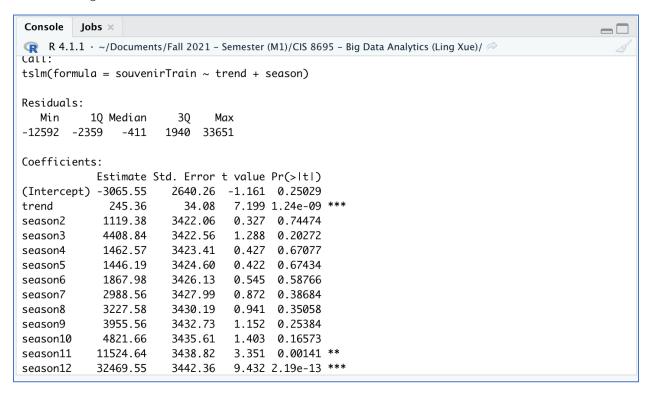
Problem-1: Time Series (Souvenir Sales)

- 17.6 b. Run a regression model with Sales (in Australian dollars) as the outcome variable and with a linear trend and monthly seasonality. Remember to fit only the training period. Call this model A.
- i. Examine the coefficients: Which month tends to have the highest average sales during the year? Why is this reasonable?
- ii. The estimated trend coefficient in model A is 245.36. What does this mean?

Ans 17.6 b i: December has the highest average sales during the year, followed by November. This seems reasonable as sales increase near the holiday season, when people purchase more. Also, Decembers in Australia are generally warmer, and it may be assumed that tourists are flocking that time of the year.

Ans 17.6 b ii. The estimated trend coefficient (245.36) in model A indicates "average linear increase" over each month, holding other variables constant.



- c. Run a regression model with an exponential trend and multiplicative seasonality. Remember to fit only training data. Call this model B.
- i. Fitting a model to log(Sales) with a linear trend is equivalent to fitting a model to Sales (in dollars) with what type of trend?
- ii. The estimated trend coefficient in model B is 0.02 What does this mean?
- iii. Use this model to forecast the sales in February 2002.

Ans 17.6 c i: Fitting a model to log(Sales) with linear trend is equivalent to fitting a model to Sales (in dollars) with "exponential trend".

Ans 17.6 c ii: The estimated trend coefficient (0.02) in model B indicates an overall increase "~2.1% each month" for every successive time period.

Ans 17.6 c iii. The forecasted sales in February 2002 using this model are "\$17062.99".

```
Console lobs
R 4.1.1 · ~/Documents/Fall 2021 - Semester (M1)/CIS 8695 - Big Data Analytics (Ling Xue)/
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 7.646363    0.084120    90.898    < 2e-16 ***
                      0.001086 19.449 < 2e-16 ***
           0.021120
season2
            0.282015
                      0.109028 2.587 0.012178 *
            0.694998
                      0.109044 6.374 3.08e-08 ***
season3
season4
            0.373873
                      0.109071
                                 3.428 0.001115 **
                      0.109109 3.865 0.000279 ***
season5
            0.421710
season6
            0.447046
                      0.109158 4.095 0.000130 ***
                                 5.341 1.55e-06 ***
season7
            0.583380
                      0.109217
season8
            0.546897
                      0.109287
                                 5.004 5.37e-06 ***
                      0.109368 5.811 2.65e-07 ***
season9
            0.635565
                      0.109460 6.664 9.98e-09 ***
season10
           0.729490
           1.200954
                      0.109562 10.961 7.38e-16 ***
season11
           1.952202 0.109675 17.800 < 2e-16 ***
season12
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' '1
Residual standard error: 0.1888 on 59 degrees of freedom
Multiple R-squared: 0.9424.
                              Adjusted R-squared: 0.9306
F-statistic: 80.4 on 12 and 59 DF, p-value: < 2.2e-16
> # Forecasting Sales for February 2022
 > febForecast <- modelB$coefficients["(Intercept)"] + modelB$coefficients["trend"]*86 + modelB$coe</p>
fficients["season2"]
 exp(febForecast)
(Intercept)
   17062.99
```

d. Compare the two regression models (A and B) in terms of forecast performance. Which model is preferable for forecasting? Mention at least two reasons based on the information in the outputs.

Ans 17.6 d: When comparing both models in terms of forecast performance, "model B" is preferrable as it appears to be a better fit. Looking at accuracy, we can find that MAPE (Mean Absolute Percentage Error) for model B is 15.5191 whereas model A is 26.6656. When we compare how well they explain data, model B has adjusted R-Squared value of "0.9306" (93%) while model A has "0.7476" (75%).

```
Console Jobs
R 4.1.1 · ~/Documents/Fall 2021 – Semester (M1)/CIS 8695 – Big Data Analytics (Ling Xue)/
season6
           0.447046
                      0.109158 4.095 0.000130 ***
season7
           0.583380
                      0.109217
                                 5.341 1.55e-06 ***
                      0.109287 5.004 5.37e-06 ***
season8
           0.546897
season9
           0.635565
                      0.109368
                                 5.811 2.65e-07 ***
                      0.109460 6.664 9.98e-09 ***
season10
           0.729490
                      0.109562 10.961 7.38e-16 ***
season11
           1.200954
           1.952202 0.109675 17.800 < 2e-16 ***
season12
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.1888 on 59 degrees of freedom
Multiple R-squared: 0.9424,
                              Adjusted R-squared: 0.9306
F-statistic: 80.4 on 12 and 59 DF, p-value: < 2.2e-16
> # Forecasting Sales for February 2022
> febForecast <- modelB$coefficients["(Intercept)"] + modelB$coefficients["trend"]*86 + modelB$coe
fficients["season2"]
> exp(febForecast)
(Intercept)
  17062.99
> modelAForecast <- forecast(modelA, h = validLength)</pre>
> accuracy(modelAForecast$mean, souvenirValid)
                    RMSE
                              MAE
                                                MAPE
              ME
                                       MPE
                                                          ACF1 Theil's U
Test set 8251.513 17451.55 10055.28 10.53397 26.66568 0.3206228 0.9075924
> modelBForecast <- forecast(modelB, h = validLength)</pre>
> accuracy(modelBForecast$mean, souvenirValid)
                                              MAPE
             ME RMSE MAE MPE
                                                         ACF1 Theil's U
Test set 4824.494 7101.444 5191.669 12.35943 15.5191 0.4245018 0.4610253
```

Sample Code (Problem-1: Time Series)

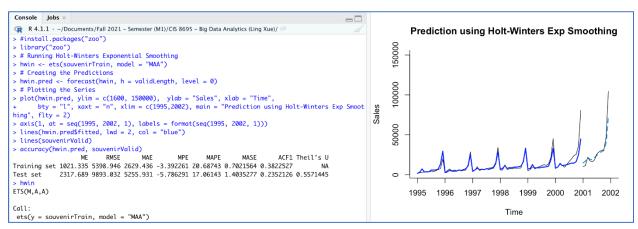


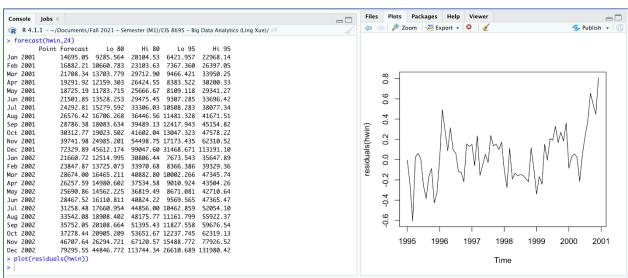
P1-Time Series.txt

Problem-2: Time Series (Souvenir Sales)

Holt Winter Exponential Smoothing

After applying the Holt Winter Exponential Smoothing method, we can see the predicted values of given data, forecasted values for 2001 and 2002 are displayed. The values are calculated at 80% and 95% confidence intervals. We can also see the residuals plot on and don't see any pattern in it, indicating forecasted values are right.





Sample Code (Problem-2: Time Series)



P2-Time Series.txt