

**SUPPLEMENTARY MATERIAL FOR:
IDENTIFICATION OF VECTOR AUTOREGRESSIVE MODELS
WITH GRANGER AND STABILITY CONSTRAINTS**

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ABSTRACT

In this document, we present experimental results that have not been included in [1] due to space constraints. The notation is the same as in the main manuscript.

In the caption of each table within the supplementary material, we write down the name of the experiment in which the results have been obtained as well as the sample size (N) used in the training phase. In all cases, the total number of trials is $N_{tr} = 200$.

In Tables 1-8, we show statistics concerning how many times each order in the set $\{1, \dots, p_{\max}\}$, where $p_{\max} = 15$, is selected by the IT criteria SBC, FPE, RNML and AICc. Note that in the first column of each table, the true VAR-order is written in bold.

In Figs. 1-2 (which are the same as Figs. 3-4 in [1]), we show boxplots for the average distances from the estimated sparsity pattern to the true one. These are graphical representations for the results in Tables 9-16. The convention in these tables is that the columns correspond to the IT criteria applied for VAR-order selection, while the rows correspond to the IT criteria used to choose the Granger sparsity pattern. For the results reported in the last column of such a table, we make the assumption that the order of the model is known. This is why the last column has the label “ORACLE”.

Amongst the three estimation methods that we investigate, only “LS with Granger constraints” does not guarantee that the model fitted to the data is stable. In Tables 17-24, we give the empirical probability that models estimated with this method are not stable. The labels for the rows and for the columns are the same as in Tables 9-16. The only difference is that there is a row labeled “ORACLE”, which corresponds to the case when the Granger sparsity pattern is assumed to be known.

For the description of the forecasting experiments, we refer to [1]. In this document, we reproduce Figs. 1-2 from [1], which are named Figs. 3-4. They are graphical representations of the results reported in Tables 25-32. Note that the

significance of the IT criteria assigned to the rows and to the columns of the Tables 25-32 is the same as in the Tables 17-24.

VAR-order estimation

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	44	0	44	0
2	56	0	63	4
3	100	200	93	196
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0
7	0	0	0	0
8	0	0	0	0
9	0	0	0	0
10	0	0	0	0
11	0	0	0	0
12	0	0	0	0
13	0	0	0	0
14	0	0	0	0
15	0	0	0	0

Table 1: Exp₁: $N = 200$

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	9	0	8	0
2	34	0	26	0
3	157	200	166	200
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0
7	0	0	0	0
8	0	0	0	0
9	0	0	0	0
10	0	0	0	0
11	0	0	0	0
12	0	0	0	0
13	0	0	0	0
14	0	0	0	0
15	0	0	0	0

Table 2: Exp₁: $N = 300$

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	3	0	3	0
2	9	0	5	0
3	188	200	192	200
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0
7	0	0	0	0
8	0	0	0	0
9	0	0	0	0
10	0	0	0	0
11	0	0	0	0
12	0	0	0	0
13	0	0	0	0
14	0	0	0	0
15	0	0	0	0

Table 3: Exp₁: $N = 400$

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	0	0	1	0
2	5	0	1	0
3	195	200	198	200
4	0	0	0	0
5	0	0	0	0
6	0	0	0	0
7	0	0	0	0
8	0	0	0	0
9	0	0	0	0
10	0	0	0	0
11	0	0	0	0
12	0	0	0	0
13	0	0	0	0
14	0	0	0	0
15	0	0	0	0

Table 4: Exp₁: $N = 500$

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	7	0	3	0
2	9	0	10	0
3	20	0	20	0
4	15	0	11	0
5	29	0	22	4
6	19	1	19	1
7	22	4	23	14
8	24	12	24	20
9	24	26	26	39
10	31	147	42	122
11	0	7	0	0
12	0	2	0	0
13	0	0	0	0
14	0	0	0	0
15	0	1	0	0

Table 5: Exp₂: $N = 200$

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	0	0	0	0
2	1	0	0	0
3	3	0	1	0
4	5	0	2	0
5	10	0	8	0
6	9	0	6	0
7	16	0	9	0
8	18	3	21	4
9	39	9	32	15
10	99	183	121	180
11	0	3	0	1
12	0	2	0	0
13	0	0	0	0
14	0	0	0	0
15	0	0	0	0

Table 7: Exp₂: $N = 400$

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	2	0	1	0
2	4	0	2	0
3	8	0	5	0
4	7	0	6	0
5	14	0	15	0
6	17	0	9	0
7	22	1	15	3
8	23	4	21	10
9	38	19	38	26
10	65	169	88	161
11	0	5	0	0
12	0	1	0	0
13	0	1	0	0
14	0	0	0	0
15	0	0	0	0

Table 6: Exp₂: $N = 300$

VAR Order	IT Criterion			
	SBC	FPE	RNML	AICc
1	0	0	0	0
2	0	0	0	0
3	2	0	1	0
4	1	0	1	0
5	5	0	6	0
6	4	0	4	0
7	17	0	8	0
8	17	1	17	3
9	29	9	26	8
10	125	178	137	185
11	0	8	0	4
12	0	2	0	0
13	0	1	0	0
14	0	1	0	0
15	0	0	0	0

Table 8: Exp₂: $N = 500$

Average distance from the estimated sparsity pattern $\hat{\mathcal{G}}$ to the true sparsity pattern \mathcal{G}

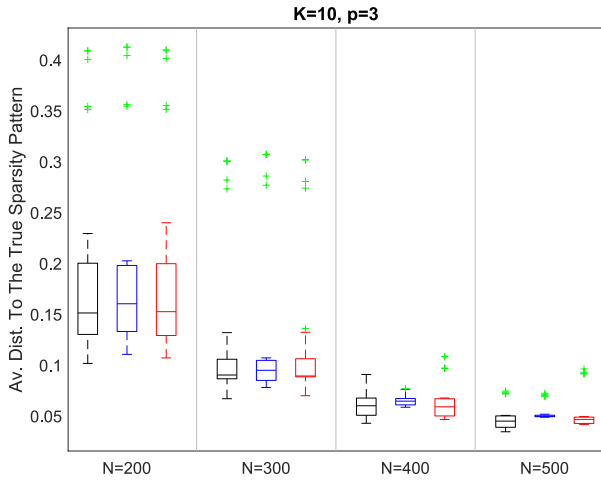


Fig. 1: Exp₁: Average distances from the estimated sparsity pattern to the true one. For each sample size (N), a different color is used for each boxplot, depending on the convex optimization problem solved in the training phase: the first boxplot (in black) is for LS with Granger constraints, the second one (in blue) is for the estimation method from [2] and the last one (in red) is for the iterative estimation method.

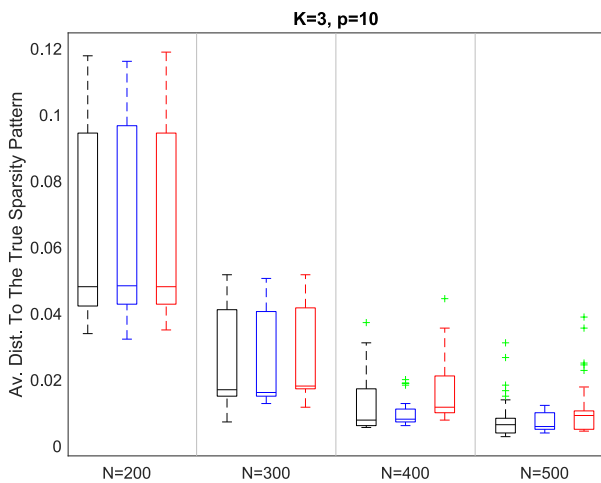


Fig. 2: Exp₂: Average distances from the estimated sparsity pattern to the true one. All graphical conventions are the same as in Fig. 1.

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.156	0.102	0.174	0.103	0.102
FPE	0.215	0.130	0.229	0.131	0.130
RNML	0.158	0.105	0.171	0.105	0.105
AICc	0.351	0.409	0.354	0.400	0.409
EBIC	0.187	0.145	0.198	0.145	0.145
ERNML	0.189	0.146	0.200	0.146	0.146
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.176	0.133	0.190	0.133	0.133
FPE	0.173	0.111	0.198	0.112	0.111
RNML	0.175	0.132	0.188	0.132	0.132
AICc	0.355	0.412	0.356	0.404	0.412
EBIC	0.191	0.146	0.203	0.146	0.146
ERNML	0.192	0.147	0.202	0.147	0.147
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.161	0.107	0.179	0.108	0.107
FPE	0.215	0.129	0.240	0.130	0.129
RNML	0.162	0.108	0.175	0.108	0.108
AICc	0.351	0.410	0.355	0.401	0.410
EBIC	0.188	0.143	0.198	0.143	0.143
ERNML	0.189	0.144	0.200	0.144	0.144

Table 9: Exp₁: $N = 200$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.088	0.067	0.088	0.067	0.067
FPE	0.132	0.103	0.127	0.103	0.103
RNML	0.087	0.067	0.085	0.067	0.067
AICc	0.273	0.300	0.282	0.300	0.300
EBIC	0.105	0.089	0.103	0.089	0.089
ERNML	0.106	0.090	0.103	0.090	0.090
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.102	0.085	0.102	0.085	0.085
FPE	0.101	0.078	0.101	0.078	0.078
RNML	0.100	0.084	0.099	0.084	0.084
AICc	0.277	0.307	0.286	0.307	0.307
EBIC	0.106	0.090	0.104	0.090	0.090
ERNML	0.107	0.091	0.105	0.091	0.091
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.089	0.070	0.088	0.070	0.070
FPE	0.132	0.106	0.136	0.106	0.106
RNML	0.089	0.070	0.087	0.070	0.070
AICc	0.274	0.302	0.281	0.302	0.302
EBIC	0.105	0.088	0.102	0.088	0.088
ERNML	0.106	0.089	0.102	0.089	0.089

Table 10: Exp₁: $N = 300$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.056	0.050	0.056	0.050	0.050
FPE	0.091	0.082	0.090	0.082	0.082
RNML	0.056	0.051	0.056	0.051	0.051
AICc	0.053	0.043	0.052	0.043	0.043
EBIC	0.069	0.064	0.068	0.064	0.064
ERNML	0.068	0.064	0.067	0.064	0.064
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.066	0.061	0.066	0.061	0.061
FPE	0.076	0.067	0.077	0.067	0.067
RNML	0.065	0.060	0.065	0.060	0.060
AICc	0.067	0.059	0.067	0.059	0.059
EBIC	0.069	0.064	0.068	0.064	0.064
ERNML	0.069	0.064	0.068	0.064	0.064
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.055	0.050	0.055	0.050	0.050
FPE	0.108	0.097	0.108	0.097	0.097
RNML	0.056	0.050	0.055	0.050	0.050
AICc	0.055	0.047	0.053	0.047	0.047
EBIC	0.067	0.062	0.067	0.062	0.062
ERNML	0.068	0.063	0.067	0.063	0.063

Table 11: Exp₁: $N = 400$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.040	0.040	0.041	0.040	0.040
FPE	0.073	0.072	0.074	0.072	0.072
RNML	0.040	0.039	0.040	0.039	0.039
AICc	0.036	0.034	0.037	0.034	0.034
EBIC	0.050	0.049	0.050	0.049	0.049
ERNML	0.051	0.049	0.050	0.049	0.049
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.051	0.050	0.051	0.050	0.050
FPE	0.071	0.070	0.072	0.070	0.070
RNML	0.050	0.049	0.050	0.049	0.049
AICc	0.051	0.049	0.052	0.049	0.049
EBIC	0.050	0.049	0.050	0.049	0.049
ERNML	0.051	0.049	0.050	0.049	0.049
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.045	0.044	0.045	0.044	0.044
FPE	0.093	0.092	0.096	0.092	0.092
RNML	0.042	0.042	0.043	0.042	0.042
AICc	0.044	0.042	0.045	0.042	0.042
EBIC	0.049	0.048	0.049	0.048	0.048
ERNML	0.049	0.048	0.049	0.048	0.048

Table 12: Exp₁: $N = 500$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.098	0.043	0.091	0.049	0.044
FPE	0.109	0.037	0.098	0.038	0.034
RNML	0.094	0.041	0.091	0.047	0.042
AICc	0.081	0.034	0.073	0.036	0.038
EBIC	0.118	0.044	0.102	0.050	0.044
ERNML	0.117	0.044	0.102	0.051	0.045
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.107	0.043	0.097	0.048	0.043
FPE	0.084	0.032	0.078	0.033	0.035
RNML	0.103	0.041	0.099	0.048	0.042
AICc	0.088	0.039	0.083	0.043	0.041
EBIC	0.116	0.044	0.102	0.051	0.045
ERNML	0.116	0.044	0.103	0.051	0.045
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.101	0.044	0.097	0.050	0.045
FPE	0.108	0.036	0.094	0.036	0.035
RNML	0.094	0.042	0.094	0.046	0.043
AICc	0.083	0.037	0.078	0.038	0.042
EBIC	0.119	0.046	0.105	0.051	0.046
ERNML	0.119	0.046	0.104	0.052	0.046

Table 13: Exp₂: $N = 200$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.046	0.014	0.038	0.016	0.015
FPE	0.052	0.019	0.043	0.019	0.018
RNML	0.046	0.014	0.038	0.015	0.015
AICc	0.032	0.008	0.027	0.007	0.009
EBIC	0.051	0.015	0.041	0.016	0.016
ERNML	0.051	0.015	0.042	0.016	0.016
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.047	0.016	0.041	0.017	0.016
FPE	0.034	0.013	0.029	0.015	0.013
RNML	0.047	0.015	0.041	0.016	0.016
AICc	0.033	0.014	0.029	0.015	0.014
EBIC	0.051	0.015	0.041	0.016	0.016
ERNML	0.051	0.016	0.041	0.017	0.016
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.047	0.017	0.039	0.018	0.017
FPE	0.050	0.021	0.044	0.020	0.018
RNML	0.047	0.017	0.040	0.017	0.017
AICc	0.035	0.013	0.031	0.012	0.013
EBIC	0.052	0.017	0.042	0.017	0.017
ERNML	0.050	0.017	0.042	0.017	0.017

Table 14: Exp₂: $N = 300$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.017	0.006	0.009	0.006	0.006
FPE	0.037	0.017	0.031	0.017	0.017
RNML	0.016	0.006	0.009	0.006	0.006
AICc	0.021	0.006	0.011	0.006	0.006
EBIC	0.019	0.006	0.010	0.006	0.006
ERNML	0.019	0.006	0.010	0.006	0.006
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.018	0.008	0.011	0.008	0.008
FPE	0.020	0.009	0.013	0.009	0.008
RNML	0.019	0.007	0.011	0.007	0.007
AICc	0.019	0.008	0.011	0.008	0.007
EBIC	0.019	0.006	0.010	0.006	0.006
ERNML	0.019	0.006	0.010	0.006	0.006
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.022	0.010	0.014	0.012	0.010
FPE	0.044	0.021	0.036	0.023	0.021
RNML	0.022	0.010	0.014	0.012	0.010
AICc	0.027	0.009	0.017	0.011	0.009
EBIC	0.022	0.010	0.013	0.010	0.010
ERNML	0.020	0.008	0.011	0.008	0.008

Table 15: Exp₂: $N = 400$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.007	0.005	0.008	0.004	0.003
FPE	0.031	0.018	0.027	0.017	0.015
RNML	0.008	0.004	0.008	0.003	0.003
AICc	0.014	0.006	0.012	0.006	0.006
EBIC	0.008	0.005	0.008	0.004	0.003
ERNML	0.008	0.005	0.008	0.004	0.003
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.010	0.007	0.010	0.006	0.005
FPE	0.012	0.006	0.011	0.004	0.004
RNML	0.011	0.007	0.011	0.006	0.005
AICc	0.010	0.006	0.010	0.004	0.004
EBIC	0.009	0.006	0.009	0.005	0.004
ERNML	0.009	0.006	0.009	0.004	0.004
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.009	0.006	0.010	0.005	0.004
FPE	0.039	0.025	0.036	0.024	0.023
RNML	0.011	0.006	0.011	0.005	0.004
AICc	0.018	0.009	0.015	0.008	0.008
EBIC	0.011	0.006	0.011	0.005	0.004
ERNML	0.010	0.006	0.010	0.005	0.004

Table 16: Exp₂: $N = 500$

Empirical probability that the estimated model is not stable

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.105	0.170	0.095	0.175	0.170
FPE	0.110	0.165	0.100	0.165	0.165
RNML	0.100	0.165	0.095	0.170	0.165
AICc	0.080	0.140	0.075	0.140	0.140
EBIC	0.135	0.175	0.115	0.180	0.175
ERNML	0.150	0.180	0.140	0.185	0.180
ORACLE	0.100	0.160	0.095	0.160	0.160

Table 17: Exp₁: $N = 200$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.115	0.135	0.115	0.135	0.135
FPE	0.085	0.110	0.095	0.110	0.110
RNML	0.095	0.115	0.105	0.115	0.115
AICc	0.095	0.115	0.100	0.115	0.115
EBIC	0.100	0.125	0.105	0.125	0.125
ERNML	0.095	0.125	0.105	0.125	0.125
ORACLE	0.075	0.105	0.085	0.105	0.105

Table 18: Exp₁: $N = 300$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.110	0.110	0.110	0.110	0.110
FPE	0.090	0.095	0.090	0.095	0.095
RNML	0.115	0.110	0.110	0.110	0.110
AICc	0.095	0.100	0.095	0.100	0.100
EBIC	0.115	0.115	0.115	0.115	0.115
ERNML	0.115	0.115	0.115	0.115	0.115
ORACLE	0.080	0.085	0.075	0.085	0.085

Table 19: Exp₁: $N = 400$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.085	0.085	0.080	0.085	0.085
FPE	0.075	0.075	0.070	0.075	0.075
RNML	0.095	0.095	0.085	0.095	0.095
AICc	0.080	0.080	0.075	0.080	0.080
EBIC	0.100	0.100	0.090	0.100	0.100
ERNML	0.100	0.100	0.090	0.100	0.100
ORACLE	0.070	0.070	0.065	0.070	0.070

Table 20: Exp₁: $N = 500$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.140	0.305	0.150	0.305	0.325
FPE	0.155	0.300	0.170	0.300	0.325
RNML	0.140	0.305	0.150	0.305	0.325
AICc	0.140	0.305	0.150	0.305	0.325
EBIC	0.140	0.305	0.150	0.305	0.325
ERNML	0.140	0.305	0.145	0.305	0.325
ORACLE	0.140	0.295	0.150	0.295	0.315

Table 21: Exp₂: $N = 200$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.195	0.255	0.205	0.255	0.265
FPE	0.195	0.255	0.205	0.260	0.270
RNML	0.195	0.255	0.205	0.255	0.265
AICc	0.195	0.255	0.205	0.255	0.265
EBIC	0.195	0.255	0.205	0.255	0.265
ERNML	0.195	0.255	0.205	0.255	0.265
ORACLE	0.195	0.255	0.205	0.255	0.265

Table 22: Exp₂: $N = 300$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.180	0.245	0.210	0.250	0.250
FPE	0.185	0.245	0.205	0.250	0.250
RNML	0.180	0.245	0.210	0.250	0.250
AICc	0.185	0.245	0.210	0.250	0.250
EBIC	0.180	0.245	0.210	0.250	0.250
ERNML	0.180	0.245	0.210	0.250	0.250
ORACLE	0.180	0.245	0.210	0.250	0.250

Table 23: Exp₂: $N = 400$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	0.130	0.150	0.125	0.150	0.150
FPE	0.125	0.150	0.120	0.150	0.150
RNML	0.130	0.150	0.125	0.150	0.150
AICc	0.125	0.150	0.120	0.150	0.150
EBIC	0.130	0.150	0.125	0.150	0.150
ERNML	0.130	0.150	0.125	0.150	0.150
ORACLE	0.130	0.150	0.125	0.150	0.150

Table 24: Exp₂: $N = 500$

Prediction errors

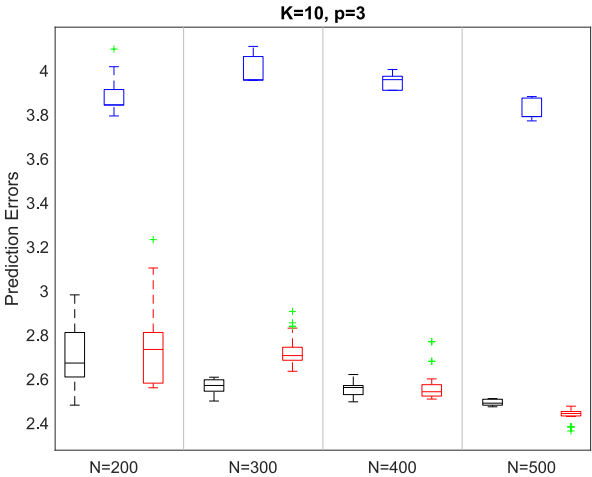


Fig. 3: Results of prediction in Exp₁. This plot is the same as the one in [1, Fig. 1].

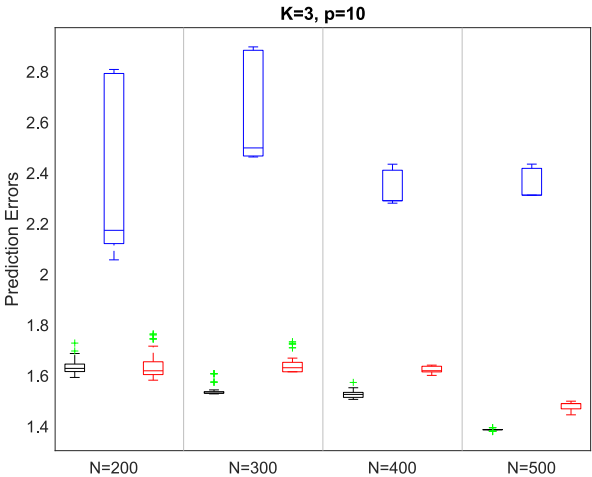


Fig. 4: Results of prediction in Exp₂. This plot is the same as the one in [1, Fig. 2].

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.668	2.549	2.690	2.549	2.549
FPE	2.711	2.666	2.757	2.666	2.666
RNML	2.672	2.597	2.694	2.597	2.597
AICc	2.675	2.642	2.656	2.642	2.642
EBIC	2.793	2.816	2.858	2.816	2.816
ERNML	2.888	2.866	2.981	2.866	2.866
ORACLE	2.694	2.481	2.695	2.481	2.481
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	3.994	3.842	3.914	3.842	3.842
FPE	3.993	3.853	3.914	3.845	3.853
RNML	3.994	3.842	3.909	3.842	3.842
AICc	3.886	3.793	3.886	3.793	3.793
EBIC	4.004	3.842	3.908	3.842	3.842
ERNML	4.004	3.842	3.908	3.842	3.842
ORACLE	4.097	3.844	4.017	3.844	3.844
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.723	2.560	2.877	2.560	2.560
FPE	2.733	2.564	2.815	2.564	2.564
RNML	2.728	2.597	2.948	2.597	2.597
AICc	2.798	2.624	2.938	2.624	2.624
EBIC	2.835	2.751	3.103	2.751	2.751
ERNML	2.860	2.766	3.232	2.766	2.766
ORACLE	2.785	2.575	3.101	2.575	2.575

Table 25: Exp₁: $N = 200$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.513	2.558	2.558	2.558	2.558
FPE	2.501	2.607	2.501	2.607	2.607
RNML	2.542	2.596	2.590	2.596	2.596
AICc	2.550	2.600	2.602	2.600	2.600
EBIC	2.550	2.570	2.550	2.570	2.570
ERNML	2.570	2.590	2.570	2.590	2.590
ORACLE	2.499	2.541	2.517	2.541	2.541
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	4.063	3.955	4.087	3.955	3.955
FPE	4.063	3.957	4.087	3.957	3.957
RNML	4.063	3.955	4.087	3.955	3.955
AICc	4.047	3.958	4.060	3.958	3.958
EBIC	4.063	3.955	4.109	3.955	3.955
ERNML	4.063	3.955	4.109	3.955	3.955
ORACLE	4.060	3.958	4.083	3.958	3.958
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.767	2.742	2.837	2.742	2.742
FPE	2.702	2.690	2.745	2.690	2.690
RNML	2.718	2.698	2.839	2.698	2.698
AICc	2.713	2.682	2.778	2.682	2.682
EBIC	2.733	2.662	2.830	2.662	2.662
ERNML	2.733	2.634	2.854	2.634	2.634
ORACLE	2.744	2.706	2.906	2.706	2.706

Table 26: Exp₁: $N = 300$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.614	2.570	2.594	2.570	2.570
FPE	2.565	2.554	2.565	2.554	2.554
RNML	2.620	2.561	2.594	2.561	2.561
AICc	2.496	2.496	2.496	2.496	2.496
EBIC	2.582	2.558	2.570	2.558	2.558
ERNML	2.602	2.570	2.582	2.570	2.570
ORACLE	2.559	2.505	2.520	2.505	2.505
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	3.973	3.910	3.942	3.910	3.910
FPE	3.989	3.958	3.989	3.958	3.958
RNML	3.973	3.910	3.942	3.910	3.910
AICc	3.989	3.958	3.989	3.958	3.958
EBIC	4.004	3.910	3.968	3.910	3.910
ERNML	4.004	3.910	3.968	3.910	3.910
ORACLE	4.003	3.958	3.988	3.958	3.958
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.588	2.522	2.559	2.522	2.522
FPE	2.579	2.508	2.544	2.508	2.508
RNML	2.588	2.534	2.563	2.534	2.534
AICc	2.563	2.522	2.540	2.522	2.522
EBIC	2.769	2.542	2.680	2.542	2.542
ERNML	2.768	2.542	2.680	2.542	2.542
ORACLE	2.600	2.522	2.577	2.522	2.522

Table 27: Exp₁: $N = 400$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.508	2.508	2.476	2.508	2.508
FPE	2.481	2.481	2.478	2.481	2.481
RNML	2.476	2.508	2.501	2.508	2.508
AICc	2.510	2.510	2.479	2.510	2.510
EBIC	2.473	2.490	2.490	2.490	2.490
ERNML	2.473	2.490	2.490	2.490	2.490
ORACLE	2.493	2.493	2.493	2.493	2.493
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	3.874	3.790	3.874	3.790	3.790
FPE	3.874	3.771	3.874	3.771	3.771
RNML	3.874	3.790	3.874	3.790	3.790
AICc	3.874	3.790	3.874	3.790	3.790
EBIC	3.881	3.790	3.881	3.790	3.790
ERNML	3.881	3.790	3.881	3.790	3.790
ORACLE	3.871	3.788	3.871	3.788	3.788
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.433	2.443	2.452	2.443	2.443
FPE	2.432	2.432	2.432	2.432	2.432
RNML	2.433	2.443	2.452	2.443	2.443
AICc	2.452	2.443	2.452	2.443	2.443
EBIC	2.429	2.462	2.476	2.462	2.462
ERNML	2.429	2.462	2.476	2.462	2.462
ORACLE	2.364	2.382	2.382	2.382	2.382

Table 28: Exp₁: $N = 500$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.675	1.605	1.624	1.635	1.616
FPE	1.728	1.601	1.628	1.611	1.612
RNML	1.667	1.615	1.634	1.645	1.616
AICc	1.696	1.606	1.634	1.635	1.616
EBIC	1.687	1.615	1.641	1.645	1.616
ERNML	1.687	1.615	1.641	1.645	1.616
ORACLE	1.681	1.592	1.631	1.611	1.612
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.792	2.120	2.792	2.173	2.056
FPE	2.792	2.120	2.800	2.145	2.056
RNML	2.792	2.120	2.792	2.173	2.056
AICc	2.802	2.120	2.808	2.173	2.056
EBIC	2.792	2.120	2.792	2.173	2.056
ERNML	2.792	2.120	2.792	2.173	2.056
ORACLE	2.794	2.119	2.793	2.143	2.056
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.743	1.598	1.623	1.618	1.616
FPE	1.745	1.594	1.632	1.598	1.581
RNML	1.715	1.603	1.642	1.628	1.616
AICc	1.745	1.598	1.624	1.605	1.610
EBIC	1.763	1.603	1.650	1.628	1.616
ERNML	1.763	1.603	1.655	1.628	1.616
ORACLE	1.759	1.592	1.658	1.601	1.599

Table 29: Exp₂: $N = 200$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.574	1.530	1.536	1.530	1.527
FPE	1.606	1.535	1.543	1.535	1.535
RNML	1.607	1.530	1.536	1.530	1.527
AICc	1.607	1.530	1.536	1.530	1.527
EBIC	1.574	1.530	1.536	1.530	1.527
ERNML	1.574	1.530	1.536	1.530	1.527
ORACLE	1.607	1.530	1.536	1.530	1.527
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.897	2.466	2.883	2.498	2.462
FPE	2.897	2.466	2.883	2.497	2.462
RNML	2.897	2.466	2.883	2.498	2.462
AICc	2.897	2.466	2.883	2.497	2.462
EBIC	2.897	2.466	2.883	2.498	2.462
ERNML	2.897	2.466	2.883	2.498	2.462
ORACLE	2.897	2.466	2.883	2.497	2.462
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.724	1.614	1.652	1.614	1.627
FPE	1.709	1.618	1.631	1.614	1.630
RNML	1.732	1.614	1.652	1.614	1.627
AICc	1.709	1.614	1.631	1.614	1.627
EBIC	1.724	1.614	1.652	1.614	1.627
ERNML	1.724	1.614	1.652	1.614	1.627
ORACLE	1.732	1.635	1.668	1.635	1.652

Table 30: Exp₂: $N = 300$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.551	1.520	1.533	1.505	1.525
FPE	1.572	1.520	1.537	1.505	1.525
RNML	1.551	1.520	1.533	1.505	1.525
AICc	1.551	1.520	1.533	1.505	1.525
EBIC	1.551	1.520	1.533	1.505	1.525
ERNML	1.551	1.520	1.533	1.505	1.525
ORACLE	1.551	1.510	1.533	1.505	1.511
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.431	2.290	2.410	2.290	2.280
FPE	2.431	2.290	2.410	2.290	2.280
RNML	2.430	2.290	2.410	2.290	2.280
AICc	2.431	2.290	2.410	2.290	2.280
EBIC	2.434	2.290	2.410	2.290	2.280
ERNML	2.434	2.290	2.410	2.290	2.280
ORACLE	2.431	2.290	2.410	2.290	2.280
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.615	1.614	1.604	1.631	1.600
FPE	1.638	1.618	1.619	1.631	1.605
RNML	1.615	1.631	1.604	1.631	1.617
AICc	1.615	1.614	1.604	1.631	1.600
EBIC	1.615	1.631	1.604	1.631	1.617
ERNML	1.641	1.641	1.619	1.641	1.640
ORACLE	1.641	1.641	1.616	1.641	1.640

Table 31: Exp₂: $N = 400$

LS with Granger constraints					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.386	1.386	1.386	1.386	1.394
FPE	1.379	1.386	1.379	1.379	1.380
RNML	1.386	1.386	1.386	1.386	1.394
AICc	1.379	1.386	1.379	1.379	1.380
EBIC	1.386	1.386	1.386	1.386	1.394
ERNML	1.386	1.386	1.386	1.386	1.394
ORACLE	1.386	1.386	1.386	1.386	1.394
Estimation method from [2]					
	SBC	FPE	RNML	AICc	ORACLE
SBC	2.434	2.312	2.417	2.312	2.312
FPE	2.434	2.312	2.417	2.312	2.312
RNML	2.434	2.312	2.417	2.312	2.312
AICc	2.434	2.312	2.417	2.312	2.312
EBIC	2.434	2.312	2.417	2.312	2.312
ERNML	2.434	2.312	2.417	2.312	2.312
ORACLE	2.434	2.312	2.417	2.312	2.312
Iterative estimation method					
	SBC	FPE	RNML	AICc	ORACLE
SBC	1.478	1.498	1.447	1.489	1.489
FPE	1.451	1.498	1.444	1.479	1.479
RNML	1.478	1.498	1.447	1.489	1.489
AICc	1.465	1.498	1.447	1.489	1.489
EBIC	1.478	1.498	1.447	1.489	1.489
ERNML	1.478	1.498	1.447	1.489	1.489
ORACLE	1.478	1.498	1.447	1.489	1.489

Table 32: Exp₂: $N = 500$

1. REFERENCES

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