



# Adventures in Bayesian Structural Time Series

## *Part 5: Analyzing SST Data With Regression*

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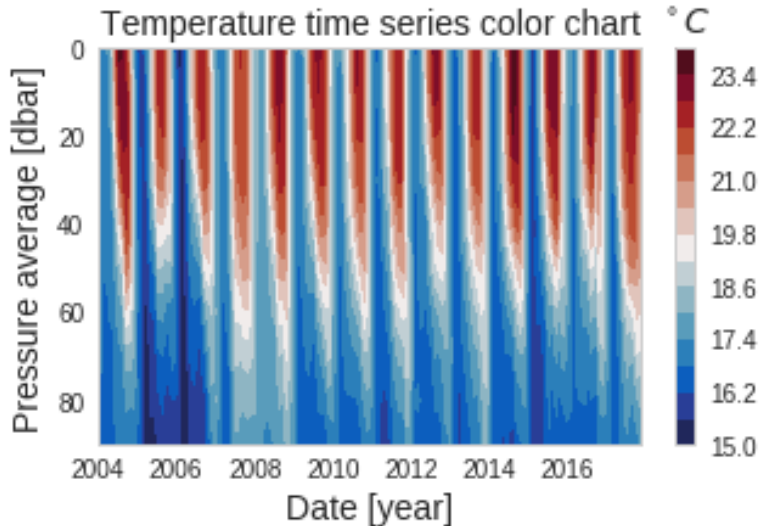




- ⊠ SST data with covariates
- ⊠ Use **bsts** to
  - ⊠ Fit structural model with regression
  - ⊠ Regression posterior
  - ⊠ Forecast
  - ⊠ Custom regresson prior



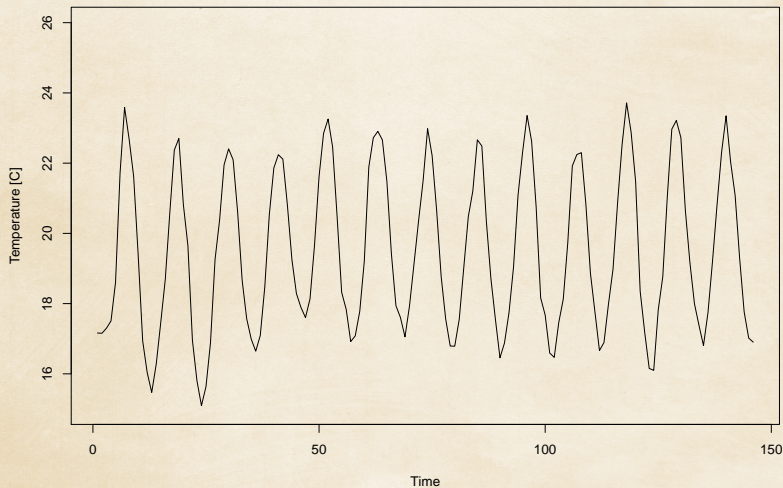
- ⊠ Sea Surface Temperature near Gibraltar
- ⊠ Aggregated monthly
- ⊠ January 2004 to April 2018
- ⊠ Covariates: depth at 10, 20, ..., 90 meters







SST of Gilbralter region





# Setup

```
library(readr)
library(bsts)

gib <- read_csv("data/gilbralter_time_series_r_2.csv",
               col_types = cols(startDate = col_skip(),
                                timeIdx = col_skip()))
names(gib) <- c('SST', '10m', '20m', '30m', '40m',
               '50m', '60m', '70m', '80m', '90m')
gib_train <- gib[1:146, ]
gib_test <- gib[147:158, ]
```



```
head(gib_test[, 1:8])
```

```
## # A tibble: 6 x 8
##   SST `10m` `20m` `30m` `40m` `50m` `60m` `70m`
##   <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
## 1  18.6  18.3  18.0  17.7  17.0  16.7  16.7  16.5
## 2  19.9  19.8  19.5  19.3  18.4  17.7  17.1  17.0
## 3  21.7  21.7  21.2  20.9  20.0  19.0  18.1  17.6
## 4  22.4  22.6  22.3  21.8  21.1  19.9  18.9  18.2
## 5  22.9  22.9  22.4  22.2  21.8  20.9  19.1  17.9
## 6  22.5  22.7  22.2  22.1  22.0  21.0  19.4  18.1
```





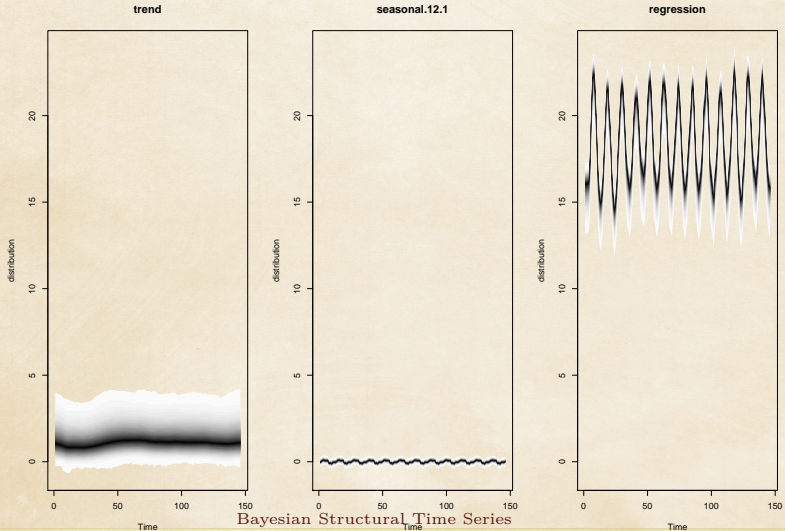
# Model Fitting

---

```
ss <- AddLocalLinearTrend(list(), gib$SST)
ss <- AddSeasonal(ss, gib$SST, nseasons = 12)
model1 <- bstts(SST ~ ., state.specification = ss,
                data = gib_train, niter = 1000, ping = 0)
```

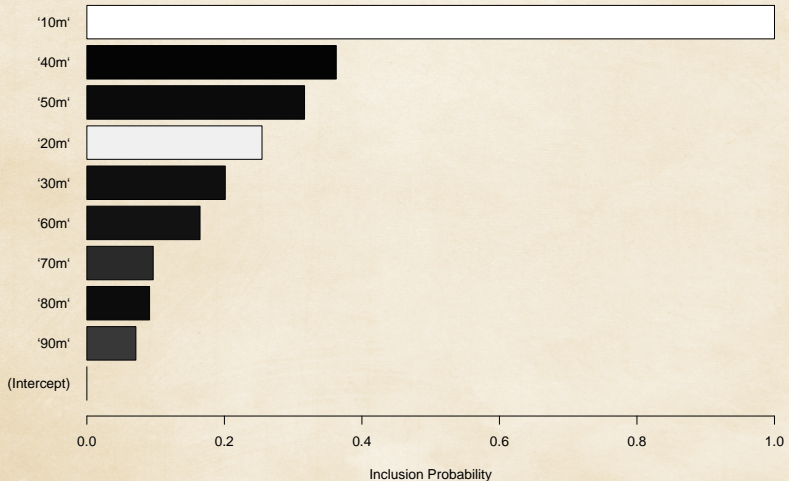


```
plot(model1, 'components')
```





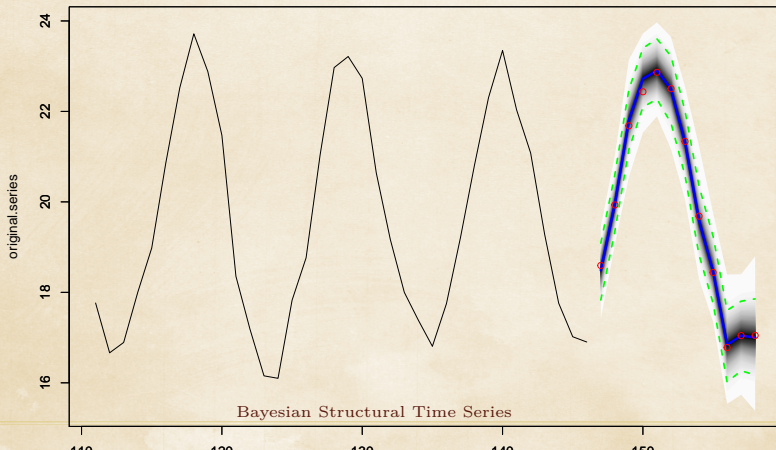
```
plot(model1, 'coefficients')
```





# Forecasting

```
model1_pred <- predict(model1, newdata = gib_test[, -1],  
                        horizon = 12)  
plot(model1_pred, plot.original = 36)  
points(147:158, gib_test$SST, col = 'red')
```



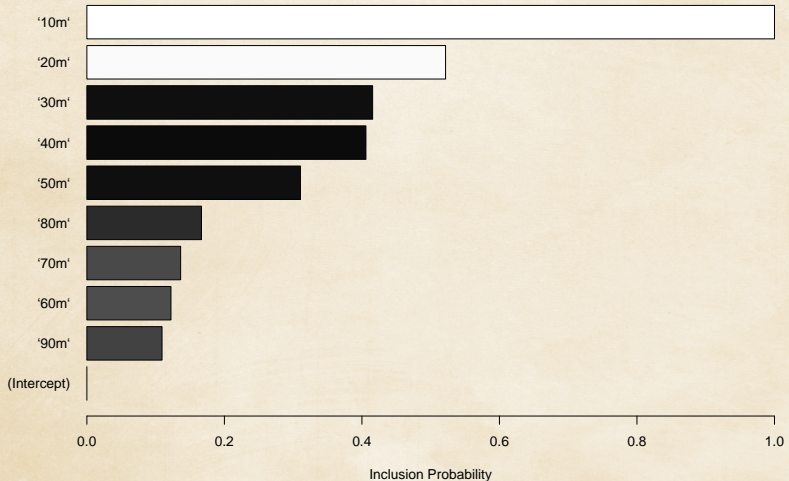


```
model2 <- bstS(SST ~., state.specification = ss,  
               data = gib_train, niter = 1000, ping = 0,  
               expected.model.size = 2)
```





```
plot(model2, 'coefficients')
```

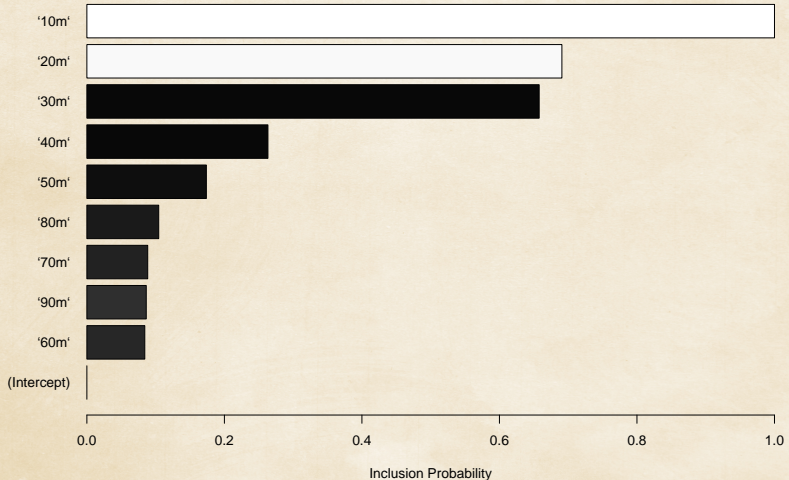




```
bp = c(.01,.5,.3,.3,.1,.1,.1,.1,.1,.1)
model3 <- bstts(SST ~., state.specification = ss,
               data = gib_train, niter = 1000, ping = 0,
               prior.inclusion.proBABILITIES = bp)
```



```
plot(model3, 'coefficients')
```





# Forecasting

```
model3_pred <- predict(model3, newdata = gib_test[, -1],  
                        horizon = 12)  
plot(model3_pred, plot.original = 36)  
points(147:158, gib_test$SST, col = 'red')
```

