



TIME SERIES MODELING & ANALYSIS

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Lab#: 9

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ABSTRACT

The main purpose of this LAB is to implement the GPAC array covered in lecture using Python program and test the accuracy of your code using an ARMA(n_a, n_b) model. Using statsmodels we simulated the ARMA process. GPAC code that generates GPAC table for various numbers of rows and columns was coded from scratch and was shown that the General Partial Autocorrelation Function (GPAC), which has recently been suggested to be used as one of a set of convenient tools for order identification in ARMA models.

INTRODUCTION

The generalized partial autocorrelation is used to estimate the order of ARMA model when $n_a \neq 0$ and $n_b \neq 0$. The generalized partial autocorrelation (GPAC) function was introduced by Woodward and Gray (1981) for purposes of model identification in the ARMA(p,q) setting. The GPAC function is an extension of the partial autocorrelation function used by Box and Jenkins (1975) in ARMA model identification. Woodward and Gray (1981) used an array to present the information in the GPAC function, and this array was shown to be related to the S-array of Gray, Kelley, and McIntire (1978). Woodward and Gray (1981) showed that the GPAC array uniquely determines p and q when the true autocorrelation is known, a property it shares with the S-array. Unique identification of p and q when the true autocorrelation function is known is only assured using the Box-Jenkins approach when either $p=0$ or $q=0$. Woodward and Gray (1981) discussed the use of the GPAC based on single, finite length realizations, and showed examples in which the model identifying pattern in the GPAC was clearly discernible. Davies and Petrucci (1984) presented simulation evidence and real data examples to argue that the sample GPAC array is unstable when applied to time series of only moderate length and that its use in detecting MA components is limited.

$$\phi_{kk}^j = \frac{\begin{vmatrix} \hat{R}_y(j) & \hat{R}_y(j-1) & \dots & \hat{R}_y(j+1) \\ \hat{R}_y(j+1) & \hat{R}_y(j) & \dots & \hat{R}_y(j+2) \\ \vdots & \vdots & \ddots & \vdots \\ \hat{R}_y(j+k-1) & \hat{R}_y(j+k-2) & \dots & \hat{R}_y(j+k) \end{vmatrix}}{\begin{vmatrix} \hat{R}_y(j) & \hat{R}_y(j-1) & \dots & \hat{R}_y(j-k+1) \\ \hat{R}_y(j+1) & \hat{R}_y(j) & \dots & \hat{R}_y(j-k+2) \\ \vdots & \vdots & \ddots & \vdots \\ \hat{R}_y(j+k-1) & \hat{R}_y(j+k-2) & \dots & \hat{R}_y(j) \end{vmatrix}}$$

$j \backslash k$	1	2	...	n_a
0	ϕ_{11}^0	ϕ_{22}^0	...	
1	ϕ_{11}^1	ϕ_{22}^1	...	
\vdots	\vdots	\vdots		
n_b				

$-a_{n_a}$	0	0	0
$-a_{n_a}$	$\frac{0}{0}$	$\frac{0}{0}$	$\frac{0}{0}$
$-a_{n_a}$	$\frac{0}{0}$	$\frac{0}{0}$	$\frac{0}{0}$

METHOD, THEORY & PROCEDURES

Method:

1. Programming Language: Python

Libraries used: Some basic libraries used for analysis & model building are mentioned below

library(Numpy) - large collection of high-level mathematical functions to operate on these arrays.

library (Pandas) – For Data manipulation and analysis

library(Matplotlib) – is a system for declaratively creating graphics

library(Math) –To Compute mathematical calculations

library (statsmodels) – Import statistical models

library (scipy) – Scientific Computations

Theory:

To Implement GPAC array.

Procedure:

I shall be looking at the results of GPAC models and Perform various plots and infer about it in my analysis.

And through my exploration I shall try to identify which methods perform better and draw inferences.

The Dataset will be explored in following stages:

1. **Data Exploration (EDA)** – looking at the models and making inferences about the data.
2. **Data Visualization** – Plotting different time series plots for the regression method and forecast accuracy.
3. **Testing** – Running Autocorrelation, Pearson correlation test to identify the correlation between errors.

ANSWERS TO QUESTIONS

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1 C:\ProgramData\Anaconda3\python.exe "C:\Program Files\
  JetBrains\PyCharm 2019.3.1\plugins\python\helpers\pydev\
  pydevconsole.py" --mode=client --port=57324
2
3 import sys; print('Python %s on %s' % (sys.version, sys.
  platform))
4 sys.path.extend(['C:\\Users\\nsree_000\\Desktop\\Python-
  Quiz', 'C:/Users/nsree_000/Desktop/Python-Quiz'])
5
6 Python 3.7.4 (default, Aug  9 2019, 18:34:13) [MSC v.1915
  64 bit (AMD64)]
7 Type 'copyright', 'credits' or 'license' for more
  information
8 IPython 7.8.0 -- An enhanced Interactive Python. Type '?'
  for help.
9 PyDev console: using IPython 7.8.0
10
11 Python 3.7.4 (default, Aug  9 2019, 18:34:13) [MSC v.1915
  64 bit (AMD64)] on win32
12 In[2]: runfile('C:/Users/nsree_000/Desktop/Python-Quiz/TIME
  SERIES/labs-911/ma.py', wdir='C:/Users/nsree_000/Desktop/
  Python-Quiz/TIME SERIES/labs-911')
13 -----x(x) - 0.5x(x - 1
  ) = x(x)-----
14 Enter the number of data samples :>? 1000
15 Enter AR order:>? 1
16 Enter MA order:>? 0
17 Enter Mean of white noise:>? 1
18 Enter variance of white noise:>? 2
19 Enter coefficient of AR a1>? -0.5
20 Is this stationary process: True
21 GPAC Table for y:
22      1      2      3      4      5      6
23 0  0.50600  0.05640  0.04877  0.02592 -0.00626 -0.00726 -0.
  01795
24 1  0.58893 -0.37989  0.01895  0.03781 -0.03704  0.00617 -0.
  03041
25 2  0.69463 -0.15370  0.49020  0.02500 -0.14286 -0.00000 -0.
  00000
26 3  0.73430  1.15102  0.24000  0.00000  0.00000      NaN
  NaN
27 4  0.64474 -0.22695 -0.50000      NaN      NaN      NaN
  NaN
28 5  0.60204 -1.60938 -1.00000      NaN      NaN      NaN
  NaN

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Page 1 of 7

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29 6 0.42373 1.28155 2.33333 NaN NaN NaN
    NaN
30 Repeat for 5000 samples
31 Enter the number of data samples:>? 5000
32 GPAC Table for y:
33      1      2      3      4      5      6
    7
34 0 0.50000 -0.01467 -0.00656 -0.00441 -0.02780 -0.01250 0
    .01358
35 1 0.47800 -0.23818 0.00271 0.03763 -0.02617 -0.04392 -0
    .00415
36 2 0.45607 -0.33969 5.00000 0.00000 -0.04348 -0.07692 0
    .00000
37 3 0.42202 -2.61798 -2.00000 NaN -0.00000 -0.00000
    NaN
38 4 -0.04348 -0.45494 0.30000 NaN NaN NaN
    NaN
39 5 11.50000 -0.47170 -1.00000 NaN NaN NaN
    NaN
40 6 0.52174 0.68000 0.33333 NaN NaN NaN
    NaN
41 Repeat for 10000 samples
42 Enter the number of data samples:>? 10000
43 GPAC Table for y:
44      1      2      3      4      5      6
    7
45 0 0.50800 -0.00682 0.00549 0.00157 -0.00878 -0.00249 -0.
    00912
46 1 0.49803 0.40119 0.00662 0.03125 -0.00752 0.03571 -0.
    01316
47 2 0.51383 -0.14778 0.50000 -0.00000 -0.00000 -0.00000 -0.
    00000
48 3 0.52308 2.83333 1.00000 NaN NaN NaN
    NaN
49 4 0.42647 -0.18824 -1.00000 NaN NaN NaN
    NaN
50 5 0.34483 -1.37500 -0.00000 NaN NaN NaN
    NaN
51 6 -0.40000 -0.13636 NaN NaN NaN NaN
    NaN
52
53 -----y(t) = e(t) + 0.5e(t-1
    )-----
54 Enter the number of data samples :>? 10000
55 Enter AR order:>? 0
56 Enter MA order:>? 1

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Page 2 of 7

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57 Enter Mean of white noise:>? 1
58 Enter variance of white noise:>? 2
59 coefficient of MA b1>? 0.5
60 Is this stationary process: True
61 GPAC Table for y:
62      1      2      3      4      5      6
      7
63 0  0.39100 -0.19228  0.10767 -0.06502  0.01600 -0.02525  0
   .01194
64 1 -0.02558  0.01860 -0.00712 -0.03864 -0.08659 -0.01764 -0
   .00291
65 2 -0.80000  0.00660 -0.11321 -0.03571 -0.03226 -0.00000 -0
   .00000
66 3 -0.50000 -8.00000 -0.00000 -0.00000 -0.00000      NaN
   NaN
67 4  4.50000 -1.50000      NaN      NaN      NaN      NaN
   NaN
68 5  1.11111 -1.12500      NaN      NaN      NaN      NaN
   NaN
69 6  0.35000 -0.11111      NaN      NaN      NaN      NaN
   NaN
70 -----ARMA (1,1): y(t) + 0.5y(t-1) = e(t) +
   0.5e(t-1)-----
71 Enter the number of data samples :>? 10000
72 Enter AR order:>? 1
73 Enter MA order:>? 1
74 Enter Mean of white noise:>? 1
75 Enter variance of white noise:>? 2
76 Enter coefficient of AR a1>? 0.5
77 coefficient of MA b1>? 0.5
78 Is this stationary process: True
79 GPAC Table for y:
80      1      2      3      4      5      6
      7
81 0  0.01300  0.01283  0.00567 -0.00531  0.00198 -0.00496  0
   .00113
82 1  1.00000  0.00701  0.01764 -0.00377 -0.01010 -0.00404  0
   .00000
83 2  0.46154 -1.11111  0.00000 -0.00000 -0.00000 -0.00000
   NaN
84 3 -0.83333 -0.10000      NaN      NaN      NaN      NaN
   NaN
85 4 -0.40000  2.00000      NaN      NaN      NaN      NaN
   NaN
86 5 -2.50000  1.00000      NaN      NaN      NaN      NaN
   NaN

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87 6 -0.20000 -0.00000 NaN NaN NaN NaN
      NaN
88 -----ARMA (2,0):  $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t)$ -----
89 Enter the number of data samples :>? 10000
90 Enter AR order:>? 2
91 Enter MA order:>? 0
92 Enter Mean of white noise:>? 1
93 Enter variance of white noise:>? 2
94 Enter coefficient of AR a1>? 0.5
95 Enter coefficient of AR a2>? 0.2
96 Is this stationary process: True
97 GPAC Table for y:
98      1      2      3      4      5      6
      7
99 0 -0.41800 -0.22019 -0.01352 -0.00208 -0.01672 -0.00293
      0.01455
100 1 0.01675 -0.19580 0.02055 0.10377 -0.01647 -0.08696
      0.01117
101 2 -12.14286 -0.19562 2.55556 0.27273 -0.00000 -0.00000 -
      0.00000
102 3 -0.44706 -0.35345 0.02174 -0.00000 NaN NaN
      NaN
103 4 0.31579 -0.33740 -2.00000 NaN NaN NaN
      NaN
104 5 -1.50000 -0.48193 -0.50000 NaN NaN NaN
      NaN
105 6 0.33333 -1.12500 -0.00000 NaN NaN NaN
      NaN
106 -----ARMA (2,1):  $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t) - 0.5e(t-1)$  -----
107 Enter the number of data samples :>? 10000
108 Enter AR order:>? 2
109 Enter MA order:>? 1
110 Enter Mean of white noise:>? 1
111 Enter variance of white noise:>? 2
112 Enter coefficient of AR a1>? 0.5
113 Enter coefficient of AR a2>? 0.2
114 coefficient of MA b1>? 0.5
115 Is this stationary process: True
116 GPAC Table for y:
117      1      2      3      4      5      6
      7
118 0 -0.02800 -0.19994 0.09353 -0.06208 0.03039 -0.02608 -0
      .00457
119 1 7.10714 -0.21253 -0.03804 -0.01644 -0.02287 -0.03144 -0

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119 .05382
120 2 -0.51256 -0.13260 0.09091 0.02151 0.01667 -0.04478 0
    .00000
121 3 -0.23529 -0.26465 0.19355 -0.00000 0.00000 -0.00000
    NaN
122 4 0.37500 -0.08725 0.50000 NaN NaN NaN
    NaN
123 5 -0.22222 1.15385 0.33333 NaN NaN NaN
    NaN
124 6 -8.50000 1.73333 1.00000 NaN NaN NaN
    NaN
125 -----ARMA (1,2):  $y(t) + 0.5y(t-1) = e(t) +$ 
     $0.5e(t-1) - 0.4e(t-2)$  -----
126 Enter the number of data samples :>? 10000
127 Enter AR order:>? 1
128 Enter MA order:>? 2
129 Enter Mean of white noise:>? 1
130 Enter variance of white noise:>? 2
131 Enter coefficient of AR a1>? 0.5
132 coefficient of MA b1>? 0.5
133 coefficient of MA b2>? -0.4
134 Is this stationary process: True
135 GPAC Table for y:
136      1      2      3      4      5      6
      7
137 0 -0.10200 -0.30053 0.10472 -0.15510 0.09856 -0.10584 0
    .06449
138 1 2.81373 -0.33289 -0.33551 -0.09015 -0.06639 -0.04652 -0
    .07007
139 2 -0.56794 -0.02778 0.04058 0.01353 0.03991 0.01027 -0
    .03509
140 3 -0.50920 -0.78545 0.04724 -0.13333 0.05556 0.33333 -0
    .00000
141 4 -0.34940 0.11111 0.50000 0.00000 0.00000 0.00000
    NaN
142 5 -0.44828 2.04167 0.66667 NaN NaN NaN
    NaN
143 6 0.84615 -0.16327 -0.50000 NaN NaN NaN
    NaN
144 -----ARMA (0,2):  $y(t) = e(t) + 0.5e(t-1) -$ 
     $0.4e(t-2)$  -----
145 Enter the number of data samples :>? 10000
146 Enter AR order:>? 0
147 Enter MA order:>? 2
148 Enter Mean of white noise:>? 1
149 Enter variance of white noise:>? 2

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150 coefficient of MA b1>? 0.5
151 coefficient of MA b2>? -0.4
152 Is this stationary process: True
153 GPAC Table for y:
154      1      2      3      4      5      6
      7
155 0  0.21800 -0.35751  0.20422 -0.17941  0.14310 -0.12807  0
    .11203
156 1 -1.34404 -0.24891 -0.09679 -0.02177 -0.01411 -0.00490  0
    .01248
157 2 -0.01706 -0.09014 -0.01407  0.03252 -0.01020 -0.04348  0
    .02632
158 3  5.20000 -0.08901 -0.18182  0.00000 -0.00000 -0.00000  0
    .00000
159 4  0.00000 -0.30882 -0.00000      NaN      NaN      NaN
      NaN
160 5      -inf -0.28571      NaN      NaN      NaN      NaN
      NaN
161 6 -0.25000 -1.33333      NaN      NaN      NaN      NaN
      NaN
162 -----ARMA (2,2): y(t)+0.5y(t-1) +0.2y(t-2
    ) = e(t)+0.5e(t-1) - 0.4e(t-2)-----
163 Enter the number of data samples :>? 10000
164 Enter AR order:>? 2
165 Enter MA order:>? 2
166 Enter Mean of white noise:>? 1
167 Enter variance of white noise:>? 2
168 Enter coefficient of AR a1>? 0.5
169 Enter coefficient of AR a2>? 0.2
170 coefficient of MA b1>? 0.5
171 coefficient of MA b2>? -0.4
172 Is this stationary process: True
173 GPAC Table for y:
174      1      2      3      4      5      6
      7
175 0 -0.12800 -0.43145  0.13119 -0.19660  0.11402 -0.12972  0
    .09517
176 1  3.18750 -0.46312 -0.50426 -0.12349 -0.10670 -0.04739 -0
    .04912
177 2 -0.57598 -0.20830  0.04492  0.03856 -0.00000  0.02326  0
    .00833
178 3 -0.14894 -0.27675  0.23932  0.03448      NaN  0.00000  0
    .00000
179 4  1.22857 -0.23389  0.03571  0.50000      NaN      NaN
      NaN
180 5 -0.53488 -0.20000  1.50000  0.00000      NaN      NaN

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180			NaN				
181	6	0.00000	-0.64151	0.33333	NaN	NaN	NaN
		NaN					
182							

1.

Develop a python code that generates ARMA (na,nb) process.

The program should be written in a way that asks a user to enter the following information.

Hint: Use statsemmodels library.

- a. Enter the number of data samples: _____
- b. Enter the mean of white noise: _____
- c. Enter the variance of the white noise: _____
- d. Enter AR order: _____
- e. Enter MA order: _____
- f. Enter the coefficients of AR (you need to include a hint how this should be entered): _____
- g. Enter the coefficients of MA (you need to include a hint how this should be entered): _____

Enter the number of data samples :>? 1000

Enter AR order:>? 1

Enter MA order:>? 0

Enter Mean of white noise:>? 1

Enter variance of white noise:>? 2

Enter coefficient of AR a1>? -0.5

Is this stationary process: True

2.

Edit the python code in step 1 that implement the GPAC table using the following equation.

The output should be the GPAC table

GPAC Table for y:

	1	2	3	4	5	6	7
0	0.44600	-0.03860	0.07675	-0.00459	-0.04416	0.02289	0.00333
1	0.37668	0.84702	0.07442	-0.74359	-0.04639	0.02989	0.32941
2	0.72619	0.08706	0.32787	0.00575	-0.02410	-0.13636	-0.10714
3	0.61475	-2.35965	0.33333	1.00000	-0.00000	-0.00000	-0.00000
4	0.02667	0.08364	0.20000	-1.00000	NaN	NaN	NaN
5	3.00000	0.02222	0.37500	-1.00000	NaN	NaN	NaN
6	2.33333	-39.00000	0.33333	-1.00000	NaN	NaN	NaN

3.

Using the developed code above, simulate ARMA(1,0) for 1000 samples as follows:

$$y(t) - 0.5y(t-1) = e(t)$$

Use statsemmodels library to simulate above ARMA (1,2) process.

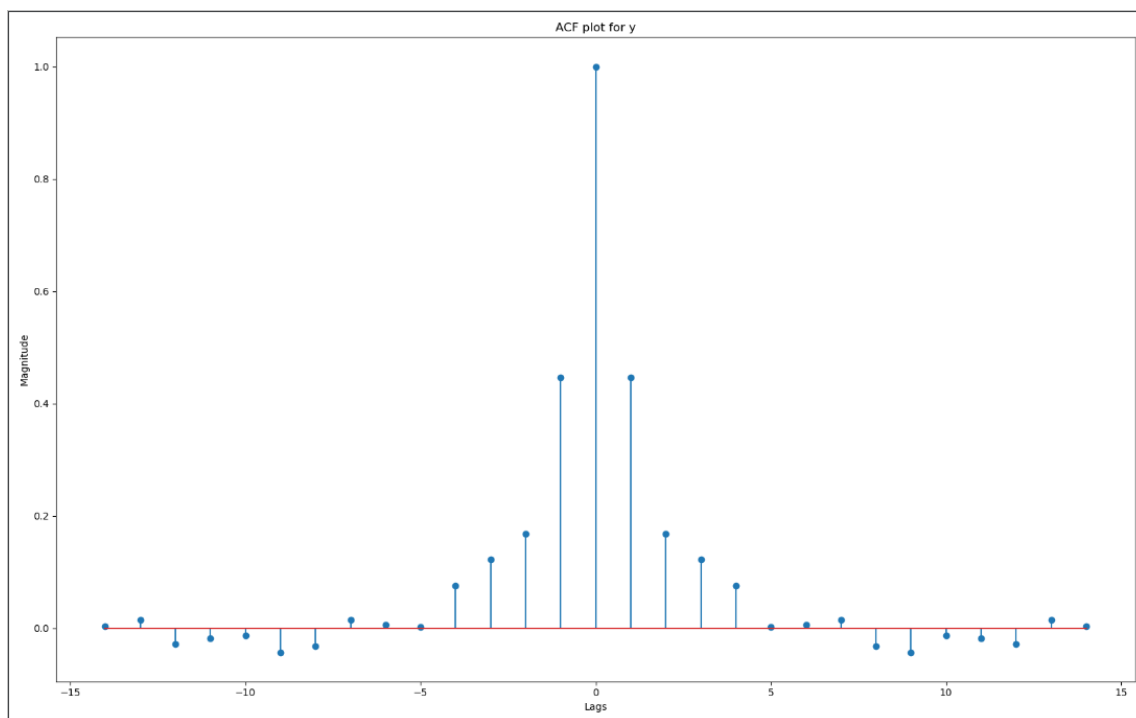
You can use the .generate_sample (# of samples, scales = std of WN noise) + mean (y) where

$$\text{mean}(y) = \mu e(1 + \sum b_i) / 1 + \sum a_i$$

4.

Using python program, estimate ACF for y(t) using the ACF function developed in previous labs.

Number of lags = 15.



5.

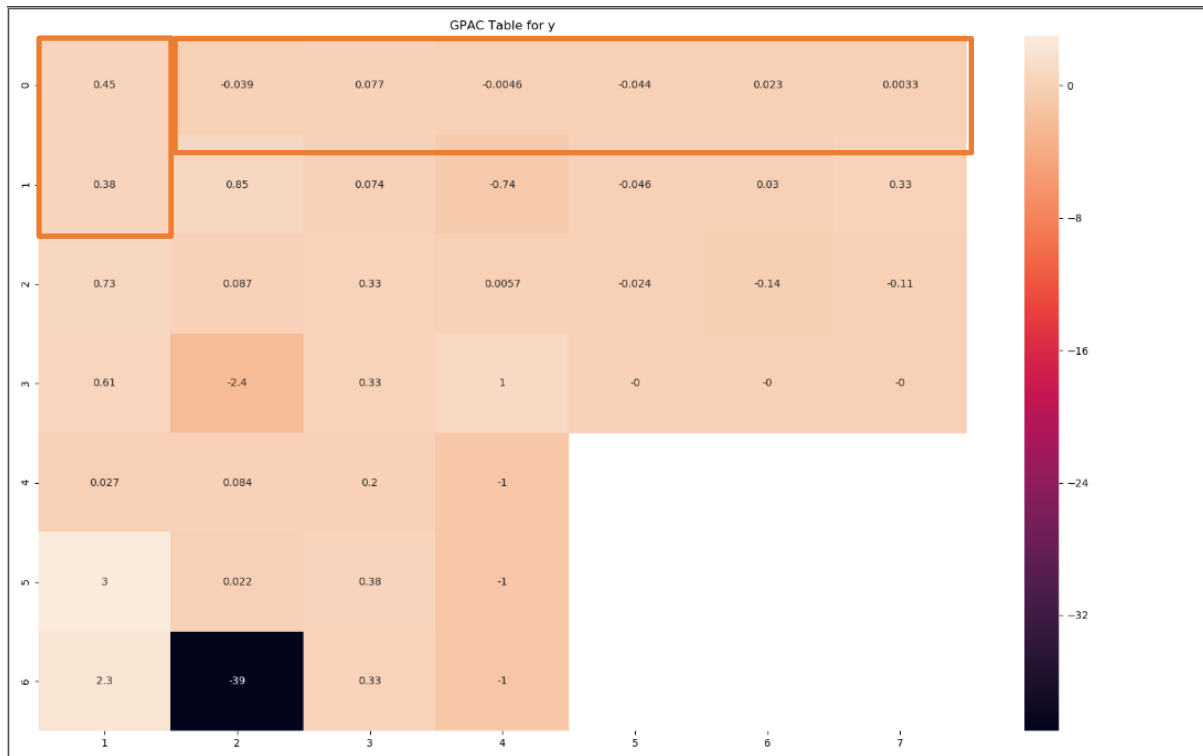
Using the estimated ACF from previous question, display GPAC table for $k=7$ and $j=7$.

Do you see a pattern of constant column 0.5 and a row of zeros?

What is the estimated na and what is the estimated nb ?

Yes I see a pattern of constant column 0.5 and a row of zeros

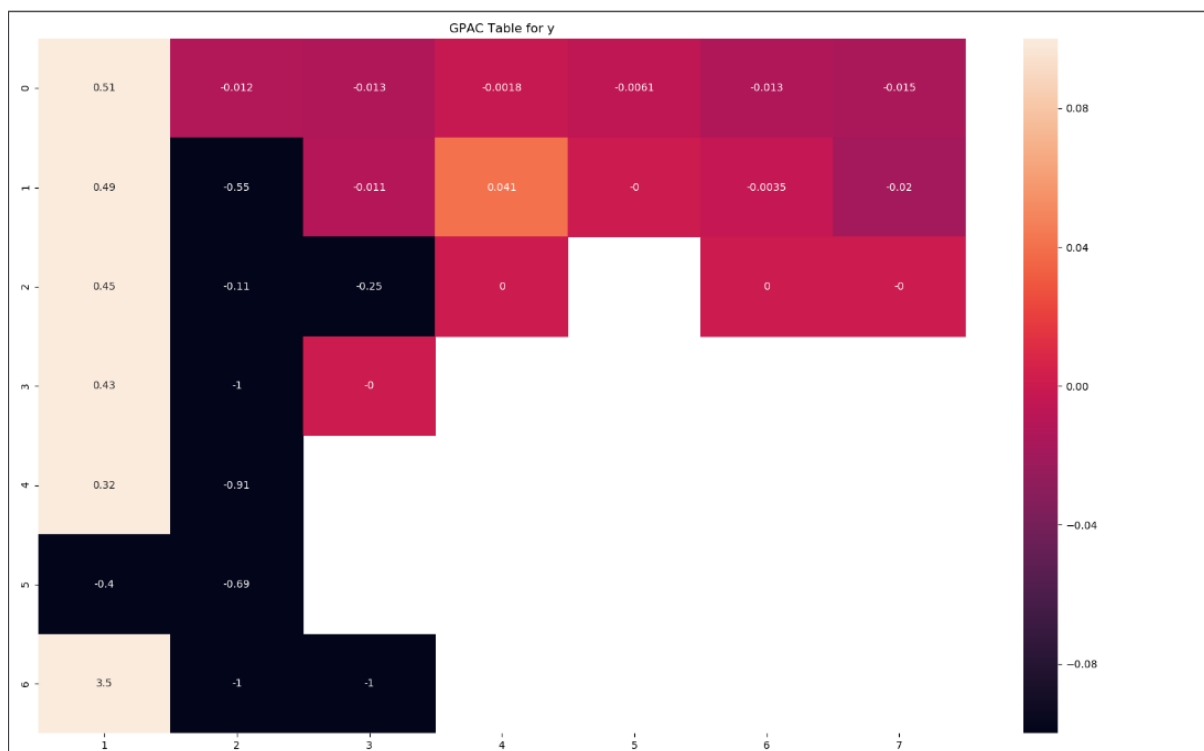
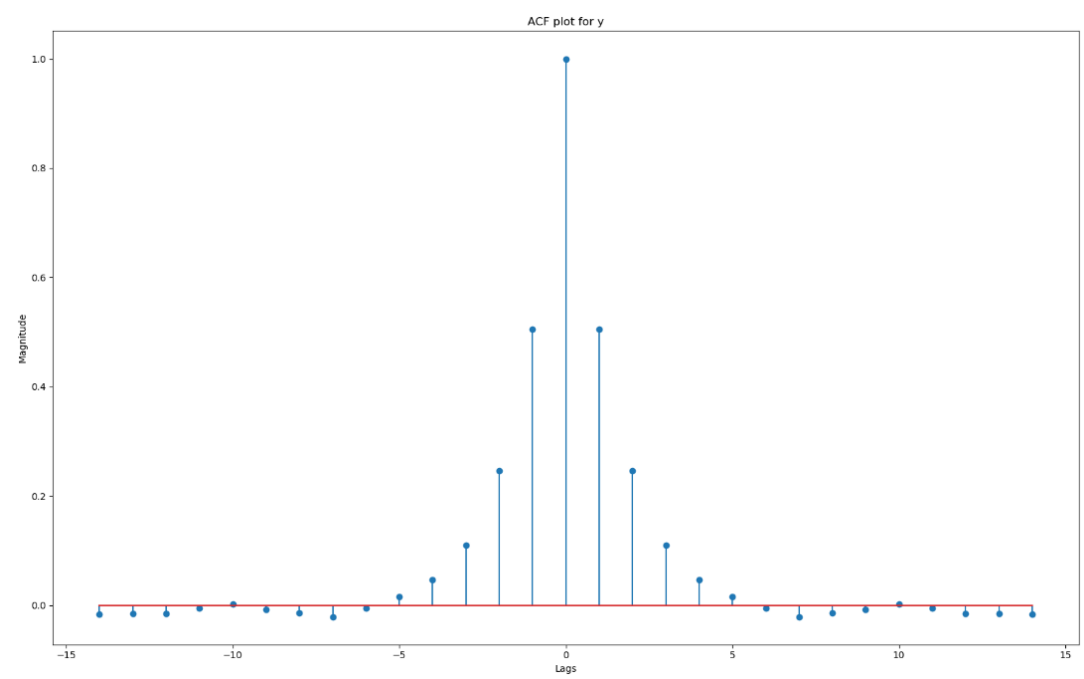
Estimated $na : 1$; $nb : 0$



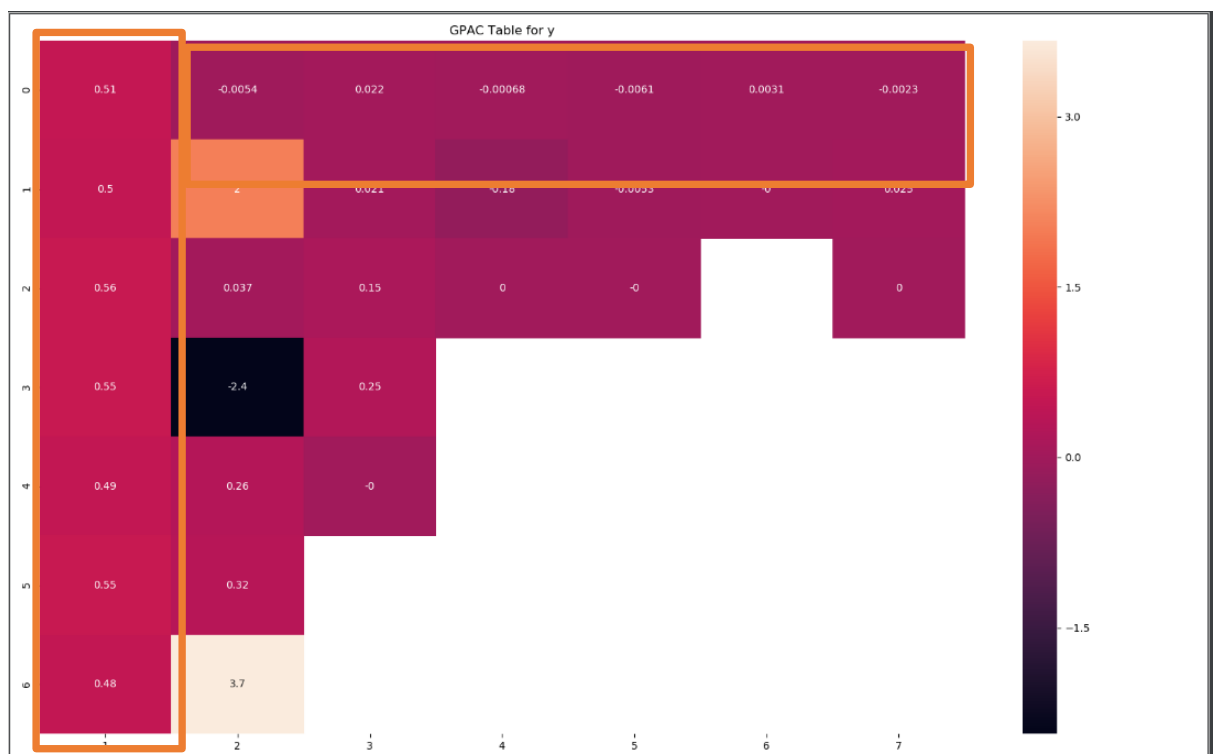
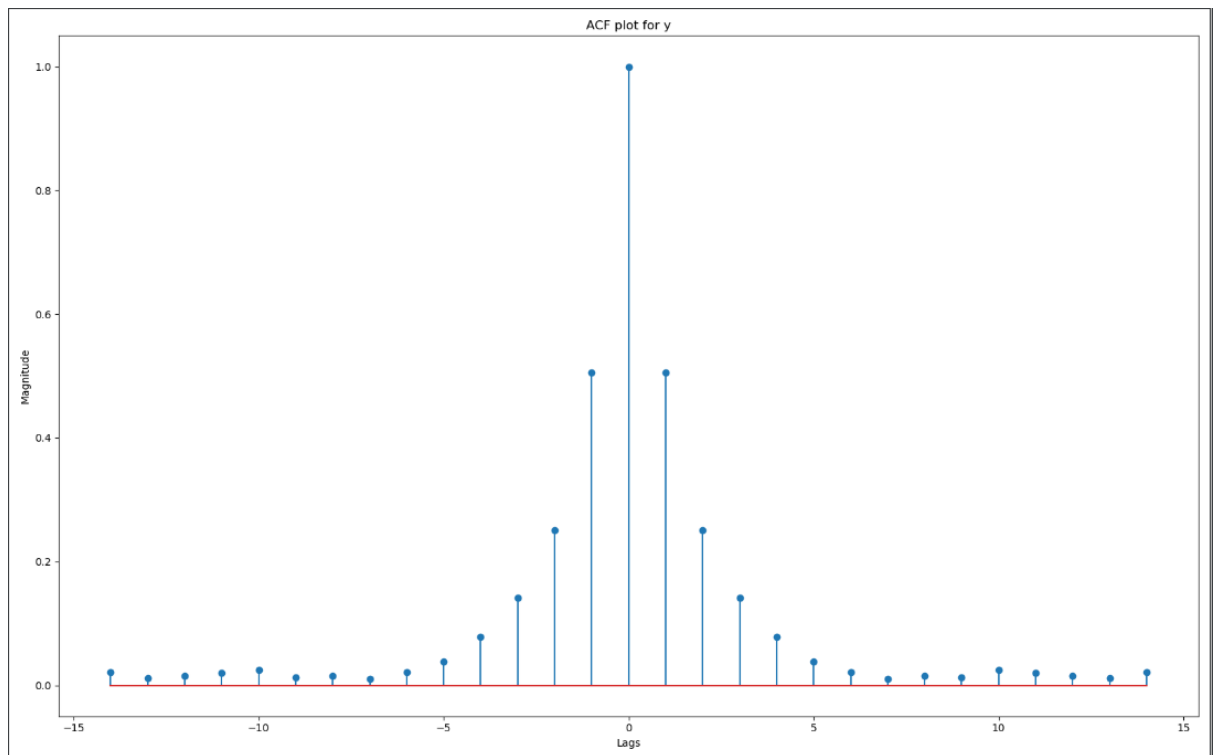
6. Increase the number of samples to 5000 and 10000. Do the numbers in the pattern converge?

Yes the numbers converge clear distinction is seen in below heatmap

Samples = 5000



Samples 10000

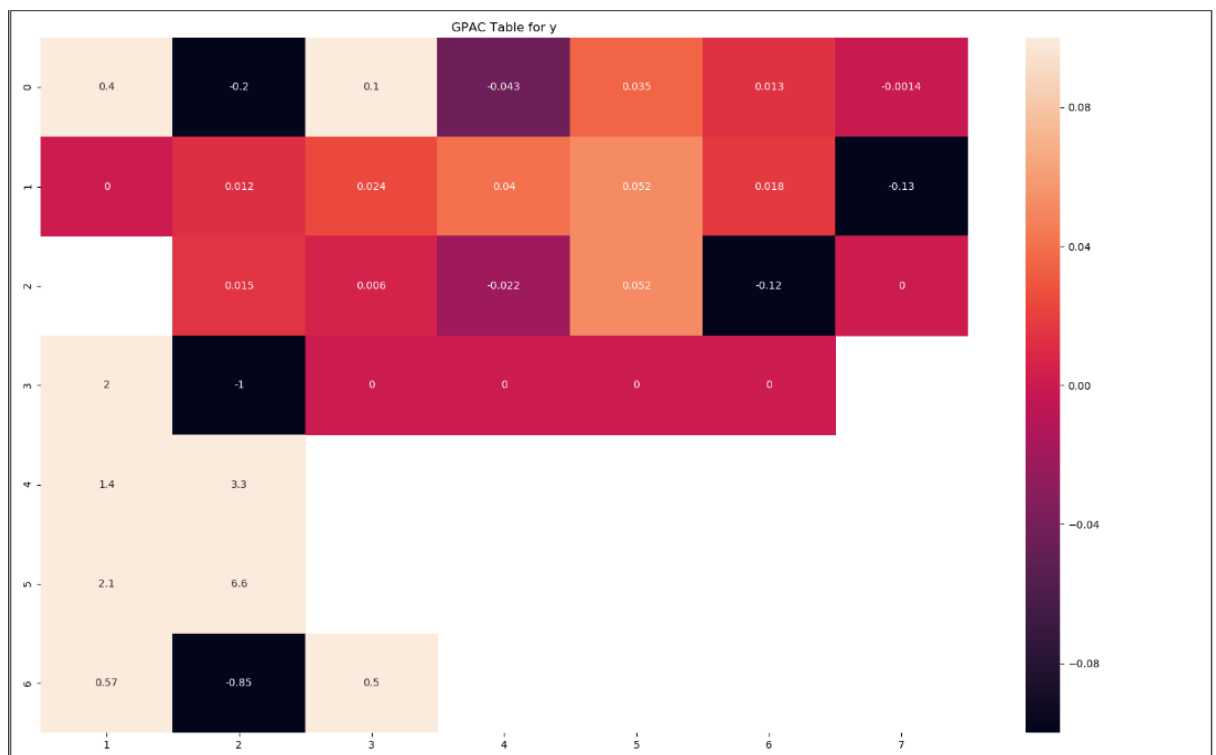
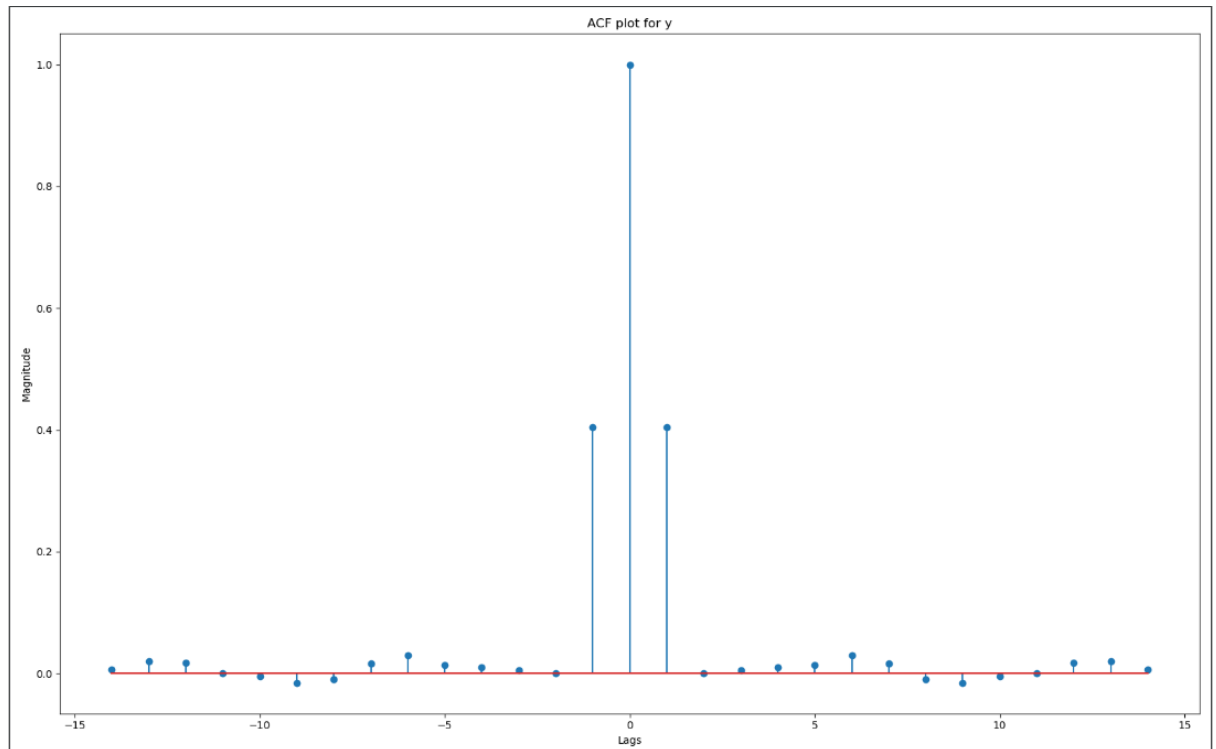


7. Repeat above steps for the following 7 examples with 10000 samples which should cover the followings:
- a. Display the GPAC table ($k=7, j=7$) in your report and highlight the pattern that read the process orders. Assign the correct label for each column (i.e. $k=1, k=2, \dots$) and each row (i.e. $j=0, j=1, \dots$)
 - b. Display the ACF for the $y(t)$ for 15 lags.
 - c. Based on the observed pattern in the GPAC table, what is the estimated n_a and estimated n_b . Compare the estimated order versus the true order. Write down your observations.
 - d. Make sure to include the .py code that can be run and verify the results.

Example 2: ARMA (0,1): $y(t) = e(t) + 0.5e(t-1)$

Estimated na: 0 Estimated nb: 1 | True Orders na:0 nb: 1.

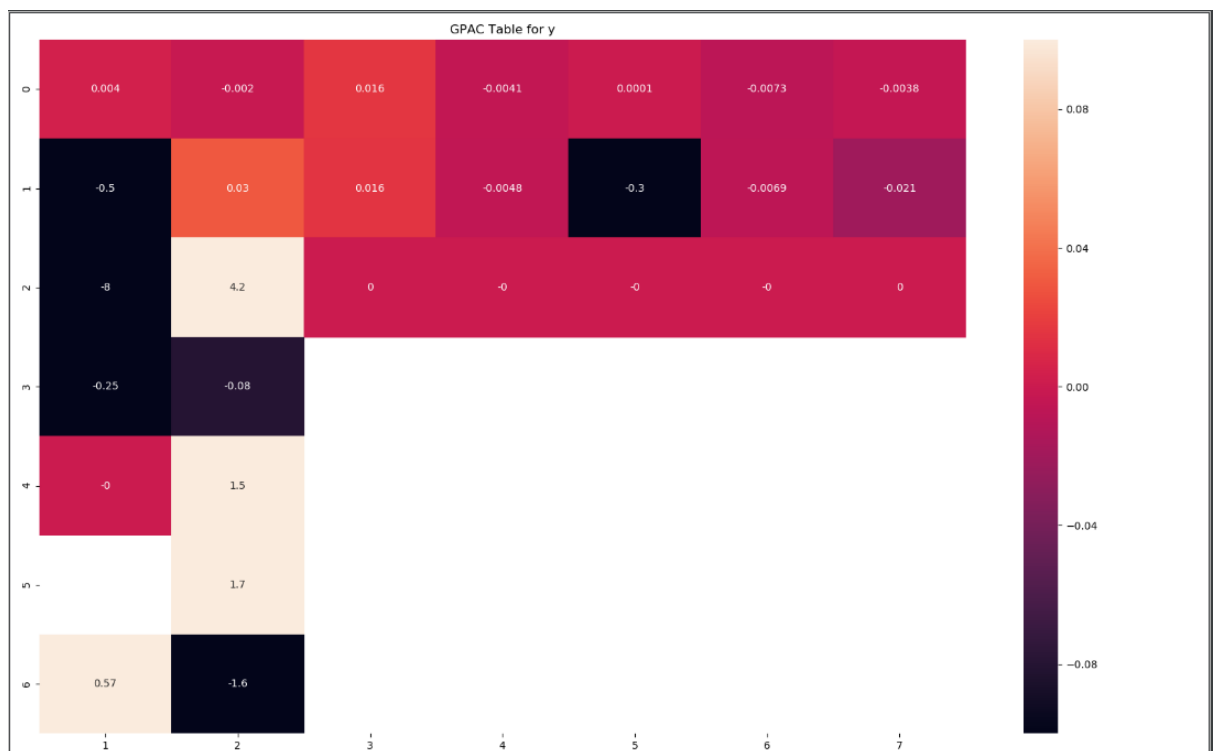
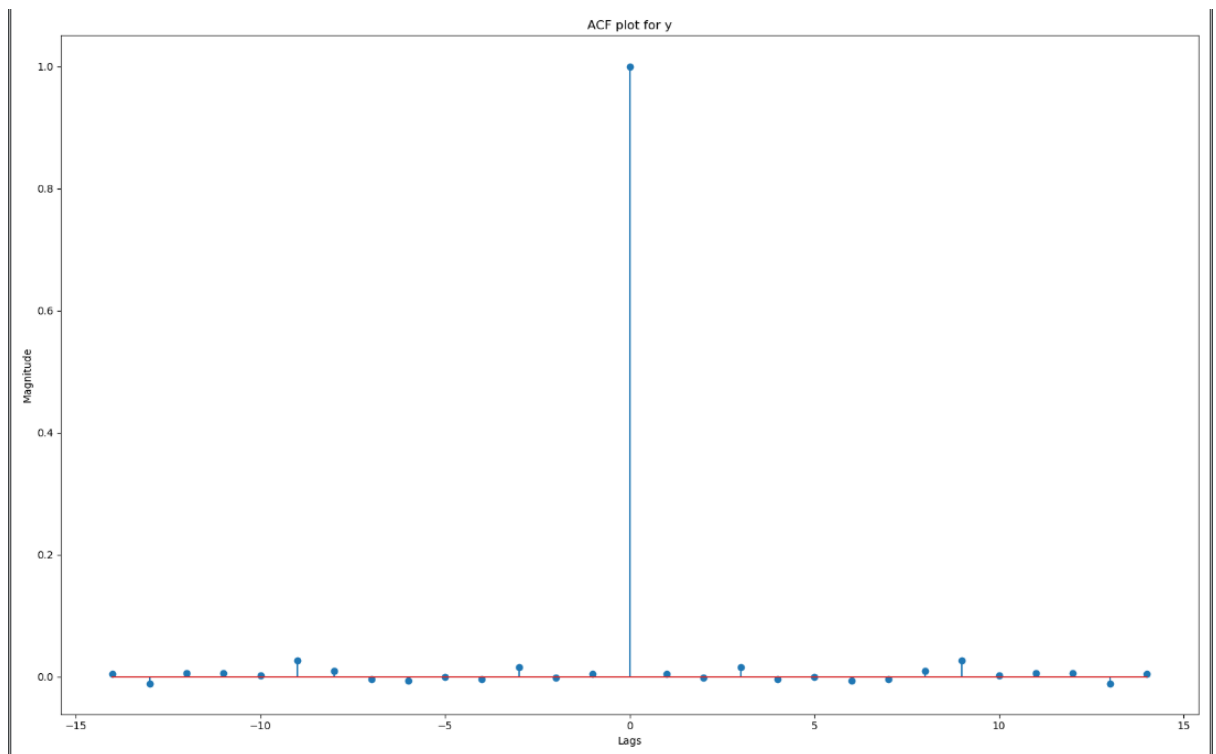
Estimated and true orders match.



Example 3: ARMA (1,1): $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t)$

Estimated na: 1 Estimated nb: 1 | True Orders na: 1 nb: 1.

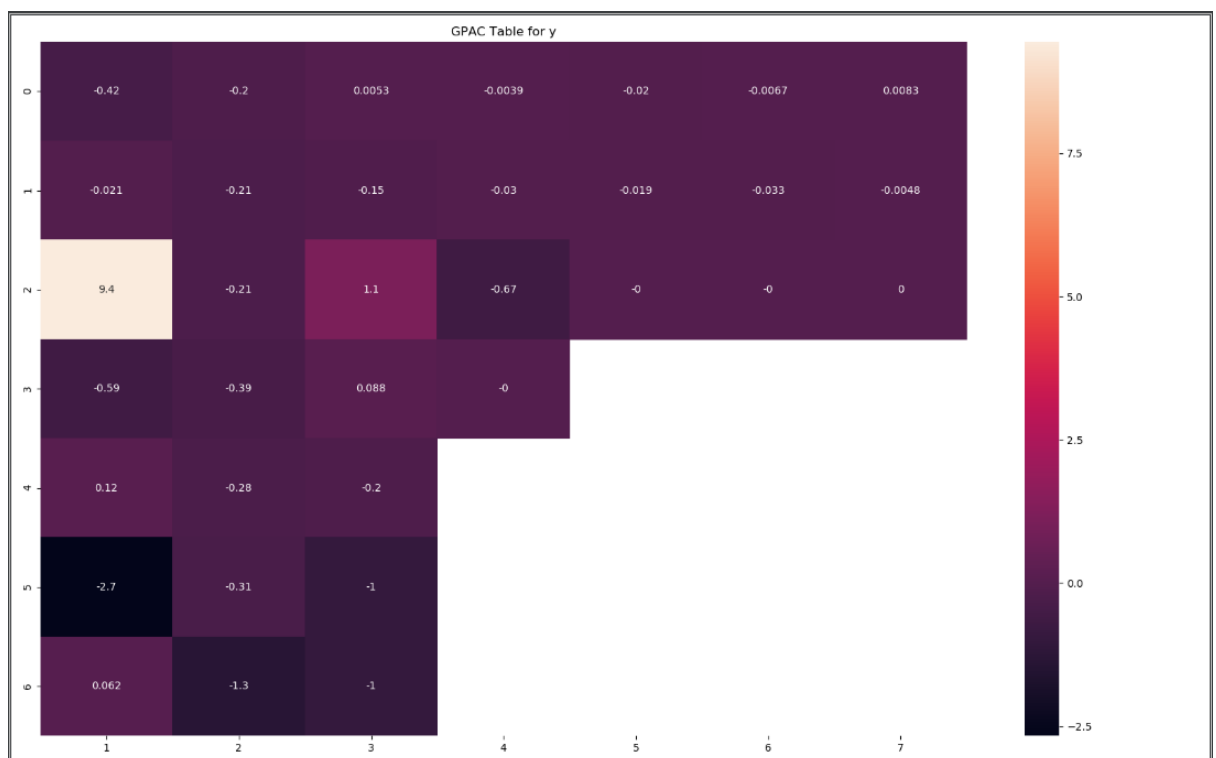
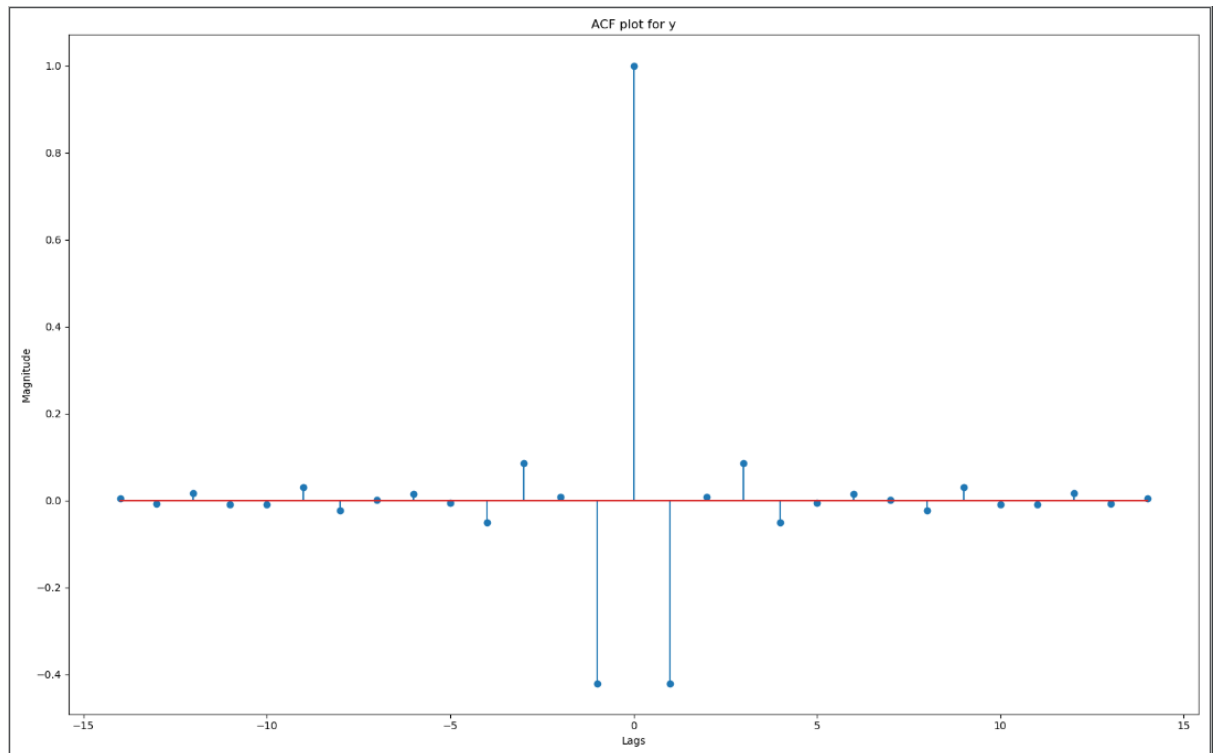
Estimated and true orders match.



Example 4: ARMA (2,0): $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t)$

Estimated na: 2 Estimated nb: 0 | True Orders na: 2 nb: 0.

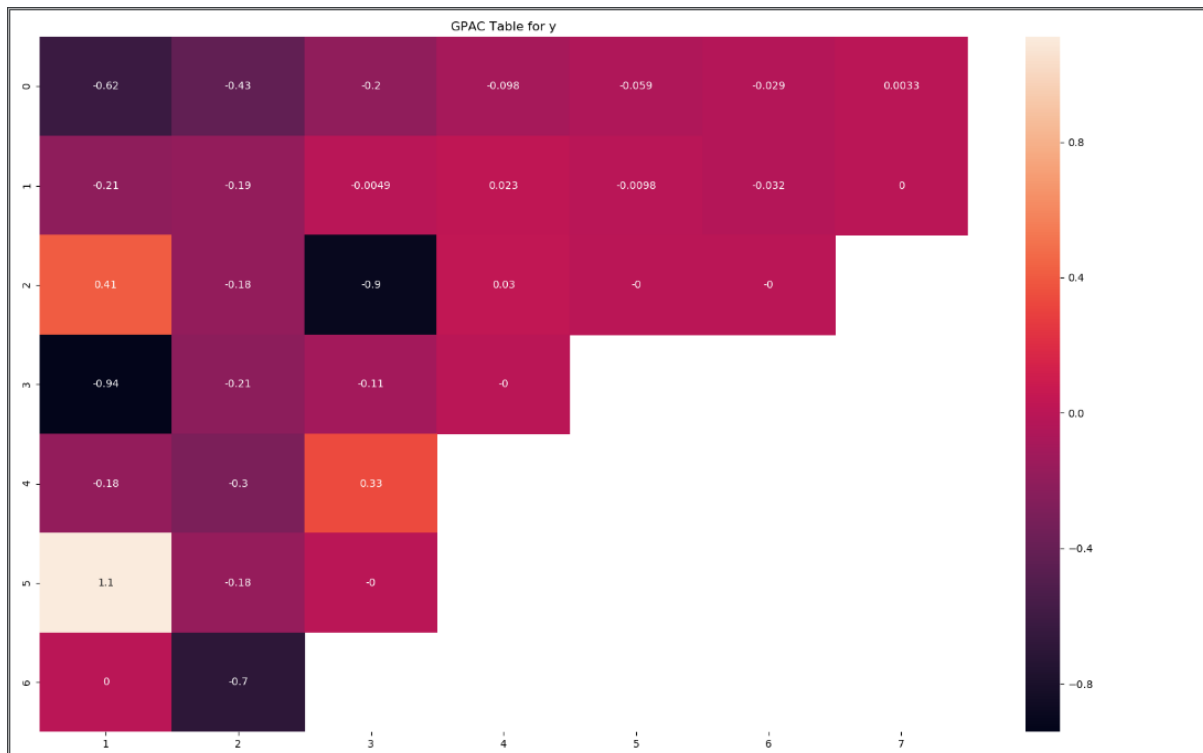
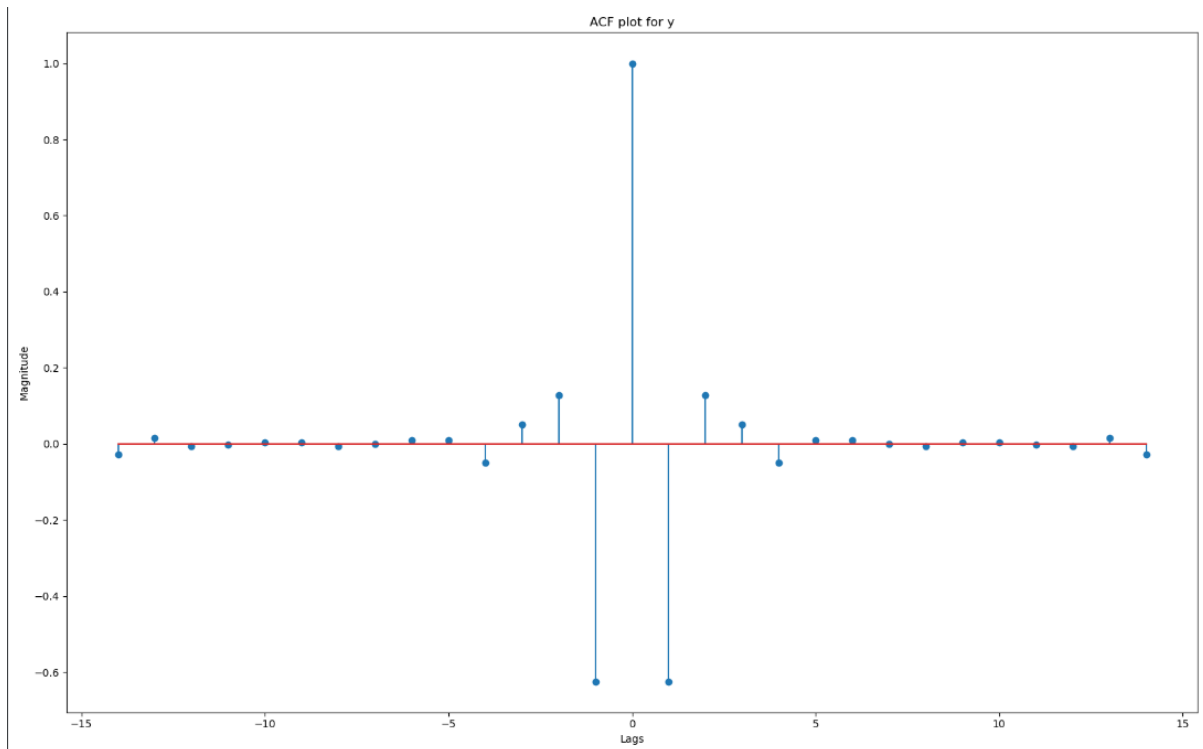
Estimated and true orders match.



Example 5: ARMA (2,1): $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t) - 0.5e(t-1)$

Estimated na: 2 Estimated nb: 1 | True Orders na: 2 nb: 1.

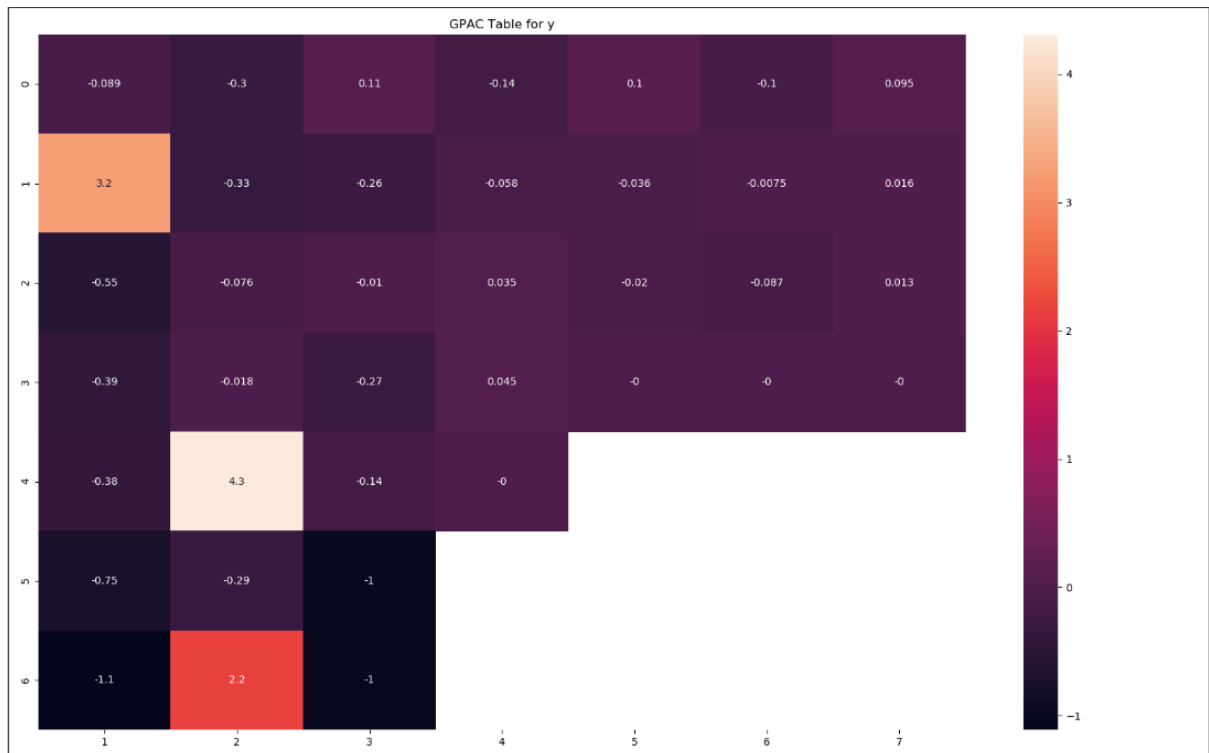
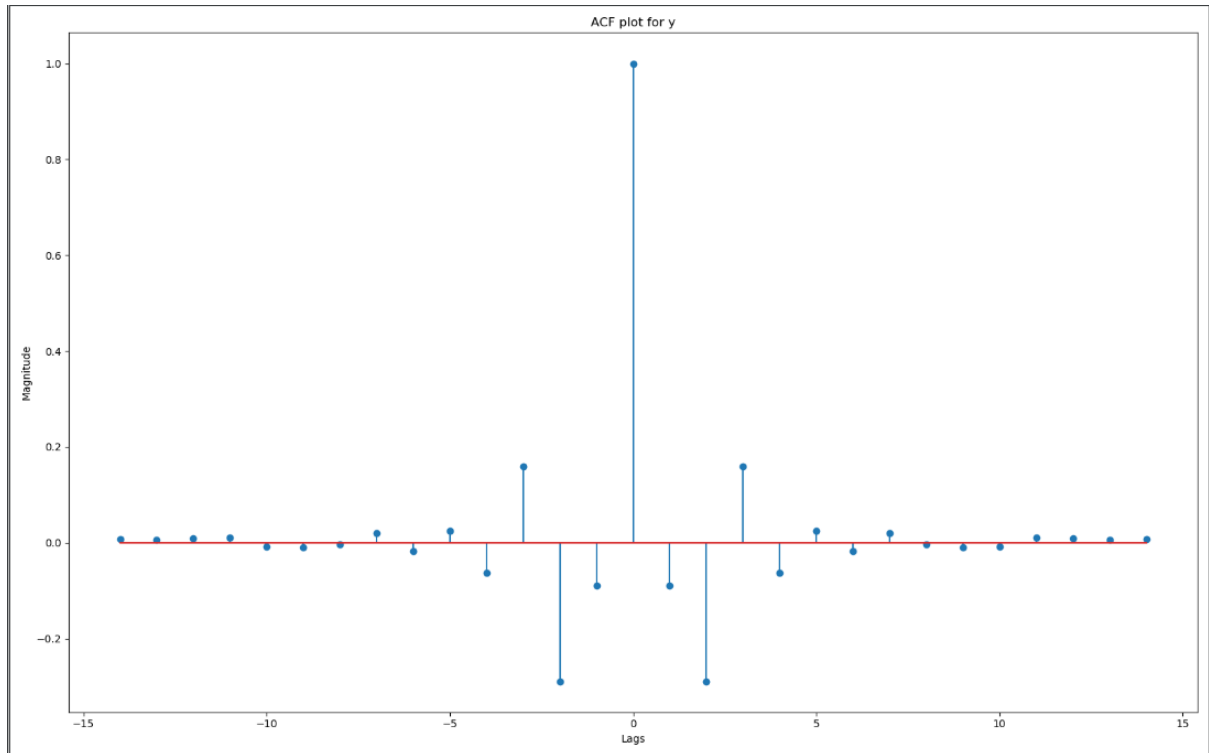
Estimated and true orders match.



Example 6: ARMA (1,2): $y(t) + 0.5y(t-1) = e(t) + 0.5e(t-1) - 0.4e(t-2)$

Estimated na: 1 Estimated nb: 2 | True Orders na: 1 nb: 2

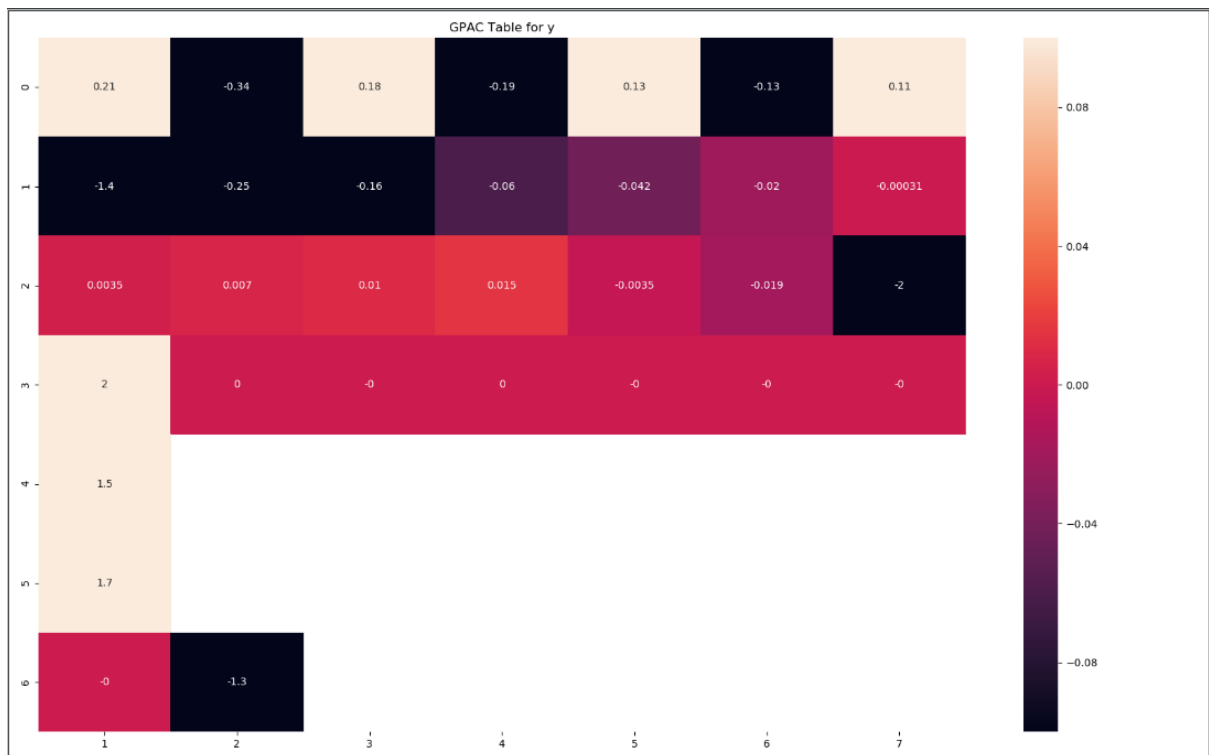
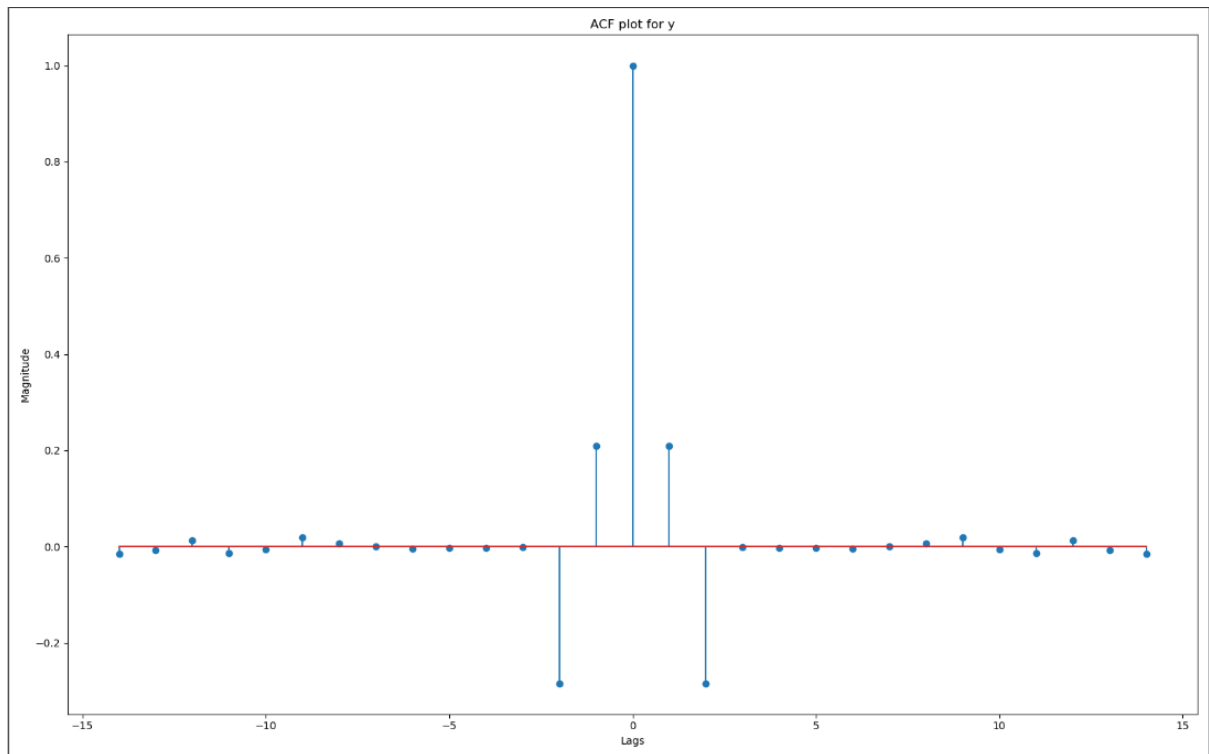
Estimated and true orders match.



Example 7: ARMA (0,2): $y(t) = e(t) + 0.5e(t-1) - 0.4e(t-2)$

Estimated na: 0 Estimated nb: 2 | True Orders na: 0 nb: 2

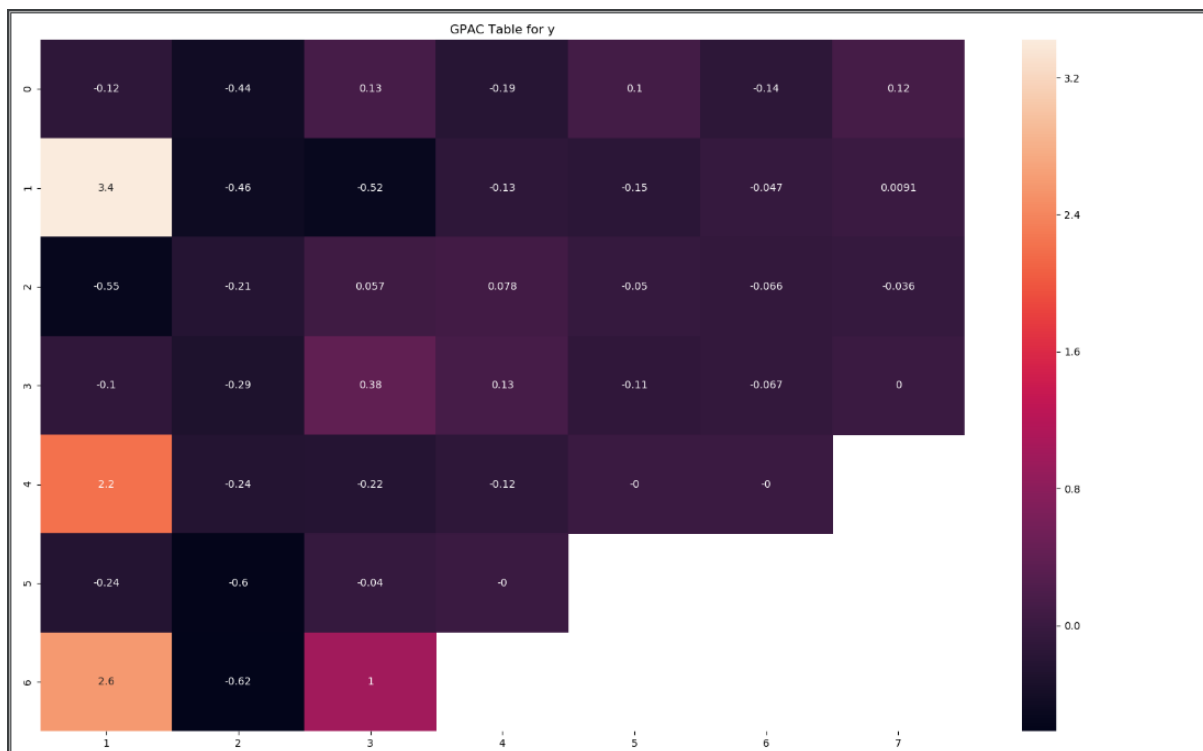
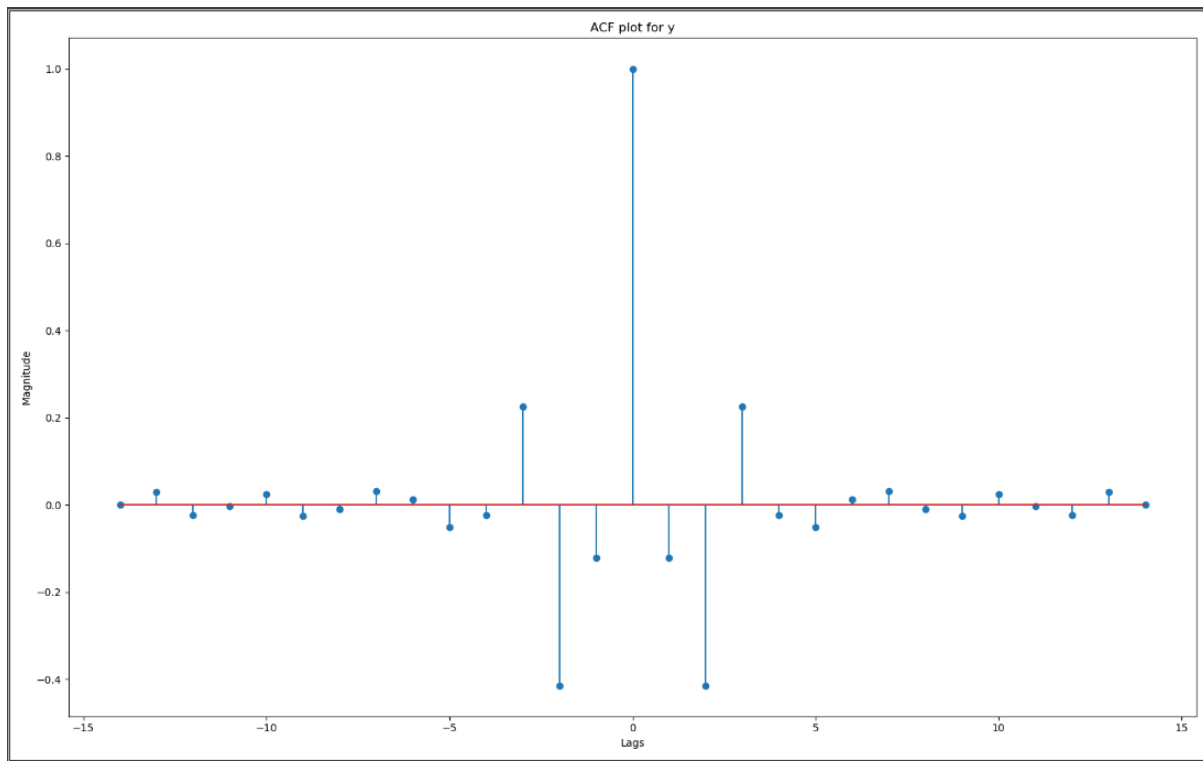
Estimated and true orders match.



Example 8: ARMA (0,2): $y(t) = e(t) + 0.5e(t-1) - 0.4e(t-2)$

Estimated na: 0 Estimated nb: 2 | True Orders na: 0 nb: 2

Estimated and true orders match.



CONCLUSION

In this LAB we implemented the GPAC array covered in lecture using Python program and tested the accuracy of code using an ARMA(n_a, n_b) model. Using statsmodels we simulated the ARMA process and GPAC code that generates GPAC table for various numbers of rows and columns was Coded. Thus we tested the above code for various equations and made a comparison with estimated orders and true orders and ACF were plotted.

CHALLENGE

GPAC Calculations was little tricky to understand in the beginning.

APPENDIX

```
import statsmodels.api as sm
import numpy as np
import matplotlib.pyplot as plt
from functions import *
import warnings
warnings.filterwarnings("ignore")

# %%=====
#1 .  $y(t) - 0.5y(t - 1) = e(t)$ 
# %%-----
print(20 * "-" + "y(t) - 0.5y(t - 1) = e(t)" + 20 * "-")
lags = 15
T = int(input("Enter the number of data samples :"))
na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y = (1*(1+1))/((1-0.5))

an = [0]*na
bn = [0]*nb

for i in range(na):
    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))
    # an_sum = np.sum(an[i])
    # print("Sum of an[i] = ",an_sum)

for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))
    # bn_sum = np.sum(bn[i])
    # print("Sum of bn[i] = ",bn_sum)

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):
    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparmas = np.array(an)
maparams = np.array(bn)

na = len(arparmas)
nb = len(maparams)
```

```

ar = np.r_[1, arparams]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y
# y = y + mean_y

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")
#gpac
j = 7
k = 7
lags = j + k

# autocorrelation of traffic volume
ry = cal_auto_correlation(y, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")

# %%-----
print(" Repeat for 5000 samples")
T1 = int(input("Enter the number of data samples:"))
y1 = arma_process.generate_sample(T1, scales) + mean_y

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y1, 15)
plot_acf(autocorrelation, "ACF plot for y")

# autocorrelation of traffic volume
ry = cal_auto_correlation(y1, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")
# %%-----
-----
print(" Repeat for 10000 samples")
T2 = int(input("Enter the number of data samples:"))
y2 = arma_process.generate_sample(T2, scales) + mean_y

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y2, 15)
plot_acf(autocorrelation, "ACF plot for y")

# autocorrelation of traffic volume

```

```

ry = cal_auto_correlation(y2, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")
print()

#%%=====
#2 .  $y(t) = e(t) + 0.5e(t-1)$ 
# %%-----
print(20 * "-" + "y(t) = e(t) + 0.5e(t-1)" + 20 * "-")
T = int(input("Enter the number of data samples :"))
na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y2 = (1*(1+0.5))/((1))

an = [0]*na
bn = [0]*nb

for i in range(na):
    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))
    # an_sum = np.sum(an[i])
    # print("Sum of an[i] = ",an_sum)

for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))
    # bn_sum = np.sum(bn[i])
    # print("Sum of bn[i] = ",bn_sum)

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):
    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparams = np.array(an)
maparams = np.array(bn)

na = len(arparams)
nb = len(maparams)

ar = np.r_[1, arparams]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y2

```

```

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")

#gpac
j = 7
k = 7
lags = j + k

# autocorrelation of traffic volume
ry = cal_auto_correlation(y, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")

#
%%=====
=====
#3 . ARMA (1,1):  $y(t) + 0.5y(t-1) = e(t) + 0.5e(t-1)$ 
# %%-----
-----
print(20 * "-" + "ARMA (1,1):  $y(t) + 0.5y(t-1) = e(t) + 0.5e(t-1)$ " + 20 * "-")
T = int(input("Enter the number of data samples :"))
na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y3 = (1*(1+0.5))/((1+0.5))

an = [0]*na
bn = [0]*nb

for i in range(na):
    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))

for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):
    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparams = np.array(an)
maparams = np.array(bn)

na = len(arparams)

```

```

nb = len(maparams)

ar = np.r_[1, arparams]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y3

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")

#gpac
j = 7
k = 7
lags = j + k

# autocorrelation
ry = cal_auto_correlation(y, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")

#
%%=====
=====
#4 . ARMA (2,0):  $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t)$ 
# %%-----
-----
print(20 * "-" + "ARMA (2,0):  $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t)$ " + 20 * "-")
T = int(input("Enter the number of data samples :"))
na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y4 = (1*(1))/((1+0.5+0.2))

an = [0]*na
bn = [0]*nb

for i in range(na):
    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))

for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):

```

```

    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparmas = np.array(an)
maparams = np.array(bn)

na = len(arparmas)
nb = len(maparams)

ar = np.r_[1, arparmas]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y4

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")

#gpac
j = 7
k = 7
lags = j + k

# autocorrelation
ry = cal_auto_correlation(y, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")
#
%%=====
=====
#5 . ARMA (2,1):  $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t) - 0.5e(t-1)$ 
# %%-----
-----
print(20 * "-" + "ARMA (2,1):  $y(t) + 0.5y(t-1) + 0.2y(t-2) = e(t) - 0.5e(t-1)$  " +
20 * "_")
T = int(input("Enter the number of data samples :"))
na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y5 = (1*(1-0.5))/((1+0.5+0.2))

an = [0]*na
bn = [0]*nb

for i in range(na):

```



```

    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))
for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):
    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparmas = np.array(an)
maparams = np.array(bn)

na = len(arparmas)
nb = len(maparams)

ar = np.r_[1, arparmas]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y5

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")

#gpac
j = 7
k = 7
lags = j + k

# autocorrelation
ry = cal_auto_correlation(y, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")

#
%%=====
=====
#6 . ARMA (1,2):  $y(t) + 0.5y(t-1) = e(t) + 0.5e(t-1) - 0.4e(t-2)$ 
# %%-----
-----
print(20 * "-" + "ARMA (1,2):  $y(t) + 0.5y(t-1) = e(t) + 0.5e(t-1) - 0.4e(t-2)$  " +
20 * "-")

T = int(input("Enter the number of data samples :"))

```

```

na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y6 = (1*(1+0.5-0.4))/((1+0.5))

an = [0]*na
bn = [0]*nb

for i in range(na):
    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))

for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):
    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparmas = np.array(an)
maparams = np.array(bn)

na = len(arparmas)
nb = len(maparams)

ar = np.r_[1, arparmas]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y6

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")

#gpac
j = 7
k = 7
lags = j + k

# autocorrelation
ry = cal_auto_correlation(y, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")

```

```

# %%=====
=====
#7 .ARMA (0,2):  $y(t) = e(t) + 0.5e(t-1) - 0.4e(t-2)$ 
# %%-----
-----
print(20 * "-" + "ARMA (0,2):  $y(t) = e(t) + 0.5e(t-1) - 0.4e(t-2)$ " + 20 * "-")

T = int(input("Enter the number of data samples :"))
na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y7 = (1*(1+0.5-0.4))/((1))

an = [0]*na
bn = [0]*nb

for i in range(na):
    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))

for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):
    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparmas = np.array(an)
maparams = np.array(bn)

na = len(arparmas)
nb = len(maparams)

ar = np.r_[1, arparmas]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y7

# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")

#gpac
j = 7
k = 7
lags = j + k

# autocorrelation
ry = cal_auto_correlation(y, lags)

```

```

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")

#
%%=====
=====
#8. ARMA (2,2):  $y(t)+0.5y(t-1) +0.2y(t-2) = e(t)+0.5e(t-1) - 0.4e(t-2)$ 
# %%-----
-----
print(20 * "-" + "ARMA (2,2):  $y(t)+0.5y(t-1) +0.2y(t-2) = e(t)+0.5e(t-1) - 0.4e(t-2)$ " + 20 * "-")

T = int(input("Enter the number of data samples :"))
na = int(input("Enter AR order:"))
nb = int(input("Enter MA order:"))
mean = int(input("Enter Mean of white noise:"))
std = int(input("Enter variance of white noise:"))
scales = np.sqrt(std)
mean_y8 = (1*(1+0.5-0.4))/((1+0.5+0.2))

an = [0]*na
bn = [0]*nb

for i in range(na):
    an[i] = float(input("Enter coefficient of AR a{}".format(i+1)))

for i in range(nb):
    bn[i] = float(input("coefficient of MA b{}".format(i+1)))

max_order = max(na,nb)
num = [0]*(max_order+1)
den = [0]*(max_order+1)
for i in range(na+1):
    if i==0:
        den[i] = 1
    else:
        den[i] = an[i-1]

arparams = np.array(an)
maparams = np.array(bn)

na = len(arparams)
nb = len(maparams)

ar = np.r_[1, arparams]
ma = np.r_[1, maparams]

arma_process = sm.tsa.ArmaProcess(ar,ma)
print("Is this stationary process: ", arma_process.isstationary)

y = arma_process.generate_sample(T, scales) + mean_y8

```

```
# ACF of the dependent variable.
autocorrelation = cal_auto_correlation(y, 15)
plot_acf(autocorrelation, "ACF plot for y")

#gpac
j = 7
k = 7
lags = j + k

# autocorrelation
ry = cal_auto_correlation(y, lags)

# create GPAC Table
gpac_table = create_gpac_table(j, k, ry)
print("GPAC Table for y:")
print(gpac_table.to_string())

# heatmap
plot_heatmap(gpac_table, "GPAC Table for y")
```

REFERENCES

<https://otexts.com/fpp2/#>