

Tutorial 1

IM532 3.0 Applied Time Series Forecasting

07/04/2020

Data

Use data in `colmozzie` package in R.

Package installation

```
install.packages("colmozzie")
```

```
library(colmozzie)
head(colmozzie)
```

	Cases	Year	Week	TEM	TMAX	Tm	SLP	H	PP	VV	V
1	44	2009	1	27.300	32.7000	23.60	1010.7	68.0000	0.00	13.0000	11.1000
2	39	2009	2	26.400	29.7667	23.90	1010.7	78.6667	0.00	18.3333	5.8333
3	57	2009	3	27.140	32.0200	23.52	1012.58	67.0000	0.00	20.0000	6.1200
4	53	2009	4	26.800	31.0000	23.35	1009.9	68.0000	0.00	20.0000	7.1500
5	29	2009	5	26.775	30.0750	23.60	1010.1	78.0000	17.21	18.9750	3.2000
6	45	2009	6	26.850	30.5500	23.00	1012.05	72.5000	0.00	20.0000	5.5500

VM

1	20.600
2	9.200
3	10.460
4	11.100
5	6.825
6	10.300

To view the description of variables type,

```
?colmozzie
```

Other packages

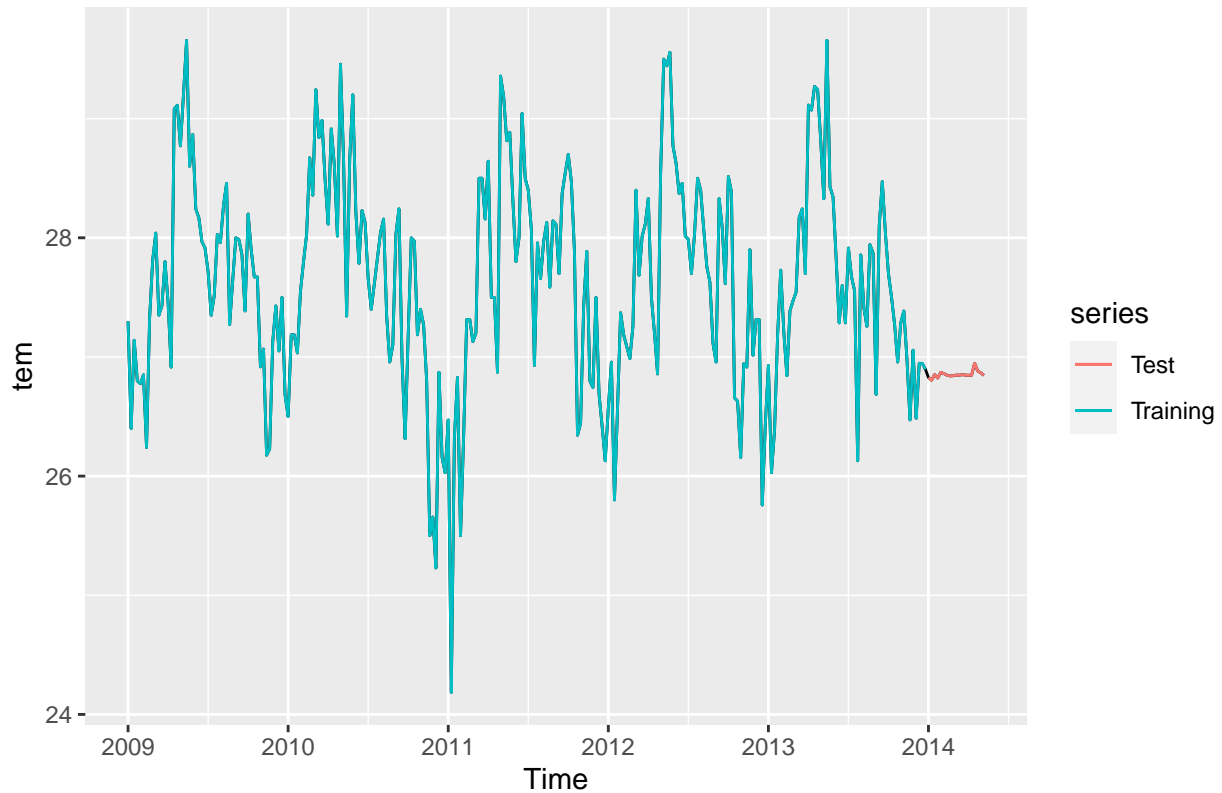
```
install.packages("forecast")
library(forecast)
```

1. Use TEM: average temperature of the week (Celcius) in `colmozzie` data set. Split the data into two parts: i) training set (2009 to 2013 data) and ii) test set (2014 data).

```
tem <- ts(colmozzie$TEM, start=c(2009, 1), frequency = 52)
tem.train <- window(tem, end=c(2013, 52))
tem.test <- window(tem, start=c(2014, 1))
```

2. The following code can be used to check whether time series has been split correctly.

```
library(forecast)
autoplot(tem) +
  autolayer(tem.train, series="Training") +
  autolayer(tem.test, series="Test")
```



4. Calculate forecasts using naive method applied to `tem.train` .

```
forecast.naive <- naive(tem.train, h=length(tem.test))
forecast.naive
```

	Point	Forecast	Lo 80	Hi 80	Lo 95	Hi 95
	2014.000	26.89685	26.01636	27.77735	25.55025	28.24346
	2014.019	26.89685	25.65164	28.14207	24.99247	28.80124
	2014.038	26.89685	25.37179	28.42192	24.56446	29.22924
	2014.058	26.89685	25.13586	28.65785	24.20364	29.59006
	2014.077	26.89685	24.92800	28.86571	23.88575	29.90796
	2014.096	26.89685	24.74008	29.05362	23.59836	30.19535
	2014.115	26.89685	24.56727	29.22643	23.33407	30.45964
	2014.135	26.89685	24.40643	29.38728	23.08808	30.70563
	2014.154	26.89685	24.25536	29.53835	22.85704	30.93667
	2014.173	26.89685	24.11247	29.68123	22.63851	31.15519
	2014.192	26.89685	23.97657	29.81714	22.43067	31.36304
	2014.212	26.89685	23.84672	29.94699	22.23207	31.56163
	2014.231	26.89685	23.72217	30.07153	22.04160	31.75211
	2014.250	26.89685	23.60233	30.19138	21.85832	31.93539
	2014.269	26.89685	23.48670	30.30701	21.68147	32.11223
	2014.288	26.89685	23.37486	30.41885	21.51043	32.28328

```

2014.308      26.89685 23.26647 30.52724 21.34466 32.44905
2014.327      26.89685 23.16122 30.63249 21.18369 32.61002
2014.346      26.89685 23.05885 30.73486 21.02714 32.76657

```

5. Compare the accuracy on the results using the test set.

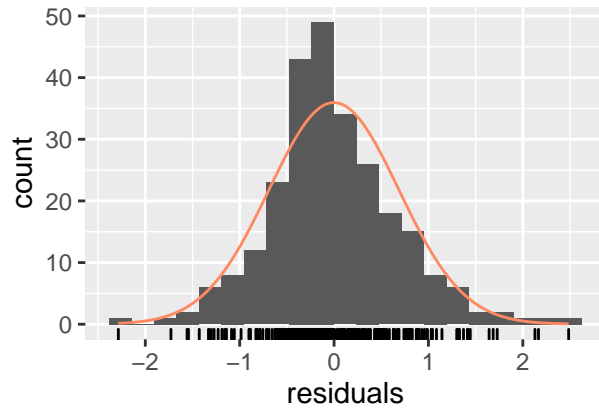
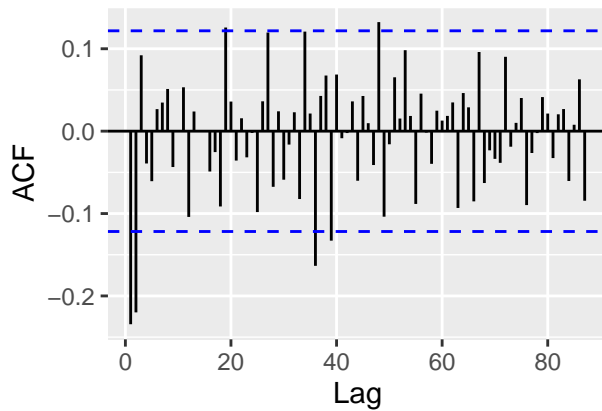
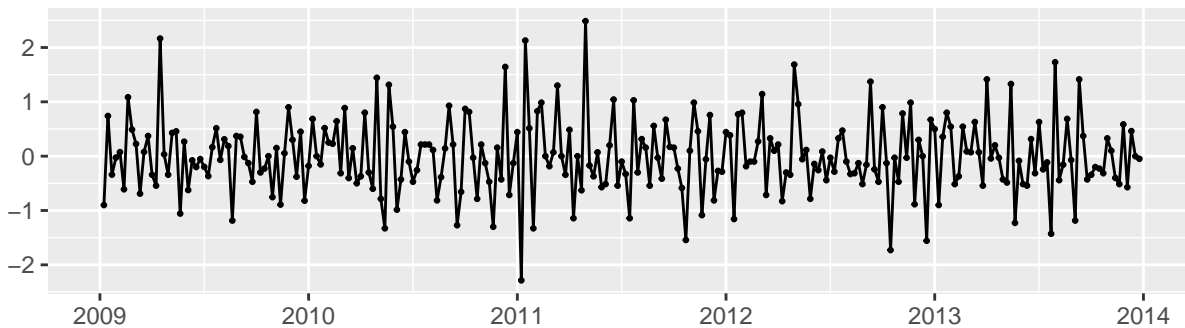
```
accuracy(forecast.naive, tem.test)
```

	ME	RMSE	MAE	MPE	MAPE
Training set	-0.001556551	0.68705633	0.51753041	-0.03729597	1.8815583
Test set	-0.044448907	0.05243908	0.04971848	-0.16563780	0.1851932
	MASE	ACF1	Theil's U		
Training set	0.74948896	-0.2343118	NA		
Test set	0.07200244	0.2525745	1.492231		

6. Check the residuals. Do they resemble white noise?

```
checkresiduals(forecast.naive)
```

Residuals from Naive method



Ljung-Box test

```

data: Residuals from Naive method
Q* = 92.034, df = 52, p-value = 0.0005217

```

```
Model df: 0. Total lags used: 52
```

7. Use other benchmark methods such as snaive, average method and random walk with drift methods to forecast the training set and compare the results on the test set.

8. Which do you think is best to forecast average temperature in Colombo district? Give reasons for your answer.
9. Repeat the exercise for the other climate variable (TMAX, Tm, SLP, H, PP, VV, V, VM) in `colmozzie` dataset.

Note:

The above exercise is prepared based on Hyndman, R. J., & Athanasopoulos, G. (2018). *Forecasting: principles and practice*. OTexts. adapting to `colmozzie` data set available on `colmozzie` package in R written by Thiyanga S. Talagala.