Forecasting issues

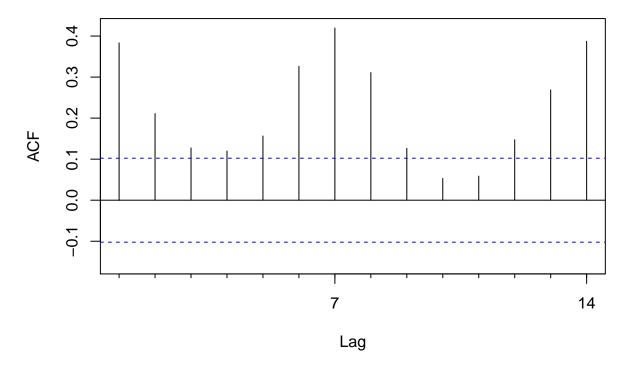
Forecast Padawan 2 November 17, 2016

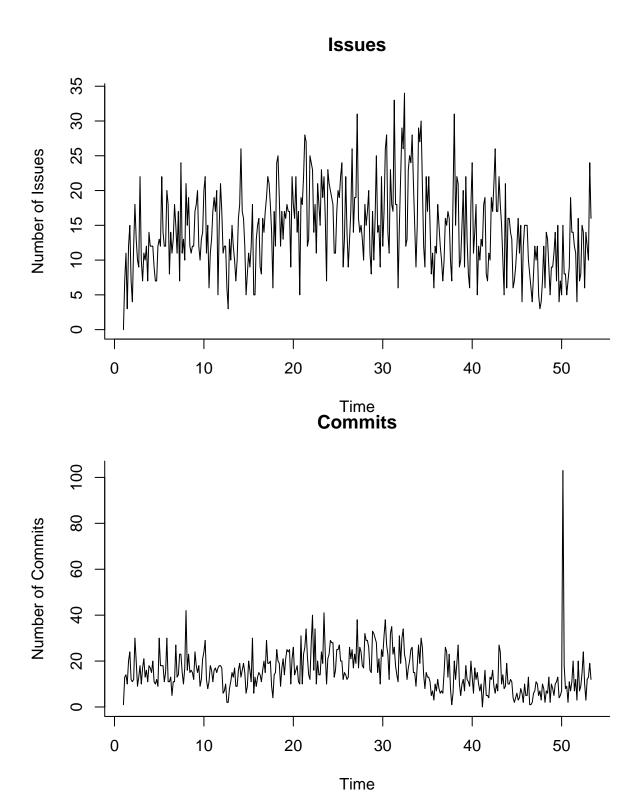
The goal of this experiment is to design the best model to forcaste the number of issue in the per day in the comming two weeks. We think that this could help Open Source organisation to manage there human ressources.

Load the data

```
#install.packages('forecast')
library('forecast')
library(knitr)
#load the data frame
repository.csv <- read.csv("time_series/julialang_julia_daily.csv")
repository.csv$date = as.POSIXlt(as.Date(repository.csv$date,format='%Y-%m-%d'))</pre>
```

keep the last 12 months





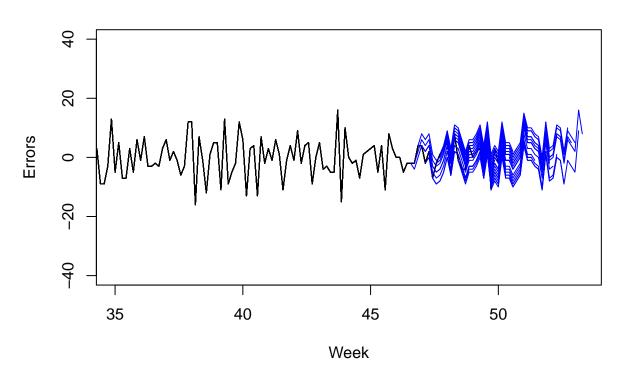
Naive Forecast

Naive

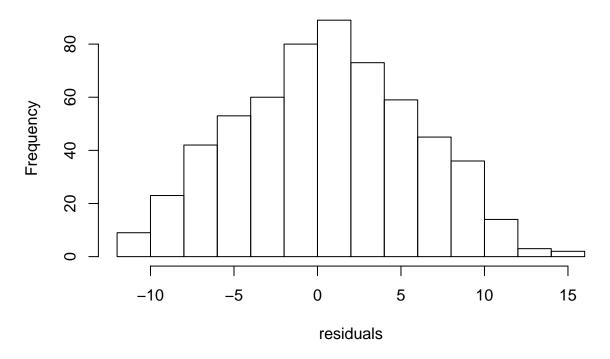
	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	0.0260345	6.693201	5.292482	-13.742514	43.88979	1.0482411	-0.3437756	NA
Test set	1.0578231	5.351101	4.442177	-8.511755	52.79771	0.8805386	-0.0010628	0.9754286
[](forecast_is	$sues_files/f$	igure-latex	/unnamed-ch	unk-6-1.pdf) <	!>			

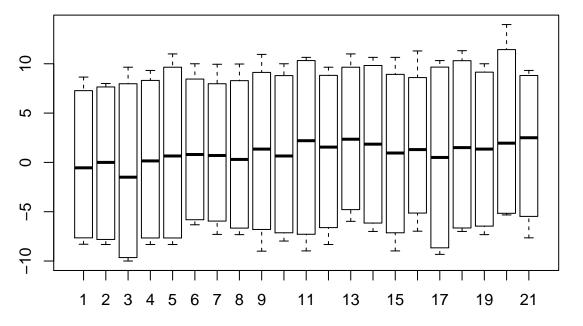
NULL

Residuals



Histogram of residuals





Seasonal Naive

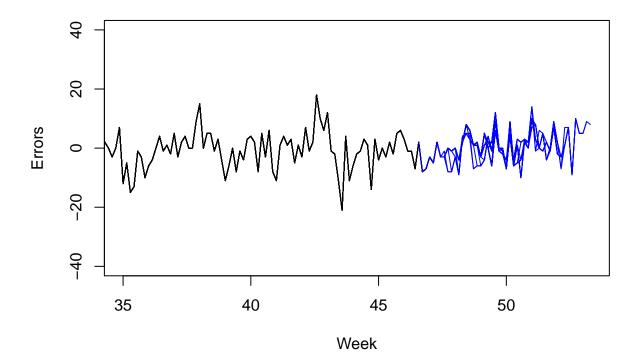
```
snaive.forecast <- function(sample) {
  results <- list()
  results$train <- sample$train.ts
  results$valid <- sample$valid.ts
  results$pred <- snaive(sample$train.ts, h=n.valid)
  results$fitted <- results$pred$fitted
  results$residual <- sample$valid.ts - results$pred$mean
  results$summary <- accuracy(results$pred, sample$valid.ts)

  return(results)
}
all.snaive.forecast <- sapply(1:n.sample, function(i) return(snaive.forecast(all.issues[,i])))
kable(mean.all.accuracy(all.snaive.forecast))</pre>
```

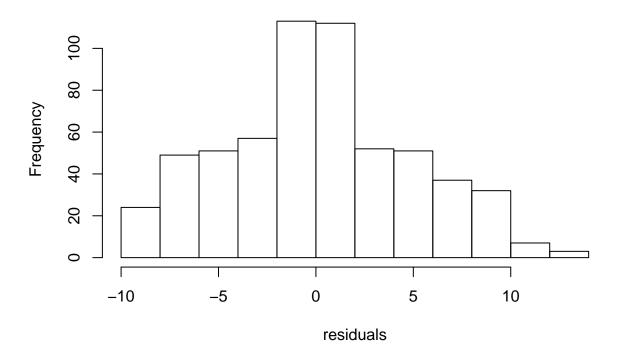
	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	0.0258080	6.436605	5.049082	-11.319282	39.94165	1.0000000	0.1250244	NA
Test set	0.6445578	4.846830	3.869048	-7.453078	43.62280	0.7670362	-0.0941361	0.8859009
$[](forecast_is$	$sues_files/f$	igure-latex	/unnamed-ch	unk-11-1.pdf				

NULL

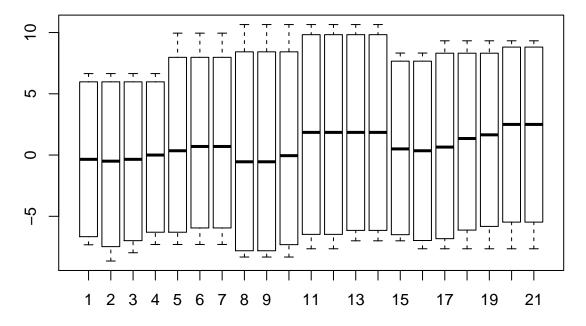
Residuals



Histogram of residuals



97.5% 95% 5% 2.5% ## 10 9 -7 -9



97.5% 95% 5% 2.5% ## 10 9 -7 -9

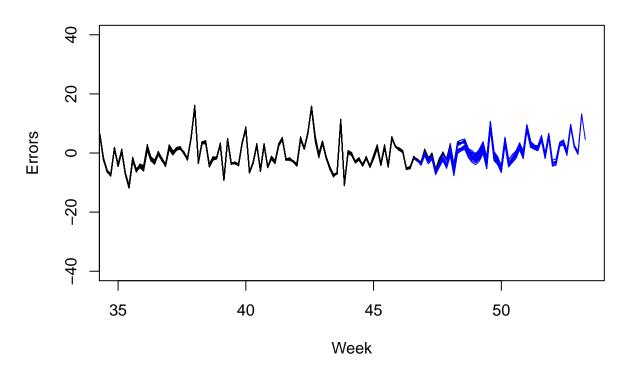
Smoothing

Exponential smoothing ZAA

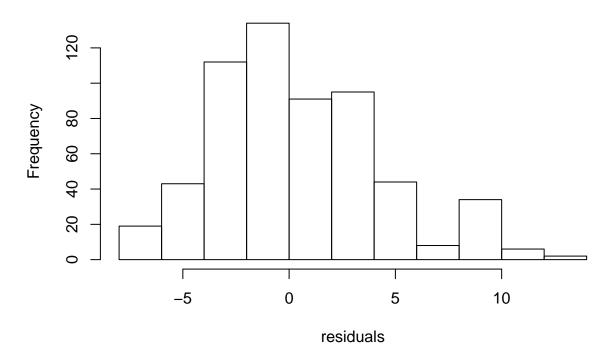
	3.65	DATOR	3.54.5	MDE	3.64.DE	3.5.4.CE	1 OT54	GD1 111 TT
	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	0.0162352	4.794487	3.808861	-Inf	Inf	0.7543855	0.0652222	NA
Test set	0.3894835	3.920933	3.091101	-8.900475	34.96387	0.6126821	-0.1436503	0.7219754
$[](forecast_is$	$sues_files/f$	igure-latex	/unnamed-ch	unk-16-1.pdf)			

NULL

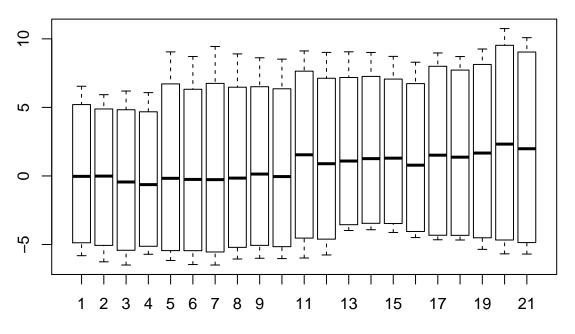
Residuals



Histogram of residuals

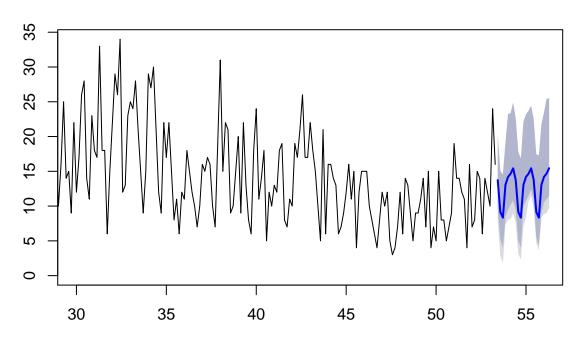


97.5% 95% 5% 2.5% ## 9.538827 8.534905 -5.217258 -6.223327



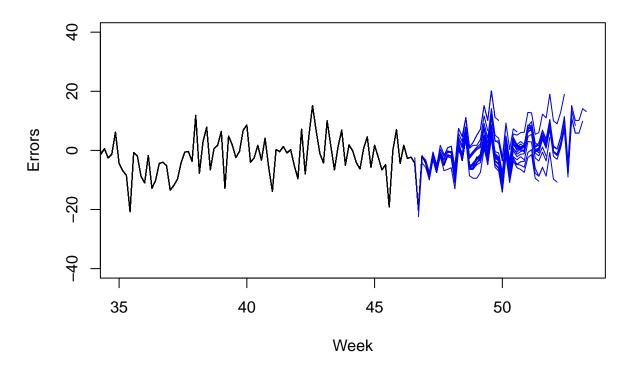
97.5% 95% 5% 2.5% ## 9.538827 8.534905 -5.217258 -6.223327

Forecasts from



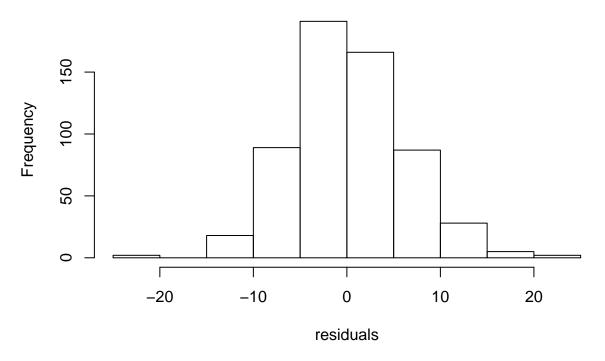
Double differencing

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	-0.0691779	6.246668	4.862882	-12.26012	37.71729	0.9564487	0.1526361	NA
Test set	0.2201166	5.804296	4.688047	-12.86606	57.22284	0.9219840	-0.0075625	1.010501
$[](forecast_is$	$sues_files/fi$	gure-latex/	unnamed-chu	nk-22-1.pdf				

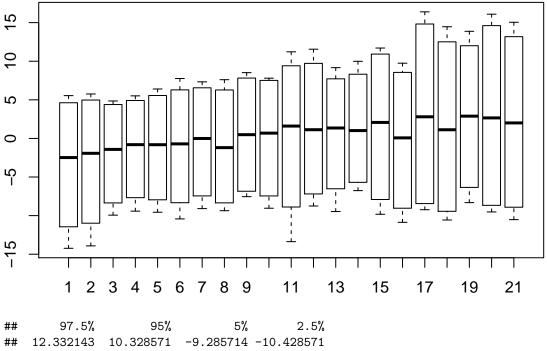


NULL

Histogram of residuals



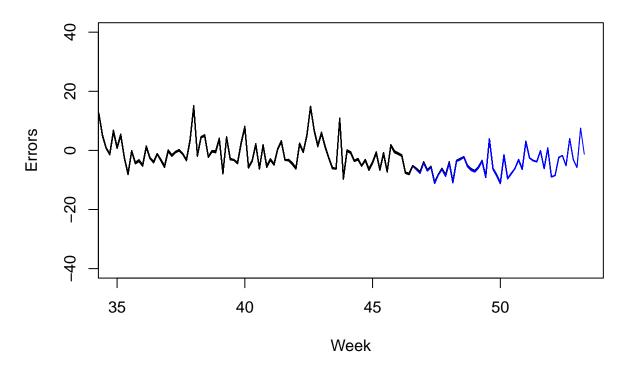
97.5% 95% 5% 2.5% ## 12.332143 10.328571 -9.285714 -10.428571



Regression

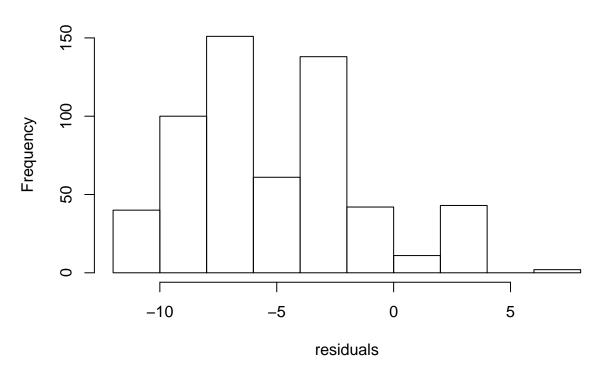
Linear additive regression season

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	0.000000	5.400961	4.274193	-Inf	Inf	0.8466724	0.3527956	NA
Test set	-5.050462	6.286722	5.641483	-77.00581	80.59837	1.1162375	-0.1390164	1.167219
$[](forecast_is$	$sues_files/f$	igure-latex	/unnamed-ch	unk-27-1.pdf)			

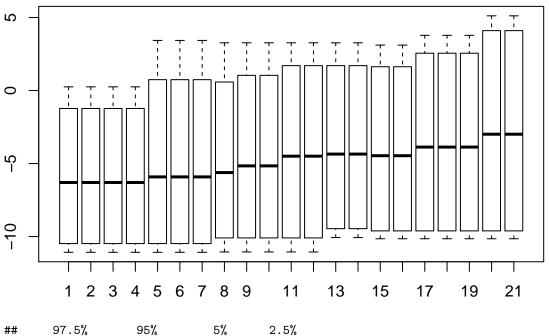


NULL

Histogram of residuals



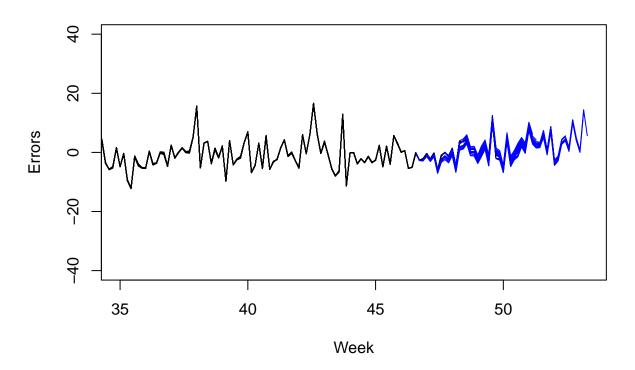
97.5% 95% 5% 2.5% ## 3.87234 3.20000 -11.00000 -11.16667



97.5% 95% 5% 2.5% ## 3.87234 3.20000 -11.00000 -11.16667

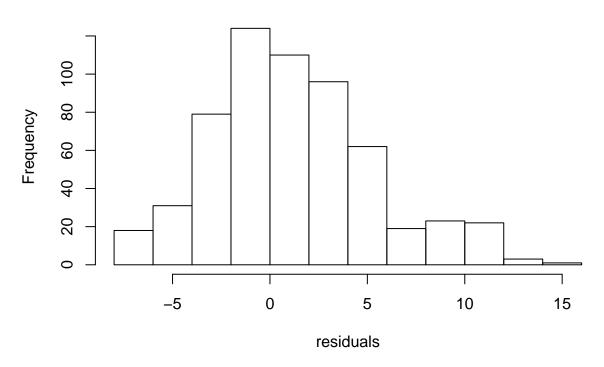
Arima (order=c(1,1,1), seasonal=c(1,1,1))

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	-0.4302686	4.853986	3.765734	-15.297233	31.46267	0.7458404	-0.0144115	NA
Test set	1.2040254	4.247021	3.290344	1.050514	35.26485	0.6525639	-0.1116419	0.7717494
$[](forecast_is$	$sues_files/fi$	gure-latex/	unnamed-chu	nk-32-1.pdf)<	!>			

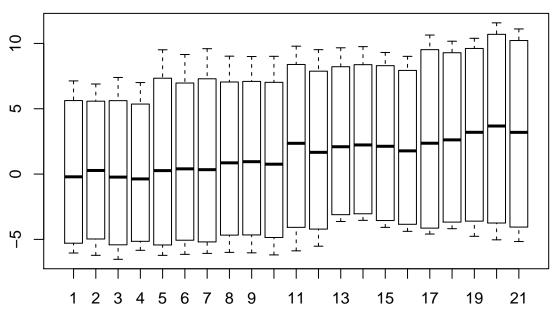


NULL

Histogram of residuals



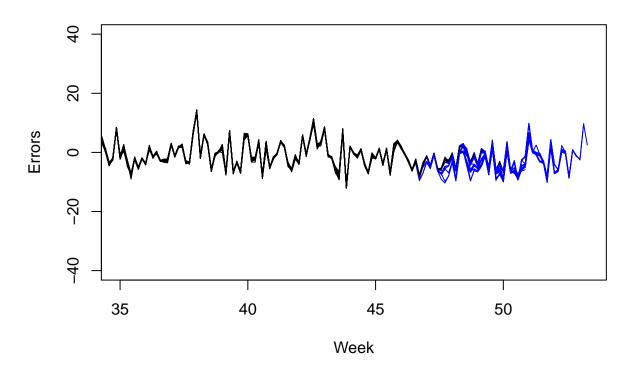
97.5% 95% 5% 2.5% ## 10.483473 9.404964 -5.003919 -6.098532



97.5% 95% 5% 2.5% ## 10.483473 9.404964 -5.003919 -6.098532

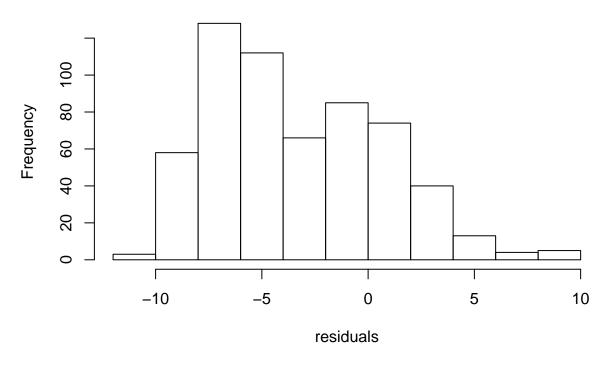
Neural Network (repeats = 20, p=1, P=1, size=7)

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	0.0051319	4.817521	3.869326	-14.78911	33.35744	0.7663914	-0.0416209	NA
Test set	-3.3259210	5.186514	4.331163	-60.67342	66.81815	0.8574047	-0.0903059	0.9670458
$[](forecast_is$	$sues_files/fi$	gure-latex/	unnamed-chu	nk-37-1.pdf				

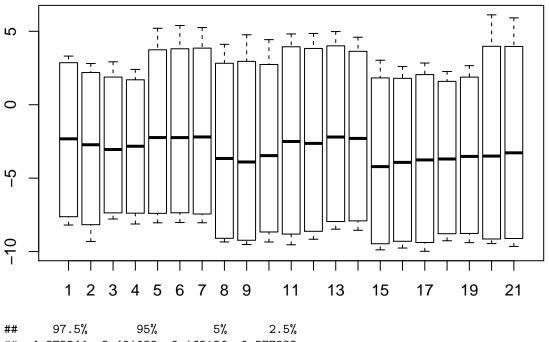


NULL

Histogram of residuals



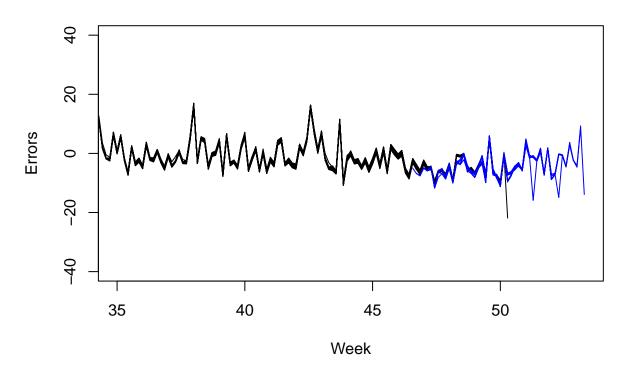
97.5% 95% 5% 2.5% ## 4.375544 3.491038 -9.163136 -9.577223



4.375544 3.491038 -9.163136 -9.577223

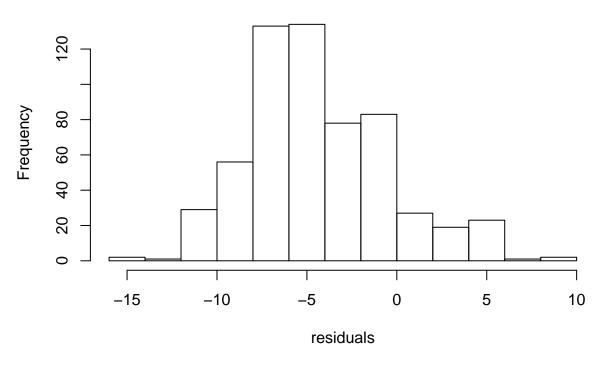
External info Numerical using regression model

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	0.000000	5.023168	3.997864	-14.91884	34.07674	0.7913971	0.1080813	NA
Test set	-4.273161	5.866936	5.100787	-68.25978	73.47325	1.0084897	-0.1745868	1.092292
$[](forecast_is$	$sues_files/f$	igure-latex	/unnamed-ch	unk-42-1.pdf)			

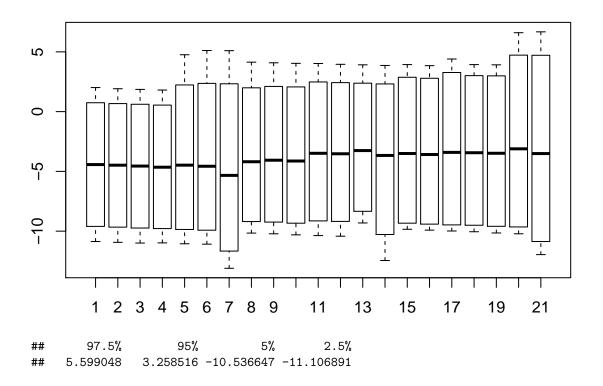


NULL

Histogram of residuals

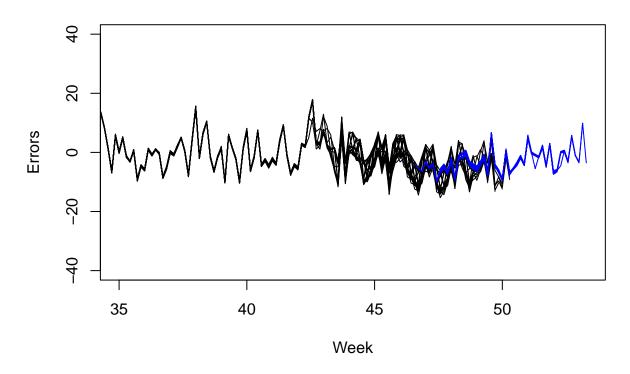


97.5% 95% 5% 2.5% ## 5.599048 3.258516 -10.536647 -11.106891



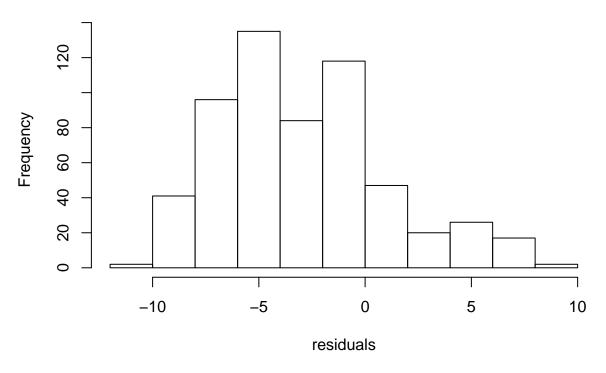
 $Ensemble \ (all.regr.ext.forecast[,i], \ all.regr.add.forecast[,i], \ all.hw.forecast[,i])$

	ME	RMSE	MAE	MPE	MAPE	MASE	ACF1	Theil's U
Training set	0.2337461	6.000984	4.687606	-17.34201	40.70483	0.9219997	0.2788951	NA
Test set	-2.9780467	4.854122	4.149092	-51.38869	58.92879	0.8151462	-0.1550379	0.9094339
$[](forecast_is$	$sues_files/fi$	gure-latex/	unnamed-chu	nk-47-1.pdf				

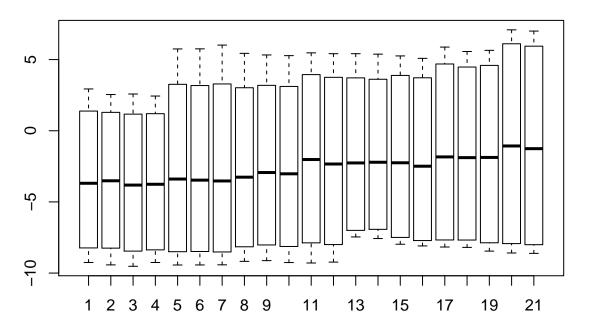


NULL

Histogram of residuals



97.5% 95% 5% 2.5% ## 6.275036 5.234703 -8.967777 -9.330981



97.5% 95% 5% 2.5% ## 6.275036 5.234703 -8.967777 -9.330981