

```

HireRate.df <- read.csv("HireRate.csv", header = T)
View(HireRate.df)
library(forecast)

# Plotting the entire ts
HireRate.ts <- ts(HireRate.df$Hire.rate, start = c(2005,1), end = c(2015,6), frequency = 12)
plot(HireRate.ts,xlab="Time",ylab="Hire Rate")

# Partitioning
Valid <- 24
Training <- length(HireRate.ts)- Valid

train.h <- window(HireRate.ts,start=c(2005,1),end=c(2005,Training))
valid.h <- window(HireRate.ts,start=c(2005,Training+1),end=c(2005,Training+Valid))

# Naïve Forecast with seasonality
naive.forecast <- snaive(train.h, h = Valid, level = 0)
plot(naive.forecast,xlab="Time", ylab ="Hire Rate")
lines(valid.h)
summary(naive.forecast)

```

```
Forecast method: Seasonal naive method
```

```
Model Information:
```

```
Call: snaive(y = train.h, h = valid, level = 0)
```

```
Residual sd: 2.7085
```

```
Error measures:
```

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1
Training set	-0.8888889	2.836273	2.088889	-11.02824	22.00302	1	0.709137

```
Forecasts:
```

	Point Forecast	Lo 0	Hi 0
Jul 2013	15	15	15
Aug 2013	17	17	17
Sep 2013	12	12	12
Oct 2013	14	14	14
Nov 2013	10	10	10
Dec 2013	3	3	3
Jan 2014	12	12	12
Feb 2014	8	8	8
Mar 2014	10	10	10
Apr 2014	16	16	16
May 2014	17	17	17
Jun 2014	17	17	17
Jul 2014	15	15	15
Aug 2014	17	17	17
Sep 2014	12	12	12
Oct 2014	14	14	14
Nov 2014	10	10	10
Dec 2014	3	3	3
Jan 2015	12	12	12
Feb 2015	8	8	8
Mar 2015	10	10	10
Apr 2015	16	16	16
May 2015	17	17	17
Jun 2015	17	17	17

```
# Linear trend with seasonality
HireRate.lm <- tslm(train.h ~ trend + season)
linear.forecast <- forecast(HireRate.lm, h=Valid, level=0)
```

```
plot(linear.forecast, xlab="Time", ylab="Hire Rate")
lines(valid.h)
summary(linear.forecast)
```

```
Forecast method: Linear regression model
```

```
Model Information:
```

```
Call:
tslm(formula = train.h ~ trend + season)
```

```
Coefficients:
```

```
(Intercept)          trend      season2      season3      season4      season5
season6      season7      season8      season9      season10      season11
17.76362      -0.09041      -5.02070      -0.81917      4.38235      5.02832
6.11874      4.41667      4.38208      2.22249      2.43791      -2.97168
season12
-8.88126
```

```
Error measures:
```

	ME	RMSE	MAE	MPE	MAPE	MASE
ACF1						
Training set	1.235225e-16	2.474595	1.978149	-6.054011	21.76669	0.9469864
33729						0.83

```
Forecasts:
```

	Point Forecast	Lo 0	Hi 0
Jul 2013	12.8676471	12.8676471	12.8676471
Aug 2013	12.7426471	12.7426471	12.7426471
Sep 2013	10.4926471	10.4926471	10.4926471
Oct 2013	10.6176471	10.6176471	10.6176471
Nov 2013	5.1176471	5.1176471	5.1176471
Dec 2013	-0.8823529	-0.8823529	-0.8823529
Jan 2014	7.9084967	7.9084967	7.9084967
Feb 2014	2.7973856	2.7973856	2.7973856
Mar 2014	6.9084967	6.9084967	6.9084967
Apr 2014	12.0196078	12.0196078	12.0196078
May 2014	12.5751634	12.5751634	12.5751634
Jun 2014	13.5751634	13.5751634	13.5751634
Jul 2014	11.7826797	11.7826797	11.7826797
Aug 2014	11.6576797	11.6576797	11.6576797
Sep 2014	9.4076797	9.4076797	9.4076797
Oct 2014	9.5326797	9.5326797	9.5326797
Nov 2014	4.0326797	4.0326797	4.0326797
Dec 2014	-1.9673203	-1.9673203	-1.9673203
Jan 2015	6.8235294	6.8235294	6.8235294
Feb 2015	1.7124183	1.7124183	1.7124183
Mar 2015	5.8235294	5.8235294	5.8235294
Apr 2015	10.9346405	10.9346405	10.9346405
May 2015	11.4901961	11.4901961	11.4901961
Jun 2015	12.4901961	12.4901961	12.4901961

```
# Quadratic trend with seasonality
```

```
HireRate.quad <- tslm(train.h ~ trend + I(trend^2) + season)
quadratic.forecast <- forecast(HireRate.quad, h=Valid, level=0)
```

```
plot(quadratic.forecast, xlab="Time", ylab="Hire Rate")
lines(valid.h)
summary(quadratic.forecast)
```

```

Forecast method: Linear regression model

Model Information:

Call:
tslm(formula = train.h ~ trend + I(trend^2) + season)

Coefficients:
(Intercept)          trend      I(trend^2)         season2         season3         season4
season5          season6      season7         season8         season9        season10
  21.494689      -0.318350      0.002213      -5.011845      -0.805894      4.395631
5.037174      6.118736      4.868113      4.842379      2.687219      2.902633
  season11          season12
 -2.511379      -8.429817

Error measures:

              ME      RMSE      MAE      MPE      MAPE      MASE      A
CF1
Training set 2.338475e-18 1.797464 1.398552 -2.72295 13.4309 0.6695195 0.6862
318

Forecasts:
      Point Forecast      Lo 0      Hi 0
Jul 2013      17.050166 17.050166 17.050166
Aug 2013      17.164167 17.164167 17.164167
Sep 2013      15.153168 15.153168 15.153168
Oct 2013      15.517169 15.517169 15.517169
Nov 2013      10.256170 10.256170 10.256170
Dec 2013       4.495171  4.495171  4.495171
Jan 2014      13.086853 13.086853 13.086853
Feb 2014       8.241299  8.241299  8.241299
Mar 2014      12.617967 12.617967 12.617967
Apr 2014      17.994634 17.994634 17.994634
May 2014      18.815747 18.815747 18.815747
Jun 2014      20.081303 20.081303 20.081303
Jul 2014      19.019101 19.019101 19.019101
Aug 2014      19.186213 19.186213 19.186213
Sep 2014      17.228326 17.228326 17.228326
Oct 2014      17.645438 17.645438 17.645438
Nov 2014      12.437550 12.437550 12.437550
Dec 2014       6.729663  6.729663  6.729663
Jan 2015      15.374456 15.374456 15.374456
Feb 2015      10.582013 10.582013 10.582013
Mar 2015      15.011792 15.011792 15.011792
Apr 2015      20.441572 20.441572 20.441572
May 2015      21.315795 21.315795 21.315795
Jun 2015      22.634463 22.634463 22.634463

```

#Forecasting the 4-month Hire Rate

```

HireRate.quad.full <- tslm(HireRate.ts~ trend + I(trend^2) + season)
HireRate.forecast <- forecast(HireRate.quad.full, h = 4, level = 0)

```

HireRate.forecast

```

      Point Forecast      Lo 0      Hi 0
Jul 2015      20.99703 20.99703 20.99703
Aug 2015      21.27607 21.27607 21.27607
Sep 2015      19.45512 19.45512 19.45512
Oct 2015      19.73416 19.73416 19.73416

```