Low-Order Perturbation Analysis

of the Complete Markets Case

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This discussion represents the views of the author and should not be interpreted as reflecting those of the Board of Governors of the Federal Reserve System or any other person associated with the Federal Reserve System.

- Algorithm for Perturbation
 - Split between State and Control Variables
 - Split Not Needed
 - * versatility
 - * interpretation

- Output of Perturbation: Nonlinear Difference Equation
 - How to use it?
 - This paper is using the pruning approach presented in KKSS.
 - * stability of the quadratic solution
 - \ast correct quadratic approximation of x_{t+2} w.r.t. x_t

• Results (to be continued)

• ACCURACY HIGHEST CLOSE TO STEADY STATE

• APPROXIMATION ERRORS INCREASING IN VOLATILITY OF PRODUCTIVTY

• <u>ACCURACY GAIN FROM QUADRATIC APPROX.</u> (COMPARED TO LINEAR APPROX) LARGER FOR BIG SHOCKS CASE ($\sigma = 0.01$) THAN FOR SMALL SHOCKS CASE ($\sigma = 0.001$).

ACROSS ALL SPECIFICATIONS WITH $\sigma = 0.01$, MEAN ACCURACY GAIN FOR S_{max} STATISTIC ($S_{\text{max}}^L - S_{\text{max}}^Q$): 0.80%

ACROSS CASES WITH $\sigma = 0.001$ mean $S_{\text{max}}^L - S_{\text{max}}^Q$: 0.001%.

• Approximation errors increasing in number of countries, risk aversion; decreasing in capital adjustment cost

• Simulated den Haan-Marcet (DM) statistics favor very slightly quadratic approximation (over linear approx.), but only when $\sigma = 0.01$.

KKK (2005) found that DM more clearly in favor when $\sigma = 0.1$.

• Computing speed: N=10 model solved in a few seconds

	Li	near ap	proxima	tion	2	2nd order approximation						
	Test	1	Test 2			Test 1	Test 2					
T _{.01}	$T_{.1}$	T _{.3}	$S_{_{ m max}}$	S _{mean}	T 01	$T_{.1}$	T .3	$S_{_{ m max}}$	S _{mean}			
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)	(10)			
All specifications												
-2.57	-1.40	-0.59	-1.63	-2.80	-4.41	-2.53	-1.13	-2.50	-4.03			
Small shocks ($\sigma = 0.001$)												
- 3.95	-1.98	-1.10	-4.65	-5.31	-6.01	-3.06	-1.70	-6.90	-7.85			
Big shocks ($\sigma = 0.01$)												
-2.57	-1.40	-0.59	-1.63	-2.50	-4.41	-2.53	-1.13	-2.50	-3.73			

Table 1. Accuracy tests: results for sets of specifications

Table shows \log_{10} of test statistics

Table 2. Relating accuracy to model parameters: regression results

$log_{10}(S_{\max}^{L,i})$	=	-4.93 (51.25)	+ 0.005 N^{i} (0.60)	-0.46 γ ⁱ	+315.8 σ ⁱ (49.16)	-0.02 φ^{i} (4.52)	$R^2 = .96;$
$log_{10}(S_{\max}^{Q,i})$	=	-7.94 (49.54)	$+ 0.002 N^{i}$ (0.15)	-0.67 γ^{i} (4.66)	$+455.7 \sigma^{i}$ (46.74)	-0.06 <i>q</i> ⁱ (6.96)	$R^2 = .96;$
$log_{10}(S_{\max}^{L,i}-S_{\max}^{Q,i})$	=	-4.94 (52.32)	$+ 0.005 N^{i}$ (0.57)	•	+312.5 σ ⁱ (49.61)	$-0.02 \pmb{\varphi}^i$	$R^2 = .96;$

In parentheses: absolute t-stats.

						L	Linear approximation							2nd order approximation							
Mo=						Test 1	Те	Test 2		Test 3		Test 1		Test 2		Test 3					
del	N	γ	η	χ	μ	$T_{.01}$ $T_{.3}$	S _{max}	S _{mean}	P .05	P .5	P .95	T 01	T .3	S _{max}	S _{mean}	P .05	P .5	P .95			
(1)	(2)	(3)	(4)	(5)	(6)	(7) (8)	(9)	(10)	(11)	(12) (13)	(14)	(15)	(16)	(17)	(19)	(20)	(21)			
(a) Small shocks: $\sigma = 0.001$																					
A1	2	1				-4.18 -1.3	3 -5.36	-6.01	.04	.49 .	96	-6.33	-2.01	-7.61	-8.32	.04	.49	.96			
A1	6	1				-4.70 -1.8	5 -5.49	-5.84	.99	.991.	00	-7.01	-2.66	-7.65	-8.22	.99	.991	.00			
(b) Big shocks: $\sigma = 0.01$																					
A1	2	1			-	-3.26 -0.9	9 -2.34	-3.02	.03	.43 .9	94	-5.40	-1.71	-3.66	-4.44	.04	.45	.94			
A1	6	1				-3.15 -1.5	1 -2.35	-2.80	.00	.06 .	63	-5.41	-2.19	-3.69	-4.23	.00	.06	.63			
$\mathbf{T}_{\mathbf{A}}$	L 1~	ah		a 1a	20	of T	r c														

Table 3. Accuracy tests: results for selected specifications

Table shows \log_{10} of T_r, S_{\max}, S_{mean} .