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\*

:

, ARMAX ARX :

BJ OE

ARX(1,7,3)

(4)

## A Comparison between the Prediction of State Space models and Stochastic Dynamic Linear Systems with Application

### Abstract:

This Paper is concerned with the synchronizing between numbers of dynamic linear system models with different parameters with their two different kinds. The first kind represented the equation error model which included ARMAX and ARX, also the second kind was the output error model which contained BJ and OE , and, another state space models, After the application of the above models on the real data ,the comparison has been done, we choose the ARX(1,7,3) and the state space model with 4 parameters which gave the minimum statistical

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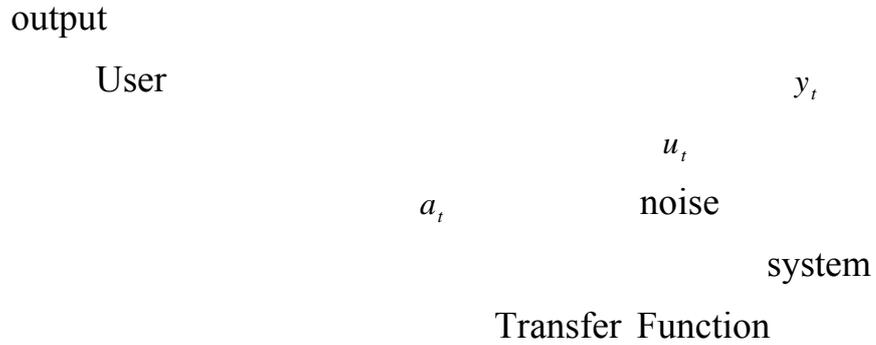
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criteria. These two models are also used in forecasting and another comparison has been done between two forecasted models, and we conclude that the forecasted dynamic linear system model is the best which gave the minimum statistical criteria.

**Introduction .1**



system Identification

Dynamic Systems  
(Ljung,1999)

(1981 )

**Objective research .2**

Linear State Space  
Stochastic Dynamic Systems

(Box and Jenkins(1976)

## State-Space Models

.3

internal model

.(Kanjilal, 1995)

(2002 ) (Wei,1990) .

$$\begin{array}{r}
 U_t \quad X_t \quad X_{t+1} \quad (t+1) \\
 Y_t \quad \text{State Equation} \\
 U_t \quad X_t \\
 \text{Observation Equation}
 \end{array}$$

-: SISO (Single Input Single Output)

$$X_{t+1} = AX_t + BU_t \quad \dots (1a)$$

$$Y_t = CX_t + DU_t \quad \dots (1b)$$

-:

$$A : (n \times n)$$

$$B : (n \times n)$$

$$D : (n \times 1)$$

.(Nelles,2001)

(Makridakis,1998)

Recursive

Equation

Kalman Filter

.(Kalman,1960)

**Stochastic Linear**

**.4**

**Dynamic Systems**

General Linear Model

Linear Models

$y_t$

$G(q)$

Input Transfer Function

$H(q)$

$v_t$

Noise Transfer Function

:

$$y_t = G(q)u_t + H(q)v_t \quad \dots \quad (2)$$

Linear

:

Combination

$$y_t = \frac{B(q)}{F(q)A(q)}u_t + \frac{C(q)}{D(q)A(q)}v_t \quad \dots \quad (3)$$

:

$$A(q) = 1 + a_1q^{-1} + a_2q^{-2} + \dots + a_naq^{-na}$$

$$B(q) = b_1q^{-1} + b_2q^{-2} + \dots + b_n bq^{-nb}$$

$$C(q) = 1 + c_1q^{-1} + c_2q^{-2} + \dots + c_{nc}q^{-nc}$$

$$D(q) = 1 + d_1q^{-1} + d_2q^{-2} + \dots + d_{nd}q^{-nd}$$

$$F(q) = 1 + f_1q^{-1} + f_2q^{-2} + \dots + f_{nf}q^{-nf}$$

$nf, nd, nc, nb, na$

$$q^{-1} \quad F(q), D(q), C(q), B(q), A(q)$$

: Backword Shift

$$q^{-1}u_t = u_{t-1}$$

$$q^{-2}u_t = q(qu_t)$$

$$= q(u_{t-1})$$

$$= u_{t-2}$$

:

**Equation Error Models**

:

ARX

ARMAX

$$1/A(q)$$

: ARX

$$y_t = \frac{B(q)}{A(q)} u_t + \frac{1}{A(q)} v_t \quad \dots \quad (4)$$

:

ARMAX

$$y_t = \frac{B(q)}{A(q)} u_t + \frac{C(q)}{A(q)} v_t \quad \dots \quad (5)$$

**Output Error Models**

:

BJ

-

OE

.(Ljung & Söderström,1983)

(Nelles,2001)

$$y_t = \frac{B(q)}{F(q)} u_t + v_t \quad \dots \quad (6)$$

$$y_t = \frac{B(q)}{F(q)} u_t + \frac{C(q)}{D(q)} v_t \quad \dots \quad (7)$$

Goodness of

Accuracy

Fit

(2005 ) (Makridakis,1998)

**Mean Square Error .1**

$$S_e^2 = \frac{1}{n-m} \sum_{i=1}^n e_i^2$$

$$= \frac{1}{n-m} \sum_{i=1}^n \left( Y_i - \hat{Y}_i \right)^2 \quad \dots (8)$$

**Mean Absolute Error .2**

$$MAE = \sum_{i=1}^n |e_i| / n - m$$

$$e_t = Y_t - F_t \quad \dots (9)$$

:  $F_t$

:  $Y_t$

:  $e_t$

**Mean Absolute Percentage Error .3**

:

$$MAPE = \frac{1}{n - m} \sum_{t=1}^n PE_t$$

$$PE = \left( \frac{Y_t - F_t}{Y_t} \right) * 100 \quad \dots(10)$$

**Mean Percentage Error .4**

:

$$MPE = \frac{1}{n - m} \sum_{t=1}^n PE_t \quad \dots(11)$$

**.6**

$U_t$  Box and Jenkins(1976)

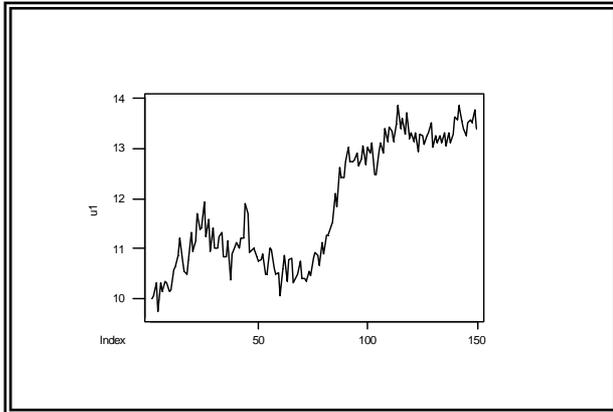
Sales

$Y_t$

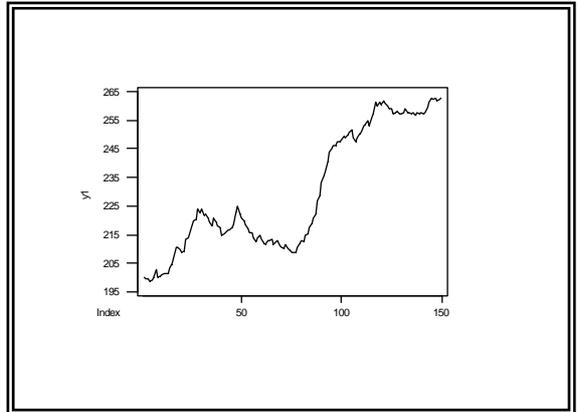
Leading Indicator

(150)

:



:(2)



:(1)

Box (2) (1)  
130 and Jenkins(1976)

( )  
OE , BJ ARMAX , ARX

AIC , FPE , Loss Function

. (1)

: (1)

	na	nb	nc	nd	nf	nk	AIC	Loss fun.	FPE
<b>ARX</b>	<b>1</b>	<b>7</b>	<b>-</b>	<b>-</b>	<b>-</b>	<b>3</b>	<b>-1.0393</b>	<b>0.3013</b>	<b>0.35380</b>
<b>ARMAX</b>	<b>7</b>	<b>2</b>	<b>3</b>	<b>-</b>	<b>-</b>	<b>3</b>	<b>-0.6945</b>	<b>0.3215</b>	<b>0.5029</b>
<b>OE</b>	<b>-</b>	<b>3</b>	<b>-</b>	<b>-</b>	<b>7</b>	<b>3</b>	<b>-0.7830</b>	<b>0.3252</b>	<b>0.4585</b>
<b>BJ</b>	<b>-</b>	<b>5</b>	<b>2</b>	<b>2</b>	<b>1</b>	<b>3</b>	<b>-0.83481</b>	<b>0.3088</b>	<b>0.4354</b>

(1)

FPE , AIC , Loss function

ARX

:

ARX

$$y_t = 0.6819y_{t-1} + 0.01098y_{t-2} + 0.007196y_{t-3} + 0.01836y_{t-4} + 0.121 u_{t-1} - 0.04565 u_{t-2} + 4.653 u_{t-3} + 0.2658 u_{t-4} + a_t - 0.8056 a_{t-1} - 0.02864 a_{t-2} - 0.0189 a_{t-3} + 0.0535 a_{t-4} \quad \dots (12)$$

( 2 )

( 8 )

: (13)

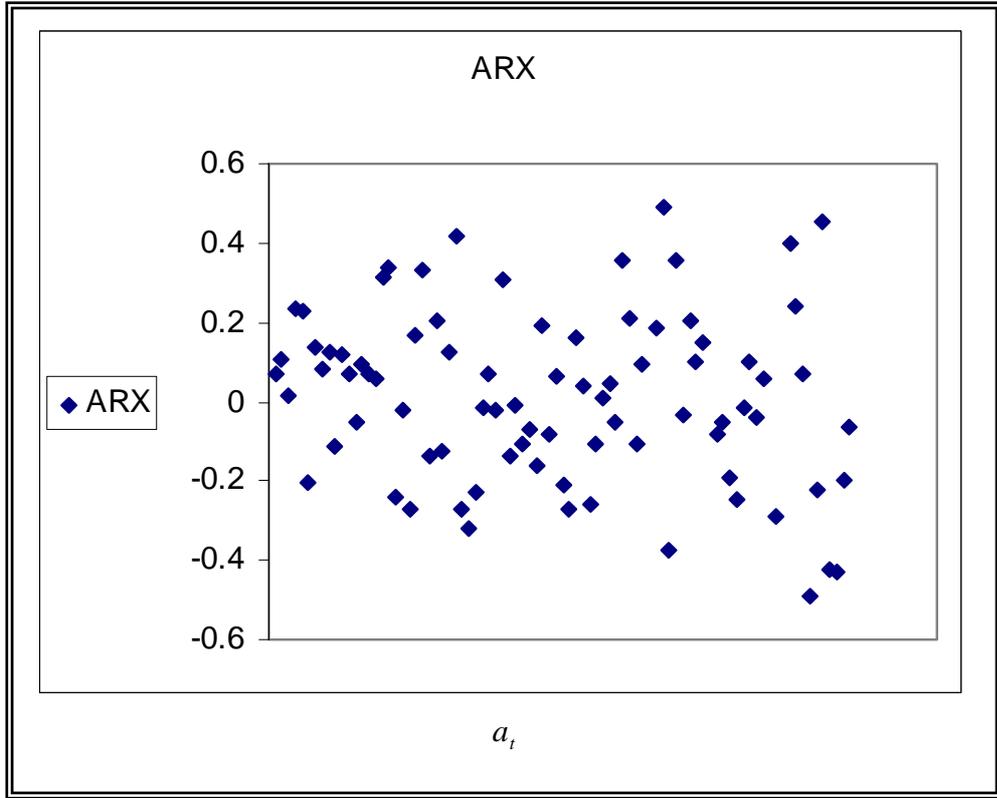
$$\begin{aligned}
 a_t = & y_t - 0.6819y_{t-1} - 0.01098y_{t-2} - 0.007196y_{t-3} - 0.01836y_{t-4} - 0.121 u_{t-1} \\
 & + 0.04565 u_{t-2} - 4.653 u_{t-3} - 0.2658 u_{t-4} + 0.8056 a_{t-1} \quad \dots (13) \\
 & + 0.02864 a_{t-2} + 0.0189 a_{t-3} - 0.0535 a_{t-4}
 \end{aligned}$$

.  $a(1:4) = 0$  :

**ARX**

: ( 2 )

$t$	$a_t$										
1	-0.17254	23	-0.12304	45	-0.04797	67	0.35780	89	0.28545	111	0.306950
2	0.34448	24	-0.94577	46	-0.42959	68	-0.00080	90	0.53696	112	-0.226230
3	0.31620	25	1.79610	47	0.19045	69	-0.18349	91	0.18213	113	0.676050
4	-0.13044	26	0.82894	48	-0.46106	70	-0.10805	92	0.06933	114	-0.092710
5	-0.04108	27	-0.32214	49	0.00622	71	0.30056	93	0.45053	115	0.295510
6	0.10972	28	-0.53113	50	0.08611	72	0.04960	94	-0.11658	116	-0.008975
7	0.59461	29	-0.07669	51	-0.06747	73	0.09913	95	-0.49104	117	0.287360
8	-0.14229	30	0.29457	52	3.76290	74	0.17128	96	0.23184	118	0.056140
$t$	$a_t$										
9	-0.13771	31	0.03684	53	-0.40609	75	0.25055	97	0.27768	119	-0.153610
10	-0.29126	32	-0.06331	54	-0.94125	76	-0.07910	98	-0.23061	120	-0.329300
11	-0.13992	33	-0.15640	55	-0.49222	77	0.50567	99	0.37420	121	0.306950
12	0.64144	34	-0.00650	56	-0.12004	78	0.27438	100	0.43369	122	-0.226230
13	0.46192	35	0.18569	57	-0.41645	79	0.10557	101	0.31384	123	0.676050
14	0.89296	36	-0.07208	58	-0.90451	80	0.19271	102	-0.90155		
15	0.30279	37	0.31138	59	-0.01766	81	-0.48758	103	-2.96870		
16	-0.41914	38	-0.23996	60	-0.33586	82	0.86591	104	2.84010		
17	-0.38453	39	0.05149	61	0.19740	83	-0.07682	105	0.50301		
18	0.42565	40	-0.21594	62	-0.64806	84	0.57393	106	0.93782		
19	0.10797	41	-0.20168	63	0.67939	85	0.21069	107	-0.70005		
20	0.12077	42	-0.01657	64	0.02651	86	0.15853	108	0.71305		
21	0.32501	43	0.68354	65	-0.84308	87	0.11728	109	0.30713		
22	0.50361	44	0.37717	66	-0.09759	88	0.37552	110	-0.28792		



**ARX**

**:(3)**

**:** (3)

$$\begin{aligned} \hat{y}_{131} = & 0.6819y_{130} + 0.01098y_{129} + 0.007196y_{128} + 0.01836y_{127} + 0.121 u_{130} \\ & -0.04565 u_{129} + 4.653 u_{128} + 0.2658 u_{127} -0.8056 a_{130} \quad \dots (14) \\ & -0.02864 a_{129} -0.0189 a_{128} + 0.0535 a_{127} \end{aligned}$$

ARX : (3)

$y$	$\hat{y}$	$y$	$\hat{y}$
1.3	1.43330	0.2	0.47086
-1.1	-1.15820	2.1	1.90840
-0.1	-0.14208	1.5	1.29420
-0.5	-0.43112	1.8	1.99490
0.3	0.53536	-0.6	0.06947
-0.7	-0.60098	0.5	-0.54899
0.7	0.31739	-1.0	-0.88468
-0.5	-0.90083	0.4	0.47787
0.6	0.85002	0.5	0.46385
-0.3	-0.54685		

: (4)

:(4)

	AIC	Loss fun.	FPE
<b>1</b>	<b>0.5310</b>	<b>1.3924</b>	<b>1.70185</b>
<b>2</b>	<b>0.2839</b>	<b>0.890379</b>	<b>1.33557</b>
<b>3</b>	<b>-0.4758</b>	<b>0.341028</b>	<b>0.633337</b>
<b>4</b>	<b>-0.9273</b>	<b>0.13246</b>	<b>0.16858</b>
<b>5</b>	<b>0.6018</b>	<b>0.481181</b>	<b>2.4059</b>

(4)

: FPE Loss function AIC

**State space model:**

$$A = \begin{bmatrix} -0.571 & 2.7689 & -0.80335 & -1.3188 & 1.1337 \\ -0.67978 & 0.87412 & -0.50666 & 0.64417 & 0.24174 \\ -2.9175 & 4.7416 & -1.9105 & -1.6517 & 2.1577 \\ -0.29977 & -0.3237 & 0.48917 & -0.25331 & -0.82795 \\ 1.4372 & -2.2158 & 0.76012 & 1.571 & -1.1798 \end{bmatrix}$$

$$B = \begin{bmatrix} 102.53 \\ 20.271 \\ 159.53 \\ 3.2534 \\ -93.239 \end{bmatrix}, \quad K = \begin{bmatrix} 25.972 \\ 7.8994 \\ 46.546 \\ -4.1389 \\ -22.47 \end{bmatrix}, \quad X(0) = \begin{bmatrix} 3.2701 \\ 1.2645 \\ 13.872 \\ -7.7074 \\ -1.3809 \end{bmatrix}$$

$$C = [1.504 \quad 3.603 \quad 2.9059 \quad -0.76076 \quad -3.1034]$$

MATLAB

D=0

: ARMAX

$$y_t = -0.6819 y_{t-1} + 0.01098 y_{t-2} + 0.007196 y_{t-3} + 0.01836 y_{t-4} + 0.121 u_{t-1} - 0.04565 u_{t-2} + 4.653 u_{t-3} + 0.2658 u_{t-4} + a_t - 0.8056 a_{t-1} - 0.02864 a_{t-2} - 0.0189 a_{t-3} + 0.05354 a_{t-4} \quad \dots (15)$$

(5)

(15)

:

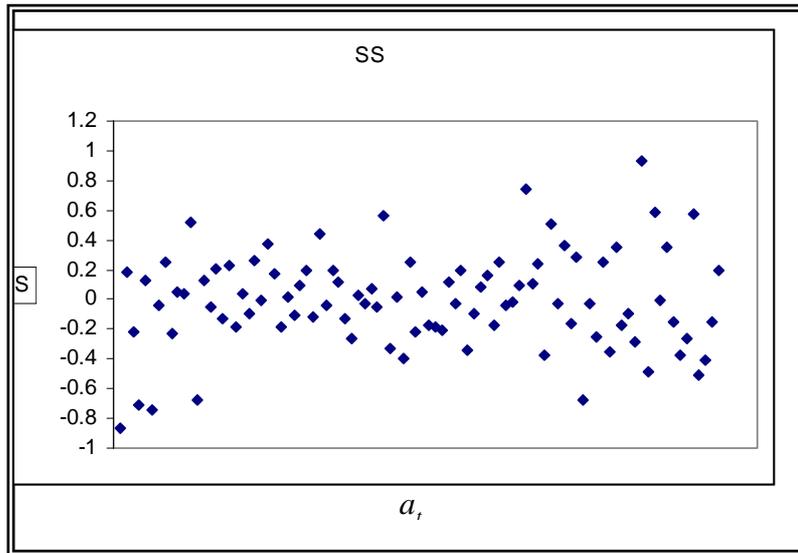
$$\begin{aligned}
 a_t = & y_t + 0.6819 y_{t-1} - 0.01098 y_{t-2} - 0.007196 y_{t-3} - 0.01836 y_{t-4} \\
 & - 0.121 u_{t-1} + 0.04565 u_{t-2} - 4.653 u_{t-3} - 0.2658 u_{t-4} \\
 & + 0.8056 a_{t-1} + 0.02864 a_{t-2} + 0.0189 a_{t-3} - 0.05354 a_{t-4} \quad \dots (16)
 \end{aligned}$$

$$a(1:4) = 0 :$$

: (5)

$t$	$a_t$										
1	-0.12636	23	0.14871	45	0.02452	67	0.15356	89	-0.24872	111	0.033978
2	4.12070	24	-0.01480	46	0.14558	68	-0.52765	90	0.21228	112	0.969810
3	-3.50950	25	0.09140	47	0.05279	69	0.27359	91	-0.10010	113	-0.632910
4	0.11630	26	0.17956	48	-0.28068	70	-0.14511	92	-0.48746	114	0.521690
5	-0.23336	27	0.21065	49	-0.09014	71	-0.53290	93	-0.04342	115	0.108240
6	-0.60129	28	-0.07345	50	0.07001	72	0.00620	94	0.08297	116	-0.176520
7	0.20720	29	-0.29614	51	0.18744	73	0.14212	95	-0.26855	117	0.547170
8	0.11612	30	-1.49690	52	-0.35183	74	-0.29061	96	-0.04436	118	-0.526560
9	0.09286	31	1.82680	53	0.30978	75	-0.10780	97	0.07614	119	0.290440
10	-0.18412	32	0.50195	54	-0.15893	76	-0.04461	98	-0.21878	120	-0.032632
11	0.01448	33	-0.00365	55	0.12836	77	0.34710	99	0.33044	121	0.196460
12	0.22695	34	-0.35922	56	0.08048	78	0.16916	100	-0.20587	122	0.026626
13	-0.16314	35	0.43058	57	0.06390	79	0.14948	101	-0.44311	123	0.057200
14	0.02740	36	0.27113	58	4.37440	80	-0.14421	102	0.10289	124	0.269610
15	-0.20105	37	-0.14200	59	-0.11795	81	-0.03861	103	0.05590	125	-0.245900
16	-0.10703	38	-0.01945	60	-0.80631	82	-0.23464	104	-0.08881	126	-0.122140
17	0.43211	39	-0.08221	61	-0.27407	83	0.37776	105	0.10239		
18	-0.15434	40	-0.13830	62	-0.33671	84	0.04565	106	-0.06068		
19	0.15843	41	0.24408	63	-0.40476	85	-0.02520	107	0.15095		
20	0.15392	42	0.51206	64	-0.55500	86	-0.11924	108	-0.41427		
21	-0.10191	43	0.33479	65	0.22126	87	-0.62588	109	-2.76470		
22	-0.12406	44	-0.24401	66	-0.47512	88	0.58583	110	2.65740		

:



:(4)

(6)

$$\hat{y}_{131} = -0.6819 y_{130} + 0.01098 y_{129} + 0.007196 y_{128} + 0.01836 y_{127} + 0.121 u_{130} - 0.04565 u_{129} + 4.653 u_{128} + 0.2658 u_{127} - 0.8056 a_{130} - 0.02864 a_{129} - 0.0189 a_{128} + 0.05354 a_{127} \quad \dots (17)$$

$a_{t+l}$

:(6)

$y$	$\hat{y}$	$y$	$\hat{y}$
1.3	1.35330	0.2	0.39330
-1.1	-1.35810	2.1	2.05840
-0.1	0.03841	1.5	1.18150
-0.5	-0.54831	1.8	2.15670
0.3	0.28831	-0.6	-0.01317
-0.7	-0.53254	0.5	-0.55018
0.7	0.52341	-1.0	-1.11620
-0.5	-0.76164	0.4	0.38092
0.6	0.66445	0.5	0.71664
-0.3	-0.47883		

ARX

ARX

, FPE AIC

:

:(7)

	MSE	MAE	MAPE	MSF	MPE
ARX(1,7,3)	0.045282 6	0.42982 8	86.498 7	0.20754 8	5.48728 -
SS(4)	0.100519	0.60825 9	122.65 7	0.29010 4	27.9940

ARX(1, 7, 3)

SS(4)

.6

:

.1

( 3 )

(6)

.2

- .7
- ":(2002) .1
- .2
- "(1981) .3
- ( ) "
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