SVM regression anomaly detection: Model-deviation detection using ML regression: an SVM forecaster predicts the next value from a sliding window; large prediction errors are flagged as anomalies. Implemented via DALToolbox regressors and thresholded with harutils().

Objectives: This Rmd demonstrates anomaly detection in a univariate time series using a Support Vector Machine regressor via harbinger’s hanr\_ml + ts\_svm. The model learns short-range dynamics to predict the next value; large residuals are flagged as anomalies using an adaptive threshold. The workflow is: load packages and data, visualize the series, define and fit the model, run detection, evaluate, and plot results (including residual scores).

# Install Harbinger (only once, if needed)  
#install.packages("harbinger")

# Load required packages  
library(daltoolbox)  
library(harbinger)   
library(tspredit)

# Load example datasets bundled with harbinger  
data(examples\_anomalies)

# Select a simple synthetic time series with labeled anomalies  
dataset <- examples\_anomalies$simple  
head(dataset)

## serie event  
## 1 1.0000000 FALSE  
## 2 0.9689124 FALSE  
## 3 0.8775826 FALSE  
## 4 0.7316889 FALSE  
## 5 0.5403023 FALSE  
## 6 0.3153224 FALSE

# Plot the time series  
har\_plot(harbinger(), dataset$serie)



# Define SVM-based regressor for anomaly detection (hanr\_ml + ts\_svm)  
# - input\_size=4 sets the window length used for prediction  
# - kernel="radial" uses the RBF kernel; tune if needed  
 model <- hanr\_ml(ts\_svm(ts\_norm\_gminmax(), input\_size=4, kernel = "radial"))

# Fit the model  
 model <- fit(model, dataset$serie)

# Detect anomalies (compute residuals and events)  
 detection <- detect(model, dataset$serie)

# Show only timestamps flagged as events  
 print(detection |> dplyr::filter(event==TRUE))

## idx event type  
## 1 50 TRUE anomaly

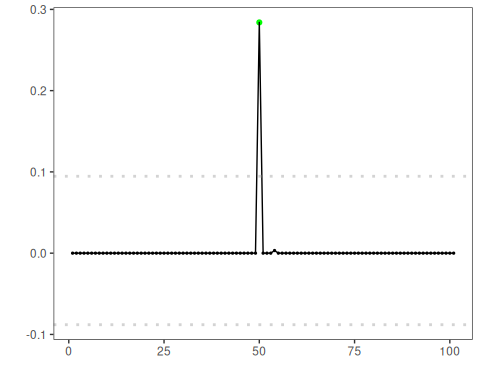
# Evaluate detections against ground-truth labels  
 evaluation <- evaluate(model, detection$event, dataset$event)  
 print(evaluation$confMatrix)

## event   
## detection TRUE FALSE  
## TRUE 1 0   
## FALSE 0 100

# Plot detections over the series  
 har\_plot(model, dataset$serie, detection, dataset$event)



# Plot residual scores and threshold  
 har\_plot(model, attr(detection, "res"), detection, dataset$event, yline = attr(detection, "threshold"))



References - Hyndman, R. J., Athanasopoulos, G. (2021). Forecasting: Principles and Practice. OTexts.