

Average Run Length Computations of Autoregressive and Moving Average Process using the Extended EWMA Procedure

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Abstract: - In the past, the control chart served as a statistical tool for detecting process changes. The Exponentially Weighted Moving Average (EWMA) control chart is highly effective for detecting small changes. This research introduces the Extended Exponentially Weighted Moving Average (Extended EWMA) control chart for the Autoregressive and Moving average process with order $p = 1$ and $q = 1$ (ARMA(1,1)) The Extended EWMA control chart incorporates two smoothing parameters (λ_1 and λ_2) derived from the EWMA control chart. A comparative analysis of the performance of the EWMA control chart. The Average Run Length (ARL) value as determined by the explicit formulas in this research, serves as a metric for evaluating the performance of the Extended EWMA control chart and the EWMA control chart. The Numerical Integral Equation (NIE) method is used to verify the accuracy of the ARL for the explicit formulas of the two control charts which has not been before discovered. The effectiveness of control charts can also be evaluated by analyzing SDRL, ARL, MRL, RMI, AEQL, and PCI values as metrics for various design parameter values. After analyzing the results of the ARL and all five performance meters, it was determined that the Extended EWMA control chart is better than the EWMA control chart at all shift sizes of process changes. Finally, the assessment of the ARMA process is being conducted to evaluate the ARL using a dataset on PM2.5 dust levels in Bangkok, Thailand during January and February of 2024.

Key-Words: - Average Run Length, Autoregressive and Moving Average Process, Extended Exponentially Weighted Moving Average control chart, Explicit Formula, Numerical Integral Equation, ARL, NIE method.

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1 Introduction

The statistically important tool is the control chart. [1], invented the control chart, which detects the alteration of the control diagram, which is commonly used in the manufacturing industry. The Shewhart control chart is very efficient with small change detection as well. [2], introduced an EWMA control chart for better detection of small changes. The Cumulative Sum (CUSUM) control chart, developed by [3], is extensively utilized in statistical control charting. In 2017, [4], developed and presented a more efficient, Modified EWMA control chart than the EWMA control chart in the detection of minor changes. [5], presented The Extended EWMA control chart as a powerful chart designed to detect minor changes in the process being examined. The effectiveness of control charts can be assessed by utilizing the ARL, [6]. The ARL

is divided into two values, [7], The ARL_0 is the number of expected observations required before a process is under control, and The ARL_1 refers to the amount of observations expected from an uncontrollable process and should be reduced. In 1990, [8], are presented a quantitative analysis comparing the EWMA control chart and the CUSUM control chart. [9], examined the design of the optimal EWMA control chart and compared it to the CUSUM control chart. [10], are presented, and the autocorrelation data will be linked to a statistical model. The ARMA process is a frequently employed model in real-world data analysis. [11], have introduced datasets with exponential white noise to the control chart. [12], shows the performance of the CUSUM control chart for autocorrelated seasonal consistency of trends using the Midpoint Rule method with exponential white noise. [13], presented an

approximating of the ARL of changes in the mean of a Seasonal Time Series Model with exponential white noise running on a CUSUM control chart. [14], presented the explicit formulas and NIE of ARL when observations were seasonal autoregressive models with an exogenous variable SARX(P,r)_L with exponential white noise based on the CUSUM control chart. [15], proposed an explicit formula for the ARL using the Fredholm integral equation method in the EWMA control chart on the MAX(q,r) process. [16], are improving the CUSUM control chart for monitoring a change in processes based on a seasonal autoregressive model with one exogenous variable. [17], introduced a Modified EWMA control chart that is derived from its particular examples. [18], developed the explicit formula for the ARL on a Modified EWMA control chart for the AR(1) process. Subsequently, [19], presents analytical explicit formulas of the ARL of Homogenously Weighted Moving Average control chart (HEWMA) based on a MAX process.

However, the derivation of the explicit formula for the ARL on the Extended EWMA control chart for the Autoregressive and Moving Average process with parameters $p = 1$ and $q = 1$ (ARMA(1,1)), when there are two smoothing parameters (λ_1 and λ_2) for the ARMA(1,1) process has not been reported previously. The objective of this research is to determine the explicit formula for the ARL on the Extended EWMA control chart for the ARMA(1,1). The explicit formula for the ARL was compared with the NIE method. This research differs from that of other researchers, using five additional statistical performance measurements, Standard of Deviation Run Length (SDRL), Median Run Length (MRL), [20], Related Mean Index (RMI), [21], Average Extra Quadratic Loss (AEQL), [22] and Performance Comparison Index (PCI). The comparison of performance between the Extended EWMA control chart and the EWMA control chart using a dataset on PM2.5 dust levels in Bangkok, Thailand during January and February of 2024.

2 Materials and Methods

2.1 The Exponential Weighted Moving Average Control Chart

This research describes the properties of the EWMA control chart for ARMA(p,q) processes. The EWMA control chart is defined by a recursive equation, [2].

$$Z_t = (1 - \lambda)Z_{t-1} + \lambda W_t, t = 1, 2, 3, \dots \quad (1)$$

where W_t is a sequence of an ARMA(p,q) processes with exponential white noise and $0 < \lambda \leq 1$, W_0 is the initial value of the EWMA statistics, $Z_0 = u$. The control limits of the EWMA control chart consist of the upper control limit (UCL) and the lower control limit (LCL) are:

$$\begin{aligned} UCL &= \mu_0 + \gamma\sigma\sqrt{\frac{\lambda}{2-\lambda}} \\ LCL &= \mu_0 - \gamma\sigma\sqrt{\frac{\lambda}{2-\lambda}} \end{aligned} \quad (2)$$

where μ_0 is process mean of ARMA(p,q) process, σ is the process standard deviation parameter, γ is a suitable control limit width. Let b be an UCL on Z_t . The stopping time (τ_b) of the Extended EWMA control chart is defined as

$$\tau_b = \inf\{t > 0; Z_t \geq b\}$$

3 The Exact Solutions of the ARL on the Extended EWMA Control Chart

3.1 The Explicit Formula of the ARL on the Extended EWMA Control Chart for ARMA(p,q) process

The ARMA(p,q) process are defined by the following recursion:

$$\begin{aligned} W_t &= \phi_0 + \phi_1 W_{t-1} + \phi_2 W_{t-2} + \dots + \phi_p W_{t-p} \\ &+ \theta_1 \mathcal{G}_{t-1} - \theta_2 \mathcal{G}_{t-2} - \dots - \theta_q \mathcal{G}_{t-q} \end{aligned} \quad (3)$$

where W_t is a sequence of the ARMA(p,q) processes with exponential white noise, ϕ_i is autoregressive parameter, $W_0 = u$ is the initial value, where $u = [0, b]$ and b is UCL of the Extended EWMA control chart

[6], presented The Extended EWMA control chart. The performance control chart is highly efficient in detecting small changes in the monitored process. The Extended EWMA statistic can be derived as:

$$\omega_t = (1 - \lambda_1 + \lambda_2)\omega_{t-1} + \lambda_1 W_t - \lambda_2 W_{t-1}, t = 1, 2, 3, \dots \quad (4)$$

where W_t is a process with mean, λ_1 and λ_2 are exponential smoothing parameters with

$0 \leq \lambda_2 < \lambda_1 < 1$ and ω_0 is the initial value of the Extended EWMA statistics, $\omega_0 = u$ and $\mathcal{G}_0 = v$. The UCL and the LCL are:

$$UCL = \mu_0 + \gamma\sigma \sqrt{\frac{\lambda_1^2 + \lambda_2^2 - 2\lambda_1\lambda_2(1 - \lambda_1 + \lambda_2)}{2(\lambda_1 - \lambda_2) - (\lambda_1 - \lambda_2)^2}}$$

$$LCL = \mu_0 - \gamma\sigma \sqrt{\frac{\lambda_1^2 + \lambda_2^2 - 2\lambda_1\lambda_2(1 - \lambda_1 + \lambda_2)}{2(\lambda_1 - \lambda_2) - (\lambda_1 - \lambda_2)^2}}, \quad (5)$$

where μ_0 is the process mean of moving average process, σ is the process standard deviation parameter and γ is a suitable control limit width, and $W_t = \phi_0 + \phi_1 W_{t-1} + \phi_2 W_{t-2} + \dots + \phi_p W_{t-p} + \mathcal{G}_t - \theta_1 \mathcal{G}_{t-1} - \theta_2 \mathcal{G}_{t-2} - \dots - \theta_q \mathcal{G}_{t-q}$ (6)

Hence, the formulation of the Extended EWMA control chart for the ARMA(p,q) process is as.
 $\omega_t = (1 - \lambda_1 + \lambda_2)\omega_{t-1} + \lambda_1(\phi_0 + \phi_1 W_{t-1} + \phi_2 W_{t-2} + \dots + \phi_p W_{t-p} + \mathcal{G}_t - \theta_1 \mathcal{G}_{t-1} - \theta_2 \mathcal{G}_{t-2} - \dots - \theta_q \mathcal{G}_{t-q}) - \lambda_2 W_{t-1}, t = 1, 2, 3, \dots$

When $t = 1$
 $\omega_1 = (1 - \lambda_1 + \lambda_2)\omega_0 + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_{-1} + \dots + \phi_p W_{1-p} - \theta_1 \mathcal{G}_0 - \theta_2 \mathcal{G}_{-1} - \dots - \theta_q \mathcal{G}_{1-q}) - \lambda_2 W_0 + \mathcal{G}_1 \lambda_1, t = 1, 2, 3, \dots$

Let $\Delta = \phi_0 + \phi_1 W_0 + \phi_2 W_{-1} + \dots + \phi_p W_{1-p} - \theta_1 \mathcal{G}_0 - \theta_2 \mathcal{G}_{-1} - \dots - \theta_q \mathcal{G}_{1-q}$

So
 $\omega_t = (1 - \lambda_1 + \lambda_2)\omega_0 + \lambda_1 \Delta - \lambda_2 u + \mathcal{G}_1 \lambda_1, t = 1, 2, 3, \dots$ (7)

Let's examine the in-control process, where the UCL = b and the LCL = 0 .

$$0 < \omega_1 < b$$

$$0 < (1 - \lambda_1 + \lambda_2)\omega_0 + \lambda_1 \Delta - \lambda_2 u + \mathcal{G}_1 \lambda_1 < b$$

$$0 < \mathcal{G}_1 \lambda_1 < b - (1 - \lambda_1 + \lambda_2)\omega_0 - \lambda_1 \Delta + \lambda_2 u$$

$$0 < \mathcal{G}_1 < \frac{b - (1 - \lambda_1 + \lambda_2)\omega_0 - \lambda_1 \Delta + \lambda_2 u}{\lambda_1}$$

$$0 < \mathcal{G}_1 < \frac{b - (1 - \lambda_1 + \lambda_2)u - \lambda_1 \Delta + \lambda_2 u}{\lambda_1}$$

The function $\rho(u)$ can be obtained by Fredholm integral equation of the second kind as follows;

$$\rho(u) = 1 + \int_0^b \rho(\omega_1) f(\mathcal{G}) d\mathcal{G} \quad (8)$$

$$\rho(u) = 1 + \int_0^b \rho((1 - \lambda_1 + \lambda_2)\omega_0 + \lambda_1 \Delta - \lambda_2 u + \mathcal{G}_1) f(\mathcal{G}) d\mathcal{G}$$

Therefore, the function $\rho(u)$ is obtained as follows:

$$\rho(u) = 1 + \frac{1}{\lambda_1} \int_0^b \rho(k) f\left(\frac{b - (1 - \lambda_1 + \lambda_2)u - \lambda_1 \Delta + \lambda_2 u + \mathcal{G}_1}{\lambda_1}\right) dk$$

Given that $\varepsilon_t \sim \text{Exp}(\alpha)$ is determined,

$$f(k) = \frac{1}{\alpha} e^{-\frac{k}{\alpha}}$$

$$\rho(u) = 1 + \frac{e^{-\frac{b - (1 - \lambda_1 + \lambda_2)u - \lambda_1 \Delta + \lambda_2 u + \mathcal{G}_1}{\lambda_1 \alpha}}}{\lambda_1 \alpha} \int_a^b \rho(k) e^{-\frac{k}{\lambda_1 \alpha}} dk$$

Setting $\Gamma(u) = e^{-\frac{b - (1 - \lambda_1 + \lambda_2)u - \lambda_1 \Delta + \lambda_2 u + \mathcal{G}_1}{\lambda_1 \alpha}}$ and $\Pi = \int_0^b \rho(k) e^{-\frac{k}{\lambda_1 \alpha}} dk$

$$\text{Thus } \rho(u) = 1 + \frac{\Gamma(u)}{\lambda_1 \alpha} \Pi \quad (9)$$

Consider Π and take run $\rho(k)$

$$\Pi = \frac{-\lambda_1 \alpha (e^{-\frac{b}{\lambda_1 \alpha}} - 1)}{1 + \frac{1}{(\lambda_1 - \lambda_2)} e^{-\frac{b - (1 - \lambda_1 + \lambda_2)u - \lambda_1 \Delta + \lambda_2 u + \mathcal{G}_1}{\lambda_1 \alpha}} (e^{-\frac{-(\lambda_1 - \lambda_2)b}{\lambda_1 \alpha}} - 1)} \quad (10)$$

By substituting the value of an equal to α_1 into $\rho(u)$, we can get the explicit formula of the ARL_1 on the Extended EWMA control chart as follows:

$$ARL_1 = 1 - \frac{(\lambda_1 - \lambda_2) e^{-\frac{(1 - \lambda_1 + \lambda_2)u}{\lambda_1 \alpha_1}} (e^{-\frac{b}{\lambda_1 \alpha_1}} - 1)}{(\lambda_1 - \lambda_2) e^{-\frac{(1 - \lambda_1 + \lambda_2)u + \lambda_1 \Delta - \lambda_2 u + \mathcal{G}_1}{\lambda_1 \alpha_1}} + (e^{-\frac{-(\lambda_1 - \lambda_2)b}{\lambda_1 \alpha_1}} - 1)}$$

$$ARL_1 = 1 - \frac{(\lambda_1 - \lambda_2) e^{-\frac{(1 - \lambda_1 + \lambda_2)u}{\lambda_1 \alpha_1}} (e^{-\frac{b}{\lambda_1 \alpha_1}} - 1)}{(\lambda_1 - \lambda_2) e^{-\frac{(1 - \lambda_1 + \lambda_2)u + \lambda_1 \Delta - \lambda_2 u + \mathcal{G}_1}{\lambda_1 \alpha_1}} + (e^{-\frac{-(\lambda_1 - \lambda_2)b}{\lambda_1 \alpha_1}} - 1)}$$

$$ARL_1 = 1 - \frac{(\lambda_1 - \lambda_2) e^{-\frac{(1 - \lambda_1 + \lambda_2)u}{\lambda_1 \alpha_1}} (e^{-\frac{b}{\lambda_1 \alpha_1}} - 1)}{(\lambda_1 - \lambda_2) e^{-\frac{(1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_{-1} + \dots + \phi_p W_{1-p} - \theta_1 \mathcal{G}_0 - \theta_2 \mathcal{G}_{-1} - \dots - \theta_q \mathcal{G}_{1-q}) - \lambda_2 u + \mathcal{G}_1}{\lambda_1 \alpha_1}} + (e^{-\frac{-(\lambda_1 - \lambda_2)b}{\lambda_1 \alpha_1}} - 1)}$$

From Equation (11), the ARL finished explicit formula of the Extended EWMA control chart for the ARMA(p,q) process is to be compared to the EWMA control chart.

3.2 The Numerical Integral Equation of the ARL on the Extended EWMA Control Chart of ARMA(p,q) process.

Let $\psi(u)$ denote the estimated value of the ARL determined from the m linear equation systems using the composite midpoint quadrature rule.

The evaluation of the ARL approaching NIE on the Extended EWMA control chart is carried out in the following manner:

$$\psi(u) = \int_0^b \rho(\omega_i) f(k) dk \approx \sum_{j=1}^s c_j f(x_j) \quad (12)$$

The system of s linear equations is represented as

$$L_{s \times 1} = 1_{s \times 1} + R_{s \times s} L_{s \times 1} \text{ or } L_{s \times 1} = (I_s - R_{s \times s})^{-1} 1_{s \times 1}$$

$$L_{s \times 1} = [L_{NIE}(x_1), L_{NIE}(x_2), \dots, L_{NIE}(x_s)]^T,$$

$$I_s = \text{diag}(1, 1, \dots, 1) \text{ and } 1_{s \times 1} = [1, 1, \dots, 1]^T.$$

Let $R_{s \times s}$ be a matrix. The s to s^{th} element matrix R is defined as follows:

$$[R_{ij}] \approx \frac{1}{\lambda_1} c_j f\left(\frac{k_j - (1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_1 + \dots + \phi_p W_{1-p} - \theta_1 g_0 - \theta_2 g_1 - \dots - \theta_q g_{1-q}) - \lambda_2 u + g_1}{\lambda_1}\right)$$

The answer to the NIE can be succinctly expressed as.

$$\psi(u) = 1 + \frac{1}{\lambda_1} \sum_{j=1}^s c_j f\left(\frac{k_j - (1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_1 + \dots + \phi_p W_{1-p} - \theta_1 g_0 - \theta_2 g_1 - \dots - \theta_q g_{1-q}) - \lambda_2 u + g_1}{\lambda_1}\right) \quad (13)$$

where k_j is a set of the division point on the interval $[0, b]$ as $k_j = \left(j - \frac{1}{2}\right) c_j, j = 1, 2, \dots, s$. c_j is a weight of composite midpoint formula $c_j = \frac{b}{s}$.

From Equation (13), the NIE method is a comparative criterion to the explicit formulas that the explicit formula is accurate. As a result, both ARL values are similar.

4 Existence and Uniqueness of ARL

The answer is obtained from the explicit formula using the ARL of the existence of the NIE, as proven by Banach's fixed-point theorem, [23]. In this study, let T denote an operation on the set of all continuous functions that are defined.

$$T(\rho(u)) = 1 + \frac{1}{\lambda_1} \sum_{j=1}^s c_j \quad (14)$$

$$f\left(\frac{k_j - (1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_1 + \dots + \phi_p W_{1-p} - \theta_1 g_0 - \theta_2 g_1 - \dots - \theta_q g_{1-q}) - \lambda_2 u + g_1}{\lambda_1}\right) dk$$

According to Banach's fixed-point theorem, if an operator T is guaranteed to satisfy the criterion of being a contraction, $T(\rho(u)) = \rho(u)$ has a one solution, as previously stated. For Equation (14) to have a solution that is both present and unique, the Banach fixed-point theorem can be utilized. The Banach fixed-point theorem, also referred to as the contraction mapping theorem, was initially introduced concretely in Banach's.

Typically, it is employed to determine the existence of a solution to an integral problem. Following this, [24], have extensively utilized this tool to address numerous problems related to the presence of solutions in diverse mathematical domains, due to its straightforwardness and practicality. The following information provides the specific details.

Theorem 1 Banach's Fixed-point Theorem :

Assume that $D: X \rightarrow X$ is a contraction mapping with contraction constant $0 \leq S < 1$, such that $\|D(\rho_1) - D(\rho_2)\| \leq S \|\rho_1 - \rho_2\| \forall \rho_1, \rho_2 \in X$, meets this criterion. [25], have established the existence of a single unique $a. \rho(\cdot) \in X$ such that $D(\rho(u)) = \rho(u)$ has a unique fixed point in X .

Proof: To demonstrate the value of T , as determined by the equation $T(\rho(u))$ is a contraction mapping for $\rho_1, \rho_2 \in \Omega[0, b]$. that $\|D(\rho_1) - D(\rho_2)\| \leq S \|\rho_1 - \rho_2\|, \forall \rho_1, \rho_2 \in \Omega[0, b]$. with $0 \leq S < 1$ under the norm $\|\rho_\infty\| \leq \sup_{u \in [0, b]} \|\rho(u)\|$ From

$$\begin{aligned} & \rho(u) \text{ and } D(\rho(u)). \\ & \|D(\rho_1) - D(\rho_2)\|_\infty \\ & = \sup_{u \in [0, b]} \left| \frac{1}{\lambda_1 \alpha} e^{\frac{k_j - (1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_1 + \dots + \phi_p W_{1-p} - \theta_1 g_0 - \theta_2 g_1 - \dots - \theta_q g_{1-q}) - \lambda_2 u + g_1}{\lambda_1}} \int_0^b (\rho_1(k) - \rho_2(k)) e^{-\frac{b}{\lambda_1 \alpha} k} dk \right| \\ & \leq \sup_{u \in [0, b]} \left| \frac{1}{\lambda_1 \alpha} e^{\frac{k_j - (1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_1 + \dots + \phi_p W_{1-p} - \theta_1 g_0 - \theta_2 g_1 - \dots - \theta_q g_{1-q}) - \lambda_2 u + g_1}{\lambda_1}} \int_0^b (\rho_1(k) - \rho_2(k)) e^{-\frac{b}{\lambda_1 \alpha} k} dk \right| \end{aligned}$$

$$= \|\rho_1 - \rho_2\|_\infty \sup_{u \in [0, b]} \left| e^{\frac{k_j - (1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_1 + \dots + \phi_p W_{1-p} - \theta_1 \delta_0 - \theta_2 \delta_1 - \dots - \theta_q \delta_{1-q}) - \lambda_2 u + \delta_1}{\lambda_1}} \right|$$

$$\left| 1 - \left[e^{-\frac{b}{\lambda_1 \alpha}} - 1 \right] \right| \leq S \|\rho_1 - \rho_2\|_\infty$$

where

$$S = \sup_{u \in [0, b]} \left| e^{\frac{k_j - (1 - \lambda_1 + \lambda_2)u + \lambda_1(\phi_0 + \phi_1 W_0 + \phi_2 W_1 + \dots + \phi_p W_{1-p} - \theta_1 \delta_0 - \theta_2 \delta_1 - \dots - \theta_q \delta_{1-q}) - \lambda_2 u + \delta_1}{\lambda_1}} \right|$$

$$\left| 1 - \left[e^{-\frac{b}{\lambda_1 \alpha}} - 1 \right] \right|$$

, $0 \leq S < 1$, The uniqueness of the solution is ensured by Banach's fixed-point theorem.

5 Numerical Results

In this research, we evaluate the ARL_0 and ARL_1 by employing explicit formulas and the NIE for an ARMA(p,q) process on the Extended EWMA control chart. In addition, performance indicators such as the SDRL and MRL are used to assess the effectiveness of control charts. The computation for SDRL and MRL for the in-control process is as follows.

$$ARL_0 = \frac{1}{\beta_0}, SDRL_0 = \sqrt{\frac{1 - \beta_0}{(\beta_0)^2}}, MRL_0 = \frac{\log(\frac{1}{2})}{\log(1 - \beta_0)} \tag{15}$$

where β_0 represents an error of type I. This research analysis determined that $ARL_0 = 370$. The value of the ARL_0 can be computed using Equation (15) as $SDRL_0$ and MRL_0 with an approximate value. Conversely, ARL_1 , $SDRL_1$ and MRL_1 are calculated using Equation (16).

$$ARL_1 = \frac{1}{\beta_1}, SDRL_1 = \sqrt{\frac{1 - \beta_1}{(\beta_1)^2}}, MRL_1 = \frac{\log(\frac{1}{2})}{\log(1 - \beta_1)} \tag{16}$$

where β_1 represents an error of type II

The minimum values of the ARL_1 , $SDRL_1$ and MRL_1 indicate a higher ability to promptly detect variations in the process mean. To conduct a comparison analysis, we will examine the Extended EWMA control charts and the EWMA control charts for the ARMA(p,q) process.

RMI is employed to assess the efficacy of the Extended EWMA control chart. RMI can be computed.

$$RMI = \frac{1}{n} \sum_{i=1}^n \left[\frac{ARL_i(MAX) - ARL_i(MIN)}{ARL_i(MIN)} \right], \tag{17}$$

where $ARL_i(MAX)$ is the ARL_i of row i on the control chart under examination. $ARL_i(MIN)$ is the minimum of the ARL_i for row i . A control chart is deemed more effective when it has a lower RMI.

Furthermore, performance measurements can be utilized to evaluate the effectiveness of control charts across a range of modifications. In addition, the AEQL may pertain to costs that have been accrued as a result of an unmanageable situation. This comparison may entail the utilization of various control chart kinds to determine the most efficient strategy for a specific procedure. These include the study model, the research data set, the appropriate parameter value, the control chart that the research is introduced, as well as the application of the actual data to make the chart of this research result as desired.

The AEQL can be determined by using the following formula.

$$AEQL = \frac{1}{\Phi} \sum_{shift_i = shift_{min}}^{shift_{max}} (shift_i^2 \times ARL(shift_i)) \tag{18}$$

where $shift$ refers to a distinct change in the process. Φ denotes the aggregate of number of divisions from $shift_{min}(\delta_{min})$ to $shift_{max}(\delta_{max})$. In this research, $\Phi = 10$, $\delta_{min} = 0.01$ and $\delta_{max} = 3.00$. The most effective control chart is the one with the minimum AEQL value.

Additionally, the examination of control chart performance can be carried out by utilizing the performance evaluation criteria of the PCI. The determination of the PCI value entails comparing the AEQL of a certain control chart to the AEQL of the control chart with the minimum value. This helps identify the control chart that has the highest level of efficiency. The PCI can be computed:

$$PCI = \frac{AEQL}{AEQL_{min}} \tag{19}$$

The ARL was approximated by NIE using the composite midpoint rule on the Extended EWMA control chart for the ARMA(p,q) process with a sample size of 1,000 nodes. When $ARL_0 = 370$, $\phi_0 = 0.5$, $\phi_1 = 0.1$, $\theta_1 = -0.1$ and 0.1 , $\theta_2 = -0.2$

and 0.2, $\theta_3 = 0.3$, $\lambda_1 = 0.05, 0.10$, $\lambda_2 = 0.01$, and $\delta = 0.01, 0.03, 0.05, 0.10, 0.30, 0.50, 1.00, 2.00$ and 3.00 , The in-control process is $\alpha_0 = 1$. The results indicate that the ARL of the explicit formulas and the NIE are very similar for the ARMA(1,1), ARMA(1,2), and ARMA(1,3) processes in Table 1, Table 2 and Table 3 in Appendix. The results of the ARL of the Extended EWMA control chart using an explicit formula are shown in Table 4, Table 5 and Table 6 in Appendix. These Tables compare the performance of the Extended EWMA control chart against the EWMA control chart for ARMA(1,1), ARMA(1,2), and ARMA(1,3) processes. Based on the results, the Extended EWMA control chart outperformed the EWMA control chart in terms of the ARL, SDRL, and MRL for λ_1 values of 0.05 and 0.10. In addition, the results suggest that the Extended EWMA control chart with $\lambda_1 = 0.10$ exhibits the minimum values for RMI, AEQL, and PCI.

Therefore, it can be inferred that the Extended EWMA control chart exhibits greater performance when compared to the EWMA control chart. Moreover, the RMI, AEQL, and PCI derived from each control chart are utilized to assess the effectiveness of the aforementioned control charts. The Extended EWM control chart exhibited superior performance. Based on the minimal values for RMI, AEQL, and PCI, all of them were equal to one.

6 Application to Real-world Data

In this study, the explicit formulas of the ARL on the Extended EWMA control chart for the ARMA(1,1) process were applied to the dataset on PM2.5 dust levels in Bangkok, Thailand during January and February of 2024 and generated a forecasting process. The ARL was calculated using explicit formulas on the Extended EWMA control chart with $ARL_0 = 370$ for $\lambda_1 = 0.05, 0.10$ and $\lambda_2 = 0.01$, shift (δ) equal to 0.01, 0.03, 0.05, 0.10, 0.30, 0.50, 1.00, 2.00, 3.00 and sample size = 1,000 nodes. The performance of the control chart was evaluated by comparing it to the EWMA control chart using a dataset on PM2.5 dust levels in Bangkok, Thailand during January and February of 2024. The coefficient parameters estimated for the ARMA(1,1) process are determined using maximum likelihood estimation: $\alpha_0 = 27.3401$, $\phi_1 = 0.947$, $\theta_1 = 0.616$. The ARMA(1,1) process can

be defined by utilizing the parameter of this forecasting process.

$$\hat{W}_t = 0.947W_{t-1} - 0.616g_{t-1}$$

Using the explicit formula, we compare the ARL values of the ARMA(1,1) process on the Extended EWMA control chart with the ARL, SDRL, and MRL of the EWMA control charts. This comparison evaluates their efficiency. The results are presented in Table 7 (Appendix) and Figure 1 (Appendix), demonstrating a clear agreement with the findings seen in Table 4, Table 5 and Table 6 in Appendix. Figure 2 (Appendix) displays a comparison of the RMI, AEQL, and PCI derived from each control chart. The purpose is to assess the effectiveness of the control charts.

In this research, the performance of the ARL of the Extended EWMA control chart is assessed and contrasted with that of the EWMA control chart. The findings suggest that the Extended EWMA control charts are superior to the EWMA control chart for the ARMA(1,1) process. Additionally, the Extended EWMA control chart, with $\lambda_1 = 0.10$, better than all three control charts.

7 Conclusions

In this study, the formula was successful in finding the ARL value and the accuracy of the Extended EWMA control chart for the ARMA(1,1) compared to the EWMA control chart. The efficacy of control charts was evaluated for the ARL by utilizing the NIE, the explicit formula is subjected to comparison, And use all five measurements as an additional criterion to compare the performance of the two control charts. Both methods demonstrate that the ARL values are similar. The Extended EWMA control chart for the ARMA(1,1) process has superior performance compared to the EWMA control chart. When assessing the comparative efficacy of the ARL under different smoothing factors, it is recommended to utilize a smoothing parameter of $\lambda_1 = 0.10$. The simulation research and the real-world dataset on PM2.5 dust levels in Bangkok, Thailand during January and February of 2024, ultimately, the outcomes were the same. Further research, the extended EWMA control chart can be applied to other aspects, such as health or economics, as well as using an NIE comparison method with the explicit formulas. Several other methods of comparison will generate new control charts.

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Contribution of Individual Authors to the Creation of a Scientific Article (Ghostwriting Policy)

- Phunsa Mongkoltawat carried out the writing-original draft preparation and simulation.
- Yupaporn Areepong has organized the conceptualization, writing-review and editing, and validation
- Saowanit Sukparungsee has implemented the methodology and software.

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Conflicts of Interest

The authors declare no conflict of interest.

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APPENDIX

Table 1. ARL comparison of the Extended EWMA control chart for ARMA(p,q) using explicit formulas against NIE method when $\phi_0 = 0.5$. $\phi_1 = 0.1$, $\theta_1 = 0.1$, $\alpha_0 = 1$ for $ARL_0 = 370$

δ		$\lambda_2 = 0.01$						
		θ	$\theta_1 = 0.1$		$\theta_1 = 0.1, \theta_2 = 0.2$		$\theta_1 = 0.1, \theta_2 = 0.2, \theta_3 = 0.3$	
			Process	ARMA(1,1)		ARMA(1,2)		ARMA(1,3)
		λ_1	0.05	0.10	0.05	0.10	0.05	0.10
		b	2.96646	3.63105	2.96808	3.63040	2.96850	3.63300
0.00	Explicit		370.00200	370.55058	370.50773	370.04983	370.54021	370.59006
	NIE		370.00200	370.55058	370.50773	370.04983	370.54021	370.59006
0.01	Explicit		358.74338	357.14330	359.22690	356.66069	359.25546	357.17068
	NIE		358.74338	357.14330	359.22690	356.66069	359.25546	357.17068
0.03	Explicit		337.80512	332.44859	338.24730	331.99912	338.26862	332.45412
	NIE		337.80512	332.44859	338.24730	331.99912	338.26862	332.45412
0.05	Explicit		318.76091	310.26908	319.16611	309.84910	319.18091	310.25553
	NIE		318.76091	310.26908	319.16611	309.84910	319.18091	310.25553
0.10	Explicit		278.06230	263.82544	278.39071	263.46626	278.39183	263.77384
	NIE		278.06230	263.82544	278.39071	263.46626	278.39183	263.77384
0.30	Explicit		177.21156	155.21847	177.36603	154.99343	177.33527	155.09284
	NIE		177.21156	155.21847	177.36603	154.99343	177.33527	155.09284
0.50	Explicit		125.76868	104.21405	125.84709	104.05419	125.80204	104.06690
	NIE		125.76868	104.21405	125.84709	104.05419	125.80204	104.06690
1.00	Explicit		69.34975	52.92752	69.36142	52.83141	69.30529	52.78121
	NIE		69.34975	52.92752	69.36142	52.83141	69.30529	52.78121
2.00	Explicit		34.94470	24.95056	34.93253	24.89444	34.87852	24.83357
	NIE		34.94470	24.95056	34.93253	24.89444	34.87852	24.83357
3.00	Explicit		3.16420	1.17294	3.14954	1.13249	3.10190	1.07843
	NIE		3.16420	1.17294	3.14954	1.13249	3.10190	1.07843

Table 2. ARL comparison of the Extended EWMA control chart for ARMA(p,q) using explicit formulas against NIE method when $\phi_0 = 0.5$. $\phi_1 = 0.1$, $\theta_1 = -0.1$, $\alpha_0 = 1$ for $ARL_0 = 370$

δ		$\lambda_2 = 0.01$						
		θ	$\theta_1 = -0.1$		$\theta_1 = -0.1, \theta_2 = 0.2$		$\theta_1 = -0.1, \theta_2 = 0.2, \theta_3 = 0.3$	
			Process	ARMA(1,1)		ARMA(1,2)		ARMA(1,3)
		λ_1	0.05	0.10	0.05	0.10	0.05	0.10
		b	2.96750	2.96808	2.96900	3.63200	2.96850	3.63300
0.00	Explicit		370.50578	370.50773	370.91904	370.77768	370.56128	370.67882
	NIE		370.50578	370.50773	370.91904	370.77768	370.56128	370.67882
0.01	Explicit		359.23026	359.22690	359.62363	357.35786	359.27603	357.25643
	NIE		359.23026	359.22690	359.62363	357.35786	359.27603	357.25643
0.03	Explicit		338.26024	338.24730	338.61718	332.64039	338.28826	332.53439
	NIE		338.26024	338.24730	338.61718	332.64039	338.28826	332.53439
0.05	Explicit		319.18744	319.16611	319.51189	310.44085	319.19975	310.33096
	NIE		319.18744	319.16611	319.51189	310.44085	319.19975	310.33096
0.10	Explicit		278.42876	263.80566	278.68608	263.95661	278.40903	263.83937
	NIE		278.42876	263.80566	278.68608	263.95661	278.40903	263.83937
0.30	Explicit		177.43595	155.23175	177.54395	155.26439	177.34909	155.13671
	NIE		177.21156	155.23175	177.54395	154.26439	177.34909	155.13671
0.50	Explicit		125.92494	104.23758	125.97011	104.22719	125.81467	104.10135
	NIE		125.92494	104.23758	125.97011	104.22719	125.81467	104.10135
1.00	Explicit		69.43484	52.95400	69.42895	52.91705	69.31705	52.80609
	NIE		69.43484	52.95400	69.42895	52.91705	69.31705	52.80609
2.00	Explicit		34.98837	24.97123	34.96830	24.93601	34.88929	24.85146
	NIE		34.98837	24.97123	34.96830	24.93601	34.88929	24.85146
3.00	Explicit		3.19345	1.18909	3.17426	1.16019	3.11154	1.09275
	NIE		3.19345	1.18909	3.17426	1.16019	3.11154	1.09275

Table 3. ARL comparison of the Extended EWMA control chart for ARMA(p,q) using explicit formulas against NIE method when $\phi_0 = 0.5$. $\phi_1 = 0.1$, $\theta_1 = -0.1$, $\theta_2 = -0.2$, $\alpha_0 = 1$ for $ARL_0 = 370$

δ	θ	$\lambda_2 = 0.01$						
		$\theta_1 = -0.1$		$\theta_1 = -0.1, \theta_2 = -0.2$		$\theta_1 = -0.1, \theta_2 = -0.2, \theta_3 = 0.3$		
		Process	ARMA(1,1)		ARMA(1,2)		ARMA(1,3)	
		λ_1	0.05	0.10	0.05	0.10	0.05	0.10
	b	2.96750	2.96808	2.96655	3.62855	2.96801	3.63012	
0.00	Explicit	370.50578	370.50773	370.48056	370.10566	370.50499	370.00294	
	NIE	370.50578	370.50773	370.48056	370.10566	370.50499	370.00294	
0.01	Explicit	359.23026	359.22690	359.21279	356.72779	359.22487	356.61687	
	NIE	359.23026	359.22690	359.21279	356.72779	359.22487	357.61687	
0.03	Explicit	338.26024	338.24730	338.25548	332.08623	338.24655	331.96088	
	NIE	338.26024	338.24730	338.25548	332.08623	338.24655	331.96088	
0.05	Explicit	319.18744	319.16611	319.19400	309.95325	319.16650	309.81575	
	NIE	319.18744	319.16611	319.19400	309.95325	319.16650	309.81575	
0.10	Explicit	278.42876	263.80566	278.45731	263.60281	278.39342	263.44280	
	NIE	278.42876	263.80566	278.45731	263.60281	278.39342	263.44280	
0.30	Explicit	177.43595	155.23175	177.50268	155.18361	177.37365	154.99338	
	NIE	177.21156	155.23175	177.50268	155.18361	177.37365	154.99338	
0.50	Explicit	125.92494	104.23758	125.99841	104.24435	125.85641	104.05877	
	NIE	125.92494	104.23758	125.99841	104.24435	125.85641	104.05877	
1.00	Explicit	69.43484	52.95400	69.49913	52.99346	69.37115	52.84059	
	NIE	69.43484	52.95400	69.49913	52.99346	69.37115	52.84059	
2.00	Explicit	34.98837	24.97123	35.03238	25.00843	34.94060	24.90295	
	NIE	34.98837	24.97123	35.03238	25.00843	34.94060	24.90295	
3.00	Explicit	3.19345	1.18909	2.22606	1.21908	3.15615	1.13953	
	NIE	3.19345	1.18909	2.22606	1.21908	3.15615	1.13953	

Table 4. ARL comparison of the Extended EWMA control chart for ARMA(1,1) against EWMA control charts when $\phi_0 = 0.5, \phi_1 = 0.1, \theta_1 = 0.1, \lambda_2 = 0.01, \alpha_0 = 1$ for $ARL_0 = 370$

δ	Control Chart	Extended EWMA	Extended EWMA	EWMA	EWMA
		$\lambda_1 = 0.05$	$\lambda_1 = 0.10$	$\lambda_1 = 0.05$	$\lambda_1 = 0.10$
	UCL	2.96646	3.63105	2.96920	3.63200
0.00	ARL_0	370.00200	370.55058	370.96100	370.68423
	$SDRL_0$	370.00200	370.55058	370.96100	370.68423
	MRL_0	370.00200	370.55058	370.96100	370.68423
0.01	ARL_1	358.74338	357.14330	359.66153	359.26077
	$SDRL_1$	358.64338	357.04330	359.56153	359.16770
	MRL_1	247.30530	246.20230	247.93820	247.66200
0.03	ARL_1	337.80512	332.44859	339.64755	334.53658
	$SDRL_1$	337.66369	332.30716	339.50612	334.39515
	MRL_1	232.87120	229.17860	234.14130	230.61800
0.05	ARL_1	318.76091	310.26908	319.53546	312.33100
	$SDRL_1$	318.53730	310.04547	319.31185	312.10739
	MRL_1	219.74280	213.88880	220.27670	215.31020
0.10	ARL_1	278.06230	263.82544	279.69520	265.83410
	$SDRL_1$	277.74607	263.50921	279.37897	265.51787
	MRL_1	191.68660	181.87220	192.81220	183.25690
0.30	ARL_1	177.21156	155.21847	179.51763	157.11150
	$SDRL_1$	176.66383	154.67074	178.96990	156.56377
	MRL_1	122.16350	107.00230	123.75330	108.3072
0.50	ARL_1	125.76868	104.21405	127.92548	106.05878
	$SDRL_1$	125.06157	103.50694	127.21837	105.35167
	MRL_1	86.70059	71.84157	88.18741	73.11326
1.00	ARL_1	69.34975	52.92752	69.36277	54.73000
	$SDRL_1$	68.34975	51.92752	68.36277	53.73000
	MRL_1	47.80732	36.48641	47.81630	37.72897
2.00	ARL_1	34.94470	24.95056	36.88625	26.73442
	$SDRL_1$	33.53048	23.53634	35.47203	25.32020
	MRL_1	24.08967	17.20005	25.42811	18.42979
3.00	ARL_1	3.16420	1.17294	23.08540	15.95210
	$SDRL_1$	1.43214	0.55911	21.35334	14.22004
	MRL_1	1.18129	0.50858	15.42811	10.99683
	RMI	0.31766	0.0000	2.21926	1.41800
	AEQL	28.89150	20.70582	47.67518	34.89650
	PCI	1.39533	1.00000	2.30250	1.68534

Table 5. ARL comparison of the Extended EWMA control chart for ARMA(1,2) against EWMA control charts when $\phi_0 = 0.5, \phi_1 = 0.1, \theta_1 = 0.1, \theta_2 = 0.2, \lambda_2 = 0.01, \alpha_0 = 1$ for $ARL_0 = 370$

Control	Extended EWMA	Extended EWMA	EWMA	EWMA
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δ	Chart	$\lambda_1 = 0.05$	$\lambda_1 = 0.10$	$\lambda_1 = 0.05$	$\lambda_1 = 0.10$
		UCL	3.63040	3.63040	2.96880
0.00	ARL ₀	370.50773	370.04983	370.73988	370.46651
	SDRL ₀	370.50773	370.04983	370.73988	370.46651
	MRL ₀	370.50773	370.04983	370.73988	370.46651
0.01	ARL ₁	359.22690	356.66069	359.44785	359.05123
	SDRL ₁	359.12690	356.56069	359.34785	358.95123
	MRL ₁	247.63860	245.86960	247.79090	247.51750
0.03	ARL ₁	338.24730	331.99912	339.44759	334.34198
	SDRL ₁	338.07409	331.82591	339.27438	334.16877
	MRL ₁	233.17600	228.86870	234.00350	230.48380
0.05	ARL ₁	319.16611	309.84910	319.34781	312.14968
	SDRL ₁	318.94250	309.62549	319.12420	311.92607
	MRL ₁	220.02210	213.59930	220.14740	215.18520
0.10	ARL ₁	278.39071	263.46626	279.53339	265.68011
	SDRL ₁	278.07448	263.15003	279.21716	265.36388
	MRL ₁	191.91290	181.62450	192.70070	183.15070
0.30	ARL ₁	177.36603	154.99343	179.41662	157.01868
	SDRL ₁	176.81830	154.44570	178.86889	156.47095
	MRL ₁	122.27000	106.84710	123.68360	108.24330
0.50	ARL ₁	125.84709	104.05419	127.85347	106.99335
	SDRL ₁	125.13998	103.34708	127.14636	106.28624
	MRL ₁	86.75464	71.73137	88.13777	73.75752
1.00	ARL ₁	69.36142	52.83141	69.42107	54.69192
	SDRL ₁	68.36142	51.83141	68.42107	53.69192
	MRL ₁	47.81537	36.42015	47.85649	37.70272
2.00	ARL ₁	34.93253	24.89444	36.86304	26.71276
	SDRL ₁	33.51831	23.48022	35.44882	25.29854
	MRL ₁	24.08128	17.16137	25.41211	18.41485
3.00	ARL ₁	3.14954	1.13249	23.06902	15.93662
	SDRL ₁	1.14174	0.59556	21.33696	14.20456
	MRL ₁	1.14118	0.58700	15.90299	10.98616
	RMI	0.32930	0.00000	2.29702	1.47237
	AEQL	28.87843	20.63083	28.88843	34.96265
	PCI	1.39977	1.00000	1.40025	1.69467

Table 6. ARL comparison of the Extended EWMA control chart for ARMA(1,3) against EWMA control charts when $\phi_0 = 0.5$, $\phi_1 = 0.1$, $\theta_1 = 0.1$, $\theta_2 = 0.2$, $\theta_3 = 0.3$, $\lambda_2 = 0.01$, $\alpha_0 = 1$ for $ARL_0 = 370$

Control	Extended EWMA	Extended EWMA	EWMA	EWMA
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δ	Chart	$\lambda_1 = 0.05$	$\lambda_1 = 0.10$	$\lambda_1 = 0.05$	$\lambda_1 = 0.10$
		UCL	3.63300	2.96950	3.63400
0.00	ARL ₀	370.54021	370.59006	370.92196	370.90797
	SDRL ₀	370.54021	370.59006	370.92196	370.90797
	MRL ₀	370.54021	370.59006	370.92196	370.90797
0.01	ARL ₁	359.25546	357.17068	359.62196	359.47073
	SDRL ₁	359.15546	357.07068	359.52196	359.37073
	MRL ₁	247.65830	246.22110	247.91100	247.80670
0.03	ARL ₁	338.26862	332.45412	339.60703	334.72153
	SDRL ₁	338.09541	332.28091	339.43382	334.54832
	MRL ₁	233.19070	229.18240	234.11340	230.74550
0.05	ARL ₁	319.18091	310.25553	319.49410	312.49391
	SDRL ₁	318.95730	310.03192	319.27049	312.27070
	MRL ₁	220.03230	213.87950	220.24820	215.42250
0.10	ARL ₁	278.39183	263.77384	279.65219	265.95230
	SDRL ₁	278.07560	263.45761	279.33596	265.63607
	MRL ₁	191.91370	181.83660	192.78260	183.33830
0.30	ARL ₁	177.33527	155.09284	179.47178	157.13560
	SDRL ₁	176.80754	154.54511	178.92405	156.58787
	MRL ₁	122.24880	106.91570	123.72170	108.32390
0.50	ARL ₁	125.80204	104.06690	127.87959	106.04689
	SDRL ₁	125.09493	103.35979	127.17248	105.33978
	MRL ₁	86.72358	71.74013	88.15578	73.10507
1.00	ARL ₁	69.30529	52.78121	69.42012	54.69363
	SDRL ₁	68.30529	51.78121	68.42012	53.69363
	MRL ₁	47.77667	36.38554	47.85583	37.70390
2.00	ARL ₁	34.87852	24.83357	36.85123	26.69774
	SDRL ₁	33.46430	23.41935	35.43701	25.28352
	MRL ₁	24.04405	17.11941	25.40397	18.40450
3.00	ARL ₁	3.10190	1.07843	23.05620	15.92080
	SDRL ₁	1.36984	0.65362	21.32414	14.18874
	MRL ₁	1.13834	0.64343	15.89415	10.97525
	RMI	0.33929	0.00000	2.40922	1.54835
	AEQL	28.80694	20.55447	47.63901	34.92037
	PCI	1.40149	1.00000	2.31769	1.69891

Table 7. ARL comparison of the Extended EWMA control chart for ARMA(1,1) using NIE against EWMA control chart when $\alpha_0 = 27.3401$, $\phi_1 = 0.947$, $\theta_1 = 0.616$ for $ARL_0 = 370$

δ	Control Chart	Extended EWMA	Extended EWMA	EWMA	EWMA
		$\lambda_1 = 0.05$	$\lambda_1 = 0.10$	$\lambda_1 = 0.05$	$\lambda_1 = 0.10$
	UCL	81.14050	99.36650	81.20000	99.42950

0.00	ARL ₀	370.09745	370.04424	370.87888	370.78295
	SDRL ₀	370.09745	370.04424	370.87888	370.78295
	MRL ₀	370.09745	370.04424	370.87888	370.78295
0.01	ARL ₁	358.83352	346.65490	359.58227	357.35438
	SDRL ₁	358.73352	346.55490	354.48227	357.25438
	MRL ₁	247.36740	245.86560	247.88360	246.34780
0.03	ARL ₁	337.88490	321.99279	338.57353	332.62084
	SDRL ₁	337.71169	321.81958	338.40032	332.44763
	MRL ₁	232.92620	228.86440	233.40090	229.29730
0.05	ARL ₁	318.83135	299.84205	319.46609	310.40691
	SDRL ₁	318.60774	299.61844	319.24248	310.18330
	MRL ₁	219.79130	213.59440	220.22890	213.98380
0.10	ARL ₁	278.11315	253.45687	288.63544	273.89268
	SDRL ₁	277.79692	253.14064	288.31921	273.57645
	MRL ₁	191.72160	181.61810	198.97530	188.81220
0.30	ARL ₁	177.21618	144.97452	179.47908	155.13019
	SDRL ₁	176.66845	144.42679	178.93135	154.58246
	MRL ₁	122.16670	106.83410	123.72670	106.94140
0.50	ARL ₁	125.75122	94.01994	127.89571	106.05868
	SDRL ₁	125.04411	93.31283	127.18860	105.35157
	MRL ₁	86.68855	71.70776	88.16689	73.11319
1.00	ARL ₁	69.30894	42.77308	79.33972	62.70964
	SDRL ₁	68.30894	41.77308	78.33972	61.70964
	MRL ₁	47.77919	36.37994	54.69406	43.22986
2.00	ARL ₁	34.89093	14.80931	36.86497	26.70099
	SDRL ₁	33.47671	13.39509	35.45075	25.28677
	MRL ₁	24.05260	17.10268	25.41344	18.40674
3.00	ARL ₁	3.10585	1.03310	23.06385	15.91344
	SDRL ₁	1.37379	0.69895	21.33179	14.18138
	MRL ₁	1.14106	0.61218	15.89942	10.97018
	RMI	0.35411	0.00000	2.53962	1.63711
	AEQL	28.81308	20.50045	48.65270	35.70610
	PCI	1.40548	1.0000	2.37325	1.74172

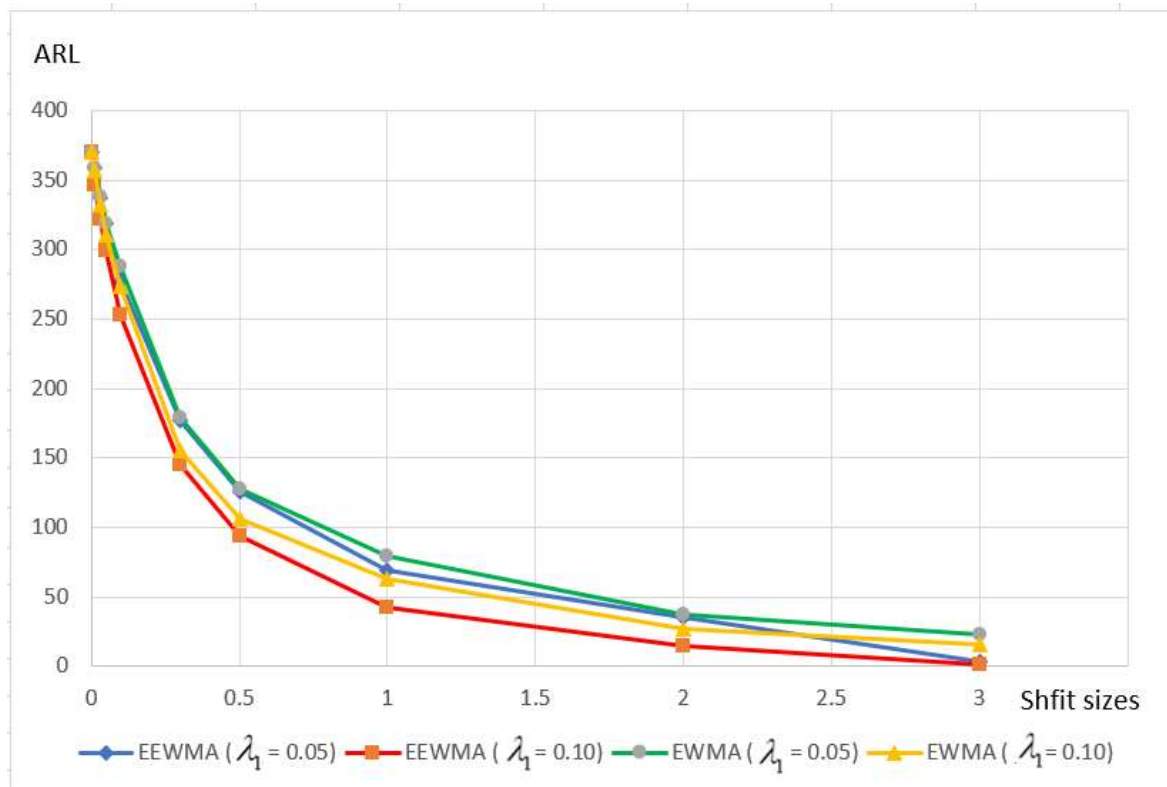


Fig. 1: The ARL₁ values on the control chart using a real-world dataset

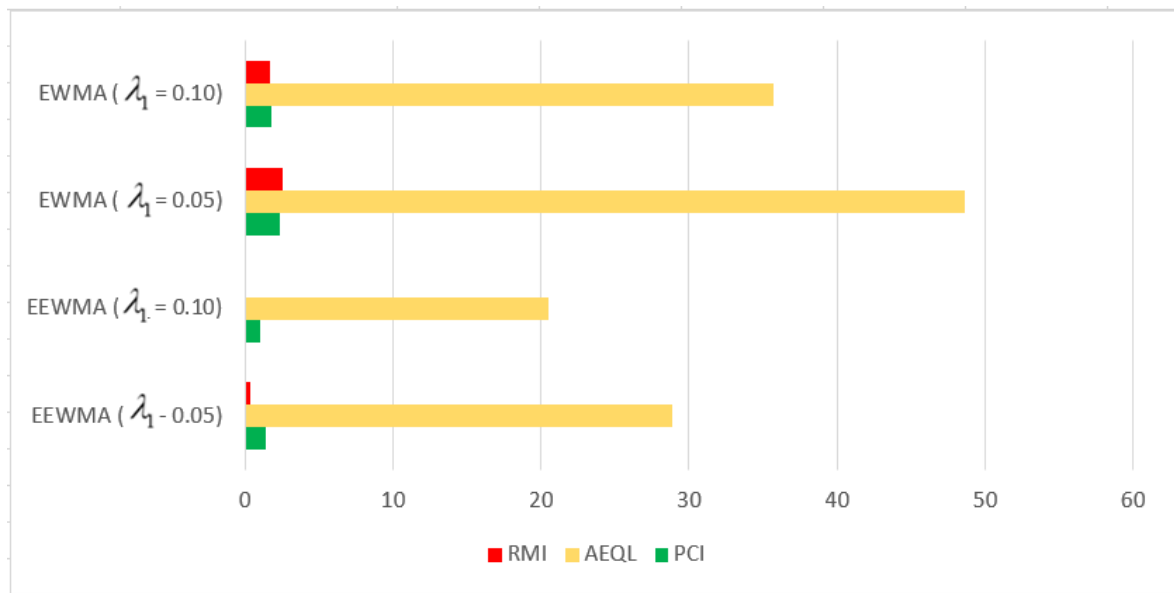


Fig. 2: Comparison the RMI, AEQL, and PCI values with the Extended EWMA control chart and the EWMA control chart for $\lambda_1 = 0.05, 0.10$