

Covid19

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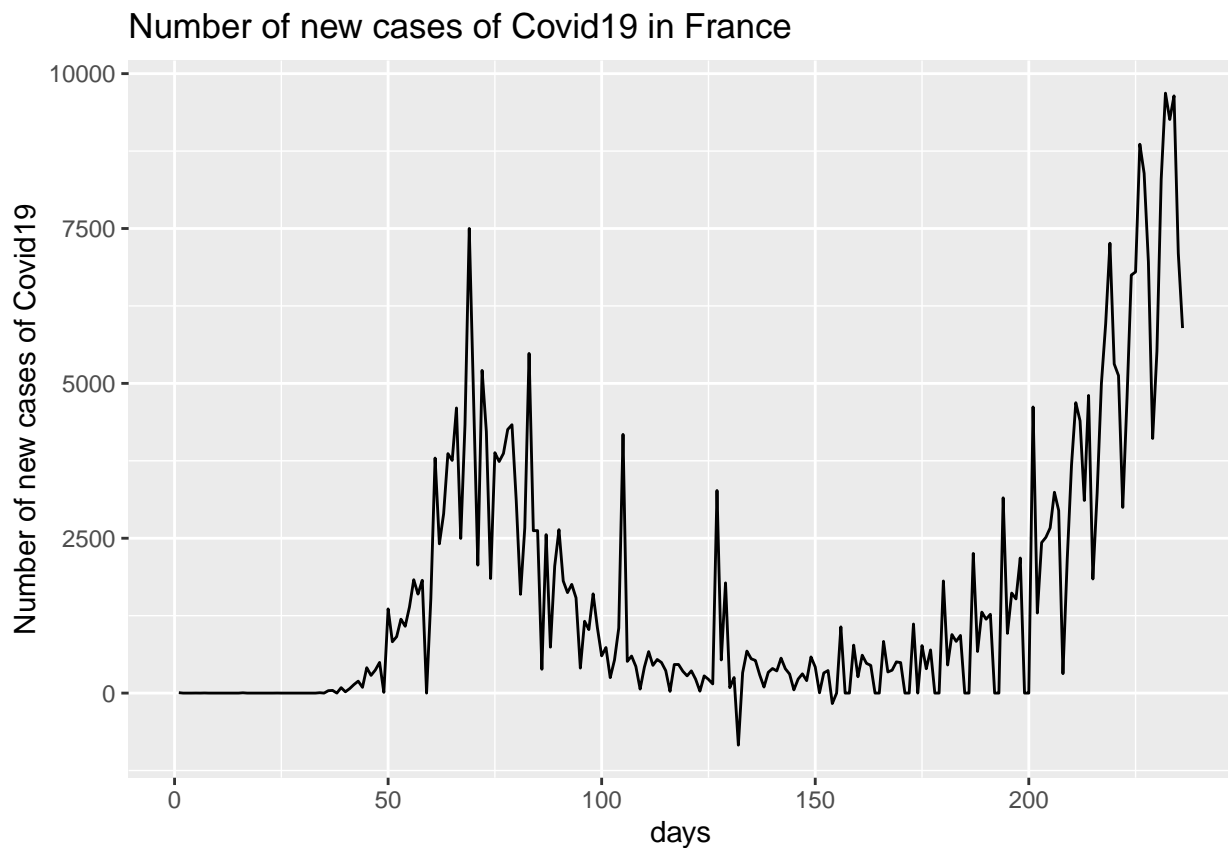
```
covid19 = read.csv("~/Enseignement/Formations/Times series/data/WHO-COVID-19-global-data.csv")
covid19_F=covid19[covid19$Country=="France",]
covid19_F_nc=ts(covid19_F$New_cases)
covid19_F_nd=ts(covid19_F$New_deaths)
```

We plot the number of new cases

```
library(forecast)
library(ggplot2)
```

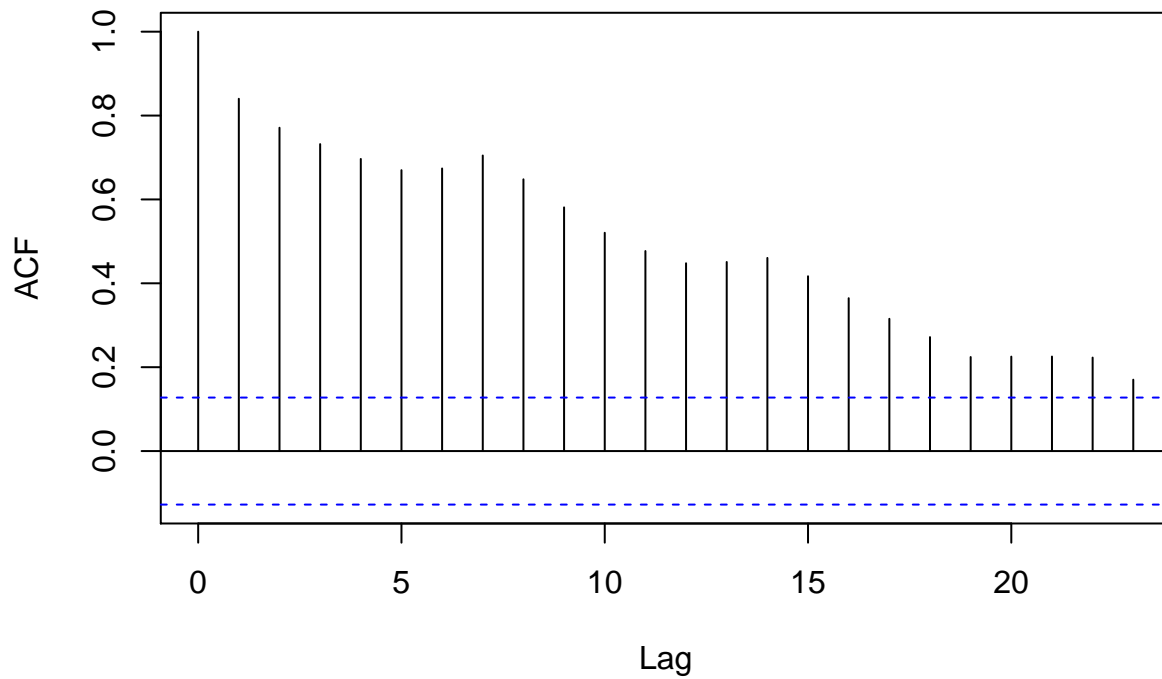
```
## Warning: package 'ggplot2' was built under R version 3.5.2
```

```
autoplot(covid19_F_nc) +
  ggtitle('Number of new cases of Covid19 in France') + xlab('days') +
  ylab('Number of new cases of Covid19')
```



```
acf(covid19_F_nc)
```

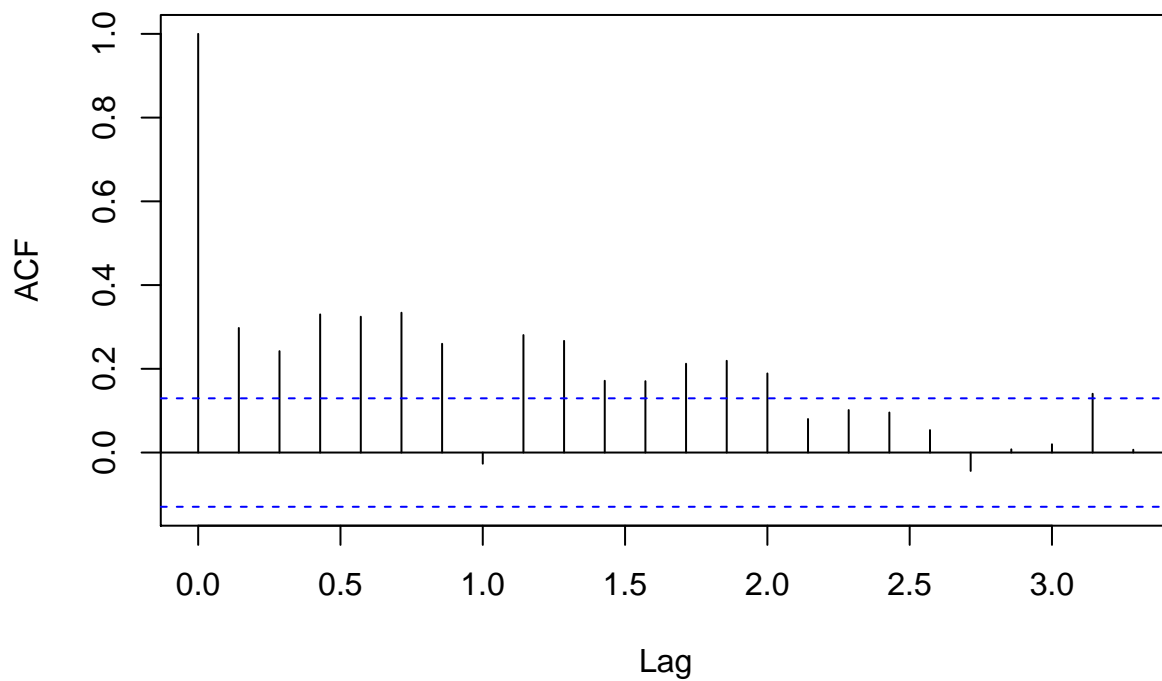
Series covid19_F_nc



It looks like if there is a trend and a seasonnal pattern of period equal to 7

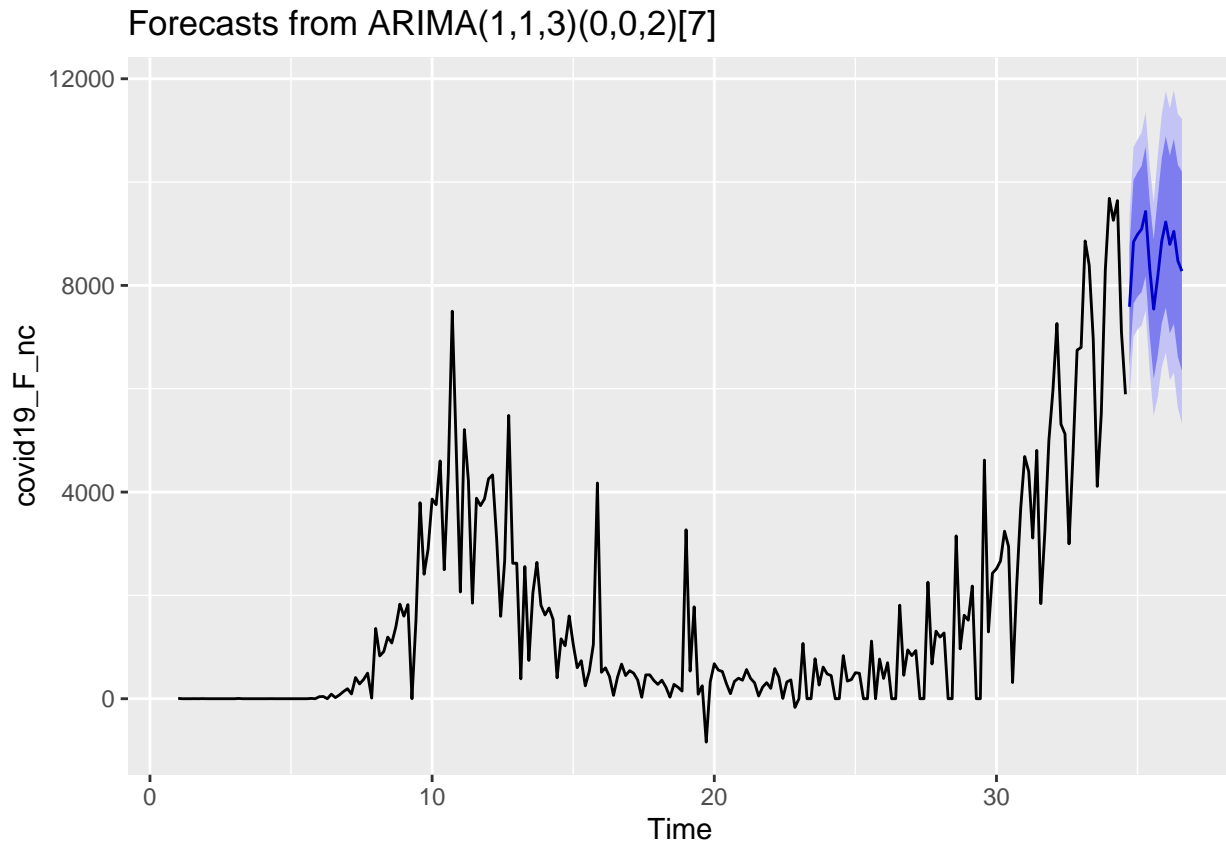
```
covid19_F_nc=ts(covid19_F_nc,freq=7)
acf(diff(covid19_F_nc,lag=7))
```

Series diff(covid19_F_nc, lag = 7)



Forecasting of the number of new cases

```
fit=auto.arima(covid19_F_nc)
autoplot(forecast(fit,h=14))
```



```
print(forecast(fit,h=14))
```

##	Point Forecast	Lo 80	Hi 80	Lo 95	Hi 95
## 34.71429	7586.987	6435.536	8738.438	5825.994	9347.980
## 34.85714	8842.091	7645.119	10039.062	7011.481	10672.701
## 35.00000	8986.545	7784.993	10188.097	7148.929	10824.161
## 35.14286	9089.448	7868.535	10310.362	7222.223	10956.674
## 35.28571	9432.374	8177.930	10686.818	7513.867	11350.881
## 35.42857	8340.200	7042.081	9638.319	6354.898	10325.502
## 35.57143	7541.099	6193.126	8889.072	5479.552	9602.645
## 35.71429	8164.517	6642.949	9686.085	5837.480	10491.554
## 35.85714	8859.255	7261.977	10456.534	6416.429	11302.082
## 36.00000	9228.608	7574.163	10883.053	6698.352	11758.864
## 36.14286	8793.009	7073.781	10512.236	6163.677	11422.340
## 36.28571	9045.476	7257.573	10833.379	6311.115	11779.837
## 36.42857	8475.186	6617.152	10333.220	5633.568	11316.804
## 36.57143	8275.703	6347.564	10203.841	5326.869	11224.536

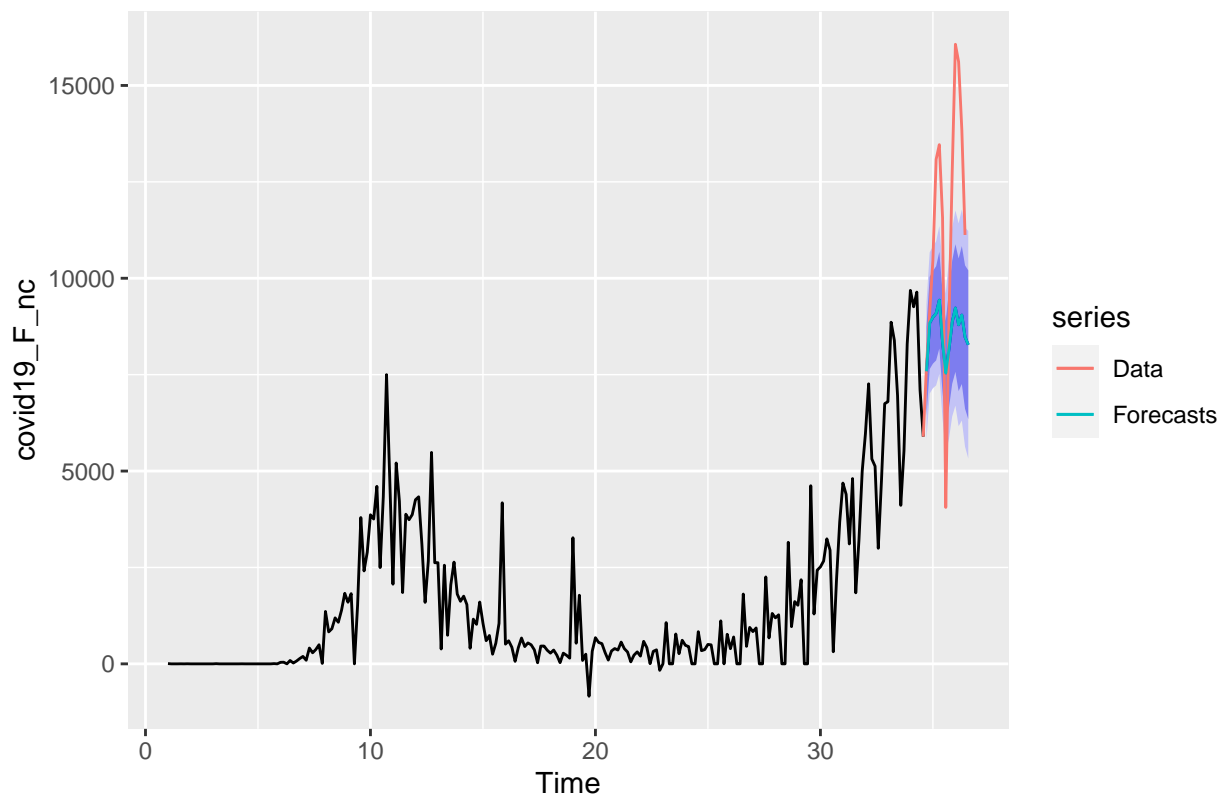
We can compare with the true data observed during the 2 course sessions

```
covid19_2 = read.csv("~/Enseignement/Formations/Times series/data/WHO-COVID-19-global-data2.csv")
covid19_F_2=covid19_2[covid19_2$Country=="France",]
covid19_F_2=covid19_F_2[-(1:21),]
```

```
covid19_F_2_nc=ts(covid19_F_2$New_cases)
covid19_F_2_nd=ts(covid19_F_2$New_deaths)
covid19_F_2_nc=ts(covid19_F_2_nc,freq=7)
```

```
prev=forecast(fit,h=14)
autoplot(prev) + autolayer(tail(covid19_F_2_nc,14), series="Data")+
  autolayer(prev$mean, series="Forecasts")
```

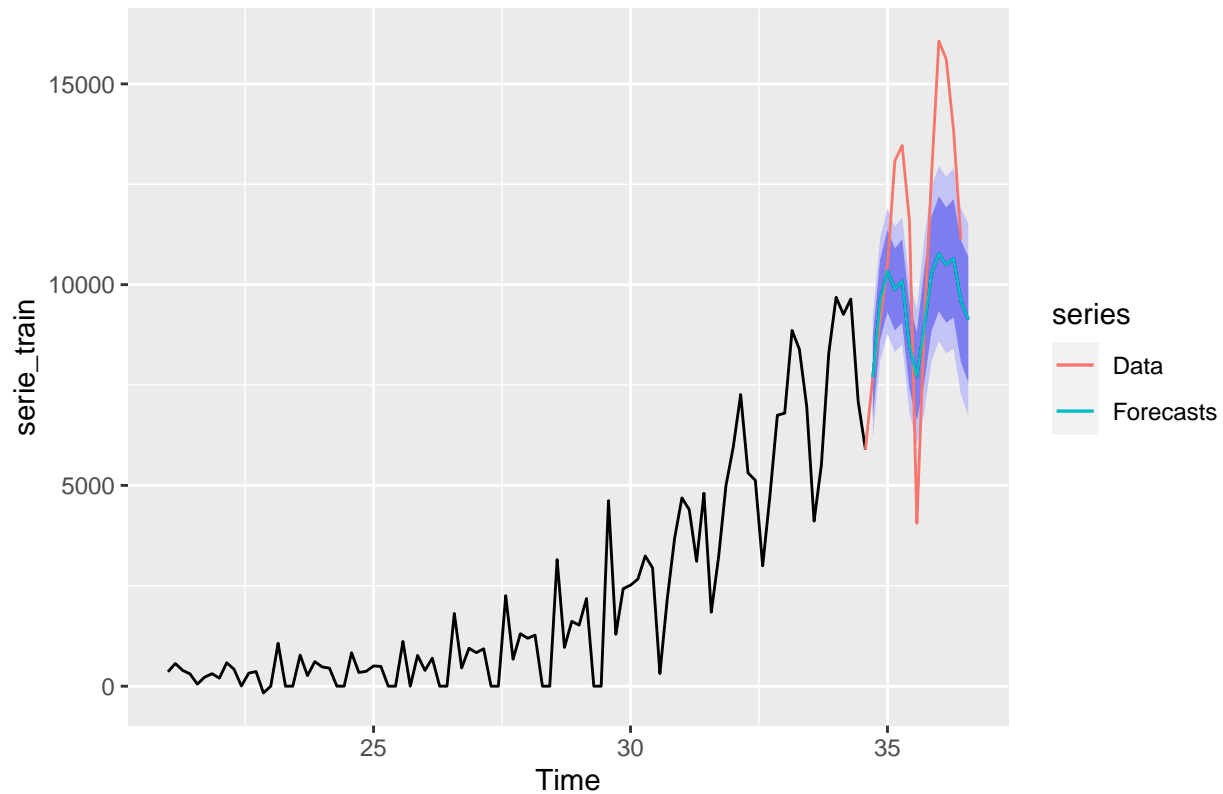
Forecasts from ARIMA(1,1,3)(0,0,2)[7]



It seems that there were a multiplicative effect of the seasonal part, that have not been taken into account. It seems also that the slope of the trend has been under-evaluated. It may be cause by the fact that the regime of the serie is different in the first 20 weeks. We can try the forecast by removing these elements.

```
serie_train=window(covid19_F_2_nc,start=c(21,1),end=c(34, 5))
serie=window(covid19_F_2_nc,start=c(21,1))
fit=auto.arima(serie_train)
prev=forecast(fit,h=14)
autoplot(prev) + autolayer(tail(covid19_F_2_nc,14), series="Data")+
  autolayer(prev$mean, series="Forecasts")
```

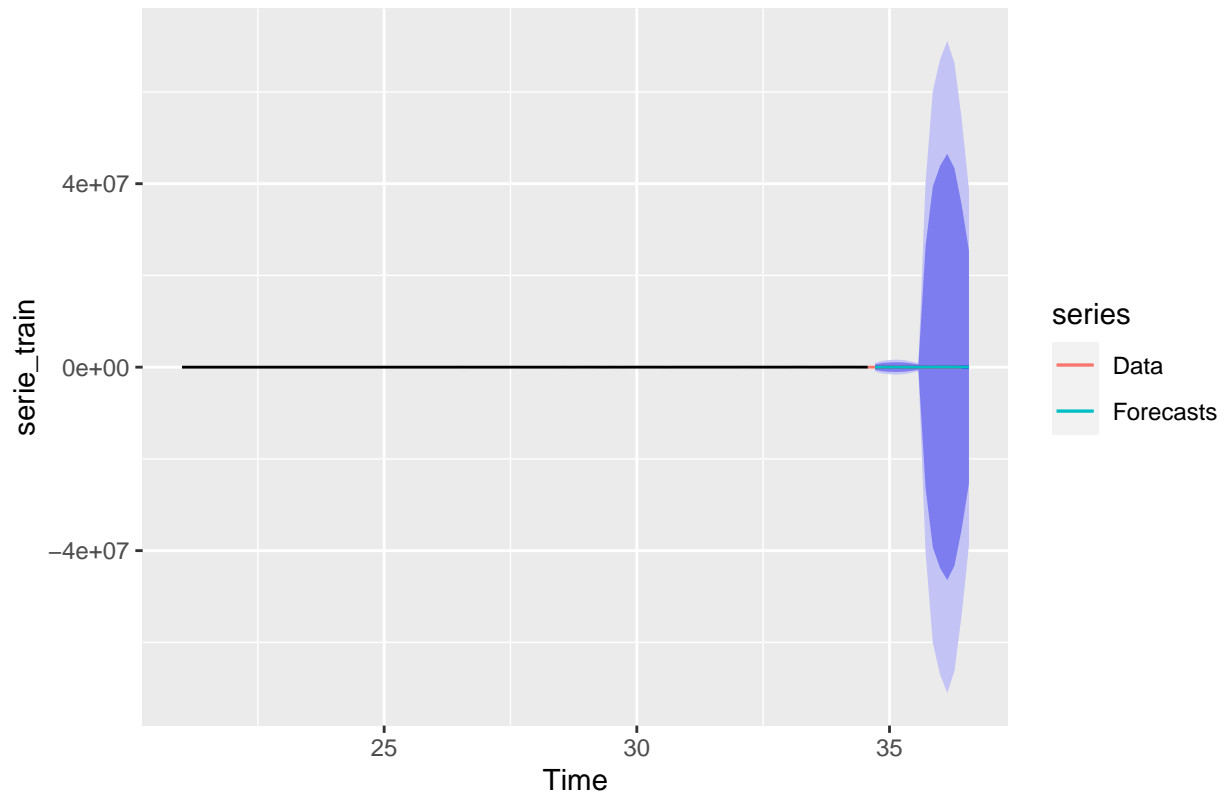
Forecasts from ARIMA(2,1,2)(1,0,0)[7] with drift



The prevision does not take into account the fact that the amplitude of the periodic pattern increase. We can try a multiplicative exponential smoothing model

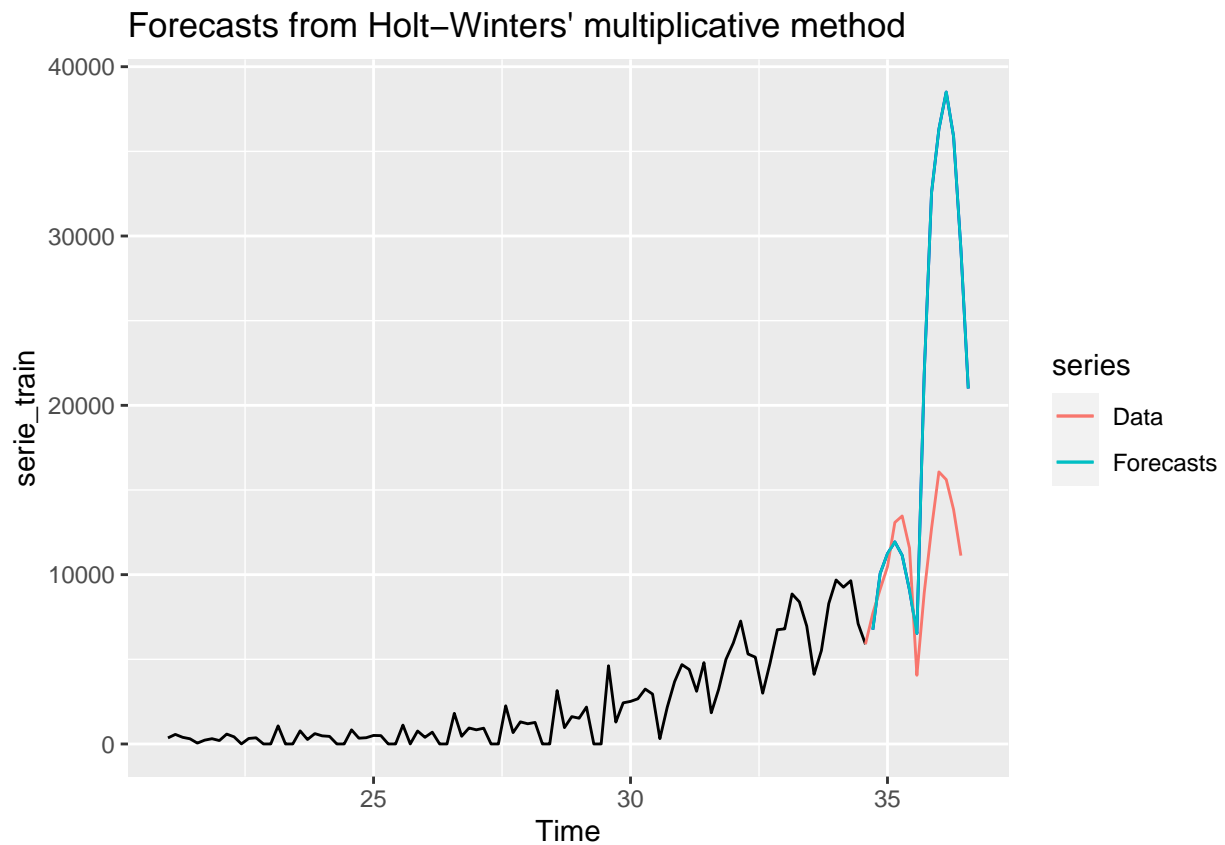
```
serie_train[serie_train<1]=0.1
fit2=hw(serie_train,seasonal='multiplicative',h=14)
autoplot(fit2)+ autolayer(tail(covid19_F_2_nc,14), series="Data")+
  autolayer(fit2$mean, series="Forecasts")
```

Forecasts from Holt–Winters' multiplicative method



we do not see nothing since the prediction interval is huge. We will plot the forecast without it:

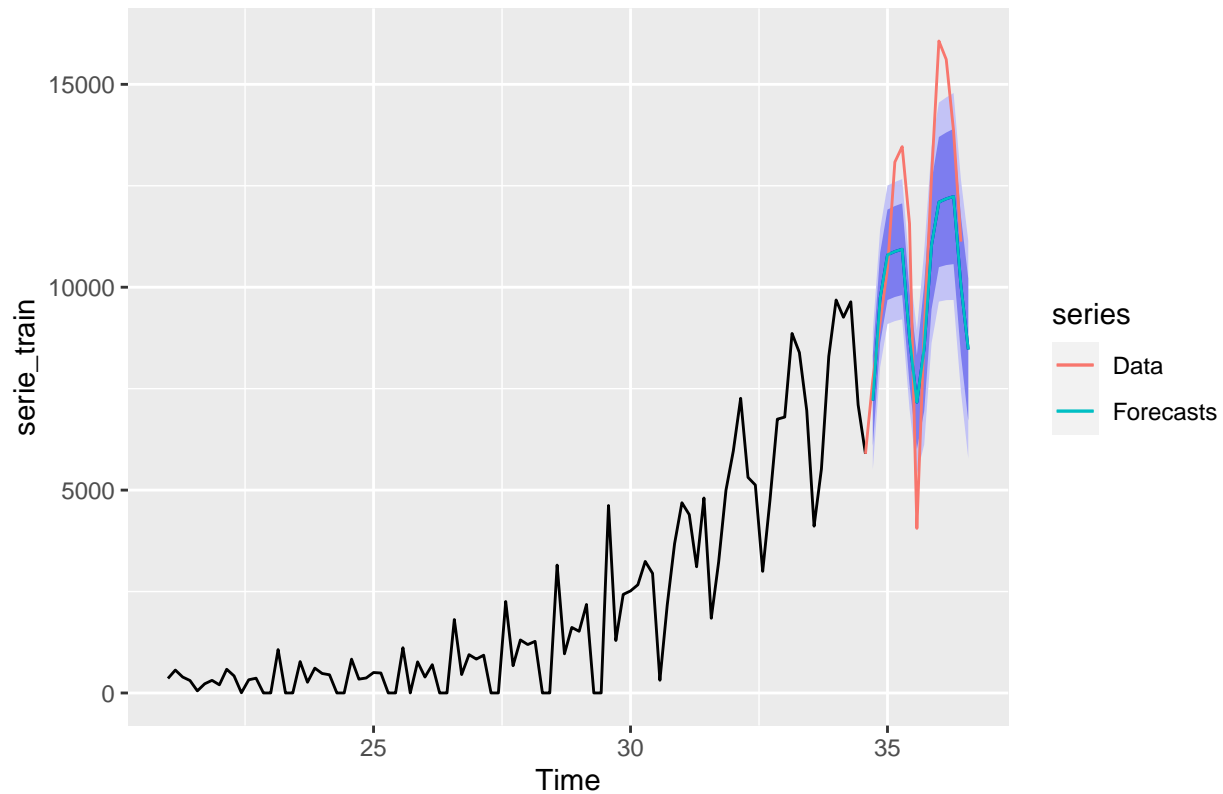
```
autoplot(fit2,PI = F)+ autolayer(tail(covid19_F_2_nc,14), series="Data")+  
  autolayer(fit2$mean, series="Forecasts")
```



and compare it to an additive :

```
fit3=hw(serie_train,seasonal='additive',h=14)
autoplot(fit3)+ autolayer(tail(covid19_F_2_nc,14), series="Data")+
  autolayer(fit3$mean, series="Forecasts")
```

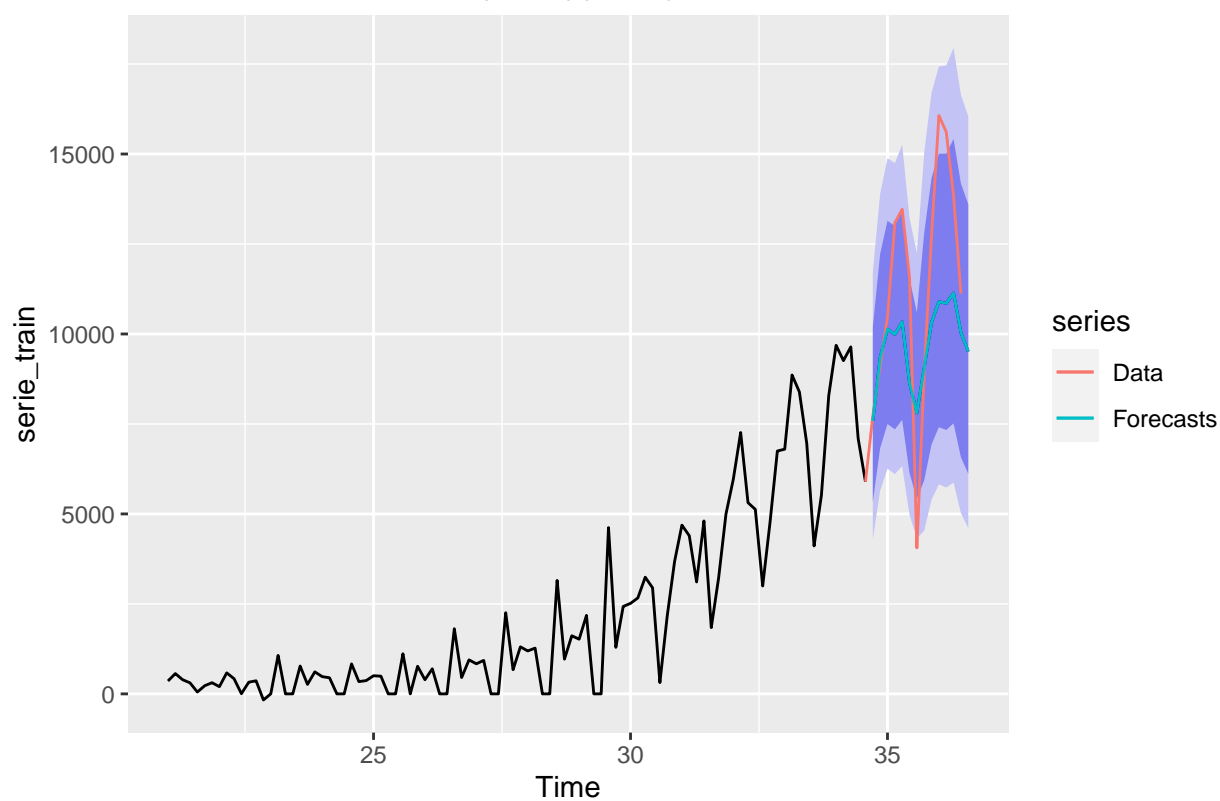
Forecasts from Holt–Winters' additive method



Finally, we can try to improve our SARIMA model by using the Box.Cox transformation

```
serie_train=window(covid19_F_2_nc,start=c(21,1),end=c(34, 5))
serie=window(covid19_F_2_nc,start=c(21,1))
fit=auto.arima(serie_train,lambda="auto")
prev=forecast(fit,h=14)
autoplot(prev) + autolayer(tail(covid19_F_2_nc,14), series="Data")+
  autolayer(prev$mean, series="Forecasts")
```


Forecasts from ARIMA(2,1,1)(1,0,0)[7] with drift



The forecast would have been better with such model, but our analyse has been influenced by observing the true future...