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Mathematical Model Analysis of a Drug Transmission with Criminal Law and Its Optimal Control

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Abstract. The Indonesian government has made various efforts in drug prevention policies through criminal law. However, law enforcement officials still contemplate the Drugs Law is oriented towards imprisonment, so drug abuse is considered as a criminal act. In fact, the government has declared 2014 as the year to save the victims of drug abuse through rehabilitation. In this paper, a drug transmission model with the criminal law aspect is presented and analyzed. The optimal control strategy is then applied in the form of rehabilitation efforts. Based on the model analysis, we found two equilibriums, namely the drugs-free equilibrium and the drug addiction equilibrium. The stability of the equilibriums depends on the basic reproduction number. The spread of drug addicts persists in the population if the basic reproduction number greater than unity. Based on the simulation, it can be seen that criminal law give a significant impact to decrease the number of mild drug addicts and also the number of heavy levels of drug addicts. Furthermore, the existence of the optimal control variable is determined through the Pontryagin's Maximum Principle method. The comparison of simulation results with and without control shows that rehabilitation efforts can reduce drug addicts transmission.

Keywords: criminal law, drugs model, optimal strategy, stability.

INTRODUCTION

In the modern era, drug abuse has become one of the main issues around the world. Drug abuse is often reported in electronic media and print media. The outbreak of drug misuse can have a negative impact on health, economy, and society. This such condition is very detrimental to individuals, society, and the state, especially the younger generation. The government has made various efforts through drug prevention policies, namely the repression of illicit drug trafficking and the prevention of drug abuse in society. Thus, the role of various parties is required to be able to fight drugs and overcome the increasing number of drugs for the future of the Indonesian nation free of drugs [1].

Indonesia is a country with the fourth most populous population in the world [2]. This is the target of drugs smuggling. In addition to the high demand for drug consumption, economic development in Indonesia is also relatively high, making it an attraction for drug syndicates. The rise of drugs smuggling to Indonesia is partly due to the vastness of Indonesian maritime with limited personnel [3].

In 2010 the Supreme Court issued Surat Edaran Mahkamah Agung (SEMA) No. 4 of 2010 concerning the placement of misuses, victims of misuse and drug addicts into rehabilitation institutions which become the reference for judges to impose rehabilitation decisions. However, so far law enforcement officials still see the Drugs Law is oriented towards imprisonment for drug users or addicts so that they are considered criminals. In fact, the government declared that 2014 is the year to save victims of drug misuse through rehabilitation [4]. Therefore, researchers are interested in examining the dynamics of the spread of drug addicts with the criminal law aspect.

The mathematical model plays an important role in understanding the behavior of epidemics and biology (see [5, 6, 7]). Drug abuse has almost identical characteristics with conventional epidemics. So far, various mathematical models describing the spread of drug abuse has been carried out (see [8, 9, 10, 11]). The author in [8] has studied the dynamic between homelessness and drug abuse. The dynamic of traditional drug users who switched into using

synthesis drugs which caused social problems have been presented in [9]. The effect of media coverage on the dissemination and control of drug addicts has been developed in [10]. The authors in [11] analyzed the effects of providing family education and public health education on the number of drug addicts. Next, the optimal control strategies have been applied to reduce the spread of drug abuse. The influence of illegal drug use in the community by paying attention to interactions that occur between individuals who are vulnerable to illegal drug users and its optimal control analysis has been investigated in [12].

In this study, we consider a mathematical model that describes the dynamics of the spread of drugs addicts that derived by authors in [11] with adding criminal penalties aspect. We also present the application of optimal control strategies to reduce the number of drug addicts. We first analyze the model without control that is the stability of the equilibriums and supported by the results of a numerical approach. This is done to find out the dynamics of drug addicts spreading. From the analysis carried out, it is expected that the results can be more relevant to the dynamics of the drug addicts spreading in Indonesia. We then apply optimal control in the form of rehabilitation efforts to reduce the population of drug addicts.

MATHEMATICAL MODEL FORMULATION

In this section, we formulate a mathematical model of drug addicts with the criminal law aspects. The assumptions used for mathematical models of drugs addicts with criminal law aspect are as follows:

- Populations that are vulnerable to drug addicts who have either received drug misuse education or have not received drug-addicted education can be mild level drug addicts.
- 2. The population of heavy drug addicts cannot be mild level drug addicts.
- 3. The population of light and heavy narcotics addicts will undergo criminal law if caught
- The population of drug addicts who have undergone criminal law can return to being mild and severe drug addicts.
- 5. The population of drug addicts who stop using the drug cannot be mild or severe drug addicts.
- 6. The natural rate of birth and death of humans is considered the same
- 7. The population of narcotics addicts who can be exposed to criminal law is 14 years or older.

Description of variables and parameters used in the model can be seen in Table 1 and Table 2.

TABLE 1. Description of variables.

| Variable | Description |
|-----------------|---|
| S(t) | The number of individuals which susceptible to drug addicts at time t |
| C(t) | The number of individuals vulnerable to being a narcotic addict but has received drugs misuse education at time t |
| L(t) | The number of mild drugs addicts at time t |
| H(t) | The number of heavy level drugs addicts at time t |
| W(t) | The number of drugs addicts who undergo criminal law at time t |
| R(t) | The number of individuals who stopped using drugs at time t |

We assume that all parameters are constant and non-negative. Based on the assumptions and definitions of variables and parameters, a transmission diagram can be shown in Figure 1. Based on Figure 1, the following system of ordinary differential equations is obtained:

$$\frac{dS}{dt} = (1-q)d\Lambda - \beta_1 S L - \beta_2 S H - (d+\mu) S \tag{1}$$

$$\frac{dC}{dt} = qd\Lambda + \mu S - \beta_1 \xi CL - \beta_2 \xi CH - (d + \delta)C$$
(2)

$$\frac{dL}{dt} = \beta_1 S L + \beta_2 S H + \beta_1 \xi C L + \beta_2 \xi C H - (\gamma + \pi + d + d_1) L + \omega W$$
(3)

TABLE 2. Description of parameters.

| Parameter | Description |
|----------------|--|
| d | Natural human birth/death rate |
| Λ | Recruitment rate |
| \overline{q} | Probability of human that receiving drugs misuse education |
| μ | Progression rate from S to C |
| β_1 | The direct contact rate of S with L |
| β_2 | The direct contact rate of S with H |
| ξ | Probability factors that resulting in direct contact levels decrease |
| d_1 | The death rate of mild drugs addicts |
| d_2 | The death rate of heavy drug addicts |
| γ | The transition rate from L to W |
| ω | The transition rate from W to L |
| π | Progression rate from L to H |
| δ | Progression rate from C to R |
| θ | The transition rate from H to W |
| σ | The transition rate from W to H |
| m | Progression rate from W to R |

$$\frac{dH}{dt} = \pi L + \sigma W - (\theta + d + d_2) H \tag{4}$$

$$\frac{dW}{dt} = \theta H + \gamma L - (\omega + \sigma + m + d) W \tag{5}$$

$$\frac{dR}{dt} = mW + \delta C - dR. \tag{6}$$

To simplify the notation, we define

$$\begin{array}{rcl} m_1 & = & \gamma + \pi + d + d_1 \\ m_2 & = & \theta + d + d_2 \\ m_3 & = & \omega + \sigma + m + d. \end{array}$$

Because the R variable does not appear in another equation, we focus on the following equation.

$$\frac{dS}{dt} = (1 - q) d\Lambda - \beta_1 S L - \beta_2 S H - (d + \mu) S$$

$$\frac{dC}{dt} = q d\Lambda + \mu S - \beta_1 \xi C L - \beta_2 \xi C H - (d + \delta) C$$

$$\frac{dL}{dt} = \beta_1 S L + \beta_2 S H + \beta_1 \xi C L + \beta_2 \xi C H - m_1 L + \omega W$$

$$\frac{dH}{dt} = \pi L + \sigma W - m_2 H$$

$$\frac{dW}{dt} = \theta H + \gamma L - m_3 W.$$
(7)

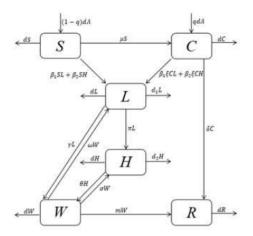


FIGURE 1. Transmission diagram of drugs addicts model with the criminal law aspect.

MODEL ANALYSIS

The mathematical model of drugs addicts with criminal law aspect has two equilibriums, namely a drugs-free (non-endemic) equilibrium and a drug addiction (endemic) equilibrium. The non-endemic equilibrium of the model is $E^0 = \left(S^0, C^0, L^0, H^0, W^0, R^0\right) = \left(\frac{(1-q)dA}{(d+\mu)}, \frac{(\mu+qd)dA}{(d+\mu)(d+\delta)}, 0, 0, 0\right)$ and endemic equilibrium $E^* = (S^*, C^*, L^*, H^*, W^*)$, where

$$\begin{split} \underline{L}^* &= \frac{m_2 m_3 - \sigma \theta}{m_3 \pi + \sigma \gamma} H^* \\ W^* &= \frac{\theta \pi + \gamma m_2}{m_3 \pi + \sigma \gamma} H^* \\ S^* &= \frac{(1 - q) (m_3 \pi + \sigma \gamma) d\Lambda}{(\beta_1 (m_2 m_3 - \sigma \theta) + \beta_2 (m_3 \pi + \sigma \gamma)) H^* + (d + \mu) (m_3 \pi + \sigma \gamma)} \\ C^* &= \frac{(q d\Lambda + \mu S^*) (m_3 \pi + \sigma \gamma)}{\beta_1 \xi (m_2 m_3 - \sigma \theta) + (\beta_2 \xi H^* + d + \delta) (m_3 \pi + \sigma \gamma)}. \end{split}$$

The compartment L^* , S^* , C^* will exist if $m_2m_3 > \sigma\theta$, while H^* is the root of the second degree polynomial as follows:

$$aH^{*2} + bH^* + c = 0. (8)$$

where

$$a = \xi(\beta_{1}(x-y) + \beta_{2})^{2} (m_{1}(x-y) - \omega z)$$

$$b = (\beta_{1}(x-y) + \beta_{2}) (m_{1}(x-y) - \omega z) (d + \delta + \xi(d + \mu)) - d\Lambda \xi(\beta_{1}(x-y) + \beta_{2})^{2},$$

$$c = d^{2}\Lambda (\beta_{1}(x-y) + \beta_{2}) (q - \xi q - 1) + d\Lambda (\beta_{1}(x-y) + \beta_{2}) (q\delta - \delta - \xi \mu) + (m_{1}(x-y) - \omega z) (d + \mu) (d + \delta),$$

$$x = \frac{m_{2}m_{3}}{m_{3}\pi + \sigma \gamma},$$

$$y = \frac{\sigma \theta}{m_{3}\pi + \sigma \gamma},$$

$$z = \frac{\pi \theta + \gamma m_{2}}{m_{2}\pi + \sigma \gamma}.$$

Based on the above conditions, the terms x > y are obtained. The Eq (8) have one positive solution if

- (i) a > 0 and c < 0, or
- (ii) a < 0 and c > 0.

The condition (ii) is not is not fulfilled because when a < 0, we have c < 0 as follows. Because $(q - \xi q - 1) < 0$ and $(q\delta - \delta - \xi \mu) < 0$, we get

$$c = d\Lambda (\beta_1 (x - y) + \beta_2) [d (q - \xi q - 1) + (q\delta - \delta - \xi \mu)] + (m_1 (x - y) - \omega z) (d + \mu) (d + \delta) < 0.$$

Hence, the endemic equilibrium exist if

- i). $m_2m_3 > \sigma\theta$,
- ii). $m_1(x-y) > \omega z$,

iii).
$$d\Lambda (\beta_1(x-y) + \beta_2) [d(1-q(1-\xi)) + (\xi\mu - \delta(q-1))] > (m_1(x-y) - \omega z) (d+\mu) (d+\delta).$$

Furthermore, it will be studied in the analysis of local stability from each equilibrium. This analysis can be used to determine the dynamics of system behavior in mathematical models of drug addicts with the criminal law aspect. In the first step, we determine the locally asymptotically stable of non-endemic equilibrium. Linearizing model (7) at the non-endemic equilibrium gives eigenvalues $\lambda_1 = -(d + \mu)$, $\lambda_2 = -(d + \delta)$ and the remainder are the roots of the following equation:

$$\lambda^3 + a_1 \lambda^2 + a_2 \lambda + a_3 = 0, (9)$$

where

$$\begin{array}{rcl} a_{1} & = & m_{1} + m_{2} + m_{3} - \beta_{1}S^{0} - \beta_{1}\xi C^{0} \\ a_{2} & = & -\beta_{1}\xi C^{0}m_{3} - \beta_{1}\xi C^{0}m_{2} - \gamma\omega - \pi\beta_{2}\xi C^{0} - \theta\sigma + m_{1}m_{2} - \beta_{1}S^{0}m_{3} - \pi\beta_{2}S^{0} + m_{1}m_{3} - \beta_{1}S^{0}m_{2} + m_{2}m_{3} \\ a_{3} & = & -m_{2}m_{3}\beta_{1}S^{0} + \theta\sigma\beta_{1}S^{0} - m_{2}m_{3}\beta_{1}\xi C^{0} + \theta\sigma\beta_{1}\xi C^{0} + m_{1}m_{2}m_{3} - \theta\sigma m_{1} - \pi m_{3}\beta_{2}S^{0} - \pi m_{3}\beta_{2}\xi C^{0} \\ & -\pi\omega\theta - \gamma\sigma\beta_{2}S^{0} - \gamma\sigma\beta_{2}\xi C^{0} - \gamma\omega m_{2}. \end{array}$$

The non-endemic equilibrium will be asymptotically stable if and only if Eq (9) has negative real parts roots. Using the Routh Hurwitz criteria, the cubic equation will have negative roots if $a_1, a_2, a_3 > 0$ and $a_1a_2 - a_3 > 0$. Hence, the non-endemic equilibrium E^0 will be locally asymptotically stable if and only if $R_i < 1$, for i = 0, 1, 2, 3, where

$$R_{0} = \frac{\beta_{1}\left(S^{0} + \xi C^{0}\right)}{m_{1} + m_{2} + m_{3}},$$

$$R_{1} = \frac{R_{0}\left(\beta_{1}\left(m_{2} + m_{3}\right) + \pi\beta_{2}\right)\left(m_{1} + m_{2} + m_{3}\right) + \beta_{1}\left(\gamma\omega + \theta\sigma\right)}{\beta_{1}\left(m_{1}m_{2} + m_{1}m_{3} + m_{2}m_{3}\right)},$$

$$R_{2} = \frac{R_{0}\left(m_{1} + m_{2} + m_{3}\right)\left(\beta_{1}m_{2}m_{3} + \beta_{2}\left(\pi m_{3} + \gamma\sigma\right)\right) + \beta_{1}\left(\theta\sigma m_{1} + \pi\omega\theta + \gamma\omega m_{2}\right)}{\beta_{1}\left(R_{0}\theta\sigma\left(m_{1} + m_{2} + m_{3}\right) + m_{1}m_{2}m_{3}\right)F + m_{3}},$$

$$R_{3} = \frac{a_{3}}{a_{1}a_{2}}.$$

This represents that if all conditions are satisfied, there is no spread of drug addicts in the population.

Then, $R_0 = \frac{\beta_1(S^0 + \xi C^0)}{m_1 + m_2 + m_3}$ is referred to as the basic reproduction number or threshold parameter of a situation that indicates the condition of drug addicts is spreading or not.

Furthermore, the stability of the endemic equilibrium is difficult to solve analytically. Therefore, the stability of the endemic equilibrium will be analyzed through numerical simulations using phase fields. The parameter values used are presented in Table 3. The phase portrait of the model in the L-W plane is given in Figure 2. Based on the three different initial values, it can be seen that the graphs tend to converge to the same point. In addition, the value of the basic reproduction ratio is $R_0 = 122,086 > 1$. Hence, the endemic equilibrium tends to be asymptotically stable if $R_0 > 1$. This shows that the population of drug addicts can transmit their consumptive behavior to an average of more than one new addict. Hence, drug addicts can spread or in other words the endemic occurrence of the drug addict.

In Figure 3, we display the effect of the criminal law aspect to the number of mild drug addicts (L) and the number of heavy level drugs addicts (H) for different values of γ and θ respectively. From Figure 3(a), it can be compared that as the parameter γ increase, the population of mild drugs addicts decreases significantly, while from Figure 3(b), it can be summarized that as the parameter θ increase, the population of heavy level drugs addicts tends to decrease.

TABLE 3. Parameters value of the model.

| Parameter | Value | Unit | Source |
|----------------|-------|------------------|---------|
| d | 0.02 | 1 year | [11] |
| Λ | 1500 | People year | Assumed |
| \overline{q} | 0.8 | 1 year | [11] |
| μ | 0.1 | $\frac{1}{year}$ | Assumed |
| β_1 | 0.7 | year | [11] |
| β_2 | 0.8 | $\frac{1}{year}$ | [11] |
| ξ | 0.9 | $\frac{1}{year}$ | [11] |
| d_1 | 0.2 | $\frac{1}{year}$ | [11] |
| d_2 | 0,. | 1 year | [11] |
| γ | 0.3 | $\frac{1}{year}$ | [11] |
| ω | 0.001 | $\frac{1}{year}$ | Assumed |
| π | 0.03 | $\frac{1}{year}$ | [11] |
| δ | 0.01 | 1 year | [11] |
| θ | 0.421 | 1 year | Assumed |
| σ | 0.7 | 1 year | [11] |
| m | 0.25 | 1 year | [11]) |
| | | | |

OPTIMAL CONTROL PROBLEM

In this study, an optimal strategy analysis will be carried out to control drug addicts on the model with the criminal law aspect. The control variable used is the rehabilitation effort (u). Rehabilitation efforts include medical and social rehabilitation. The control strategy is expected to minimize the population of drug addicts.

The mathematical models of drug addicts with criminal law accompanied by controlling strategies are as follows.

$$\frac{dS}{dt} = (1 - q) d\Lambda - \beta_1 SL - \beta_2 SH - (d + \mu)$$

$$\frac{dC}{dt} = qd\Lambda + \mu S - \beta_1 \xi CL - \beta_2 \xi CH - (d + \delta) C$$

$$\frac{dL}{dt} = \beta_1 SL + \beta_2 SH + \beta_1 \xi CL + \beta_2 \xi CH - (\gamma + \pi + d + d_1) L + \omega W - uL$$

$$\frac{dH}{dt} = \pi L + \sigma W - (\theta + d + d_2) H - uH$$

$$\frac{dW}{dt} = \theta H + \gamma L - (\omega + \sigma + m + d) W - uW$$

$$\frac{dR}{dt} = mW + \delta C - dR + uL + uH + uW.$$
(10)

The cost function of the drug addicts model with the criminal law aspect incorporates the control optimal strategy is formulated as follows:

$$Min J = \int_0^{t_f} \left(L + H + W + \frac{A}{2} u^2 \right) dt \tag{11}$$

where $u \in [0, 1]$, $t \in [0, t_f]$, and A denotes the positive constant in the form of rehabilitation costs.

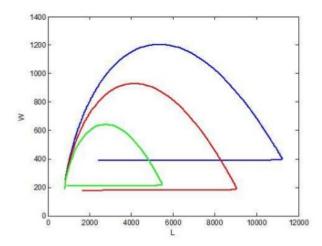


FIGURE 2. Phase portrait of the model in L-W plane.

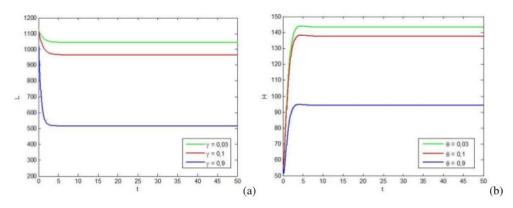


FIGURE 3. (a) Comparison of population L for various γ , (b) Comparison of population H for various θ .

The problem of optimal control in the mathematical model of drug addicts is solved by applying the Pontryagin Maximum Principle [13]. Suppose the state variable in the model with the control strategy is as follows

$$x = (S, C, L, H, W, R)^T$$
.

The initial values of all populations are assumed to be positive which are stated as follows

$$S(t_0), C(t_0), L(t_0), H(t_0), W(t_0), R(t_0) \geq 0.$$

The first step in the analysis of the problem of this control strategy is to form the Hamiltonian (\mathcal{H}) function.

$$\mathcal{H} = L + H + W + \frac{A}{2}u^2 + \sum_{i=1}^{6} \lambda_i^t f_i$$

where f_i is the right-hand side of the model (10) which is the i-th state variable equation. The variables λ_i denote the adjoint or co-state variables, for i = 1, 2, ..., 6.

The solution of the system is determined by taking the partial derivatives of the Hamiltonian function (\mathcal{H}) with respect to the associated state variables [14, 15].

Theorem 1 Given optimal control u^* be the solution of the control systems (10)-(11). Then the adjoint variables are obtained by $\dot{\lambda}_i = -\frac{\partial H}{\partial x}$, with transversality conditions $\lambda_i(t_f) = 0$, i = 1, 2, ..., 6, and

$$u^* = min\left(1, max\left(0, \frac{(\lambda_5 - \lambda_6)W}{A_2}\right)\right). \tag{12}$$

Proof. Using the condition $\dot{\lambda}_i = -\frac{\partial \mathcal{H}}{\partial x}$, we have the following adjoint system

$$\dot{\lambda}_{1} = -\frac{\partial \mathcal{H}}{\partial S} = -\left[\lambda_{1} \left(-\beta_{1} L - \beta_{2} H - (d + \mu)\right) + \lambda_{2} \mu + \lambda_{3} \left(\beta_{1} L + \beta_{2} H\right)\right]
\dot{\lambda}_{2} = -\frac{\partial \mathcal{H}}{\partial C} = -\left[\lambda_{2} \left(-\beta_{1} \xi L - \beta_{2} \xi H\right) - (d + \delta) + \lambda_{3} \left(\beta_{1} \xi L + \beta_{2} \xi H\right) + \lambda_{6} \delta\right]
\dot{\lambda}_{3} = -\frac{\partial \mathcal{H}}{\partial L} = -\left[1 - \lambda_{1} \beta_{1} S - \lambda_{2} \beta_{1} \xi C + \lambda_{3} \left(\beta_{1} S + \beta_{1} \xi C - m_{1}\right) + \lambda_{4} \pi + \lambda_{5} \gamma\right]
\dot{\lambda}_{4} = -\frac{\partial \mathcal{H}}{\partial H} = -\left[1 - \lambda_{1} \beta_{2} S - \lambda_{2} \beta_{2} \xi C + \lambda_{3} \left(\beta_{2} S + \beta_{2} \xi C\right) + \lambda_{4} m_{2} + \lambda_{5} \theta\right]
\dot{\lambda}_{5} = -\frac{\partial \mathcal{H}}{\partial W} = -\left[1 + \lambda_{3} \omega + \lambda_{4} \sigma - \lambda_{5} \left(m_{3} + u\right) + \lambda_{6} \left(m + u\right)\right]
\dot{\lambda}_{6} = -\frac{\partial \mathcal{H}}{\partial R} = \lambda_{6} d,$$
(13)

where $\lambda_i(t_f) = 0$, i = 1, 2, 3, ..., 6. Furthermore, in order to obtain optimal controls given in (12), we use the condition $\frac{\partial \mathcal{H}}{\partial u}$. \square

NUMERICAL SIMULATION

In this section, we discuss the numerical simulation of the model without and with control. The simulation is used to determine the effectiveness of rehabilitation efforts (u) as a form of strategy that can minimize the spread of drug addicts with the criminal law aspect. Simulation is done using the Runge-Kutta fourth-order scheme. The initial values used in this simulation are $(S(t_0), C(t_0), L(t_0), H(t_0), W(t_0), R(t_0)) = (135, 90, 77, 49, 15, 11)$. The time is taken in 50 years. The weighting constant for the cost of rehabilitation efforts is A = 10. The parameter values used to refer to Table 3.

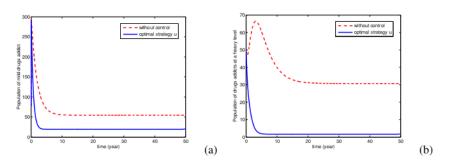


FIGURE 4. Comparison of population L (a) and H (b) with and without control

Figure 4 illustrates a comparison of the number of mild drug addicts and a heavy level of drug addicts populations before and after being given control. From Figure 4(a)-(b), it is seen that using the control strategy, the number of the population of mild-level drug addicts and also heavy-level drug addicts more decrease compared without the control. The strategy in the form of rehabilitation can reduce both populations until the end of the observation. More specifically, rehabilitation efforts can reduce mild drugs addicts and heavy drug addicts populations by 65% and 96% respectively.

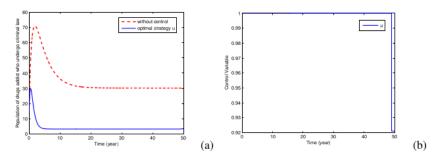


FIGURE 5. (a) Comparison of population W with and without control, (b) Control profile.

Figure 5(a) provides a comparison of the number of drug addicts who undergo criminal law population with and without the control strategy. The control strategy in the form of rehabilitation can reduce the population of drug addicts who undergo criminal law until the end of the observation. More specifically, rehabilitation efforts can reduce the population of drug addicts who undergo criminal law by 90%. Furthermore, the simulation results for the control profile is presented in Figure 5(b). From Figure 5(b) it can be seen that the rehabilitation effort should be given at the beginning of the observation until the 49th year is 100%, which means that the rehabilitation efforts carried out are fully intensive and then decreasing in the 50th year.

CONCLUSIONS

In this paper, we presented the mathematical model of drug addicts with criminal law aspect. The model has two equilibriums, namely a non-endemic equilibrium and an endemic equilibrium. Non-endemic equilibrium will be locally asymptotically stable if the basic reproduction number less than unity and complying several conditions, while the endemic equilibrium tends to be asymptotically stable if the basic reproduction number greater than unity. Based on the simulation, it is indicated that criminal law gives a significant effect to reduce the number of mild drug addicts and also the number of heavy levels drug addicts. Further, the model incorporates the control strategy in the form of rehabilitation efforts was analyzed. The optimal control problem is solved analytically by using the Pontryagin Maximum Principle. The numerical results of the model with and without control strategy show that the rehabilitation efforts can decrease the drug addicts in the population.

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REFERENCES

- [1] Badan Narkotika Nasional, *Pedoman Pencegahan Penyalahgunaan Narkotika Bagi Pecandu* (Badan Narkotika Nasional, Jakarta, 2004).
- [2] United Nations, World Population Prospects 2017 (2017) https://Population.un.org/wpp. [Accessed on 15th November, 2018].
- [3] Tjazuli., Indonesia (BUKAN) Surga Narkoba (Edisi 159, Parlementeria, Jakarta, 2018).
- [4] Badan Narkotika Nasional, Pencanangan Tahun 2014 Sebagai Tahun Penyelamatan Pengguna Narkoba (2014) www.bnn.go.id/read.pressrelease/11731/pencenangan-tahun-2014-sebagai-tahun-penyelamatan-pengguna-narkoba [Accesed on 15th November, 2018].
- [5] Fatmawati and H. Tasman, Appl. Math. Sci. 7, 33793391 (2013).
- [6] Fatmawati and H. Tasman, Math. Biosci. 262, 73-79 (2015).
- [7] Fatmawati and H. Tasman, Int. J. Math. Math. Sci. 2016, Article ID 8261208 (2016).
- [8] C. P. Bhunu, Commun. Nonlinear Sci. 19, 1908-1917 (2013).

- [9] M. Ma, S. Liu, H. Xiang and J. Li, Physica A. 491, 641-649 (2017).
- [10] M. Ma, S. Liu and J. Li, Commun. Nonlinear Sci. 50, 169-179 (2017).
- [11] J. Li and M. Ma, Infectious Disease Modelling 3, 74-84 (2018).
- [12] S. Mushayabasa and G. Tapedzesa, Comput Math Method M. 2015, Article ID 383154 (2015).
- [13] L. S. Pontryagin, V. G. Boltyanskii, R. V. Gamkrelidze and E. F. Mishchenko, *The Mathematical Theory of Optimal Processes* (Wiley, New York, 1962).
- [14] F. I. Lewis and V. L. Syrmos, Optimal Control (Willy Interscience, Canada, 2006).
- [15] D. S. Naidu, Optimal Control Systems (CRC Press, New York, 2002)

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