

SPSS Steps for Time Series Demo #2

1. Examine baseline:

a. Data → Select Cases

- Select: Based on time or case range
- Range: Select the cases that represent the baseline. Because our intervention begins at case 29, we will select a case range beginning with case 1 and ending with case 28.
- Click continue then ok
- Data view should cross out all of the cases occurring after the beginning of the intervention

b. Graphs → Chart Builder

- Choose From: Scatter/Dot
- Click on the first option and drag it up to the chart preview box
- Drag the variable Year to the X-axis
- Drag the variable InfantdeathMS to the Y-axis
- Click ok and examine the graph output
- Double Click the graph to edit it
- Right click the graph and select: add interpolation line, delete scatter, and examine line

2. Difference Baseline:

a. Transform → Create Time Series

- Drag variable InfantdeathMS to the Variable->New Name box
- Function: Difference (should already be selected)
- Click Ok
- Data view should now show the column for the differenced variable Infant_1
- Create a new graph, replacing the InfantdeathMS Y-axis variable with the new differenced variable Infant_1
- In some cases, differencing of the differenced column may be necessary, but this is rare and not necessary for our example
- If variance is still a problem after differencing, a log transformation must be used
- Find the largest negative number from the differenced column and add 1 to it for the next step

- Ex: Largest negative number for our selected cases is -729, so we would use [-730] for the next step

3. Log Transformation

a. Transform → Compute Variable

- Give transformed variable a new name, ex: Target Variable: LogInfant_1
- Select Arithmetic in the Function Group Box
- Select Lg10 from the Functions and Special Variables Box, and click the arrow to move it to the Numeric Expression Box
- Select the differenced variable and move it to the Numeric Expression Box
- The final equation should look like this: LG10(Infant_1+730)
- Create a new graph, replacing the differenced Y-axis variable Infant_1 with the log transformation variable: LogInfant_1

4. Autocorrelations

a. Analyze → Forecasting → Autocorrelations

- Select LogInfant_1 and move it to the Variable Box
- Click OK
- None of the Autocorrelations should be significant, and the graphs should show that none of the cases pass the upper or lower confidence limits
- If there are still significant autocorrelations, may need to difference again
- Because nothing is significant, move on and ID the model based on patterns (see handout)
- Based on the patterns seen in the ACFs and PACFs tables, we interpret this ARIMA as a (1,1,1) model. Once we have determined this, we have diagnosed the baseline and can now examine the predicted model based on the baseline

5. Forecasting

a. Analyze → Forecasting → Create Models

- Select the variable InfantdeathMS (the original variable before differencing and log transformation) and move it to the dependent variable box
- Select the variable Year and move it to the independent variable box
- Method: Select ARIMA
- Click the criteria box and enter what model it is in the nonseasonal column ex: we previously determined our (p,d,q) to be (1,1,1)
- Click continue and ok to view output

- Examine the Ljung-Box Q, it should not be significant
- If it is significant, you have not selected the correct ARIMA so try to determine a better model. If you still cannot determine the correct model based on patterns, try plugging in different ARIMAs such as (1,1,0) or (0,1,1) until the Ljung-Box Q is not significant
- Since our Ljung-Box Q is not significant ($p=0.255$) we know we have selected a good model

6. Determine if the intervention was significant

a. Data → Select Cases

- Select: All Cases
- Click Ok

b. Syntax

- Type the variables of interest into the syntax ex: **Arima** InfantdeathMS With WWII
- Type the model into the syntax ex: /model= (1 1 1) NOCONSTANT
- The final syntax is seen in the Demo2 Syntax File
- Highlight the entire syntax and hit play
- The output gives us the Parameter Estimates box which shows us that both the (p) autoregressive component ($t = -4.33, p < 0.001$) and (q) moving average component ($t = -2.87, p < 0.01$) as well as the intervention ($t = -2.16, p < 0.05$) are significant.