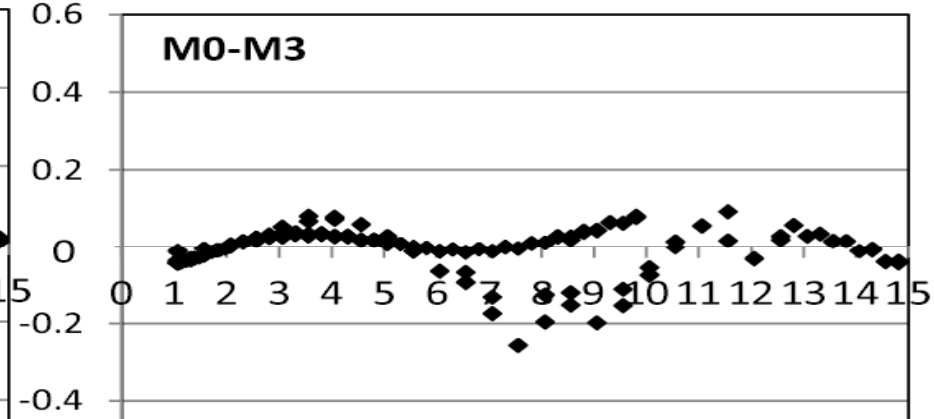
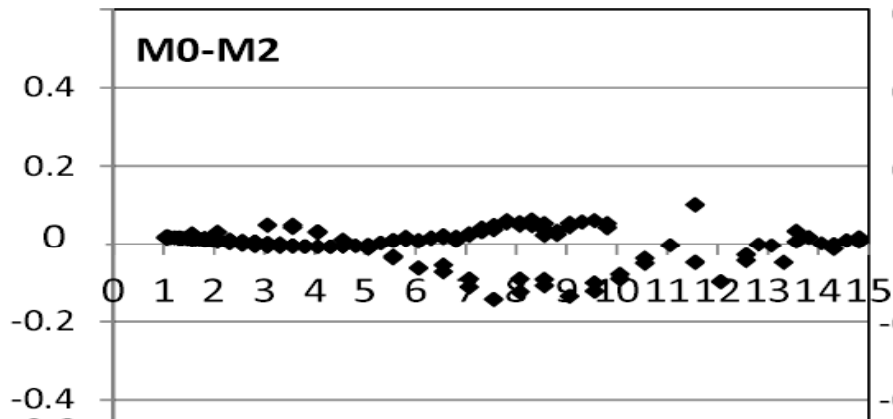
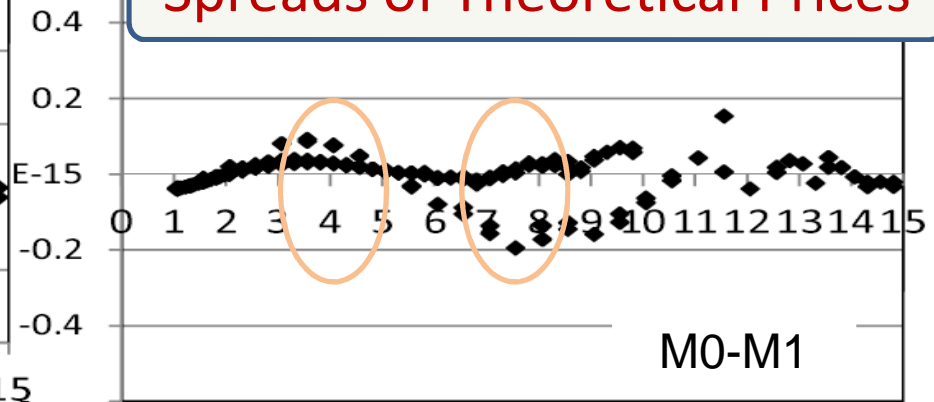
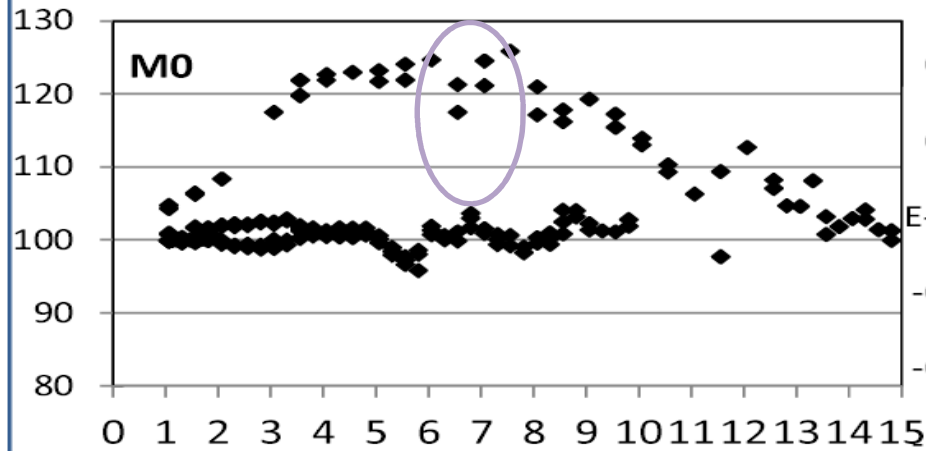
**F ratios of no attribute effects**  
**JGB**  $\hat{\psi}$

$$F \text{ ratio} = \frac{[QSR(0) - QSR(1)]/\#}{[QSR(1)/df]}$$

$$[QSR(0) - QSR(1)]/\# > 2[QSR(1)/df]$$



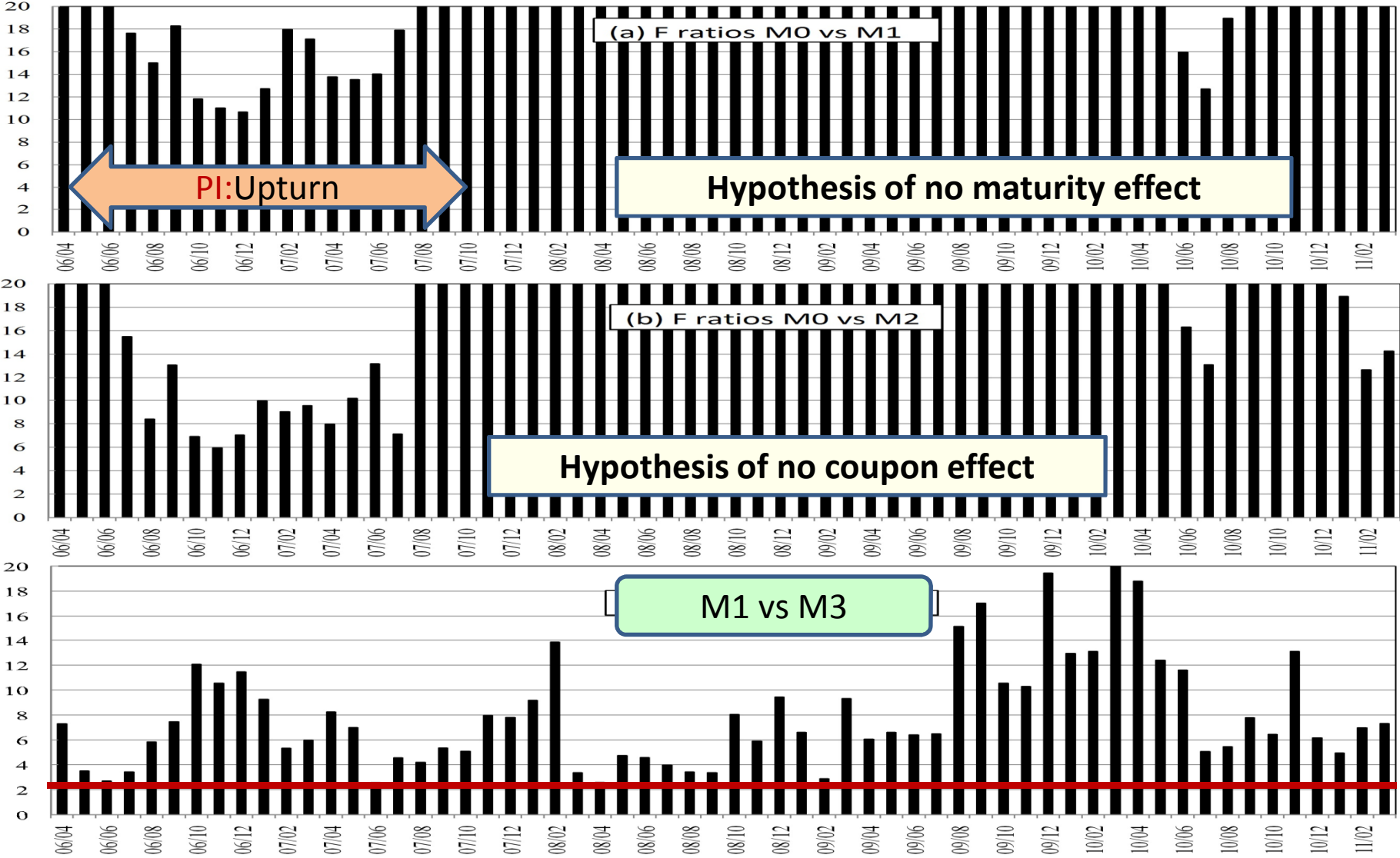
# Spreads of Theoretical Prices



**F-ratios of no attribute effects based on  $\psi$  USGB**

$$F \text{ ratio} = \frac{[QSR(0) - QSR(1)]/\#}{[QSR(1)/df]},$$

$$[QSR(0) - QSR(1)]/\# > 2[QSR(1)/df].$$



# **CRPS for individual CBs and R&I ratings**

**2010.8 Cross-section analysis**

- 1) Current CRPS Information
- 2) Do such categorical information as agency's rating and industry provide credit-homogeneous groups ?



## CRPS & S-CRPS for CBs

- Individual credit risk price spread (CRPS):  
**CRPS** = CB price — its GB-equivalent CB price  
 with the same coupon and maturity

Mean DFs  
 derived  
 From  
 GB prices



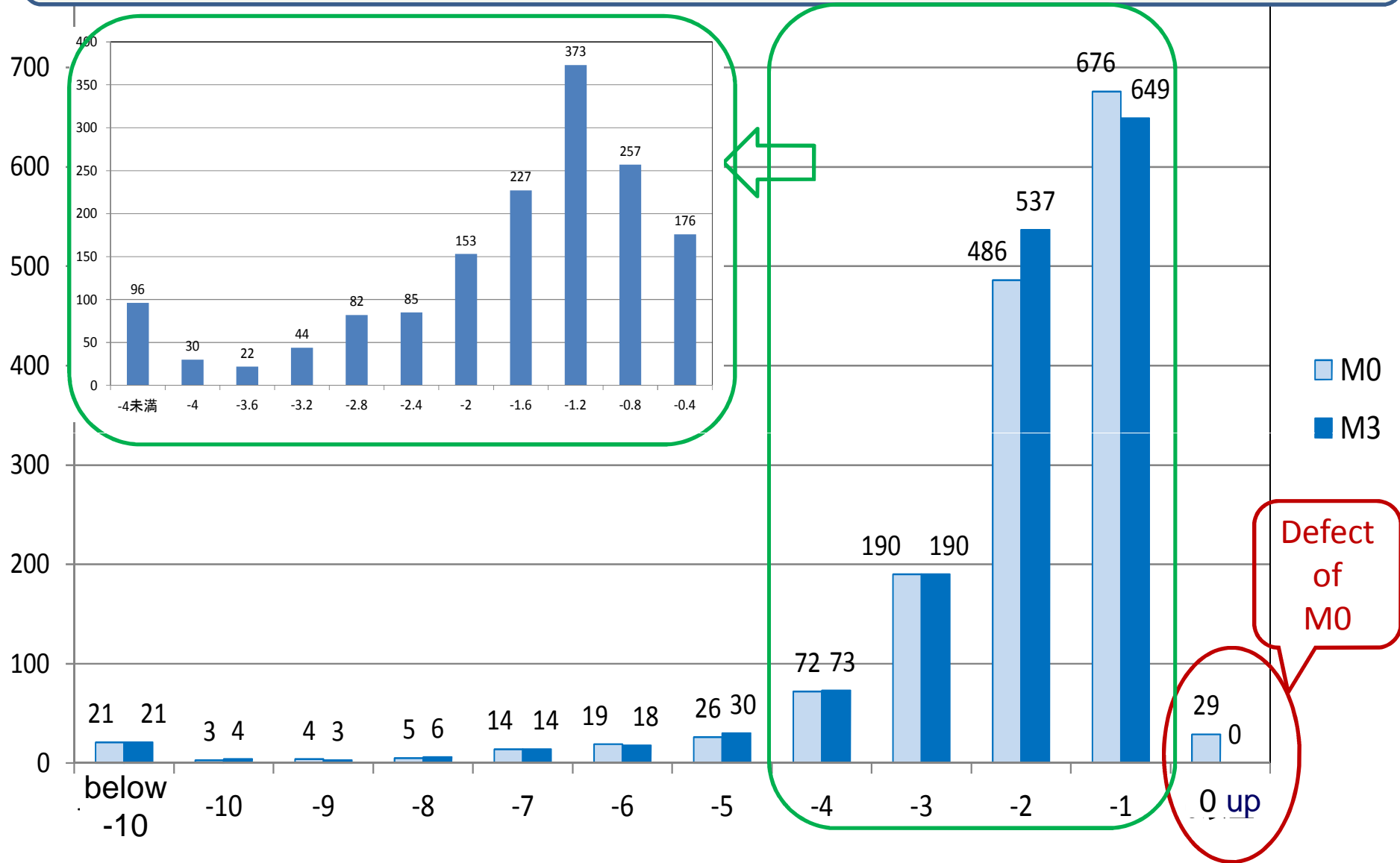
$$y_k^{(i)} = V_k - \hat{P}_k^{(i)}, \quad \hat{P}_k^{(i)} = \sum_{j=1}^{M(k)} C_k(s_{kj}) \bar{D}_k^{(i)}(s_{kj})$$

( $i = 0, 3$ ) MO Spread vs M3 Spread

- Standardized CRPS = **S-CRPS** (Standardization of uncertainty)

$$\zeta_k^{(3)} = y_k^{(3)} / s_{kM(k)}$$

# Distributions of CRPS of M0 & M3 with 1 yen interval



## Credit-homogeneous grouping

- R&I rating as of 2010.8

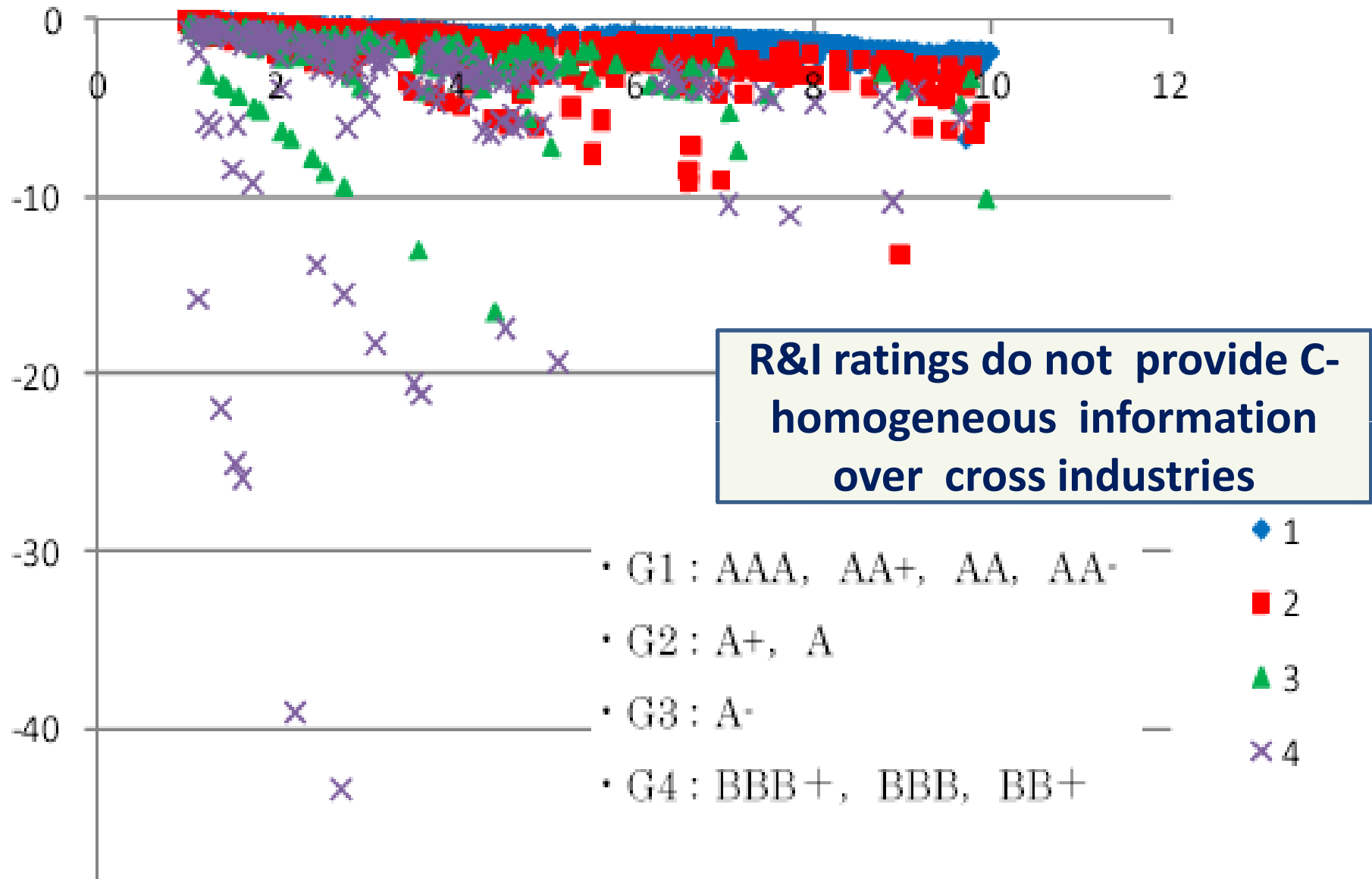
	AAA	AA+	AA	AA-	A+	A	A-	BBB+	BBB	BBB-	C+	NA	Total
#	9	497	119	172	179	152	162	95	52	1	7	100	1545
%	0.6	54.0	8.2	11.9	12.4	10.5	11.2	6.6	3.6	0.0	0.5	NA	100

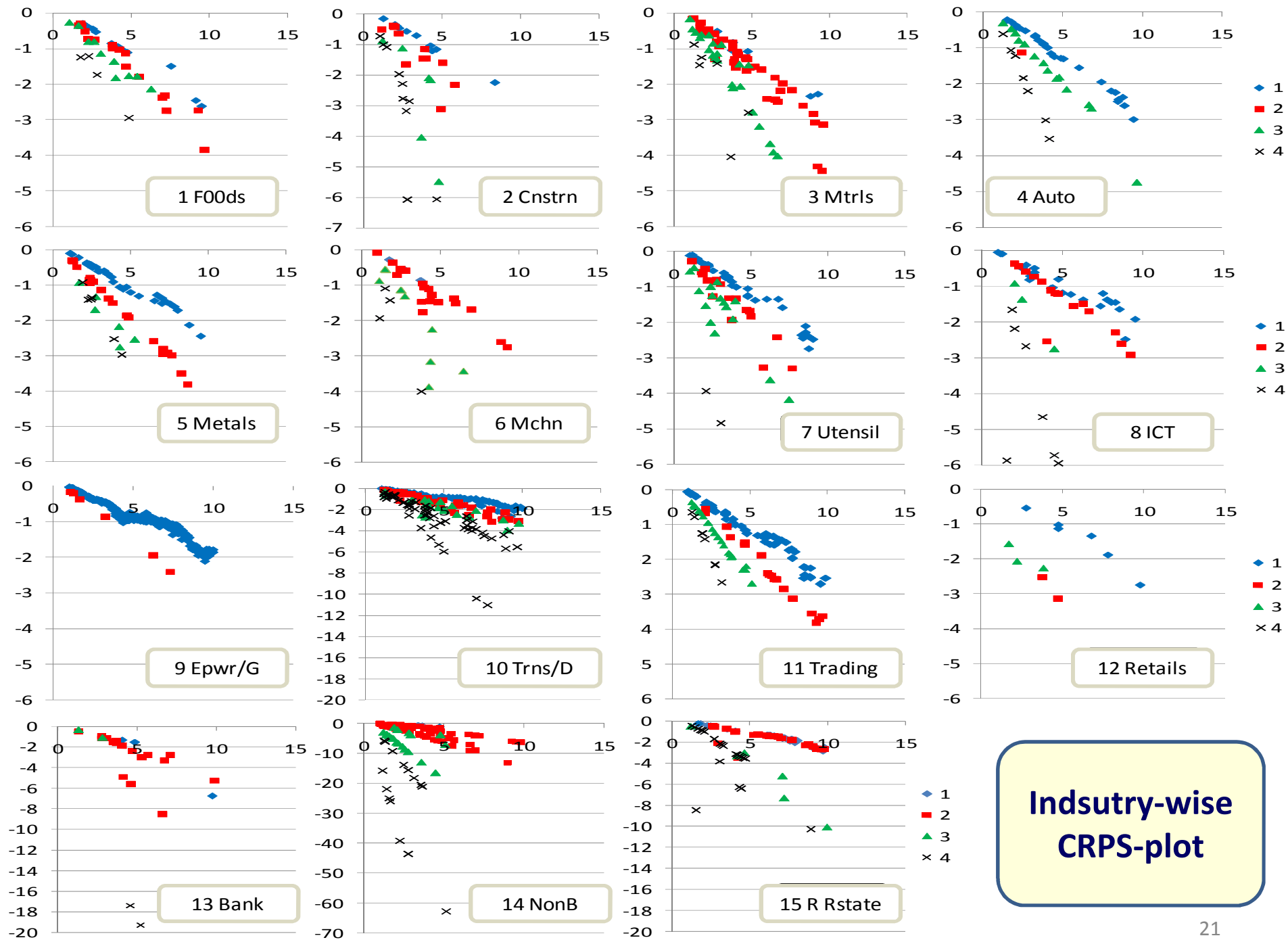
- Industry category

1 Foods, 2 Construction & its Materials, 3 Materials/Chemicals, 4 Transportation Equipments , 5 Steel /Non-steel/Mining, 6 Machinery, 7 Electric Appliances/Precision Instruments, 8 ICT /Services, 9 Electric Power/Gas, 10 Transportation/Distribution, 11 International Distribution (Trading), 12 Retails, 13 Banking, 14 Nonbank Financial Business, 15 Real Estate.

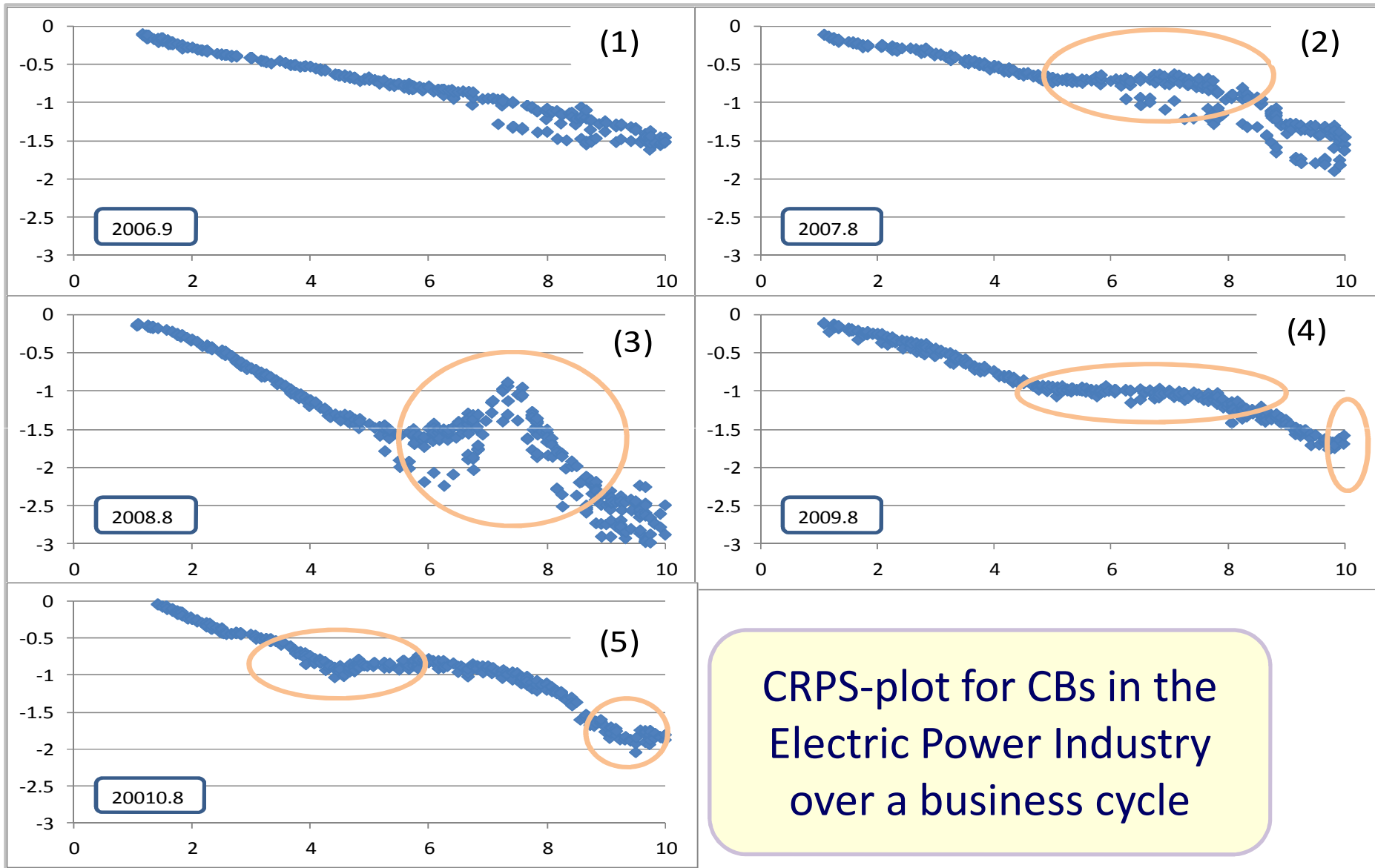
- Cluster Analysis
- **FIR (Fixed Interval Rating) via S-CRPS**

# Effectiveness of Grouping via R&I Rating

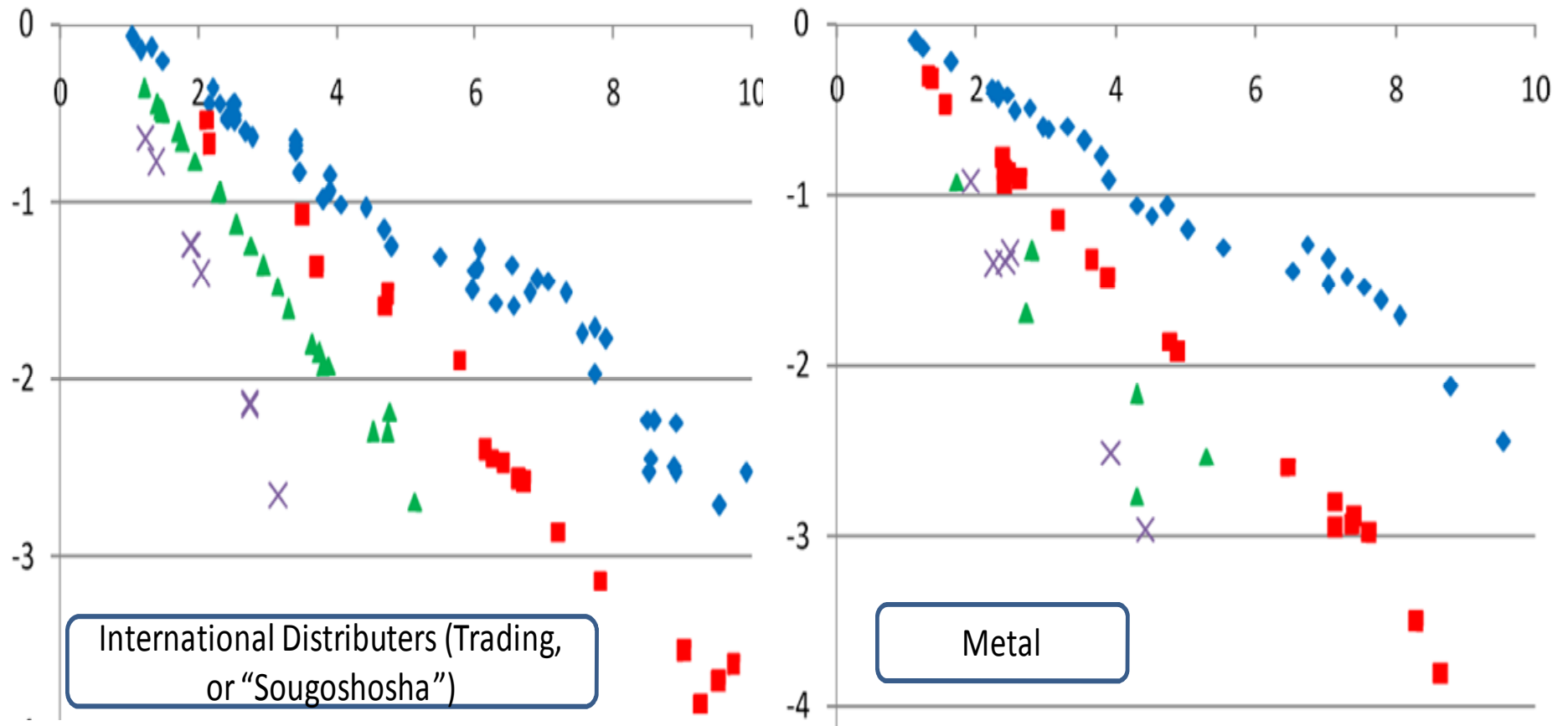


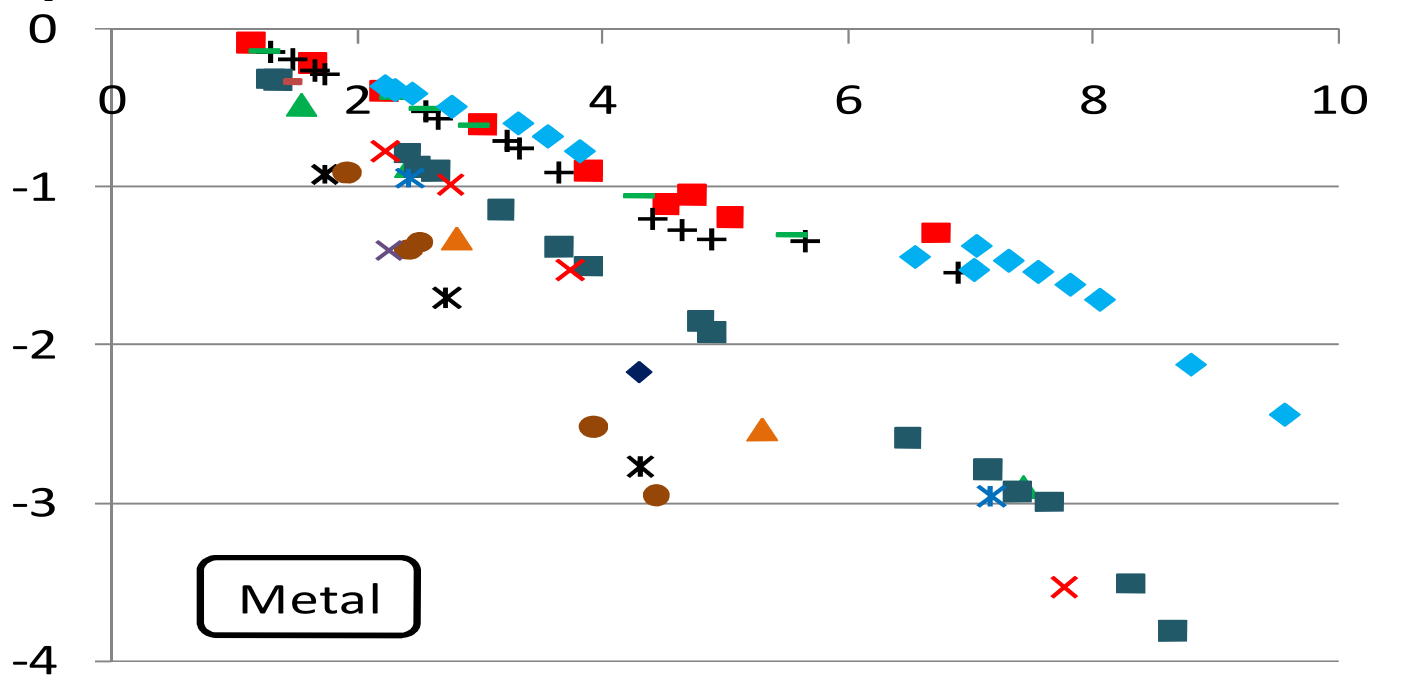
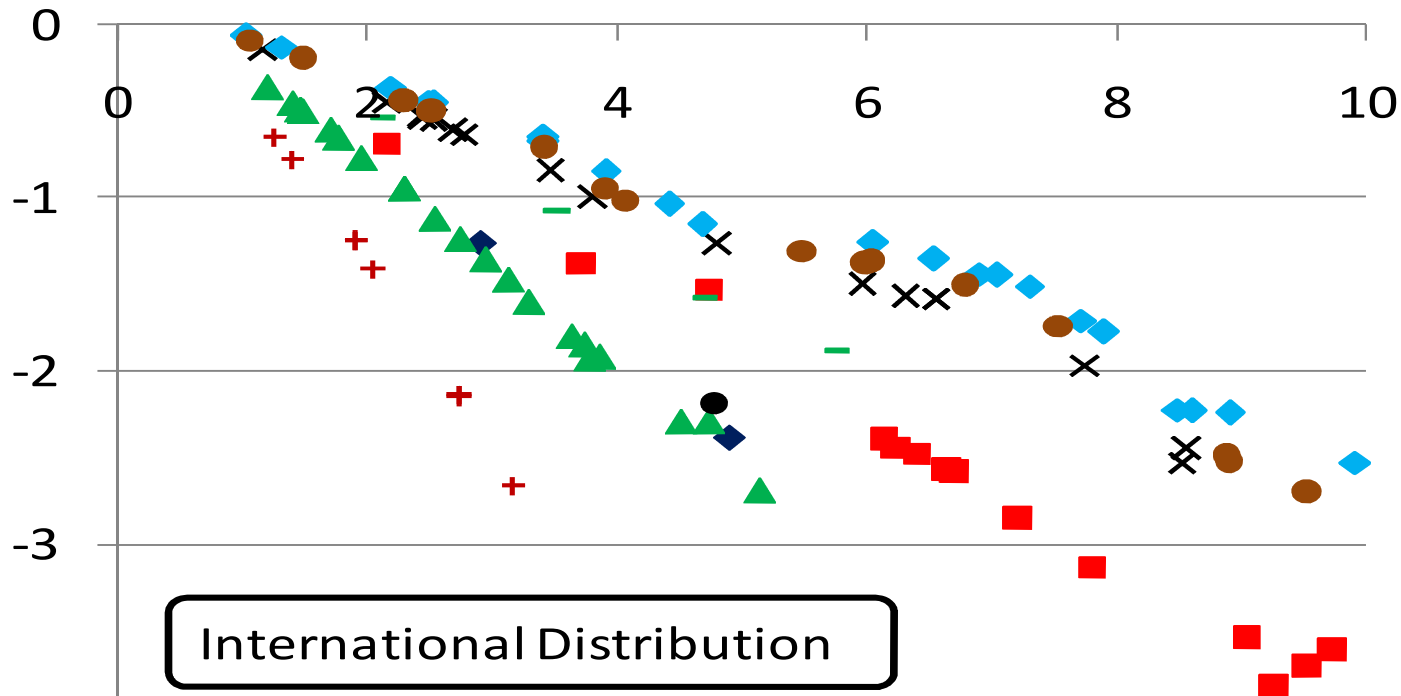


# Changes of CRPSs of E-Power Industry over business cycles



# CRPS-Plots for International Distribution and Metal Industries

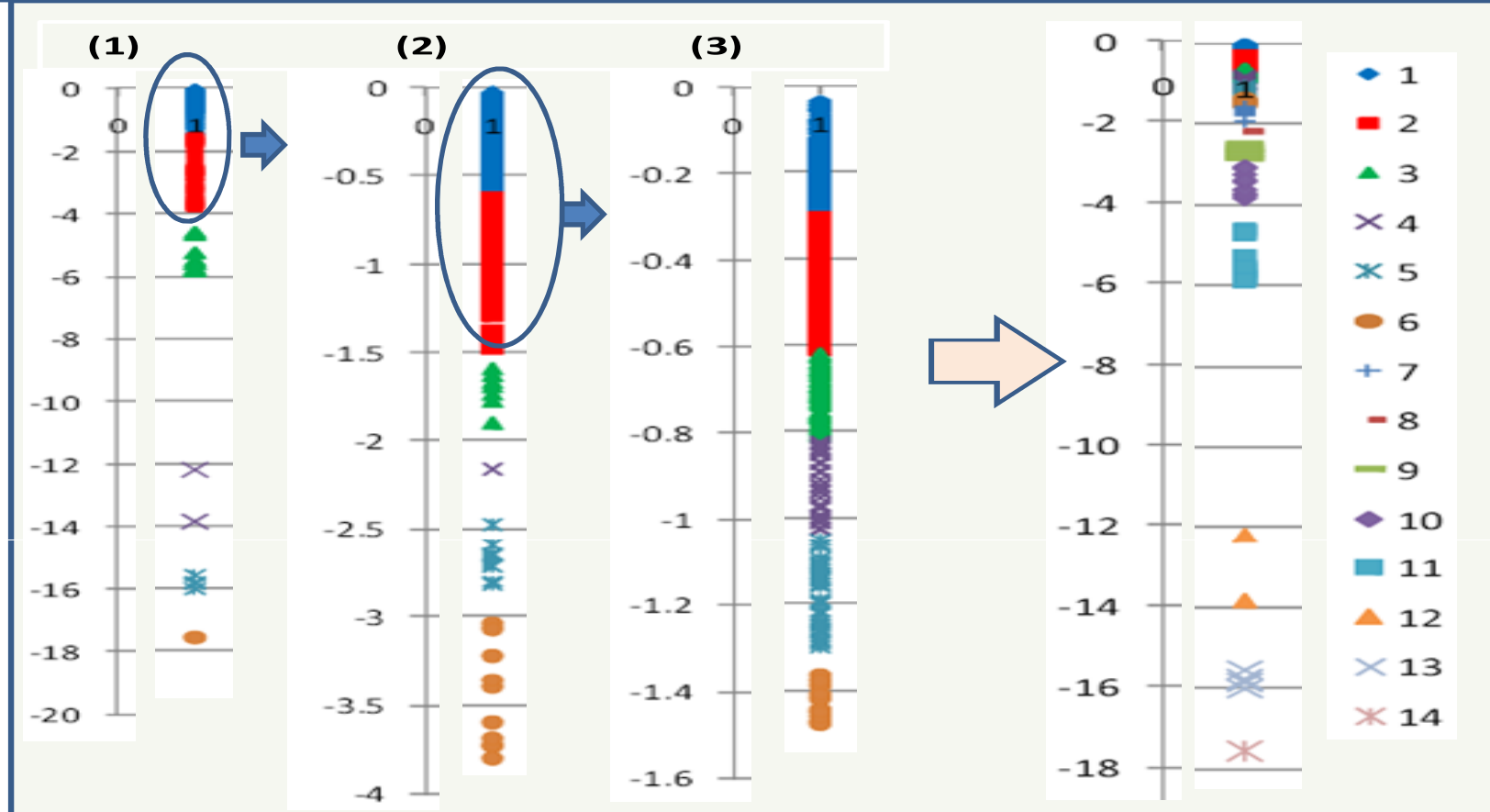




- ◆ Itochu En
- Itochu A
- ▲ Marubeni A-
- × Mitsui&Co AA-
- ◆ Mitsubishi AA-
- Sumitomo AA-
- + Sojitsu BBB
- N.Pulp A-
- Toyoda A+
- ◆ DOWA A-
- JFE AA-
- ▲ Fujikura A
- × Furukawa B3+
- \* Mitui M A-
- Mitsbsh M B3+
- + Smtm M
- Smtm Min
- Smtm E AA-
- ◆ Nppn Stl AA-
- Kobe Stl A
- ▲ Daido Stl A-
- × Nisshin Stl
- \* Hitachi M A

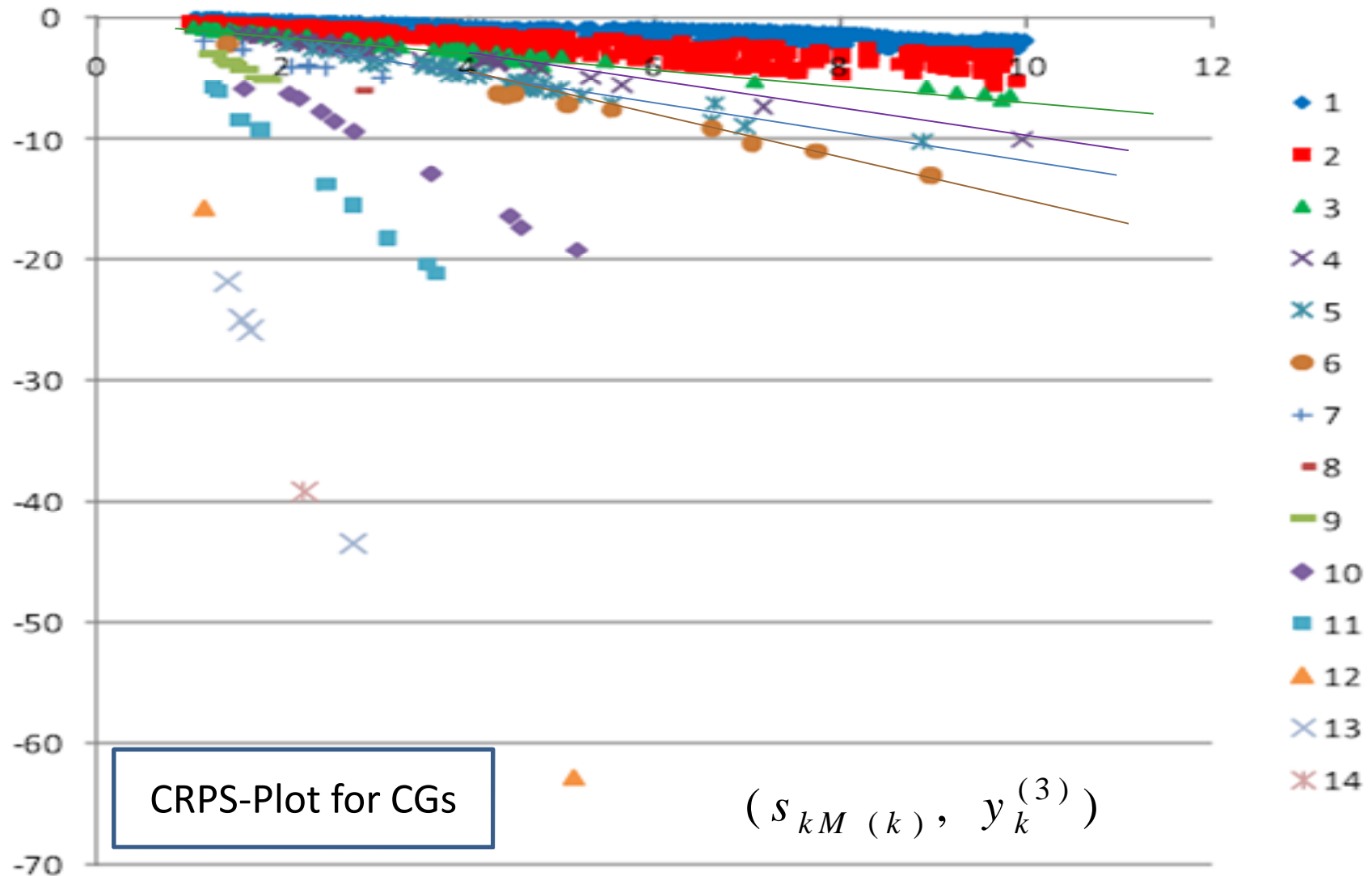


# Credit-homogeneous Groups via 3 Stage Cluster Analysis



- 1) # of clusters is fixed as 6 for each stage
- 2) Stage (1) separates 4 groups, called CG11, CG12, CG 13, CG14
- 3) Stage (2) separates 4 groups, called CG7, CG8, CG9, CG10
- 4) Stage (3) gives 6 groups;CG1-CG6, within the distance 1.5

# Credit-homogeneous groups via cluster analysis



## FIR (Fixed Interval Rating) via S-CRPS

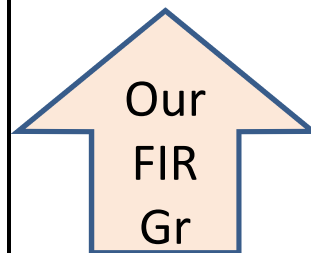
- Grouping by Cluster Analysis is a posterior grouping given CB and GB prices, which depends on the economic environment at the point of analysis
- Alternative we propose an absolute and prior criterion for grouping based on the intervals formed by **S-CRPS**

$$\zeta_k^{(3)} = y_k^{(3)} / s_{kM}(k)$$

- If  $\zeta_k$  belongs to  $I_m = (x_{m-1}, x_m]$ , CB  $k$  is assigned to group FIR  $k$ . This is an **market approach** to grouping.
- To choose an interval scheme, we consider the distribution of S-CRPS measures.

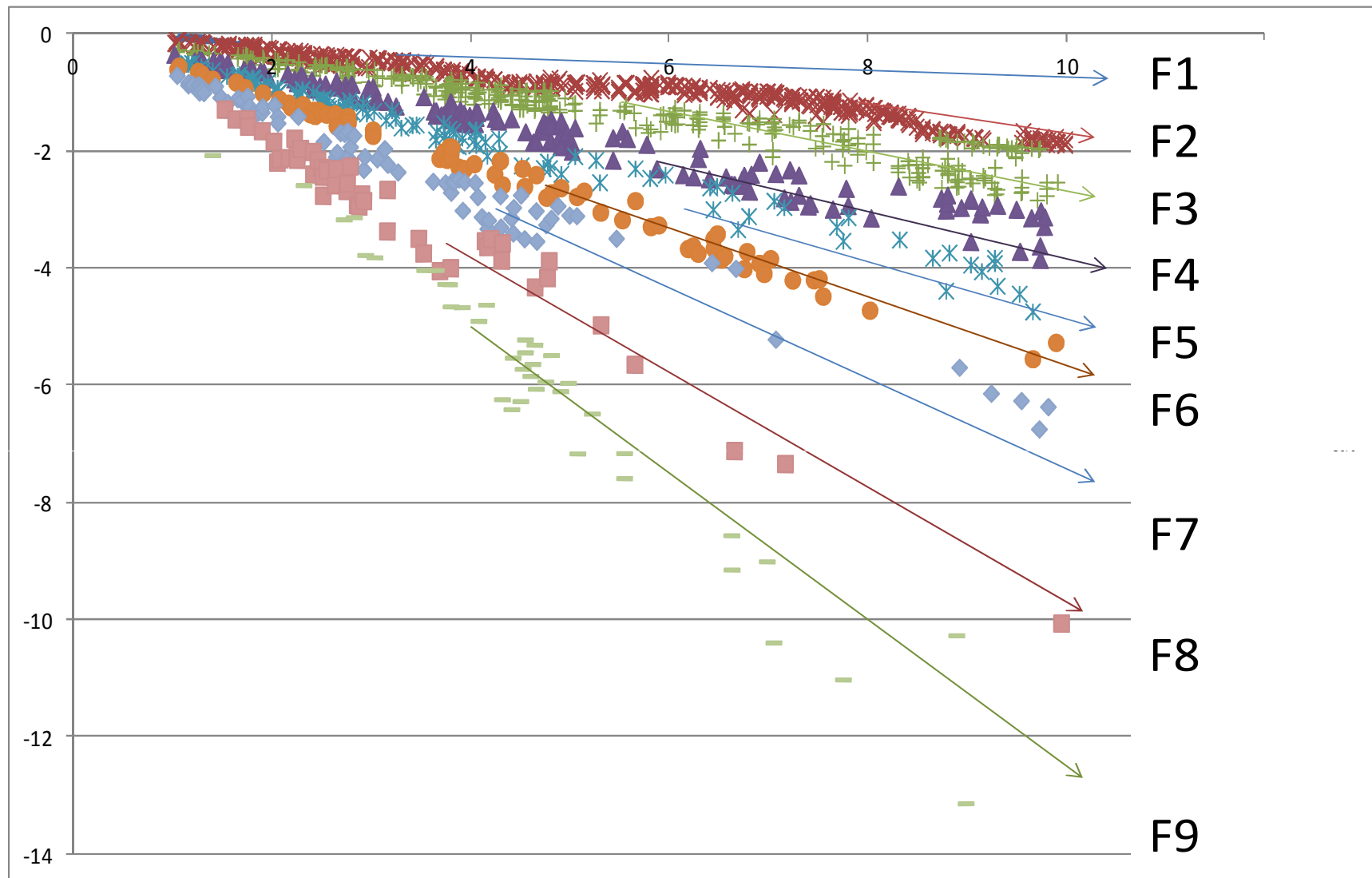
## FIR(Fixed Interval Rating) based on S-CRPS measure $\zeta$ :2010.8

FIR 1			FIR 2			FIR 3			FIR 4			FIR 5		
FG	0.5Yen	#	FG	1Yen	#	FG	M1Yen	#	FG	1.5Yen	#	FG	2Yen	#
1	[-0.5,0)	10	1	[-1,0)	42	1	[-1,0)	42	1	[-1.5,0)	213	1	[-2,0)	618
2	[-1.0,-0.5)	32	2	[-2,-1)	576	2	[-2,-1)	576	2	[-3.0,-1.5)	773	2	[-4,-2)	526
3	[-1.5,-1.0)	171	3	[-3,-2)	368	3	[-3,-2)	368	3	[-4.5,-3.0)	214	3	[-6,-4)	186
4	[-2.0,-1.5)	405	4	[-4,-3)	158	4	[-4,-3)	158	4	[-6.0,-4.5)	130	4	[-8,-6)	86
5	[-2.5,-2.0)	223	5	[-5,-4)	111	5	[-5,-4)	111	5	[-7.5,-6.0)	64	5	[-10,-8)	35
6	[-3.0,-2.5)	145	6	[-6,-5)	75	6	[-6,-5)	75	6	[-9.0,-7.5)	41	6	$(-\infty,-10)$	94
7	[-3.5,-3.0)	81	7	[-7,-6)	46	7	[-8,-6)	86	7	[-10.5,-9.0)	22			
8	[-4.0,-3.5)	77	8	[-8,-7)	40	8	[-11,-8)	48	8	$(-\infty,-10.5)$	88			
9	[-4.5,-4.0)	56	9	[-9,-8)	19	9	[-15,-11)	40						
10	[-5.0,-4.5)	55	10	[-10,-9)	16	10	$(-\infty,-15)$	41						
11	[-5.5,-5.0)	36	11	$(-\infty,-10)$	94									
12	[-6.0,-5.5)	39												
13	[-7.0,-6.0)	46												
14	[-8.0,-7.0)	40												
15	[-9.0,-8.0)	19												
16	[-10.0,-9.0)	16												
17	$(-\infty,-10)$	94												



Market Grouping via FIR 3 (Total 1545 CBs)													
R&I	F1	F2	F3	F4	F5	F6	F7	F8	F9	F10	Total	# CBs	
	(0,-1]	(-1,-2]	(-2,-3]	(-3,-4]	(-4,-5]	(-5,-6]	(-6,-8]	(-8,-11]	(-11,-15]	(-15,∞)			
AAA	0.0	22.2	77.8	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100	9	
AA+	6.6	84.9	8.5	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100	497	
AA	0.8	52.1	45.4	1.7	0.0	0.0	0.0	0.0	0.0	0.0	100	119	
AA-	3.5	27.9	66.3	1.7	0.0	0.0	0.6	0.0	0.0	0.0	100	172	
A+	0.6	12.8	52.0	14.0	7.8	6.1	3.4	3.4	0.0	0.0	100	179	
A	0.0	0.7	13.8	46.7	9.9	3.9	7.9	8.6	8.6	0.0	100	152	
A-	0.0	0.6	8.6	25.3	27.2	11.1	9.9	6.2	2.5	8.6	100	162	
BBB+	0.0	0.0	1.1	5.3	20.0	14.7	27.4	9.5	9.5	12.6	100	95	
BBB	0.0	0.0	0.0	0.0	9.6	28.8	34.6	9.6	11.5	5.8	100	52	
BBB-	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100	100	1	
CCC+	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	100	100	7	
None	1.0	17.0	22.0	11.0	14.0	11.0	7.0	5.0	8.0	4.0	100	100	
# of CBs	42	576	368	158	111	75	86	48	40	41	100	1545	

# Credit-homogeneous Groups via FIR grouping:2010.8



# **Deriving the TSDPs of C-homogeneous groups and individual firms via Kariya (2012)'s model**

Recovery rate is assumed to be zero.

- 1) Takes a long time for settlement in the creditors meeting and so uncertainty is large
- 2) Chapters 7 (liquidation) and 11 (reorganization)
- 3) Comparison

## CB Pricing Model

## Expected CFs

$$V_k = \sum_{j=1}^{M(k)} \bar{C}_k(s_{kj}) D_k(s_{kj})$$

$$\begin{aligned} \bar{C}_k(s_{mj}) &= C_k(s_{mj})[1 - p_k(s_{mj})] \\ &+ 100\gamma_k[p_k(s_{mj}) - p_k(s_{mj-1})] \end{aligned}$$

$$D_k(s) = \bar{D}_k(s) + \Delta_k(s)$$

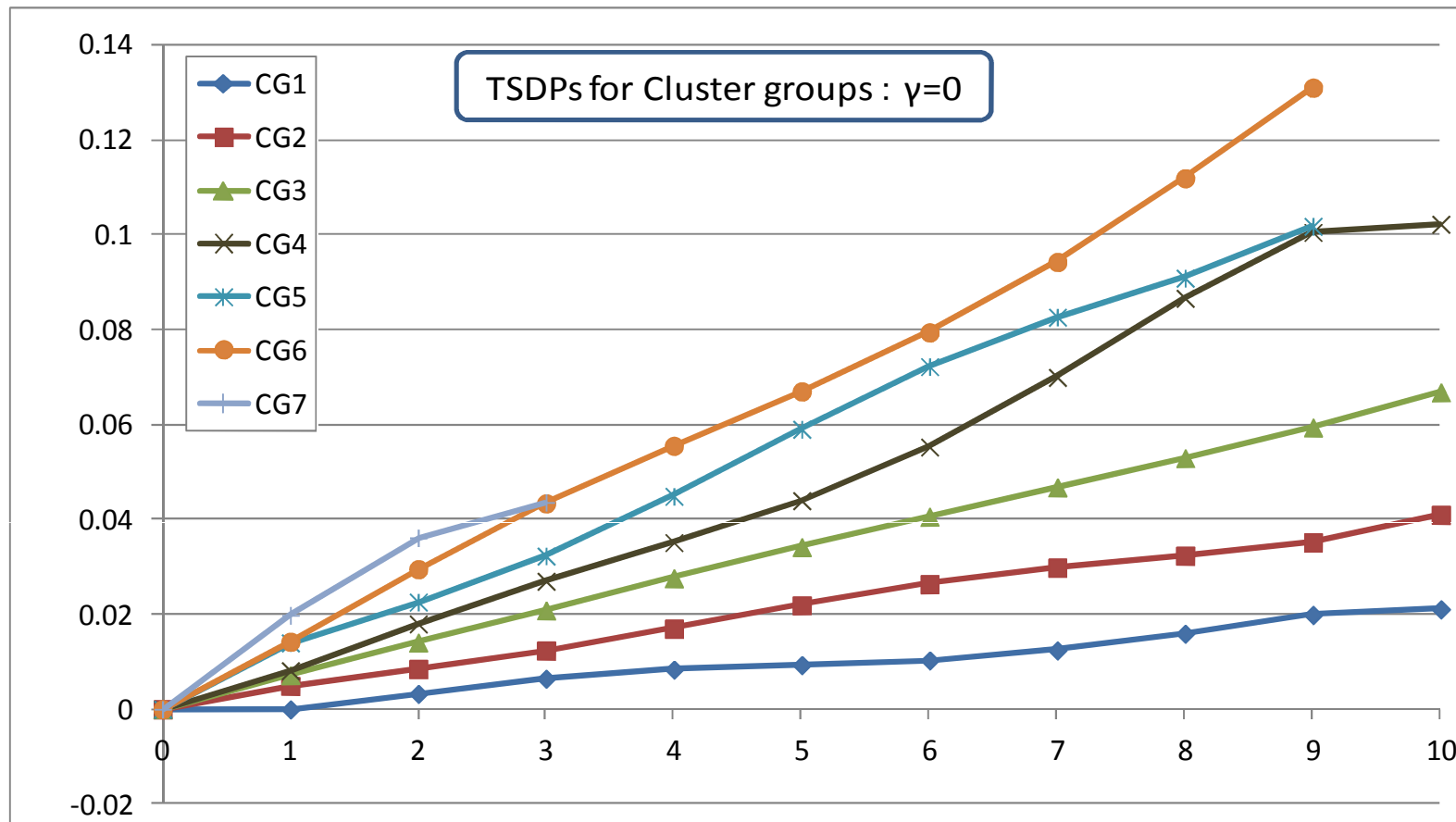
$$p_k(s : i(k)) \equiv \sum_{j=1}^J w_k(j) p(s : i(k), j)$$

$$p(s : i, j) = \alpha_1^{ij} s + \alpha_2^{ij} s^2 + \dots + \alpha_q^{ij} s^q$$

$$w_k(j) \geq 0, \quad \sum_{j=1}^J w_k(j) = 1$$

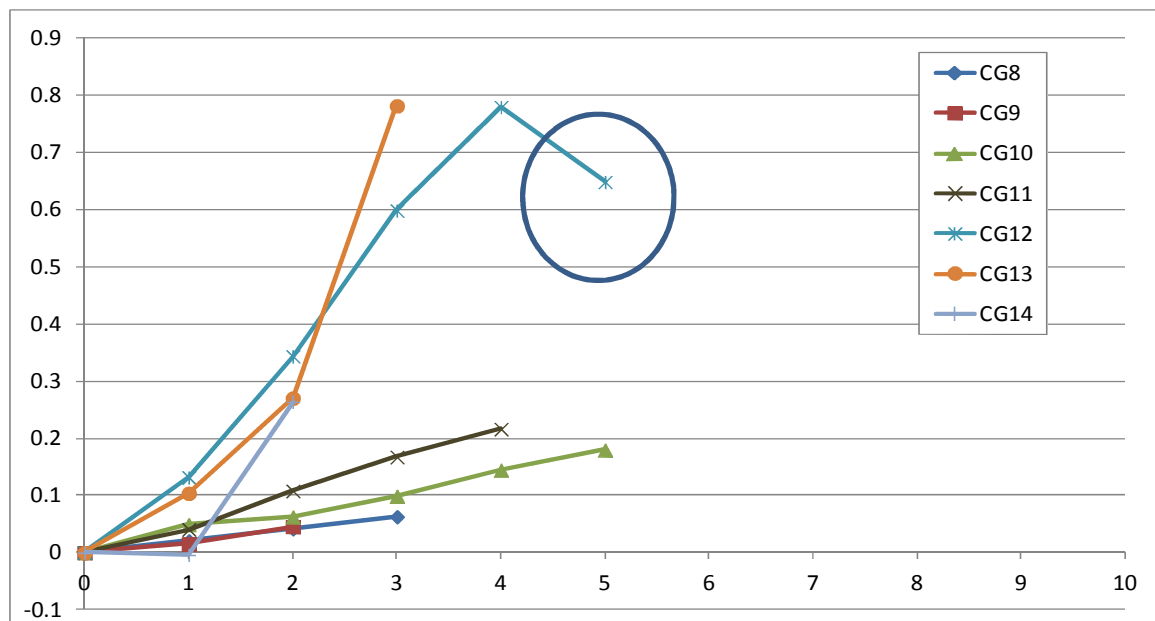


## TSDPs for Cluster Groups:2010.8



Default Probabilities for CGs

	CG1	CG2	CG3	CG4	CG5	CG6
9yrs	2.0	3.5	5.9	10.1	10.2	13.1
10yrs	2.1	4.1	6.7	10.2		

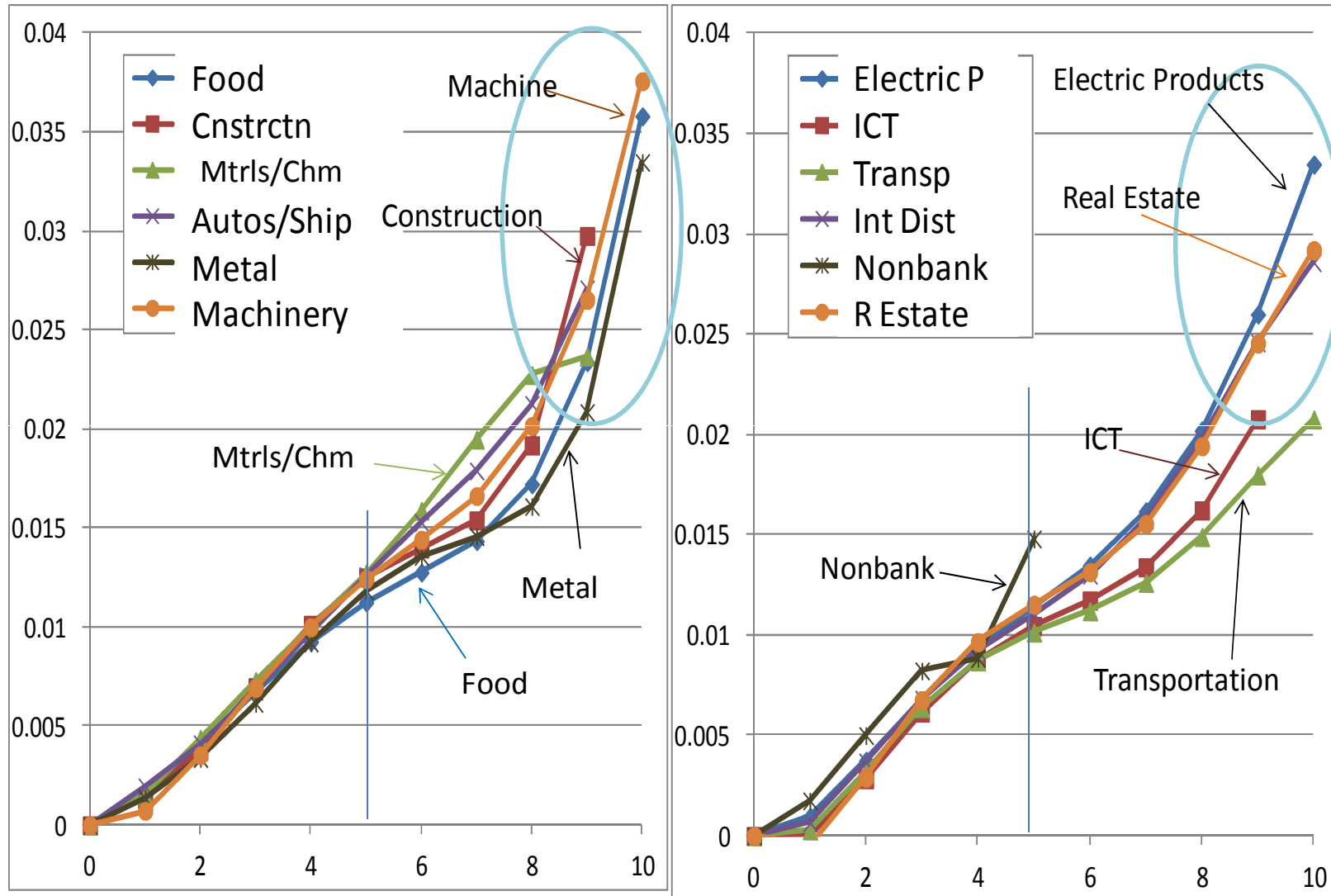


## Decomposition of CG1&2 into industry groups

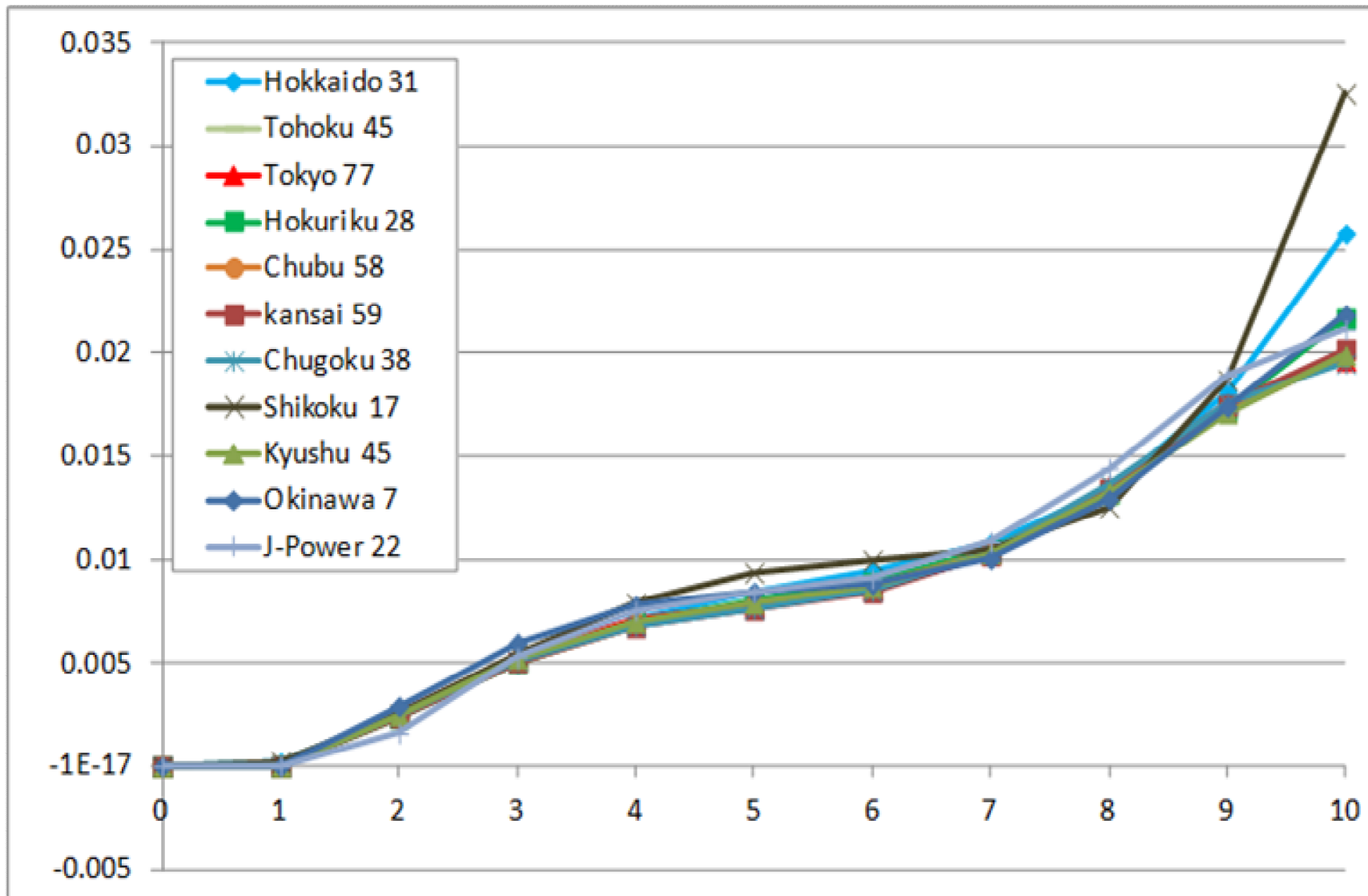
Inds	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
<b>CG1</b>	24	16	41	30	49	19	44	53	459	124	56	6	1	14	52
<b>CG2</b>	18	9	46	23	30	13	27	7	2	81	42	1	18	22	12

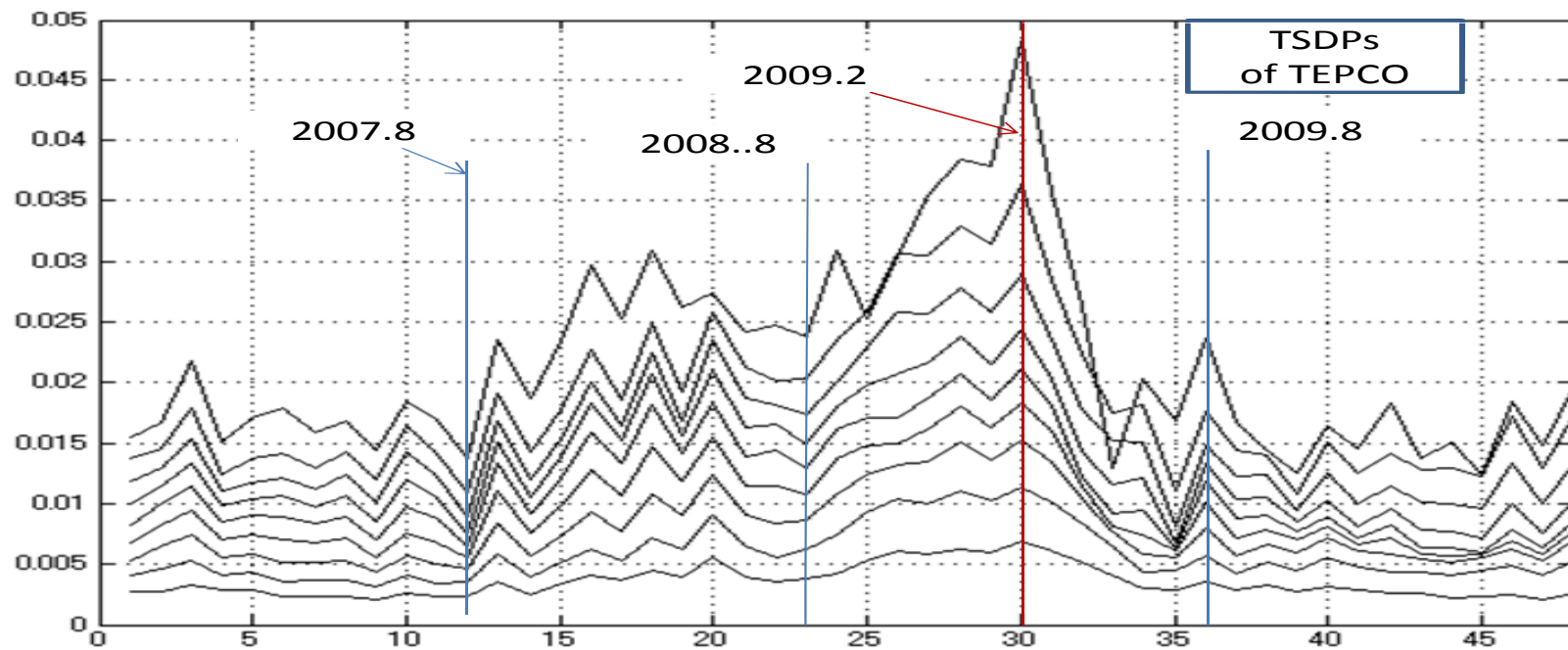
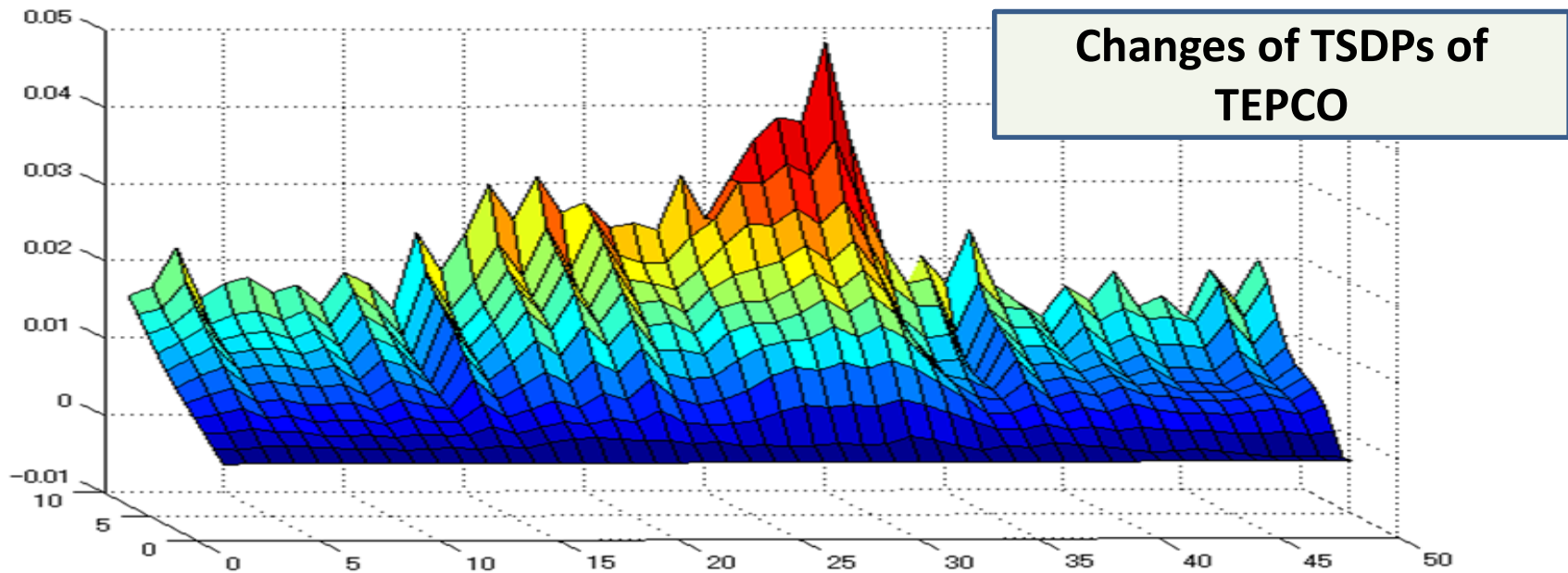
1 Foods, 2 Construction & its Materials, 3 Materials/Chemicals, 4 Transportation Equipments , 5 Steel /Non-steel/Mining, 6 Machinery, 7 Electric Appliances/Precision Instruments, 8 ICT /Services, 9 Electric Power/Gas, 10 Transportation/Distribution, 11 International Distribution (Trading), 12 Retails, 13 Banking, 14 Nonbank Financial Business, 15 Real Estate.

## Industrial Differentiation of TSDPs for CG1:2010.8

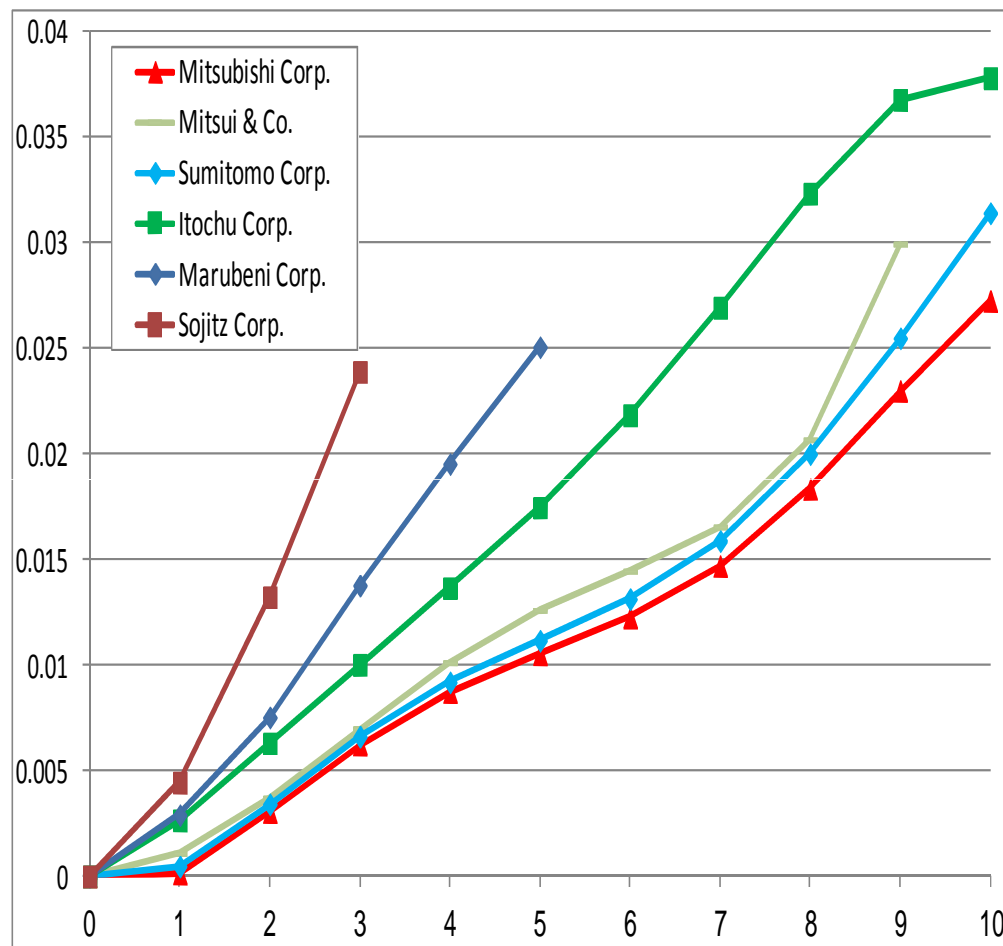


## TSDPs in the Electric Power Industry:2010.8



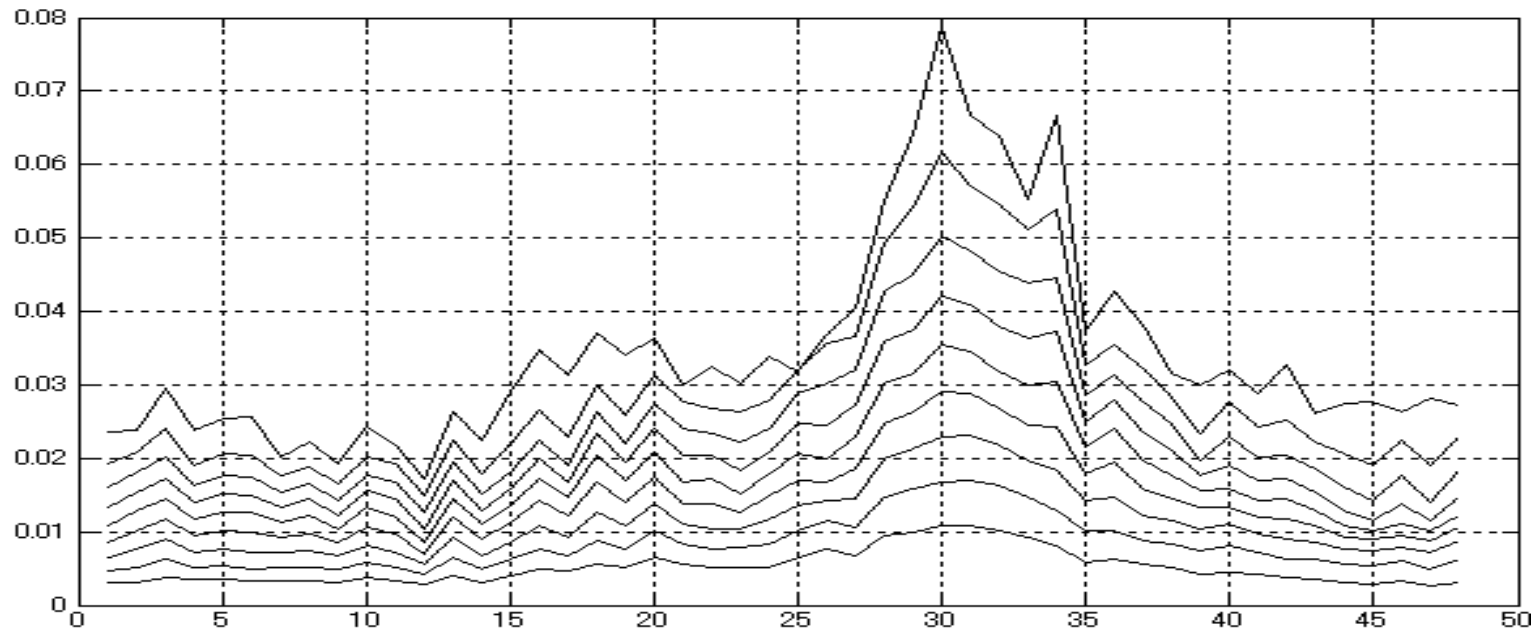
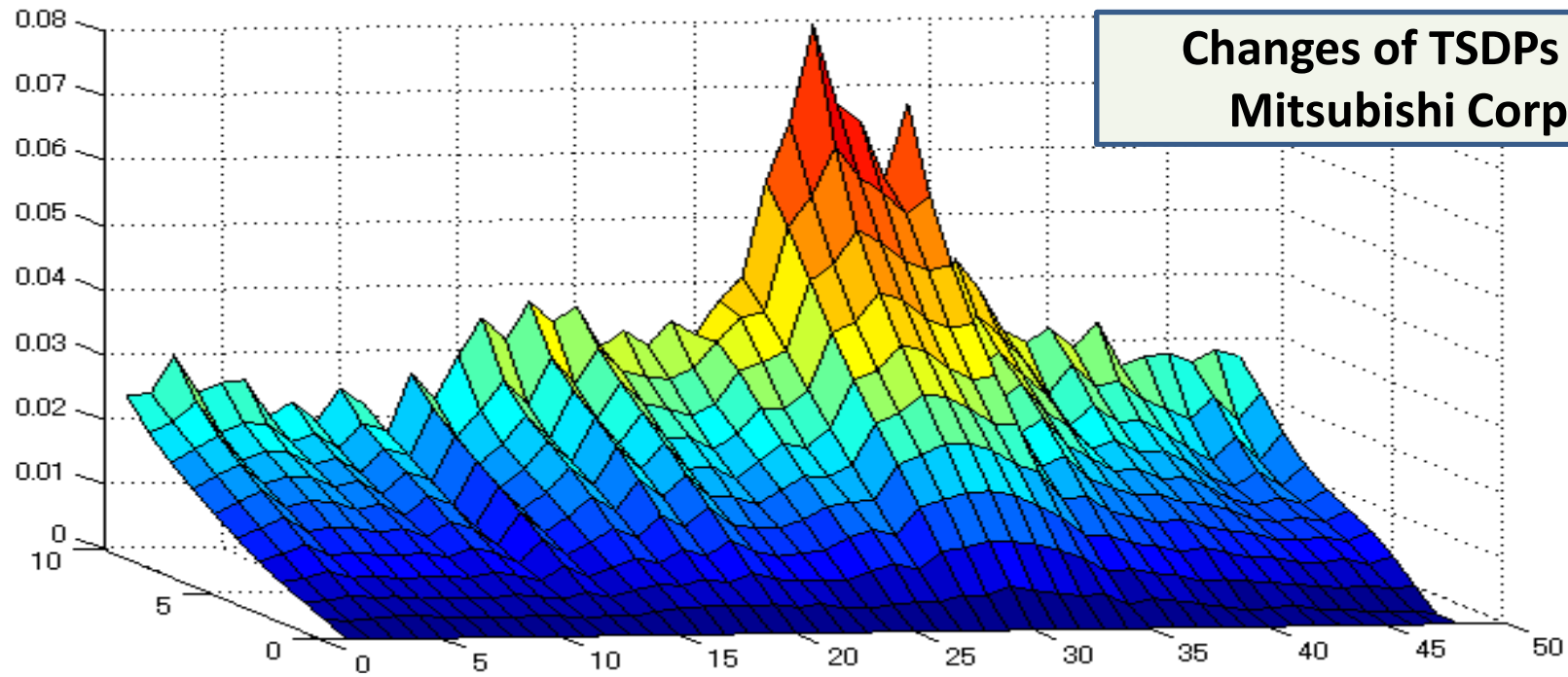


## TSDPs of individual firms in the Trading Industry:2010.8

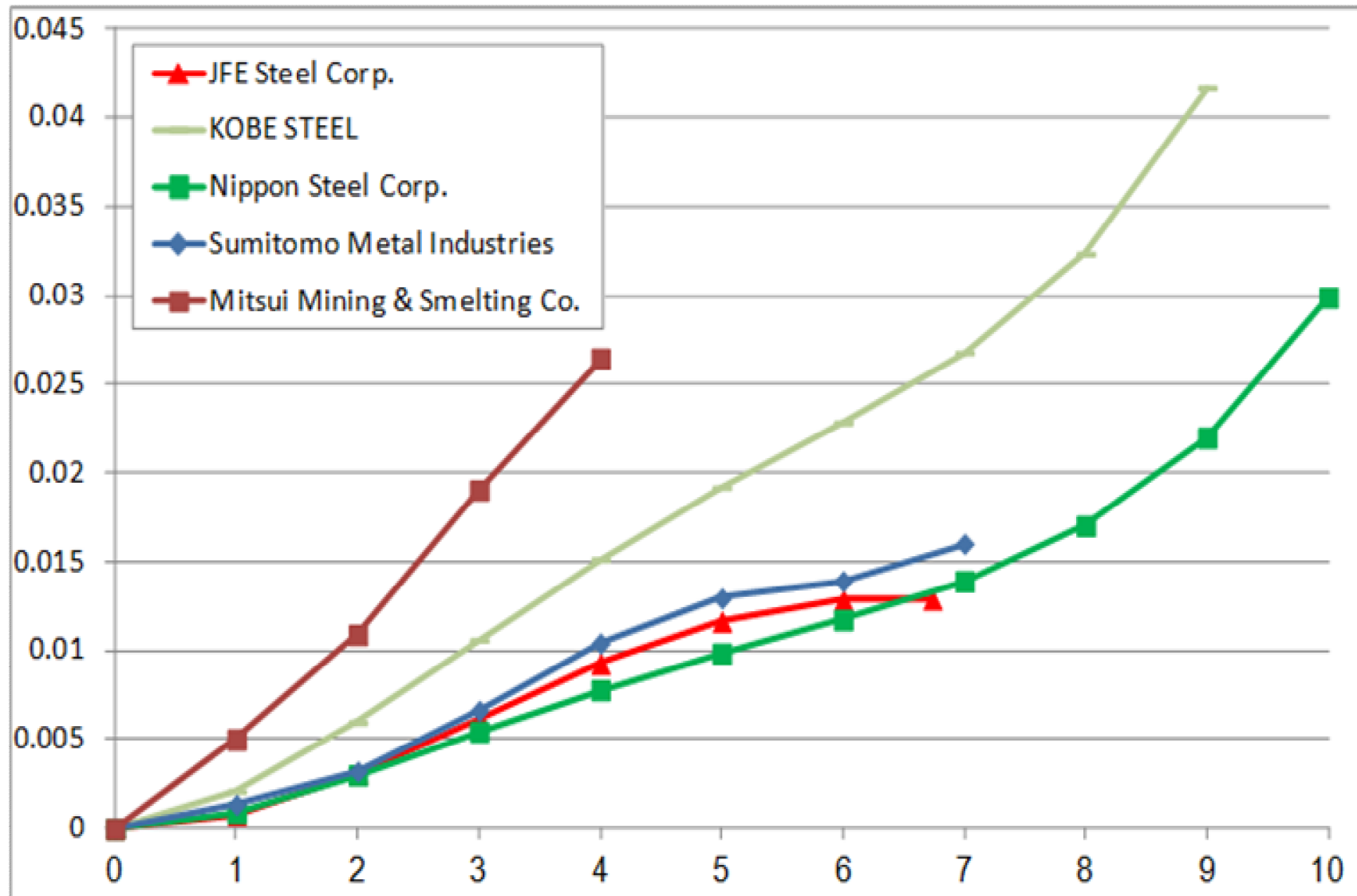


	MBS	SuS	MiT	ITo	MBN	SJT
2yrs	0.30	0.34	0.37	0.63	0.75	1.32
3yrs	0.62	0.66	0.70	1.00	1.38	2.39
4yrs	0.87	0.92	1.01	1.37	1.95	
5yrs	1.05	1.12	1.26	1.75	2.51	
6yrs	1.23	1.31	1.45	2.19		
7yrs	1.47	1.59	1.66	2.69		
8yrs	1.84	2.00	2.07	3.23		
9yrs	2.30	2.55	3.00	3.68		
10yrs	2.73	3.14		3.78		%

**Changes of TSDPs of  
Mitsubishi Corp**

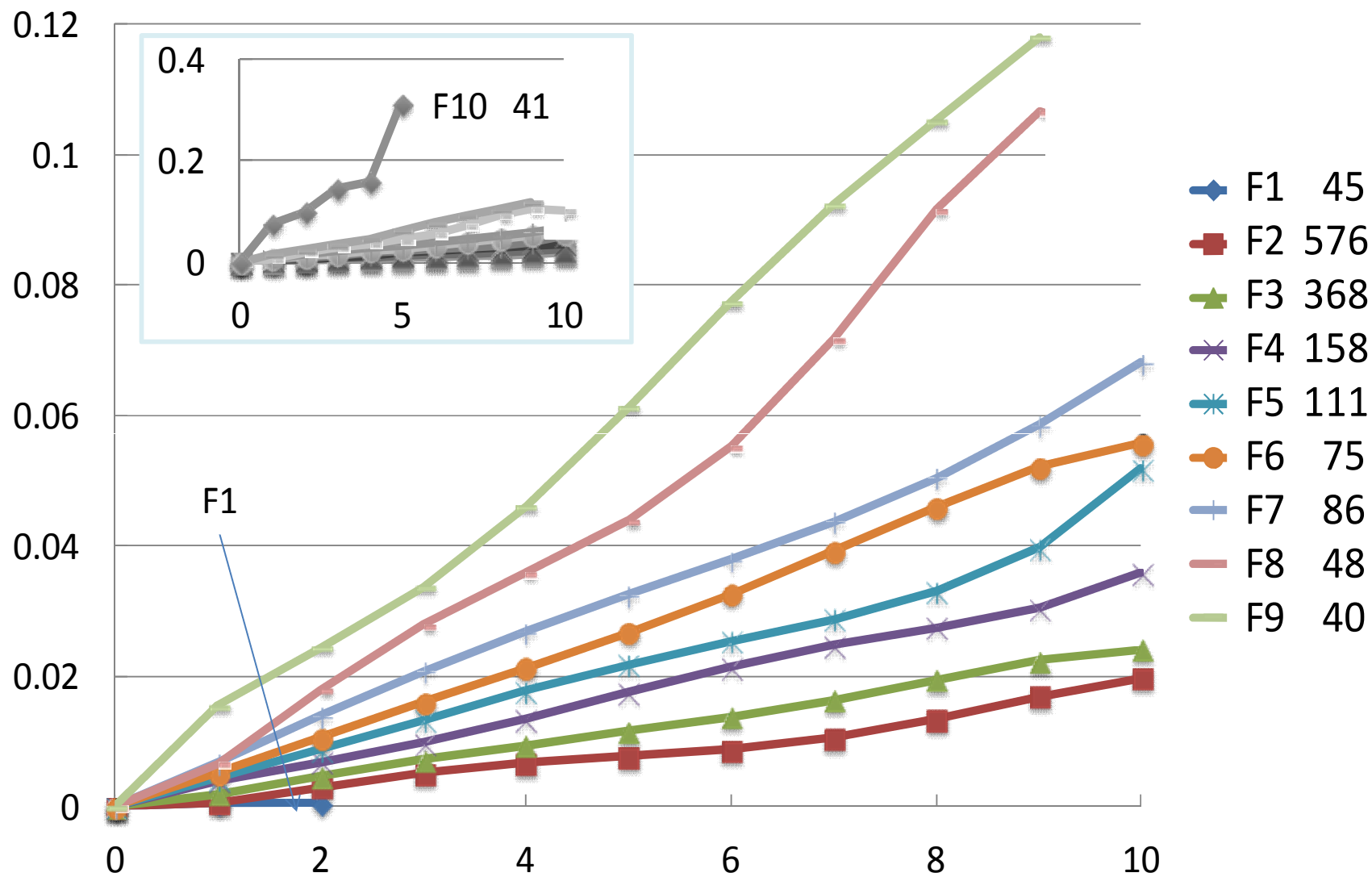


## TSDPs for individual firms in the Metal Industry:2010.8





## TSDPs for Fixed Interval Rating Groups:2010.8



# Summary

- Using Kariya, et al (2012)'s bond pricing model, we tested & **rejected H of no attribute effect**, and show the effectiveness of our GB-equivalent CB price and **CRPS measure** to analyze the C- structure.
- Showed ineffectiveness of the R&I rating scheme to analyze C-homogeneous groups in view of CRPS
- Proposed **FIR method for Market Rating of individual CBs based on the standardized CRPS**, on which Cluster groups are derived and analyzed
- TSDPs of cluster groups, FIR groups, E-power industry, TEPCO, Mitsubishi Corp are derived via Kariya(2012)'s model

# Selected References

- Kariya,T.(2012) A CB (corporate bond) pricing probabilities and recovery rates model for deriving default probabilities and recovery rates. To appear from Festschrift for Prof Morris L. Eaton , IMS monograph series
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- Kariya,T., Wang,J., Wang,Z., Doi,E., and Yamamura,Y.(2012) Empirically Effective Bond Pricing Model and Analysis on Term Structures of Implied Interest Rates in Financial Crisis *Asia-Pacific Financial Markets* 19:259–292
- Duan, J.C., J. Sun and T. Wang(2011) Multiperiod Corporate Default Prediction-A forward Intensity Approach, RMI working paper No.10/07, National University of Singapore
- Duffie, D. (2011). *Measuring Corporate Default Risk*. Clarendon Lectures in Finance, Oxford University Press

## Attribute-independent spot rate approach for given $r_0$

$$P_g(1) = \sum_{j=1}^{M(g)} C_g(s_{gj}) \overline{\overline{D}}(s_{gj}), \quad \overline{\overline{D}}(s_{gj}) = E_0[\exp(-\int_0^{s_{gj}} r_u du)] \equiv H(r_0, s_{gj}, \theta)$$

Zero yield  $R_s = -\frac{1}{s} \log H(r_0, s, \theta)$

## Attribute-dependent formulation for interest rates

$$\overline{\overline{D}}_g(s_{gj}) = E_0[\exp(-\int_0^{s_{gj}} r_{gs} ds)] \text{ conditional on } r_{g0}$$

## Attribute-independent forward rate approach

$$D(s_{gj}) = \exp(-\int_0^{s_{gj}} f_s ds)$$

## Attribute-dependent forward rate approach (our view)

$$D_g(s_{gj}) = \exp(-\int_0^{s_{gj}} f_{gs} ds)$$