

Time Series Analysis For Business

Zheng, Zhi

- 2) Do file and log file for step 1 of model selection. For total private employment and average weekly earnings for all private employees, use VSELECT to estimate and evaluate some alternative models you think are reasonable. Take the log and difference if doing so is called for, don't if it is not called for. For a reasonable set of models from VSELECT, also calculate the LOOCV. Include a simple 12-lag AR only model as a benchmark. Prepare a table summarizing the fit measures for these reasonable models.
- 1) Vselect models for forecasting All Employees: Total Private in Miami-Fort Lauderdale-West Palm Beach, FL (MSA)

After testing many different models with different predicators, the best variables to use with forecasting All Employees in Miami-Fort Lauderdale-West Palm Beach, FL (MSA) are first differenced & logged All Employees, Labor Force and Unemployment ratio in Miami-Fort Lauderdale-West Palm Beach, FL (MSA). After running the vselect to estimate and evaluate the models with 12-Lags for each predictor variables, the results are:

```
AIC
                                          AICC
# Preds
           R2ADJ
                         C
                                                     RTC
          .318734 55.40285 -1477.623 -1475.858 -1431.689
     2 .3474025 44.39182 -1487.556 -1485.531 -1438.089
     3
         .402965
                  22.42448 -1509.13 -1506.825 -1456.129
         .4136497
                  19.01953 -1512.764 -1510.16
                                               -1456.23
     5
         .4199669 17.43182 -1514.575 -1511.651 -1454.507
     6
        .4264835 15.78493 -1516.507 -1513.246 -1452.906
          .43111 14.92449 -1517.635 -1514.015 -1450.501
     7
     8 .4366408 13.71612 -1519.191 -1515.191 -1448.523
       .4393245 13.6677 -1519.487 -1515.087 -1445.286
    10 .4431539 13.17147 -1520.314 -1515.493 -1442.579
    11 .4450533 13.45513 -1520.276 -1515.013 -1439.008
    12 .4465508 13.90431 -1520.062 -1514.335 -1435.261
    13
        .4476672 14.50879 -1519.68 -1513.468 -1431.345
    14
         .449633 14.78405 -1519.694 -1512.974 -1427.826
    15
        .4504897 15.49907 -1519.205 -1511.955 -1423.804
        .4520445 15.94572 -1519.044 -1511.241 -1420.109
    16
    17
        .4520368 17.00228 -1518.167 -1509.789 -1415.699
       .4525715 17.84987 -1517.546 -1508.569 -1411.545
    18
    19
        .453195 18.66531 -1516.972 -1507.372 -1407.437
    20 .4539004 19.45159 -1516.441 -1506.194 -1403.372
    21 .4545461 20.26301 -1515.887 -1504.97 -1399.286
    22 .4532631 21.80623 -1514.446 -1502.833 -1394.31
    23 .4515671 23.50048 -1512.82 -1500.487 -1389.151
    24 .4503475 25.01033 -1511.421 -1498.342 -1384.219
    25
        .4485351 26.73717 -1509.757 -1495.906 -1379.021
    26
        .4463016
                  28.61406 -1507.908 -1493.26 -1373.639
        .4443569
                  30.37651 -1506.201 -1490.729 -1368.399
    27
    28
        .4423855 32.14207 -1504.49 -1488.167 -1363.154
    29
        .4399588 34.06756 -1502.582 -1485.382 -1357.713
        .4373684 36.04428 -1500.61 -1482.505 -1352.208
    30
    31
       .4347231 38.032 -1498.625 -1479.587 -1346.69
    32
        .4320584 40.01755 -1496.643 -1476.643 -1341.174
    33
        .4293448 42.01148 -1494.651 -1473.66 -1335.648
    34
        .4266012 44.00678 -1492.657 -1470.647 -1330.121
    35
        .4238303 46.00232 -1490.662 -1467.603 -1324.593
        .4210262
                        48 -1488.665 -1464.527 -1319.062
```

Then, I selected models from #Preds 4-10 as the best models, because model 4 have the best BIC score and model 10 have the best AIC score. Finally, I ran regression on these

models plus a simple 12-lag AR model to get LOOCV RMSE and other calculations to get a table summarization. The result is shown below:

FIT[8,4]

	df	AIC	BIC	LOOCV
Model 12-L~R	13	-1468.4923	-1422.5583	.0140099
Model 4	16	-1540.1558	-1483.3705	.01694681
Model 5	17	-1541.8356	-1481.5013	.01709896
Model 6	18	-1544.0566	-1480.1733	.01723782
Model 7	19	-1552.3921	-1484.8858	.01741191
Model 8	20	-1540.364	-1469.4605	.01839856
Model 9	21	-1537.4357	-1462.987	.01853789
Model 10	22	-1537.9288	-1459.9349	.01850438

The 12-lag AR simple model have the lowest LOOCV RMSE result, the model 7 have BIC result and AIC result. In conclusion, Model 7 are the best fit model other than the 12-lag AR model.

2) Vselect models for forecasting Average Weekly Earnings of All Employees: Total Private in Miami-Fort Lauderdale-West Palm Beach, FL (MSA)

After testing many different models with different predicators, the best variables to use with forecasting Average Weekly Earnings in Miami-Fort Lauderdale-West Palm Beach, FL (MSA) are first differenced & logged Average Weekly Earnings, Average Hourly Earnings and Average Weekly Hours in Miami-Fort Lauderdale-West Palm Beach, FL (MSA). After running the vselect to estimate and evaluate the models with 12-Lags for each predictor variables, the results are:

```
# Preds
            R2ADJ
                                   AIC
                                            AICC
                                                       BIC
                  10.20136 -995.6167 -992.8894
         .3536648
                                                  -954.928
      2 .2775691
                  29.14068 -975.8933
                                        -972.756
                                                 -932.0747
        .3433141
                   14.65922
                            -991.1124
                                        987.5334
                            -993.4939
        .3559384
                   12.72661
                                       -989.4409
         .3904361
                   5.780666
                            -1001.906
                                       -997.3457
                                                 -948.6974
         .3973962
                   5.257595 -1002.962 -997.8613 -946.6238
         .4040225
                   4.831754 -1003.954
                                       -998,2778
                                                 -944.4854
         .4117794
                    4.17009
                                                     -942.7
                            -1005.298
                                       -999.0123
                   4.715145
                            -1005.155
                                        -998,223
                   6.012603
    10
         .4134358
                            -1004.058
                                        996.4446
         .4114523
                   7.561366 -1002.641
                                       -994.3081
                                                 -930.6538
         .4107174
                   8.829029 -1001.592 -992.5012
                                                 -926.4745
         .4103552
                   10.01291 -1000.658 -990.7705
                                                 -922.4103
          .409835
                      11.23 -999.6865 -988.9631
    14
                                                 -918.3091
          .408593
                   12.60126
                            -998.5172
                                       -986.9172
                                                 -914.0099
    15
         .4080593
                     13.817
                            -997.5591 -985.0411
                                                 -909.9219
         .4089908
                   14.71946
                              -997.028
                                       -983.5498
         .4085419
                   15.91625 -996.1112
                                       -981.6295
         .4062815
    19
                   17.49155
                            -994,6868
                                       -979.1574
                                                 -897.6599
         .4041431
                   19.03456 -993.3083 -976.6861
                                                 -893.1515
                   20.53665 -991.9881 -974.2269
    21
         .4021713
                                                 -888.7014
         .3993155
                   22.21409
                            -990.4299
                                       -971.4826
                                                 -884.0134
         .3965558
                   23.86329
                            -988.9118
                                         -968.73
                                                 -879.3653
         .3936142
                   25.54085
                            -987,3559
                                        965.8902
                   27.21616
         .3906392
                            -985.8043
                                       -963.0043
          .388109
                    28.7939
                            -984.3892
                                       -960,2031
                                                  -865.453
         . 3893845
                   29.61223 -984.0368
                                       -958.4118
                                                 -861.9708
    28
         .3801133
                   32.50165 -980.7952 -953.6771
                                                 -855.5992
                   34.34401 -979.0146
          .376081
                                       -950.3479
                                                 -850.6887
         .3719672
                   36.18985
                            -977.2294
                                       -946.9574
                                                 -845.7737
         .3675216
                    38,0867
                            -975.3733
                                       -943,4378
                                                 -840.7877
         .3625563
                   40.06874
                            -973.3984
                                       -939.7399
         .3575421
                   42.04492 -971.4317
                                        -935.989
                                                 -830.5862
         .3523511
                   44.03889 -969.4401 -932.1508 -825.4647
         .3472405
                    46,0022 -967,4913 -928,2913 -820,3861
    35
         .3418577
                         48 -965.4944 -924.3179 -815.2592
```

Then, I selected models from #Preds 1-8 as the best models, because model 1 have the best BIC score and model 8 have the best AIC score. Finally, I ran regression on these models plus a simple 12-lag AR model to get LOOCV RMSE and other calculations to get a table summarization. The result is shown below:

	df	AIC	BIC	LOOCV
Model 12-L~R	13	-936.11394	-895.42525	.01530392
Model 1	13	-1001.1318	-959.76865	.01445233
Model 2	14	-1023.8161	-979.19272	.01370862
Model 3	15	-1004.0621	-956.93715	.01302046
Model 4	16	-1006.1174	-955.8508	.01288637
Model 5	17	-1007.4593	-954.15073	.01263521
Model 6	18	-1007.8937	-951.44932	.01260431
Model 7	19	-1009.6906	-950.11048	.01261056
Model 8	20	-1009.9601	-947.24411	.01255532

The Model 8 have the lowest LOOCV RMSE result, the Model 2 have BIC result and AIC result. In conclusion, Model 2 are the best fit model because it has the best information criterion scores (AIC & BIC) and the simplest model.