

Digital Signal Processing

Instituto Superior Técnico

Lab assignment 4 – Anomaly detection in energy time series data

Authors: Margarida Silveira and Jorge Salvador Marques.

Introduction

In this assignment, we will address anomaly detection in energy production data. The goal is to model the behaviour of normal energy production along time and subsequently to identify anomalies based on deviations from the normal behaviour. We will model the normal energy signal $x(n)$ as the output of an autoregressive model (AR). For example, in long term prediction, the model can be used to predict each sample from a previous sample, plus a prediction error or residual:

$$x(n) = ax(n - N) + r(n) \quad (1)$$

Alternatively, in short-term prediction, this model can be used to predict each sample from the previous P samples, plus a prediction error.

$$x(n) = \sum_{k=1}^P a_k x(n - k) + e(n) \quad (2)$$

The coefficients of the AR model can be obtained by minimizing the sum of squared residuals, using the least squares method.

Notes

In addition to your answers, any Matlab code you developed should also be submitted.

Experimental work

1. Prediction

R1.a) We will start with a model of long term prediction. Write a Matlab function that, given a vector with time series data $x(n)$ and a delay N , computes coefficient a for the following AR model

$$x(n) = ax(n - N) + r(n) \quad (3)$$

Write your own code, do not use any Matlab ready made functions.

R1.b) Load the train data `energy_train.mat` which contains energy production data from a residential solar panel, where samples were obtained every 15 minutes. Plot the data and notice the daily regularity, there is no energy production before sunrise or after sunset and maximum production occurs when the sun is at its peak. There are no anomalies in these data.

Use your code to compute a prediction using $N=96$, which means you are predicting the current sample from the sample of the previous day, at the same time. Plot the training data, the prediction you obtained and the corresponding residual. Comment on what you observe.

R1.c) Indicate the coefficient and the energy of the residual $r(n)$. Comment.

R1.d) Next, we will use a short term model for the prediction of the **residual** you obtained with the long term model

$$r(n) = x(n) - ax(n - N) \quad (4)$$

Write a Matlab function that, given a vector with time series data $r(n)$ and a number of previous samples P , computes the coefficients of the following AR model

$$r(n) = \sum_{k=1}^P a_k r(n - k) + e(n) \quad (5)$$

Again, write your own code, do not use any Matlab ready made functions.

- R1.e)** Use your code to compute a prediction for $r(n)$ using $P=6$. Use the predicted residual and the long term model to obtain a new prediction for $x(n)$. Plot the training data, the new prediction you obtained and the corresponding residual $e(n)$. Comment on what you observe and compare.
- R1.f)** Indicate the coefficients and the energy of residual $e(n)$. Comment.

2. Anomaly Detection

- R2.a)** Write a Matlab function that detects anomalies in time series data given a time series and a model that predicts normal behaviour (use additional parameters if necessary).
- R2.b)** Load the test data `energy_test.mat`. Use your function to detect anomalous periods of energy production in the test data using the previous two models you developed. Comment on your results and compare both models in what regards their ability for anomaly detection.
- R2.c)** When an anomaly occurs, the detection of the next period of energy production, which can be normal, is affected. How could you solve this problem?