

1. Introduction

1.1. Background

Structural changes are often seen in time series data. This observable behavior is highly appealing to statistical modelers who want to develop a model which is well explained. A method to detect changes in time series data when a time instant is unknown is called change point analysis (Basseville et al., 1993). It has a bottom line for discovering the time point where the changes in time series occur. Change point analysis can be referred to different kinds of name such as breakpoint and turning point. However, *change-point* is the commonly used term when the point in time series takes place. Another important term which will be used in this thesis is *regime switch* which refers to persistent changes in time series structure after the occurrence of change point (Weskamp and Hochstotter, 2010). Change point analysis has been studied over decades as it is a problem of interest in many applications which the characteristic of data is collected over time. The change should be flagged as soon as it occurs in order to be properly dealt with such changes in time and reduced any possible consequences (Sharkey and Killick, 2014). Here are some examples.

- Medical condition monitoring: Evaluate the sleep quality of patients based on their heart rate condition (Staudacher et al., 2005).
- Climate analysis: The temperature or climate variations is detected. This method gradually becomes important over the past few decades due to the effects of the global warming and the increases in greenhouse gas emissions (Reeves et al., 2007; Beaulieu et al., 2012).
- Quality control: Since industrial production is a continuous production process, in mass production process, if the product controlled value is not monitored and exceed the tolerant undetected, it could lead to the lost of a whole production lot (Page, 1954).
- Other applications: Identifying fraud transaction (Bolton and Hand, 2002), detecting anomalies in the market price (Gu et al., 2013) and detecting signal processing (Basseville et al., 1993) in streaming data as well.

In this study, change point analysis will be used to identify changes in performance of Ericsson’s software products. Many test cases have been executed for testing software packages in a simulation environment. Before launching the software products to its customers, the company needs to test and determine how each software

package performs. The performance of these software packages is evaluated by considering on CPU utilization (percentages of CPU's cycle spent on each process), memory usage and latency.

Recently, a method called hidden Markov model or Markov switching model is widely used for discovering a change point in time series. Both terms are acceptable but being called differently regarding to the different fields of study. Markov switching model uses a concept of Markov chain to model an underlying segmentation as different states and then specify a distinct change of location. Hence, the method is able to identify a switch in time series when change point occurs (Luong et al., 2012). This method is generally used in almost all current systems in speech recognition (Rabiner, 1989) and found to be important in climatology such as describing the state in the wind speed time series (Ailliot and Monbet, 2012) and in biology (Stanke and Waack, 2003) where the prediction of gene is being made. Markov switching model has been extensively applied in the field of economics and finance and has a large literature. For example, business cycle can be seen as hidden states with seasonal changes. The growth rate of gross domestic product (GDP) is modeled as a switching process to uncover business cycle phases i.e., expansion and recession. The fitting model can also be used to understand the process where there is a transition between the economic state and the duration of each period (Hamilton, 1989). In finance data, time series of returns is modeled in order to investigate stock market situation i.e., bull or bear market (Kim et al., 1998).

Markov switching model is one of the most well-known non linear time series models. This model can be applied to various time series data with dynamic behavior. The structural changes or regime shifts in data mean that constant parameter in time series model might be insufficient to capture these behaviors and describe their evolution. Markov switching model takes the presence of shifting regime in time series into account and models multiple structures that can explain these characteristics in different states at different time. The shift between states or regimes comes from the switching mechanism which is assumed to follow an unobserved Markov chain. Thus, the model is able to capture more complex dynamic patterns and also identify the change of locations and regime switch in time series.

Each software package in the system is viewed as a time point in time series and the performance of each software package is treated as an observed value. In this study, the observed value is not completely independent of each other i.e., the performance of the current software package depends on the performance from the prior version of the software package. Therefore, additional dependencies, with the first order autoregression, is taken into consideration when modeling the Markov switching model and becomes Markov switching autoregressive model.

There are two approaches, a parametric and a non-parametric analysis, for detecting the change point in the time series. The parametric analysis benefits from assuming some knowledge of data distribution and integrating it to the detection scheme. On the other hand, the non-parametric analysis is more flexible that there is no

assumption made about the distribution. It can, therefore, apply to a wider range of application and capture various kinds of changes (Sharkey and Killick, 2014). In this study, the non-parametric analysis using hierarchical estimation techniques is also implemented for identifying the change of location in the time series.

1.2. Objective

The core idea of this thesis is to reduce workload of manual inspection when the performance analysis of an update software package is required. With an increase amount of generated data from numerous test cases, the inspection becomes tedious and inefficient to be done manually. The main objective of this thesis is to implement machine learning, an algorithm that has an ability to learn from data, in order to analyze the performance of the software package. The algorithm will help indicate whether the performance of the software package is in a degrading, improving or steady state. There is also worth to mention that the performance of a particular software package can vary on different test environment. The implemented algorithm should also be able to detect when the test environment is altered. This thesis only focuses on a CPU utilization which is one of the three essential factors for evaluating the performance of the upgrade software package.

To summarize, this thesis aims to:

- Detect the state of the CPU utilization (degrading, improving or steady state)
- Detect whether there is any change in the test environment that effects the CPU utilization

The thesis is structured as follow: Chapter 2 provides details and descriptions of datasets used in the analysis. Chapter 3 presents methodology. Results from the analysis along with tables and plots are shown in Chapter 4. Chapter 5 discusses the outcomes and the obtained results. Lastly, Chapter 6 contains conclusion.