Data624 - Homework4

Amit Kapoor

2/28/2021

Contents

Exercise 7.1

Consider the `pigs` series — the number of pigs slaughtered in Victoria each month.

```
str(pigs)
```

```
##  Time-Series [1:188] from 1980 to 1996: 76378 71947 33873 96428 105084 ...
```

a)

Use the `ses()` function in R to find the optimal values of α and ℓ_0, and generate forecasts for the next four months.

```
# Using ses for pigs
pigs_ses <- ses(pigs, h=4)
```

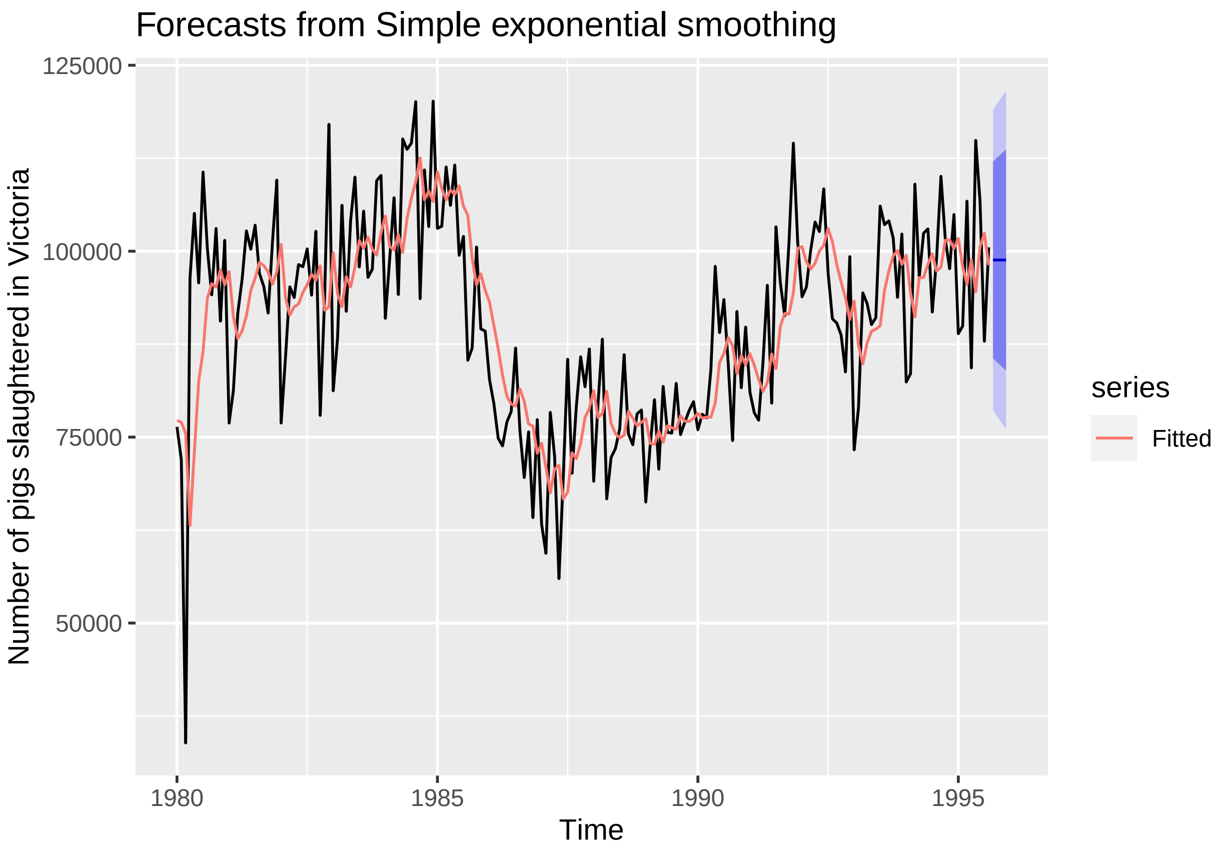
```r
#summary
summary(pigs_ses)
```

```
##
## Forecast method: Simple exponential smoothing
##
## Model Information:
## Simple exponential smoothing
##
## Call:
##  ses(y = pigs, h = 4)
##
##   Smoothing parameters:
##     alpha = 0.2971
##
##   Initial states:
##     l = 77260.0561
##
##   sigma:  10308.58
##
##      AIC     AICc      BIC
## 4462.955 4463.086 4472.665
##
## Error measures:
##                    ME    RMSE      MAE       MPE     MAPE      MASE       ACF1
## Training set 385.8721 10253.6 7961.383 -0.922652 9.274016 0.7966249 0.01282239
##
## Forecasts:
##          Point Forecast    Lo 80    Hi 80    Lo 95    Hi 95
## Sep 1995       98816.41 85605.43 112027.4 78611.97 119020.8
## Oct 1995       98816.41 85034.52 112598.3 77738.83 119894.0
## Nov 1995       98816.41 84486.34 113146.5 76900.46 120732.4
## Dec 1995       98816.41 83958.37 113674.4 76092.99 121539.8
```

Above summary shows the the optimal values of α and ℓ_0 are 0.2971 and 77260.0561 respectively. Using these values forecast is generated for next 4 months.

Next plot shows the forecast from simple exponential smoothing. Also one-step-ahead fitted values are plotted with the data over the period.

```r
autoplot(pigs_ses) +
  autolayer(fitted(pigs_ses), series="Fitted") +
  ylab("Number of pigs slaughtered in Victoria")
```



b)

Compute a 95% prediction interval for the first forecast using $\hat{y} \pm 1.96\sigma$ where σ is the standard deviation of the residuals. Compare your interval with the interval produced by R.

```
# 95% prediction interval for the first forecas
sd <- sd(residuals(pigs_ses))
ci95 <- c(lower = pigs_ses$mean[1] - 1.96*sd, upper = pigs_ses$mean[1] + 1.96*sd)
ci95
```

```
##     lower     upper
##  78679.97 118952.84
```

```
# By R
ci95_R <- c(pigs_ses$lower[1, "95%"], pigs_ses$upper[1, "95%"])
names(ci95_R) <- c("lower", "upper")
ci95_R
```

```
##     lower     upper
##  78611.97 119020.84
```

It appears the 95% prediction interval calculated by R is a little wider than the one given by the formula.

Exercise 7.5

Data set `books` contains the daily sales of paperback and hardcover books at the same store. The task is to forecast the next four days' sales for paperback and hardcover books.

a)

Plot the series and discuss the main features of the data.

b)

Use the `ses()` function to forecast each series, and plot the forecasts.

c)

Compute the RMSE values for the training data in each case.

Exercise 7.6

We will continue with the daily sales of paperback and hardcover books in data set books.

a)

Apply Holt's linear method to the `paperback` and `hardback` series and compute four-day forecasts in each case.

b)

Compare the RMSE measures of Holt's method for the two series to those of simple exponential smoothing in the previous question. (Remember that Holt's method is using one more parameter than SES.) Discuss the merits of the two forecasting methods for these data sets.

c)

Compare the forecasts for the two series using both methods. Which do you think is best?

d)

Calculate a 95% prediction interval for the first forecast for each series, using the RMSE values and assuming normal errors. Compare your intervals with those produced using `ses` and `holt`.

Exercise 7.7

For this exercise use data set `eggs`, the price of a dozen eggs in the United States from 1900–1993. Experiment with the various options in the `holt()` function to see how much the forecasts change with the damped trend, or with a Box-Cox transformation. Try to develop an intuition of what each argument is doing to the forecasts.

[Hint: use `h = 100` when calling `holt()` so you can clearly see the differences between the various options when plotting the forecasts.]

Which model gives the best RMSE?

Exercise 7.8

Recall your retail time series data (from Exercise 3 in Section 2.10).

a)

Why is multiplicative seasonality necessary for this series?

b)

Apply Holt-Winters' multiplicative method to the data. Experiment with making the trend damped.

c)

Compare the RMSE of the one-step forecasts from the two methods. Which do you prefer?

d)

Check that the residuals from the best method look like white noise.

e)

Now find the test set RMSE while training the model to the end of 2010. Can you beat the seasonal naive approach from Exercise 8 in Section 3.7?

Exercise 7.9

For the same retail data, try an STL decomposition applied to the Box-Cox transformed series, followed by ETS on the seasonally adjusted data. How does that compare with your best previous forecasts on the test set?