pp. 1119-1140

EFFECTS OF ISOLATION AND SLAUGHTER STRATEGIES IN DIFFERENT SPECIES ON EMERGING ZOONOSES

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ABSTRACT. Zoonosis is the kind of infectious disease transmitting among different species by zoonotic pathogens. Different species play different roles in zoonoses. In this paper, we established a basic model to describe the zoonotic pathogen transmission from wildlife, to domestic animals, to humans. Then we put three strategies into the basic model to control the emerging zoonoses. Three strategies are corresponding to control measures of isolation, slaughter or similar in wildlife, domestic animals and humans respectively. We analyzed the effects of these three strategies on control reproductive numbers and equilibriums and we took avian influenza epidemic in China as an example to show the impacts of the strategies on emerging zoonoses in different areas at beginning.

1. Introduction. In human history, over 70% of the emerging infectious diseases are zoonoses, which mainly originate from animal reservoirs. Zoonotic pathogens can transmit from animals to humans. And about 75% of these zoonotic pathogens originate from wildlife [28, 3, 24]. Wildlife, domestic animals and humans construct the network of pathogen transmission crossing the species barrier. Wildlife and domestic animals play important roles in the transmission of zoonotic pathogens, in spite of the fact that we always neglected them before a zoonosis emerging or reemerging [12, 4].

No matter how well the science and technology developed in human society, human is just one kind of animals, even though other animals are not equal to humans in living status. The existence of the humans has changed the relationship between humans and animals due to some anthropogenic factors. Humans domesticated wolf, which was the ancestor of dog, for hunting about tens of thousands of years ago. Later, the intimacy between humans and dogs was increased more and more by natural selection or human selection, to be precise. In the meantime rabies virus existed permanently in human life by dog-human interface maintaining, as dogs were the mainly natural reservoirs of them, especially in Asia[24, 32].

Animals are divided into wildlife and domestic animals by human selection [24]. Humans can manage domestic animals in their entire life, but they cannot control wildlife at liberty. At the same time, humans can contact with domestic animals sufficiently, but they have few opportunities to get in touch with wildlife except for some special professions, such as forest conservationists and poachers. As wildlife

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and domestic animals play different roles in human life, the zoonotic pathogen transmissions in wildlife infection, domestic animal infection and human infection would be in different styles [17, 27]. Various mathematical models have been established in the study of zoonoses [26, 1, 16, 31, 11, 29]. For example, Doctor Saenz and his partners discussed the impact of domestic animal-human interface in pathogen transmission [26] and Doctor Allen constructed several types of mathematical models to reflect the pathogen transmission in wildlife [1].

For pathogen transmission in multiple species, the multi-SIR model can be established as the form [1, 17]:

$$\begin{cases} \dot{S}_{i} = A_{i} - \sum_{j=1}^{n} \beta_{ji} I_{j} S_{i} - \mu_{i} S_{i}, \\ \dot{I}_{i} = \sum_{j=1}^{n} \beta_{ji} I_{j} S_{i} - \mu_{i} I_{i} - \gamma_{i} I_{i} - \alpha_{i} I_{i}, \\ \dot{R}_{i} = \gamma_{i} I_{i} - \mu_{i} R_{i}. \end{cases}$$
(1)

 S_i , I_i and R_i represent the number of susceptibles, infectives, and recovered individuals for species $i, i=1,2,\dots,n$. A_i is the birth or immigration rate for species i. μ_i is the natural mortality rate. α_i is the disease-induced mortality rate. γ_i is the recovery rate. And β_{ji} is the per capita incidence rate from species j to species i, which denotes the probability of I_j infecting S_i .

The basic reproduction number R_0 for model (1) is the spectral radius of $[R_{0(ji)}]_{n \times n}$, where $R_{0(ji)} = \frac{A_i \beta_{ji}}{\mu_i (\mu_i + \gamma_i + \alpha_i)}$ [10]. For $\beta_{ji} \neq 0, \forall j, i$, it is difficult to get R_0 clearly. So we can take some biological characteristics of wildlife, domestic animals and humans into account to limit the value of β_{ji} in order to simplify $[R_{0(ji)}]_{n \times n}$.

For wildlife, they are always the origin of animal-borne zoonoses [24, 12]. The pathogen transmission from wildlife to humans is often neglected due to geographic distance between them, but the globalization and urbanization has shortened this distance. The linkage between wildlife and humans is established with anthropogenic land expanding^[24]. And pathogens parasitized in different species could be transmitted to others crossing species barrier by this linkage. But for emerging zoonoses, wildlife play as the only role of natural reservoirs. The pathogen transmission from domestic animals to wildlife or from humans to wildlife could not cause emerging zoonoses. Because the pathogens parasitized in humans or domestic animals have already existed for a period of time, which could be not defined as an emerging event even if the pathogens might transmit back to humans. For example, Severe Acute Respiratory Syndromes (SARS) is defined as an emerging zoonosis, which originate from Rhinolophus, then transmit via palm civets as intermediate host to humans [8]. But for mycobacterium tuberculosis, taking humans as their reservoirs, it could not give rise to an emerging zoonosis even if it had opportunities to infect other animals [20].

That is to say, for wildlife, the zoonotic pathogens could transmit in them, $\beta_{WW} \neq 0$, but $\beta_{HW} = 0$ and $\beta_{DW} = 0$. Here we classify the hosts into three groups: wildlife, domestic animals and humans. And notation W presents wildlife, D presents domestic animals and H presents humans. In order to simplify the model further, we assume that the zoonotic pathogen transmission could not occur from humans to domestic animals in emerging event. The need of infected people is to have a rest, but not to take care of other animals [24, 26]. We assume that if an emerging zoonosis was prevalent in human life, people could be infected from other people

without the need of passing by domestic animals, then to humans. So $\beta_{HD} = 0$, but $\beta_{DD} \neq 0$ and $\beta_{WD} \neq 0$.

For the relationship between animals and humans, we assume that not all of people could have opportunities to be infected from animals. Live animals are the mainly origin of zoonotic pathogens and only part of people could contact with them including CAFO (Confined Animal Feeding Operation) workers and hunters [26]. We also take the human population heterogeneity into consideration in this paper. The human population is classified into two groups: high risk group and low risk group. High risk group has the opportunities to contact with infected animals sufficiently. But low risk group are the others. That is, high risk group can get pathogens from animals and humans, but low risk group from humans only. The emerging zoonotic pathogen transmission can be described in FIGURE 1.



FIGURE 1. Emerging zoonotic pathogen transmission from wildlife, to domestic animals, to humans

Emerging zoonotic pathogen transmission from wildlife, to domestic animals, to humans can be described as the model (2).

 S_i , I_i and R_i represent the number of susceptibles, infectives, and recovered individuals for wildlife with i = W, domestic animals with i = D, high risk group with i = HH and low risk group with i = LH. At is the birth or immigration rate for species i = W, D, HH or LH. μ_i , γ_i , and β_{ji} are defined as the same as model (1) with i = W, D, or H. Here we assume that recovered individuals could be immune in a period time when a novel zoonosis is emerging.

$$\begin{split} S_{W} &= A_{W} - \beta_{WW} I_{W} S_{W} - \mu_{W} S_{W}, \\ \dot{I}_{W} &= \beta_{WW} I_{W} S_{W} - (\mu_{W} + \gamma_{W} + \alpha_{W}) I_{W}, \\ \dot{R}_{W} &= \gamma_{W} I_{W} - \mu_{W} R_{W}, \\ \dot{S}_{D} &= A_{D} - (\beta_{WD} I_{W} + \beta_{DD} I_{D}) S_{D} - \mu_{D} S_{D}, \\ \dot{I}_{D} &= (\beta_{WD} I_{W} + \beta_{DD} I_{D}) S_{D} - (\gamma_{D} + \alpha_{D} + \mu_{D}) I_{D}, \\ \dot{R}_{D} &= \gamma_{D} I_{D} - \mu_{D} R_{D}, \\ \dot{S}_{HH} &= A_{HH} - [\beta_{WH} I_{W} + \beta_{DH} I_{D} + \beta_{HH} (I_{HH} + I_{LH})] S_{HH} - \mu_{H} S_{HH}, \quad (2) \\ \dot{I}_{HH} &= [\beta_{WH} I_{W} + \beta_{DH} I_{D} + \beta_{HH} (I_{HH} + I_{LH})] S_{HH} \\ &- (\gamma_{H} + \alpha_{H} + \mu_{H}) I_{HH}, \\ \dot{R}_{HH} &= \gamma_{H} I_{HH} - \mu_{H} R_{HH}, \\ \dot{S}_{LH} &= A_{LH} - \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - \mu_{H} S_{LH}, \\ \dot{I}_{LH} &= \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - (\gamma_{H} + \alpha_{H} + \mu_{H}) I_{LH}, \\ \dot{R}_{LH} &= \gamma_{H} I_{LH} - \mu_{H} R_{LH}. \end{split}$$

The basic model has been established to reflect the pathogen transmission from wildlife, to domestic animals, to humans as model (2). Next step, we take the isolation and slaughter strategies into consideration [22, 23, 8, 31, 2, 25, 18]. For wildlife, it is difficult to control them when a zoonosis is emerging. Lethal control, vaccination and fencing (physical barriers) are the primary approaches to limit the number of susceptibles in wildlife. In this paper, we take lethal control and fencing (physical barriers) as the strategies to compare the similar isolation and slaughter strategies in emerging zoonotic pathogen transmission.

$$\dot{S}_{W} = A_{W} - \beta_{WW} I_{W} S_{W} - (\mu_{W} + \delta_{S}) S_{W}, \dot{I}_{W} = \beta_{WW} I_{W} S_{W} - (\mu_{W} + \gamma_{W} + \alpha_{W} + \delta_{I}) I_{W}, \dot{R}_{W} = \gamma_{W} I_{W} - (\mu_{W} + \delta_{R}) R_{W}, \dot{S}_{D} = A_{D} - ((1 - \theta_{D}) \beta_{WD} I_{W} + \beta_{DD} I_{D}) S_{D} - \mu_{D} S_{D}, \dot{I}_{D} = ((1 - \theta_{D}) \beta_{WD} I_{W} + \beta_{DD} I_{D}) S_{D} - (\gamma_{D} + \alpha_{D} + \mu_{D}) I_{D}, \dot{R}_{D} = \gamma_{D} I_{D} - \mu_{D} R_{D}, \dot{S}_{HH} = A_{HH} - [(1 - \theta_{H}) \beta_{WH} I_{W} + \beta_{DH} I_{D} + \beta_{HH} (I_{HH} + I_{LH})] S_{HH} - \mu_{H} S_{HH}, \dot{I}_{HH} = [(1 - \theta_{H}) \beta_{WH} I_{W} + \beta_{DH} I_{D} + \beta_{HH} (I_{HH} + I_{LH})] S_{HH} - (\gamma_{H} + \alpha_{H} + \mu_{H}) I_{HH}, \dot{R}_{HH} = \gamma_{H} I_{HH} - \mu_{H} R_{HH}, \dot{S}_{LH} = A_{LH} - \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - \mu_{H} S_{LH}, \dot{I}_{LH} = \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - (\gamma_{H} + \alpha_{H} + \mu_{H}) I_{LH}, \dot{R}_{LH} = \gamma_{H} I_{LH} - \mu_{H} R_{LH}.$$

$$(3)$$

 δ_s , δ_I and δ_R represent lethal control or slaughter rate of susceptibles, infectives, and recovered individuals in wildlife. θ_D , θ_H represent effectiveness of fencing (physical barriers), θ_D , $\theta_H \in [0, 1]$. If $\theta_D = 1$, $\theta_H = 1$, fencing plays the best role

in the control of emerging zoonoses. If $\theta_D = 0$, $\theta_H = 0$, fencing is useless in the control of emerging zoonoses.

For domestic animals, we can manage them in their entire lives. It is no need to slaughter all of the susceptibles in domestic animals. We can quarantine all of the domestic animals, then isolate susceptibles and slaughter infectives.

$$\begin{split} S_{D} &= A_{D} - (\beta_{WD}I_{W} + \beta_{DD}I_{D})S_{D} - \mu_{D}S_{D}, \\ \dot{I}_{D} &= (\beta_{WD}I_{W} + \beta_{DD}I_{D})S_{D} - (\gamma_{D} + \alpha_{D} + \mu_{D} + \Delta_{I})I_{D}, \\ \dot{R}_{D} &= \gamma_{D}I_{D} - \mu_{D}R_{D}, \\ \dot{S}_{HH} &= A_{HH} - [\beta_{WH}I_{W} + (1 - \Theta_{H})\beta_{DH}I_{D} + \beta_{HH}(I_{HH} + I_{LH})]S_{HH} \\ &- \mu_{H}S_{HH}, \\ \dot{I}_{HH} &= [\beta_{WH}I_{W} + (1 - \Theta_{H})\beta_{DH}I_{D} + \beta_{HH}(I_{HH} + I_{LH})]S_{HH} \\ &- (\gamma_{H} + \alpha_{H} + \mu_{H})I_{HH}, \\ \dot{R}_{HH} &= \gamma_{H}I_{HH} - \mu_{H}R_{HH}, \\ \dot{S}_{LH} &= A_{LH} - \beta_{HH}(I_{HH} + I_{LH})S_{LH} - \mu_{H}S_{LH}, \\ \dot{I}_{LH} &= \beta_{HH}(I_{HH} + I_{LH})S_{LH} - (\gamma_{H} + \alpha_{H} + \mu_{H})I_{LH}, \\ \dot{R}_{LH} &= \gamma_{H}I_{LH} - \mu_{H}R_{LH}. \end{split}$$

 Δ_I represents slaughter rate of infectives in domestic animals. Θ_H represents effectiveness of isolation, $\Theta_H \in [0, 1]$. If $\Theta_H = 1$, isolation from susceptible domestic animals play the best role in the control of emerging zoonoses. If $\Theta_H = 0$, isolation in domestic animals is useless in the control of emerging zoonoses.

For humans, we could not 'slaughter' anyone no matter how serious they were infected with some kind of zoonoses. The quarantine and isolation may be the best method to limit the pathogen transmission except for vaccination. But the effect of quarantine and isolation strategies in humans are different from animals. For taking isolation strategies in animals, it is the susceptible humans, who are afraid of getting infected, to take the initiative and get away from susceptible animals. So the per capita incidence rate from animals to humans, β_{WH} and β_{DH} , is decreased by θ_H and Θ_H . But in humans, we quarantine and isolate the infected people to cut off pathogen transmission way. β_{HH} would not change at this time, but there is a new compartment O produced, which denotes the isolation compartment [25].

$$\dot{S}_{HH} = A_{HH} - [\beta_{WH}I_W + \beta_{DH}I_D + \beta_{HH}(I_{HH} + I_{LH})]S_{HH} - \mu_H S_{HH} - \varphi(I)S_{HH} + \gamma_{H1}O_{HH1}, \dot{O}_{HH1} = \varphi(I)S_{HH} - \gamma_{H1}O_{HH1} - \mu_H O_{HH1}, \dot{I}_{HH} = [\beta_{WH}I_W + \beta_{DH}I_D + \beta_{HH}(I_{HH} + I_{LH})]S_{HH} - (\gamma_H + \alpha_H + \mu_H + \sigma)I_{HH}, \dot{O}_{HH2} = \sigma I_{HH} - \gamma_{H2}O_{HH2} - \mu_H O_{HH2}, \dot{R}_{HH} = \gamma_H I_{HH} + \gamma_{H2}O_{HH2} - \mu_H R_{HH},$$
(5)
 $\dot{S}_{LH} = A_{LH} - \beta_{HH}(I_{HH} + I_{LH})S_{LH} - \mu_H S_{LH} - \varphi(I)S_{LH} + \gamma_{H1}O_{LH1}, \dot{O}_{LH1} = \varphi(I)S_{LH} - \gamma_{H1}O_{LH1} - \mu_H O_{LH1}, \dot{I}_{LH} = \beta_{HH}(I_{HH} + I_{LH})S_{LH} - (\gamma_H + \alpha_H + \mu_H + \sigma)I_{LH}, \dot{O}_{LH2} = \sigma I_{LH} - \gamma_{H2}O_{LH2} - \mu_H O_{LH2}, \dot{R}_{LH} = \gamma_H I_{LH} + \gamma_{H2}O_{LH2} - \mu_H R_{LH}.$

 O_{HH1} and O_{HH2} represent isolation compartments from susceptibles and infectives in high risk group respectively. O_{LH1} and O_{LH2} represent isolation compartments from susceptibles and infectives in low risk group. Susceptibles enter the O_{HH1} and O_{LH1} classes at the rate of $\varphi(I)S_{HH}$ and $\varphi(I)S_{LH}$, with $\varphi(I) = \rho(I_{HH} + I_{LH})$ [23]. The infectives are isolated at the constant per-capita rate of σ . γ_{H1} is the remove rate from isolation compartment to susceptible compartment. γ_{H2} is the remove rate from isolation compartment to recovery individual compartment.

In conclusion, we can get the isolation and slaughter strategies controlling model by (3), (4) and (5) in wildlife, domestic animals and humans as the form:

$$\begin{split} \dot{S}_{W} &= A_{W} - \beta_{WW} I_{W} S_{W} - (\mu_{W} + \varepsilon_{W} \delta_{S}) S_{W}, \\ \dot{I}_{W} &= \beta_{WW} I_{W} S_{W} - (\mu_{W} + \gamma_{W} + \alpha_{W} + \varepsilon_{W} \delta_{I}) I_{W}, \\ \dot{R}_{W} &= \gamma_{W} I_{W} - (\mu_{W} + \varepsilon_{W} \delta_{R}) R_{W}, \\ \dot{S}_{D} &= A_{D} - ((1 - \varepsilon_{W} \theta_{D}) \beta_{WD} I_{W} + \beta_{DD} I_{D}) S_{D} - \mu_{D} S_{D}, \\ \dot{I}_{D} &= ((1 - \varepsilon_{W} \theta_{D}) \beta_{WD} I_{W} + \beta_{DD} I_{D}) S_{D} \\ - (\gamma_{D} + \alpha_{D} + \mu_{D} + \varepsilon_{D} \Delta_{I}) I_{D}, \\ \dot{R}_{D} &= \gamma_{D} I_{D} - \mu_{D} R_{D}, \\ \dot{S}_{HH} &= A_{HH} - [(1 - \varepsilon_{W} \theta_{H}) \beta_{WH} I_{W} + (1 - \varepsilon_{D} \Theta_{H}) \beta_{DH} I_{D} + \beta_{HH} (I_{HH} \\ + I_{LH})] S_{HH} - \mu_{H} S_{HH} - \varepsilon_{H} \varphi (I) S_{HH} + \gamma_{H1} O_{HH1}, \\ \dot{O}_{HH1} &= \varepsilon_{H} \varphi (I) S_{HH} - \gamma_{H1} O_{HH1} - \mu_{H} O_{HH1}, \\ \dot{I}_{HH} &= [(1 - \varepsilon_{W} \theta_{H}) \beta_{WH} I_{W} + (1 - \varepsilon_{D} \Theta_{H}) \beta_{DH} I_{D} \\ + \beta_{HH} (I_{HH} + I_{LH})] S_{HH} - (\gamma_{H} + \alpha_{H} + \mu_{H} + \varepsilon_{H} \sigma) I_{HH}, \\ \dot{O}_{HH2} &= \varepsilon_{H} \sigma I_{HH} - \gamma_{H2} O_{HH2} - \mu_{H} O_{HH2}, \\ \dot{R}_{HH} &= \gamma_{H} I_{HH} + \gamma_{H2} O_{HH2} - \mu_{H} R_{HH}, \\ \dot{S}_{LH} &= A_{LH} - \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - \mu_{H} S_{LH} \\ - \varepsilon_{H} \varphi (I) S_{LH} - \gamma_{H1} O_{LH1} - \mu_{H} O_{LH1}, \\ \dot{O}_{LH1} &= \varepsilon_{H} \sigma I_{LH} - \gamma_{H2} O_{LH2} - \mu_{H} O_{LH2}, \\ \dot{R}_{LH} &= \gamma_{H} I_{LH} + \gamma_{H2} O_{LH2} - \mu_{H} R_{LH}. \end{split}$$

with Strategy 1,

$$\begin{cases} \varepsilon_W = 0, I_{LH} + I_{HH} < I_{WC} ,\\ \varepsilon_W = 1, I_{LH} + I_{HH} \ge I_{WC} \end{cases}$$
(7)

Strategy 2,

$$\begin{cases} \varepsilon_D = 0, I_{LH} + I_{HH} < I_{DC} ,\\ \varepsilon_D = 1, I_{LH} + I_{HH} \ge I_{DC} \end{cases}$$

$$\tag{8}$$

Strategy 3,

$$\begin{cases} \varepsilon_H = 0, I_{LH} + I_{HH} < I_{HC} ,\\ \varepsilon_H = 1, I_{LH} + I_{HH} \ge I_{HC} \end{cases}$$

$$\tag{9}$$

 $\begin{array}{l} \text{The feasible set} \Omega \ = \ \left\{ \left(S_W \left(t \right), \ I_W \left(t \right), \ R_W \left(t \right), \ S_D \left(t \right), \ I_D \left(t \right), \ R_D \left(t \right), S_{HH} \left(t \right), \\ O_{HH1} \left(t \right), \ I_{HH} \left(t \right), \ O_{HH2} \left(t \right), \ R_{HH} \left(t \right), S_{LH} \left(t \right), \ O_{LH1} \left(t \right), \ I_{LH} \left(t \right), \ O_{LH2} \left(t \right), \\ R_{LH} \left(t \right) \right) \ \left| \ S_i (t), I_i (t), R_i (t), O_j (t) \ \ge \ 0, 0 \ < \ N \ \le \ \frac{A_W}{\mu_W} + \frac{A_D}{\mu_D} + \frac{A_{HH}}{\mu_H} + \frac{A_{LH}}{\mu_H}, i \ = \ N \right\}$

W, D, HH, LH, j = HH1, HH2, LH1, LH2 is the positively invariant with respect to (6).

Total number of wildlife is $N_W = S_W + I_W + R_W$. Total number of domestic animals is $N_D = S_D + I_D + R_D$. Total number of humans is $N_H = S_{HH} + O_{HH1} + I_{HH} + O_{HH2} + R_{HH} + S_{LH} + O_{LH1} + I_{LH} + O_{LH2} + R_{LH}$. Total number of susceptible humans is $S_H = S_{HH} + S_{LH}$. Total number of infective humans is $I_H = I_{HH} + I_{LH}$. Total number of recovery humans is $R_H = R_{HH} + R_{LH}$. Total number of all is $N = N_W + N_D + N_H$.

It is difficult for us to take any strategies to control emerging zoonoses in first time. Only the infected of numbers of people would cause our attention to take some strategies to control the infectious disease. So it is assumed that if the number of infectives in human including high risk group and low risk group reached a threshold at I_{WC} , I_{DC} or I_{HC} , we would take measures as Strategy 1 in wildlife, Strategy 2 in domestic animals or Strategy 3 in humans.

2. Stability analysis. With $\varepsilon_W = 0$, $\varepsilon_D = 0$, $\varepsilon_H = 0$, we can get (2) before taking any measures in emerging zoonoses. At first, we analyze the equilibrium stability of model (2) [5, 14, 19, 16].

In (2), the wildlife class can be separated as

$$\begin{cases} \dot{S}_W = A_W - \beta_{WW} I_W S_W - \mu_W S_W, \\ \dot{I}_W = \beta_{WW} I_W S_W - \mu_W I_W - \gamma_W I_W - \alpha_W I_W. \end{cases}$$
(10)

We can get the basic reproductive number in wildlife $R_{0(WW)} = \frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)}$. The disease-free equilibrium is $E_{0(W)} = (\overline{S_W}, \overline{I_W}) = (\frac{A_W}{\mu_W}, 0)$, the epidemic equilibrium is $E^*_{(w)} = (\hat{S}_W, \hat{I}_W) = (\frac{\mu_W + \gamma_W + \alpha_W}{\beta_{WW}}, \frac{\mu_W}{\beta_{WW}} (\frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} - 1))$. The number of recovery individuals would not change the stability of the system, so we neglect it in the study of equilibriums.

Theorem 2.1. If $R_{0(WW)} < 1$, the disease-free equilibrium $E_{0(W)}$ in wildlife is stable. If $R_{0(WW)} > 1$, the epidemic equilibrium $E^*_{(W)}$ in wildlife is stable.

Proof. The next generation matrix of the vector field corresponding to system (10) at $E_{0(W)}$ is

$$J_W(E_{0(W)}) = \begin{pmatrix} -\mu_W & -\beta_{WW} \frac{A_W}{\mu_W} \\ 0 & \beta_{WW} \frac{A_W}{\mu_W} - \mu_W - \gamma_W - \alpha_W \end{pmatrix}$$

If $R_{0(WW)} = \frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} < 1$, the eigenvalues of $J_W(E_{0(W)})$ are negative and $E_{0(W)}$ is stable.

Similarly, the next generation matrix at $E_{(W)}^*$ is

$$J_W(E^*_{(W)}) = \begin{pmatrix} -\beta_{WW}\hat{I}_W - \mu_W & -\beta_{WW}\hat{S}_W \\ \beta_{WW}\hat{I}_W & \beta_{WW}\hat{S}_W - \mu_W - \gamma_W - \alpha_W \end{pmatrix}$$

The characteristic equation of $J_W(E^*_{(W)})$ is

$$f_W(\lambda) = \lambda^2 + \frac{A_W \beta_{WW}}{\mu_W + \gamma_W + \alpha_W} \lambda + \mu_W (\mu_W + \gamma_W + \alpha_W) (\frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} - 1)$$

= 0

If $R_{0(WW)} = \frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} > 1$, the real parts of eigenvalues of $J_W(E^*_{(W)})$ are negative and $E^*_{(W)}$ is stable.

In (2), the wildlife and domestic animals classes can be separated as

$$\begin{cases} \dot{S}_W = A_W - \beta_{WW} I_W S_W - \mu_W S_W, \\ \dot{I}_W = \beta_{WW} I_W S_W - \mu_W I_W - \gamma_W I_W - \alpha_W I_W, \\ \dot{S}_D = A_D - \beta_{WD} I_W S_D - \beta_{DD} I_D S_D - \mu_D S_D, \\ \dot{I}_D = \beta_{WD} I_W S_D + \beta_{DD} I_D S_D - \gamma_D I_D - \alpha_D I_D - \mu_D I_D. \end{cases}$$
(11)

We can get the basic reproductive number in domestic animals is $R_{0(DD)} = \frac{A_D \beta_{DD}}{\mu_D (\mu_D + \gamma_D + \alpha_D)}$.

The disease-free equilibrium is $E_{0(WD)} = (\overline{S_W}, \overline{I_W}, \overline{S_D}, \overline{I_D}) = (\frac{A_W}{\mu_W}, 0, \frac{A_D}{\mu_D}, 0)$, the epidemic equilibrium is

$$E^{*}_{(WD)} = (\hat{S}_{W}, \hat{I}_{W}, \hat{S}_{D}, \hat{I}_{D}) = (\frac{\mu_{W} + \gamma_{W} + \alpha_{W}}{\beta_{WW}}, \frac{\mu_{W}}{\beta_{WW}} (\frac{A_{W}\beta_{WW}}{\mu_{W}(\mu_{W} + \gamma_{W} + \alpha_{W})} - 1), \hat{S}_{D}, \hat{I}_{D})$$

Theorem 2.2. If $R_{0(WW)} < 1$ and $R_{0(DD)} < 1$, the disease-free equilibrium $E_{0(WD)}$ in system (11) is stable. If $R_{0(WW)} > 1$, there exists one unique positive epidemic equilibrium $E^*_{(WD)}$, and $E^*_{(WD)}$ is stable.

Proof. There always exists $E_{0(WD)}$ and the next generation matrix at $E_{0(WD)}$ is

$$J_{WD}(E_{0(WD)}) = \begin{pmatrix} J_W(E_{0(W)}) & 0 \\ * & J_D(E_{0(D)}) \end{pmatrix}$$

with

$$J_D(E_{0(D)}) = \begin{pmatrix} -\mu_D & -\beta_{DD}\frac{A_D}{\mu_D} \\ 0 & \beta_{DD}\frac{A_D}{\mu_D} - \mu_D - \gamma_D - \alpha_D \end{pmatrix}$$

If $R_{0(WW)} = \frac{A_W \beta_{WW}}{\mu_W(\mu_W + \gamma_W + \alpha_W)} < 1$ and $R_{0(DD)} = \frac{A_D \beta_{DD}}{\mu_D(\mu_D + \gamma_D + \alpha_D)} < 1$, the eigenvalues of $J_{WD}(E_{0(WD)})$ are negative and $E_{0(WD)}$ is stable.

In (11), the epidemic equilibrium $E^*_{(WD)} = (\hat{S}_W, \hat{I}_W, \hat{S}_D, \hat{I}_D)$ satisfies

$$A_W - \beta_{WW} \hat{I}_W \hat{S}_W - \mu_W \hat{S}_W = 0 \tag{12}$$

$$\beta_{WW}\hat{I}_W\hat{S}_W - \mu_W\hat{I}_W - \gamma_W\hat{I}_W - \alpha_W\hat{I}_W = 0 \tag{13}$$

$$A_D - \beta_{WD} \hat{I}_W \hat{S}_D - \beta_{DD} \hat{I}_D \hat{S}_D - \mu_D \hat{S}_D = 0$$
(14)

$$\beta_{WD}\hat{I}_W\hat{S}_D + \beta_{DD}\hat{I}_D\hat{S}_D - \mu_D\hat{I}_D - \gamma_D\hat{I}_D - \alpha_D\hat{I}_D = 0$$
(15)

From (12), (13), (14), (15), we can get

$$\hat{S}_W = \frac{\mu_W + \gamma_W + \alpha_W}{\beta_{WW}}$$
$$\hat{I}_W = \frac{\mu_W}{\beta_{WW}} \left(\frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} - 1\right)$$
$$\hat{I}_D = \frac{\beta_{WD} \hat{I}_W \hat{S}_D}{\mu_D + \gamma_D + \alpha_D - \beta_{DD} \hat{S}_D}$$
$$= \frac{\beta_{WD} \hat{S}_D}{\mu_D + \gamma_D + \alpha_D - \beta_{DD} \hat{S}_D} \times \frac{\mu_W}{\beta_{WW}} \left(\frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} - 1\right)$$

and

$$\hat{S}_{D} = \frac{1}{2\mu_{D}\beta_{DD}} [A_{D}\beta_{DD} + (\mu_{D} + \gamma_{D} + \alpha_{D})(\mu_{D} + \beta_{WD}\hat{I}_{W})] - \frac{1}{2\mu_{D}\beta_{DD}} [A_{D}^{2}\beta_{DD}^{2} + 2A_{D}\beta_{DD}(\mu_{D} + \gamma_{D} + \alpha_{D})(\beta_{WD}\hat{I}_{W} - \mu_{D}) + (\mu_{D} + \gamma_{D} + \alpha_{D})^{2}(\mu_{D} + \beta_{WD}\hat{I}_{W})^{2}]^{\frac{1}{2}}$$

So if $R_{0(WW)} > 1$, there exists one unique positive epidemic equilibrium $E^*_{(WD)}$. In fact, for \hat{S}_D , \hat{S}_D satisfies

$$g_1(\hat{S}_D) = \mu_D \beta_{DD} \hat{S}_D^2 - \left[A_D \beta_{DD} + (\mu_D + \gamma_D + \alpha_D)(\mu_D + \beta_{WD} \hat{I}_W) \right] \hat{S}_D + A_D(\mu_D + \gamma_D + \alpha_D) = 0$$

If $g_1(\hat{S}_D) = 0$, \hat{S}_D always has two positive roots because of $-A_D\beta_{DD} + (\mu_D + \mu_D)$ $(\gamma_D + \alpha_D)(\mu_D + \beta_{WD}\hat{I}_W) < 0$, $A_D(\mu_D + \gamma_D + \alpha_D) > 0$, and the smaller \hat{S}_D and the bigger \hat{S}_D stand on both sides of $\frac{\mu_D + \gamma_D + \alpha_D}{\beta_{DD}}$ for $g_1(\frac{\mu_D + \gamma_D + \alpha_D}{\beta_{DD}}) < 0$ and $\mu_D\beta_{DD} > 0.$

So we choose

$$\hat{S}_{D} = \frac{1}{2\mu_{D}\beta_{DD}} [A_{D}\beta_{DD} + (\mu_{D} + \gamma_{D} + \alpha_{D})(\mu_{D} + \beta_{WD}\hat{I}_{W})] - \frac{1}{2\mu_{D}\beta_{DD}} [A_{D}^{2}\beta_{DD}^{2} + 2A_{D}\beta_{DD}(\mu_{D} + \gamma_{D} + \alpha_{D})(\beta_{WD}\hat{I}_{W} - \mu_{D}) + (\mu_{D} + \gamma_{D} + \alpha_{D})^{2}(\mu_{D} + \beta_{WD}\hat{I}_{W})^{2}]^{\frac{1}{2}}$$

to guarantee $\hat{I}_D > 0$.

The next generation matrix at $E^*_{(WD)}$ is

$$J_{WD}(E^*_{(WD)}) = \begin{pmatrix} J_W(E^*_{(W)}) & 0\\ & & \\ & * & J_D(E^*_{(D)}) \end{pmatrix}$$

with

$$J_D(E^*_{(D)}) = \begin{pmatrix} -\beta_{WD}\hat{I}_W - \beta_{DD}\hat{I}_D - \mu_D & -\beta_{DD}\hat{S}_D \\ \beta_{WD}\hat{I}_W + \beta_{DD}\hat{I}_D & \beta_{DD}\hat{S}_D - \mu_D - \gamma_D - \alpha_D \end{pmatrix}$$

The characteristic equation of $J_D(E^*_{(D)})$ is $f_D(\lambda) = \lambda^2 + (\frac{A_D}{\hat{S}_D} + \frac{\beta_{WD}\hat{I}_W\hat{S}_D}{\hat{I}_D})\lambda + \frac{\beta_{WD}\hat{I}_WA_D}{\hat{I}_D} + \beta_{WD}\beta_{DD}\hat{S}_D\hat{I}_W + \beta_{DD}^2\hat{S}_D\hat{I}_D = 0$.

If
$$E_{(D)}^*$$
 exists, all of the real parts of eigenvalues of $J_D(E_{(D)}^*)$ are negative for $\frac{A_D}{\hat{S}_D} + \frac{\beta_{WD}\hat{I}_W\hat{S}_D}{\hat{I}_D} > 0$ and $\frac{\beta_{WD}\hat{I}_WA_D}{\hat{I}_D} + \beta_{WD}\beta_{DD}\hat{S}_D\hat{I}_W + \beta_{DD}^2\hat{S}_D\hat{I}_D > 0$.
In conclusion, if $R_{0(WW)} = \frac{A_W\beta_{WW}}{\mu_W(\mu_W+\gamma_W+\alpha_W)} > 1$, the real parts of eigenvalues of $J_D(E_{(D)}^*)$ are negative and $E_{(WD)}^*$ is stable.

of $J_D(E^*_{(D)})$ are negative and $E^*_{(WD)}$ is stable.

The human class in (2) can be separated as the form:

$$\dot{S}_{HH} = A_{HH} - \beta_{WH}I_WS_{HH} - \beta_{DH}I_DS_{HH} - \beta_{HH}(I_{HH} + I_{LH})S_{HH}
-\mu_HS_{HH},
\dot{I}_{HH} = \beta_{WH}I_WS_{HH} + \beta_{DH}I_DS_{HH} + \beta_{HH}(I_{HH} + I_{LH})S_{HH} - \gamma_HI_{HH}
-\alpha_HI_{HH} - \mu_HI_{HH},
\dot{S}_{LH} = A_{LH} - \beta_{HH}(I_{HH} + I_{LH})S_{LH} - \mu_HS_{LH},
\dot{I}_{LH} = \beta_{HH}(I_{HH} + I_{LH})S_{LH} - \gamma_HI_{LH} - \alpha_HI_{LH} - \mu_HI_{LH}.$$
(16)

There always exists disease-free equilibrium $E_{0(WDH)} = (\overline{S_W}, \overline{I_W}, \overline{S_D}, \overline{I_D}, \overline{S_{HH}}, \overline{S_{HH}})$ $\overline{I_{HH}}, \overline{S_{LH}}, \overline{I_{LH}}) = \left(\frac{A_W}{\mu_W}, 0, \frac{A_D}{\mu_D}, 0, \frac{A_{HH}}{\mu_{HH}}, 0, \frac{A_{LH}}{\mu_{LH}}\right) \text{ in (2).}$ The next generation matrix at $E_{0(WDH)}$ is

$$J_{WDH}(E_{0(WDH)}) = \begin{pmatrix} J_W(E_{0(W)}) & 0 & 0 \\ * & J_D(E_{0(D)}) & 0 \\ * & * & J_H(E_{0(H)}) \end{pmatrix}$$

with

$$J_H(E_{0(H)}) =$$

$$\begin{pmatrix} -\mu_{H} & -\beta_{HH} \frac{A_{HH}}{\mu_{H}} & 0 & -\beta_{HH} \frac{A_{HH}}{\mu_{H}} \\ 0 & \beta_{HH} \frac{A_{HH}}{\mu_{H}} - \mu_{H} - \gamma_{H} - \alpha_{H} & 0 & \beta_{HH} \frac{A_{HH}}{\mu_{H}} \\ 0 & -\beta_{HH} \frac{A_{LH}}{\mu_{H}} & -\mu_{H} & -\beta_{HH} \frac{A_{LH}}{\mu_{H}} \\ 0 & \beta_{HH} \frac{A_{LH}}{\mu_{H}} & 0 & \beta_{HH} \frac{A_{LH}}{\mu_{H}} - \mu_{H} - \gamma_{H} - \alpha_{H} \end{pmatrix}$$

The characteristic equation of $J_H(E_{0(H)})$ is

The characteristic equation of $J_H(E_{0(H)})$ is $f_{H1}(\lambda) = (\lambda + \mu_H)^2 (\lambda^2 - (\beta_{HH} \frac{A_{HH} + A_{LH}}{\mu_H} - 2\mu_H - 2\gamma_H - 2\alpha_H)\lambda - (\mu_H + \gamma_H + \alpha_H)(\beta_{HH} \frac{A_{HH} + A_{LH}}{\mu_H} - \mu_H - \gamma_H - \alpha_H)) = 0$ If there is $R_{0(H)} = \frac{(A_{HH} + A_{LH})\beta_{HH}}{(\mu_H + \gamma_H + \alpha_H)} < 1$, we have $\beta_{HH} \frac{A_{HH} + A_{LH}}{\mu_H} - 2\mu_H - 2\gamma_H - 2\gamma_H - 2\alpha_H < 0$ and $(\mu_H + \gamma_H + \alpha_H)(\beta_{HH} \frac{A_{HH} + A_{LH}}{\mu_H} - \mu_H - \gamma_H - \alpha_H) < 0$. Then we get all of the real parts of eigenvalues of $J_H(E_{0(H)})$ are negative, $E_{0(H)}$ is stable. At the same time, the spectral radius of $\begin{bmatrix} R_{0(HHHH)} & R_{0(LHHH)} \\ R_{0(HHLH)} & R_{0(LHLH)} \end{bmatrix}$ is $R_{0(H)} = \frac{(A_{HH} + A_{LH})\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H)}$, with $R_{0(HHHH)} = R_{0(LHHH)} = \frac{A_{HH}\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H)}$ and $R_{0(LHLH)} = R_{0(HHLH)} = \frac{A_{LH}\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H)}$.

Theorem 2.3. If $R_{0(WW)} < 1$, $R_{0(DD)} < 1$ and $R_{0(H)} < 1$, the disease-free equilibrium $E_{0(WDH)}$ in system (2) is stable. If $R_{0(WW)} > 1$, there exists epidemic equilibrium $E^*_{(WDH)}$, and $E^*_{(WDH)}$ is stable.

Proof. The next generation matrix at $E_{0(WDH)}$ is

$$J_{WDH}(E_{0(WDH)}) = \begin{pmatrix} J_W(E_{0(W)}) & 0 & 0 \\ * & J_D(E_{0(D)}) & 0 \\ * & * & J_H(E_{0(H)}) \end{pmatrix}.$$

If $R_{0(WW)} < 1$, $R_{0(DD)} < 1$ and $R_{0(H)} < 1$, all of the real parts of eigenvalues of $J_{WDH}(E_{0(WDH)})$ are negative, then we get the disease-free equilibrium $E_{0(WDH)}$ in system (2) is stable (Theorem 2.1, Theorem 2.2).

Next we prove the existence of epidemic equilibrium $E^*_{(WDH)}$ and the stability of $E^*_{(WDH)}$.

In (16), the epidemic equilibrium $E^*_{(WDH)} = (\hat{S}_W, \hat{I}_W, \hat{S}_D, \hat{I}_D, \hat{S}_{HH}, \hat{I}_{HH}, \hat{S}_{LH},$ \hat{I}_{LH}). satisfies

$$A_{HH} - \beta_{WH} \hat{I}_W \hat{S}_{HH} - \beta_{DH} \hat{I}_D \hat{S}_{HH} - \beta_{HH} (\hat{I}_{HH} + \hat{I}_{LH}) \hat{S}_{HH} - \mu_H \hat{S}_{HH} = 0 \quad (17)$$

$$\beta_{WH}I_WS_{HH} + \beta_{DH}I_DS_{HH} + \beta_{HH}(I_{HH} + I_{LH})S_{HH} - \gamma_H I_{HH} - \alpha_H \hat{I}_{HH} - \mu_H \hat{I}_{HH} = 0$$

$$(18)$$

$$A_{LH} - \beta_{HH} (\hat{I}_{HH} + \hat{I}_{LH}) \hat{S}_{LH} - \mu_H \hat{S}_{LH} = 0$$
(19)

$$\beta_{HH}(\hat{I}_{HH} + \hat{I}_{LH})\hat{S}_{LH} - \gamma_H\hat{I}_{LH} - \alpha_H\hat{I}_{LH} - \mu_H\hat{I}_{LH} = 0$$
(19)
$$\beta_{HH}(\hat{I}_{HH} + \hat{I}_{LH})\hat{S}_{LH} - \gamma_H\hat{I}_{LH} - \alpha_H\hat{I}_{LH} - \mu_H\hat{I}_{LH} = 0$$
(20)

From (17) + (19), (18) + (20), we get

$$A_{HH} + A_{LH} - \beta_{WH} \hat{I}_W \hat{S}_{HH} - \beta_{DH} \hat{I}_D \hat{S}_{HH} - \beta_{HH} (\hat{I}_{HH} + \hat{I}_{LH}) (\hat{S}_{HH} + \hat{S}_{LH}) - \mu_H (\hat{S}_{HH} + \hat{S}_{LH}) = 0$$
(21)

$$\beta_{WH}\hat{I}_W\hat{S}_{HH} + \beta_{DH}\hat{I}_D\hat{S}_{HH} + \beta_{HH}(\hat{I}_{HH} + \hat{I}_{LH})(\hat{S}_{HH} + \hat{S}_{LH}) - (\gamma_H + \alpha_H + \mu_H)(\hat{I}_{HH} + \hat{I}_{LH}) = 0$$
(22)

It is assumed that $\eta_S = \frac{\hat{S}_{HH}}{\hat{S}_{HH} + \hat{S}_{LH}} \hat{S}_H = \hat{S}_{HH} + \hat{S}_{LH}$ and $\hat{I}_H = \hat{I}_{HH} + \hat{I}_{LH}$ with $\eta_S \in (0, 1).$

Then we have

$$A_{HH} + A_{LH} - \eta_S \beta_{WH} \hat{I}_W \hat{S}_H - \eta_S \beta_{DH} \hat{I}_D \hat{S}_H - \beta_{HH} \hat{I}_H \hat{S}_H - \mu_H \hat{S}_H = 0$$
(23)

 $\eta_S \beta_{WH} \hat{I}_W \hat{S}_H + \eta_S \beta_{DH} \hat{I}_D \hat{S}_H + \beta_{HH} \hat{I}_H \hat{S}_H - (\gamma_H + \alpha_H + \mu_H) \hat{I}_H = 0$ (24)From (23), (24), we can get

$$\hat{I}_H = \frac{\eta_S \beta_{WH} \hat{I}_W \hat{S}_H + \eta_S \beta_{DH} \hat{I}_D \hat{S}_H}{\gamma_H + \alpha_H + \mu_H - \beta_{HH} \hat{S}_H}$$

and

$$\hat{S}_{H} = \frac{1}{2\mu_{H}\beta_{HH}} [(A_{HH} + A_{LH})\beta_{HH} + (\gamma_{H} + \alpha_{H} + \mu_{H})(\mu_{H} + \eta_{S}\beta_{WH}\hat{I}_{W} + \eta_{S}\beta_{DH}\hat{I}_{D})] - \frac{1}{2\mu_{H}\beta_{HH}} [(A_{HH} + A_{LH})^{2}\beta_{HH}^{2} + 2(A_{HH} + A_{LH})\beta_{HH} (\gamma_{H} + \alpha_{H} + \mu_{H})(\eta_{S}\beta_{WH}\hat{I}_{W} + \eta_{S}\beta_{DH}\hat{I}_{D} - \mu_{H}) + (\gamma_{H} + \alpha_{H} + \mu_{H})^{2} (\mu_{H} + \eta_{S}\beta_{WH}\hat{I}_{W} + \eta_{S}\beta_{DH}\hat{I}_{D})^{2}]^{\frac{1}{2}}$$

Similarly to the calculation of Theorem 2.2, we have

$$g_{2}(\hat{S}_{H}) = \mu_{H}\beta_{HH}\hat{S}_{H}^{2} + (A_{HH} + A_{LH})(\mu_{H} + \gamma_{H} + \alpha_{H}) - [(A_{HH} + A_{LH})\beta_{HH} + (\mu_{H} + \gamma_{H} + \alpha_{H})(\mu_{H} + \eta_{S}\beta_{WH}\hat{I}_{W} + \eta_{S}\beta_{DH}\hat{I}_{D})]\hat{S}_{H} = 0$$

So if $R_{0(WW)} > 1$, there exists epidemic equilibrium $E^*_{(WDH)}$ in (2). The next generation matrix at $E^*_{(WDH)}$ is

$$J_{WDH}(E^*_{(WDH)}) = \begin{pmatrix} J_W(E^*_{(W)}) & 0 & 0 \\ * & J_D(E^*_{(D)}) & 0 \\ * & * & J_H(E^*_{(H)}) \end{pmatrix}$$

with

$$J_{H}(E_{(H)}^{*}) = \begin{pmatrix} J_{11} & -\beta_{HH}\hat{S}_{HH} & 0 & -\beta_{HH}\hat{S}_{HH} \\ J_{21} & J_{22} & 0 & \beta_{HH}\hat{S}_{HH} \\ 0 & -\beta_{HH}\hat{S}_{LH} & J_{33} & -\beta_{HH}\hat{S}_{LH} \\ 0 & \beta_{HH}\hat{S}_{LH} & \beta_{HH}(\hat{I}_{HH} + \hat{I}_{LH}) & J_{44} \end{pmatrix}$$

$$J_{11} = -\beta_{WH}\hat{I}_{W} - \beta_{DH}\hat{I}_{D} - \beta_{HH}(\hat{I}_{LH} + \hat{I}_{LH}) - \mu_{H}$$

$$J_{21} = \beta_{WH}\hat{I}_{W} + \beta_{DH}\hat{I}_{D} + \beta_{HH}(\hat{I}_{LH} + \hat{I}_{LH})$$

$$J_{22} = \beta_{HH}\hat{S}_{HH} - \gamma_{H} - \alpha_{H} - \mu_{H}$$

$$J_{33} = -\beta_{HH}(\hat{I}_{HH} + \hat{I}_{LH}) - \mu_{H}$$

$$J_{44} = \beta_{HH}\hat{S}_{LH} - \gamma_{H} - \alpha_{H} - \mu_{H}$$
The characteristic equation of $J_{H}(E_{(H)}^{*})$ is

 $f_{H2}(\lambda) = \lambda^4 + a_1 \lambda^3 + a_2 \lambda^2 + a_3 \lambda + a_4 = 0$

with

$$\begin{split} a_{1} &= \frac{A_{HH}}{\hat{S}_{HH}} + \frac{A_{LH}}{\hat{S}_{LH}} - \beta_{HH}(\hat{S}_{HH} + \hat{S}_{LH}) + 2(\gamma_{H} + \alpha_{H} + \mu_{H}), \\ a_{2} &= (\frac{A_{HH}}{\hat{S}_{HH}} + \frac{A_{LH}}{\hat{S}_{LH}})(\gamma_{H} + \alpha_{H} + \mu_{H}) - \mu_{H}\beta_{HH}(\hat{S}_{HH} + \hat{S}_{LH}) \\ &- \beta_{HH}^{2}\hat{S}_{HH}\hat{S}_{LH} + (\frac{A_{HH}}{\hat{S}_{HH}} + \beta_{WH}\frac{\hat{I}_{W}}{\hat{I}_{HH}}\hat{S}_{HH} + \beta_{DH}\frac{\hat{I}_{D}}{\hat{I}_{HH}}\hat{S}_{HH} \\ &+ \beta_{HH}\frac{\hat{I}_{LH}}{\hat{I}_{HH}}\hat{S}_{HH})(\frac{A_{LH}}{\hat{S}_{LH}} + \beta_{HH}\frac{\hat{I}_{HH}}{\hat{I}_{LH}}\hat{S}_{LH}), \\ a_{3} &= \frac{A_{LH}}{\hat{S}_{LH}}(\frac{A_{HH}}{\hat{S}_{HH}} + \gamma_{H} + \alpha_{H} + \mu_{H})(\gamma_{H} + \alpha_{H} + \mu_{H}) \\ &- \mu_{H}\beta_{HH}\hat{S}_{LH}(\frac{A_{HH}}{\hat{S}_{HH}} + \gamma_{H} + \alpha_{H} + \mu_{H}) - \beta_{HH}\hat{S}_{HH}\frac{A_{LH}}{\hat{S}_{LH}}(\gamma_{H} \\ &+ \alpha_{H} + \mu_{H}) + \frac{A_{HH}}{\hat{S}_{HH}}(\frac{A_{LH}}{\hat{S}_{LH}} + \gamma_{H} + \alpha_{H} + \mu_{H}) - \beta_{HH}\hat{S}_{LH}\frac{A_{HH}}{\hat{S}_{HH}}(\gamma_{H} \\ &+ \alpha_{H} + \mu_{H}), \\ a_{4} &= \frac{A_{HH}}{\hat{S}_{HH}}\frac{A_{LH}}{\hat{S}_{LH}}(\gamma_{H} + \alpha_{H} + \mu_{H})^{2} - \mu_{H}\beta_{HH}(\hat{S}_{HH}\frac{A_{LH}}{\hat{S}_{LH}} + \hat{S}_{LH}\frac{A_{HH}}{\hat{S}_{HH}})(\gamma_{H} \\ &+ \alpha_{H} + \mu_{H}), \end{split}$$

It is assumed that $\mu_H \approx 0$ for μ_H is much smaller than other parameters. Then we get $a_1 > 0$ $a_2 > 0$ $a_3 > 0$ $a_4 > 0$ $b_1 = \frac{a_1 a_2 - a_3}{a_1} > 0$ and $c_1 = \frac{b_1 a_3 - a_1 a_4}{b_1} > 0$. So if $E^*_{(WDH)}$ exists, all of the real parts of eigenvalues of $J_H(E^*_{(H)})$ are negative according to Routh–Hurwitz stability criterion.

according to Routh–Hurwitz stability criterion. In conclusion, if $R_{0(WW)} = \frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} > 1$, the real parts of eigenvalues of $J_{WDH}(E^*_{(WDH)})$ are negative and $E^*_{(WDH)}$ is stable.

From Theorem 2.1, Theorem 2.2 and Theorem 2.3, it is more difficult to satisfy the conditions to control emerging zoonoses with the number of susceptible species increasing. But if there was an epidemic in wildlife with $R_{0(WW)} = \frac{A_W \beta_{WW}}{\mu_W(\mu_W + \gamma_W + \alpha_W)} > 1$, emerging zoonoses might be prevalent in humans.

Next we take Strategy 1, Strategy 2 and Strategy 3 into consideration in order to compare the effects of different isolation and slaughter strategies in wildlife, domestic animals and humans on emerging zoonoses.

Strategy 1.

It is assumed that $\delta = \delta_S = \delta_I = \delta_R$ with same slaughter rate in susceptibles, infectives, and recovered individuals in wildlife in order to simply the model (3).

$$\begin{split} \dot{S}_W &= A_W - \beta_{WW} I_W S_W - \mu_W S_W - \delta S_W, \\ \dot{I}_W &= \beta_{WW} I_W S_W - \mu_W I_W - \gamma_W I_W - \alpha_W I_W - \delta I_W, \\ \dot{R}_W &= \gamma_W I_W - \mu_W R_W - \delta R_W, \\ \dot{S}_D &= A_D - (1 - \theta_D) \beta_{WD} I_W S_D - \beta_{DD} I_D S_D - \mu_D S_D, \\ \dot{I}_D &= (1 - \theta_D) \beta_{WD} I_W S_D + \beta_{DD} I_D S_D - \gamma_D I_D - \alpha_D I_D - \mu_D I_D, \\ \dot{R}_D &= \gamma_D I_D - \mu_D R_D, \\ \dot{S}_{HH} &= A_{HH} - (1 - \theta_H) \beta_{WH} I_W S_{HH} - \beta_{DH} I_D S_{HH} - \beta_{HH} (I_{HH} + I_{LH}) S_{HH} - \mu_H S_{HH}, \\ \dot{I}_{HH} &= (1 - \theta_H) \beta_{WH} I_W S_{HH} + \beta_{DH} I_D S_{HH} + \beta_{HH} (I_{HH} + I_{LH}) S_{HH} \\ -\gamma_H I_{HH} - \alpha_H I_{HH} - \mu_H I_{HH}, \\ \dot{R}_{LH} &= \gamma_H I_{HH} - \mu_H R_{HH}, \\ \dot{S}_{LH} &= A_{LH} - \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - \mu_H S_{LH}, \\ \dot{I}_{LH} &= \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - \gamma_H I_{LH} - \mu_H I_{LH}, \\ \dot{R}_{LH} &= \gamma_H I_{LH} - \mu_H R_{LH}. \end{split}$$

$$(25)$$

In (25), we get the control reproductive number in wildlife is $R_{1(WW)} = \frac{A_W \beta_{WW}}{(\mu_W + \delta)}$ $\frac{1}{(\mu_W + \gamma_W + \alpha_W + \delta)}$, the control reproductive number in domestic animals is $R_{1(DD)} = \frac{A_D \beta_{DD}}{\mu_D (\mu_D + \gamma_D + \alpha_D)}$, the control reproductive number in humans is $R_{1(H)} = \frac{(A_{HH} + A_{LH})}{(\mu_H + \gamma_H + \alpha_H)}$ $\frac{\beta_{HH}}{\mu_H}$. The epidemic equilibrium of I_W , I_D and I_H are $\hat{I}_W^1 = \frac{1}{\beta_{WW}} (\frac{A_W \beta_{WW}}{\mu_W + \delta + \gamma_W + \alpha_W} - \mu_W - \delta)$, $\hat{I}_D^1 = \frac{(1 - \theta_D)\beta_{WD}\hat{I}_W^1\hat{S}_D^1}{\mu_D + \gamma_D + \alpha_D - \beta_{DD}\hat{S}_D^1}$ and $\hat{I}_H^1 = \frac{(1 - \theta_H)\eta_S\beta_{WH}\hat{I}_W^1\hat{S}_H^1 + \eta_S\beta_{DH}\hat{I}_D^1\hat{S}_H^1}{\gamma_H + \alpha_H + \mu_H - \beta_{HH}\hat{S}_H^1}$ in strategy 1.

For the epidemic equilibrium of S_W , S_D and S_H , we get $\hat{S}^1_W = \frac{\mu_W + \gamma_W + \alpha_W + \delta}{\beta_W} > \hat{S}_W$, $\hat{S}^1_D > \hat{S}_D$ and $\hat{S}^1_H > \hat{S}_H$ for $g_1(\hat{S}^1_D) < 0$ and $g_2(\hat{S}^1_H) < 0$.

Theorem 2.4. If $R_{1(WW)} < 1$, $R_{1(DD)} < 1$ and $R_{1(H)} < 1$, the disease-free equilibrium $E_{0(WDH)}^1$ in system (25) is stable. If $R_{1(WW)} > 1$, there exists epidemic equilibrium E_{WDH}^{**} , and $E_{(WDH)}^{**}$ is stable.

Strategy 2.

In (4), we get the control reproductive number in wildlife is $R_{2(WW)} = \frac{A_W}{\mu_W}$ $\frac{\beta_{WW}}{(\mu_W + \gamma_W + \alpha_W)}$, the control reproductive number in domestic animals is $R_{2(DD)} = \frac{\beta_{DD}}{(\mu_D + \gamma_D + \alpha_D + \Delta_I)} \frac{A_D}{\mu_D}$, the control reproductive number in humans is $R_{2(H)} = \frac{\beta_{HH}}{\mu_H}$

 $\frac{(A_{HH}+A_{LH})}{(\mu_H+\gamma_H+\alpha_H)}$. The epidemic equilibrium of I_W , I_D and I_H are $\hat{I}_W^2 = \frac{1}{\beta_{WW}}(-\mu_W)$ $\begin{array}{l} (\mu_{H}+\gamma_{H}+\alpha_{H}) & \text{ if it optachie equilibrium of } \mu_{W}, \mu_{D} = \mu_{H} = \mu_{W} \quad \beta_{WW} < \mu_{H} \\ + \frac{A_{W}\beta_{WW}}{\mu_{W}+\gamma_{W}+\alpha_{W}}) = \hat{I}_{W} , \hat{I}_{D}^{2} = \frac{\beta_{WD}\hat{I}_{W}^{2}\hat{S}_{D}^{2}}{\mu_{D}+\gamma_{D}+\alpha_{D}+\Delta_{I}-\beta_{DD}\hat{S}_{D}^{2}} \text{ and } \hat{I}_{H}^{2} = \frac{1}{\gamma_{H}+\alpha_{H}+\mu_{H}-\beta_{HH}\hat{S}_{H}^{2}} \\ (\eta_{S}\beta_{WH}\hat{I}_{W}^{2}\hat{S}_{H}^{2} + (1-\Theta_{H})\eta_{S}\beta_{DH}\hat{I}_{D}^{2}\hat{S}_{H}^{2}) \text{ in strategy } 2. \\ \text{For the epidemic equilibrium of } S_{W}, S_{D} \text{ and } S_{H}, \text{ we get } \hat{S}_{W}^{2} = \frac{\mu_{W}+\gamma_{W}+\alpha_{W}}{\beta_{W}} = 0 \\ \end{array}$

 $\hat{S}_W,\,\hat{S}_D^2>\hat{S}_D$ and $\hat{S}_H^2>\hat{S}_H$ for $g_2(\hat{S}_H^2)<0$.

In fact, \hat{S}_D^2 is the smaller root of

$$g_{3}(\hat{S}_{D}^{2}) = \mu_{D}\beta_{DD}(\hat{S}_{D}^{2})^{2} - \left[A_{D}\beta_{DD} + (\mu_{D} + \gamma_{D} + \alpha_{D} + \Delta_{I})(\mu_{D} + \beta_{WD}\hat{I}_{W})\right]\hat{S}_{D}^{2} + A_{D}(\mu_{D} + \gamma_{D} + \alpha_{D} + \Delta_{I}) = 0$$

So we have

$$g_3(\hat{S}_D) = \mu_D \beta_{DD} \hat{S}_D^2 - \left[A_D \beta_{DD} + (\mu_D + \gamma_D + \alpha_D + \Delta_I)(\mu_D + \beta_{WD} \hat{I}_W) \right] \hat{S}_D + A_D (\mu_D + \gamma_D + \alpha_D + \Delta_I)$$

If

$$g_1(\hat{S}_D) = \mu_D \beta_{DD} \hat{S}_D^2 - \left[A_D \beta_{DD} + (\mu_D + \gamma_D + \alpha_D)(\mu_D + \beta_{WD} \hat{I}_W) \right] \hat{S}_D + A_D (\mu_D + \gamma_D + \alpha_D)$$
$$= 0$$

and

$$A_D - \beta_{WD}\hat{I}_W\hat{S}_D - \beta_{DD}\hat{I}_D\hat{S}_D - \mu_D\hat{S}_D = 0,$$

we get

$$g_{3}(\hat{S}_{D}) = \left[A_{D}\beta_{DD} + (\mu_{D} + \gamma_{D} + \alpha_{D})(\mu_{D} + \beta_{WD}\hat{I}_{W})\right]\hat{S}_{D} - A_{D}(\mu_{D} + \gamma_{D} + \alpha_{D}) - \left[A_{D}\beta_{DD} + (\mu_{D} + \gamma_{D} + \alpha_{D} + \Delta_{I})(\mu_{D} + \beta_{WD}\hat{I}_{W})\right]\hat{S}_{D} + A_{D}(\mu_{D} + \gamma_{D} + \alpha_{D} + \Delta_{I})$$
$$= -\Delta_{I}(\mu_{D} + \beta_{WD}\hat{I}_{W})\hat{S}_{D} + \Delta_{I}A_{D}$$
$$= \Delta_{I}\beta_{DD}\hat{I}_{D}\hat{S}_{D}$$
>0.

Then we get $\hat{S}_D^2 > \hat{S}_D$ with $\mu_D \beta_{DD} > 0$.

Theorem 2.5. If $R_{2(WW)} < 1$, $R_{2(DD)} < 1$ and $R_{2(H)} < 1$, the disease-free equilibrium $E_{0(WDH)}^2$ in system (4) is stable. If $R_{2(WW)} > 1$, there exists epidemic equilibrium $E_{(WDH)}^{***}$, and $E_{(WDH)}^{***}$ is stable.

Strategy 3.

If we took quarantine and isolation strategies in humans only, the impact of wildlife and domestic animals in human epidemic would be never changed comparing to no strategy. So we select the human epidemic model (26) from (5) for further analysis. At the same time, we choose $\varphi(I) = \rho(I_{HH} + I_{LH})$ to simplify the model.

$$\begin{split} S_{HH} &= A_{HH} - \beta_{WH} I_W S_{HH} - \beta_{DH} I_D S_{HH} - \beta_{HH} (I_{HH} + I_{LH}) S_{HH} \\ & \mu_H S_{HH} - \rho (I_{HH} + I_{LH}) S_{HH} + \gamma_{H1} O_{HH1}, \\ \dot{O}_{HH1} &= \rho (I_{HH} + I_{LH}) S_{HH} - \gamma_{H1} O_{HH1} - \mu_H O_{HH1}, \\ \dot{I}_{HH} &= \beta_{WH} I_W S_{HH} + \beta_{DH} I_D S_{HH} + \beta_{HH} (I_{HH} + I_{LH}) S_{HH} \\ & -\gamma_H I_{HH} - \alpha_H I_{HH} - \mu_H I_{HH} - \sigma I_{HH}, \\ \dot{O}_{HH2} &= \sigma I_{HH} - \gamma_{H2} O_{HH2} - \mu_H O_{HH2}, \\ \dot{S}_{LH} &= A_{LH} - \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - \mu_H S_{LH} - \rho (I_{HH} + I_{LH}) S_{LH} \\ & +\gamma_{H1} O_{LH1}, \\ \dot{O}_{LH1} &= \rho (I_{HH} + I_{LH}) S_{LH} - \gamma_{H1} O_{LH1} - \mu_H O_{LH1}, \\ \dot{I}_{LH} &= \beta_{HH} (I_{HH} + I_{LH}) S_{LH} - \gamma_H I_{LH} - \alpha_H I_{LH} - \sigma I_{LH}, \\ \dot{O}_{LH2} &= \sigma I_{LH} - \gamma_{H2} O_{LH2} - \mu_H O_{LH2}. \end{split}$$

We get the control reproductive number in humans is:

$$R_{3(H)} = \frac{(A_{HH} + A_{LH})\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H + \sigma)}$$

Theorem 2.6. If $R_{0(WW)} < 1$, $R_{0(DD)} < 1$ and $R_{3(H)} < 1$, the disease-free equilibrium $E^3_{0(WDH)}$ in system (5) is stable. If $R_{0(WW)} > 1$, there exists epidemic equilibrium $E^{****}_{(WDH)}$, and $E^{****}_{(WDH)}$ is stable.

TABLE 1. Impact of different strategies on reproductive numbers

Strategies	no strategy	Strategy 1	Strategy 2	Strategy 3	
Reproductive number in wildlife	$R_{0(WW)} = \frac{R_A}{R_B}$	$R_{1(WW)} = \frac{R_A}{R_C}$	$\begin{array}{c} R_{2(WW)} = \\ R_{0(WW)} \end{array}$	$\begin{array}{c} R_{3(WW)} = \\ R_{0(WW)} \end{array}$	
Reproductive number in domestic animals	$R_{0(DD)} = \frac{R_D}{R_E}$	$\begin{array}{c} R_{1(DD)} = \\ R_{0(DD)} \end{array}$	$R_{2(DD)} = \frac{R_D}{R_F}$	$\begin{array}{c} R_{3(DD)} = \\ R_{0(DD)} \end{array}$	
Reproductive number in humans	$R_{0(H)} = \frac{R_G}{R_H}$	$R_{1(H)} = R_{0(H)}$	$R_{2(H)} = R_{0(H)}$	$R_{3(H)} = \frac{R_G}{R_I}$	
$R_A = A_W \beta_{WW}, R_B = \mu_W (\mu_W + \gamma_W + \alpha_W),$					
$R_C = (\mu_W + \delta)(\mu_W + \gamma_W + \alpha_W + \delta),$					
$R_D = A_D \beta_{DD}, R_E = \mu_D (\mu_D + \gamma_D + \alpha_D),$					
$R_F = \mu_D(\mu_D + \gamma_D + \alpha_D + \Delta_I), R_G = (A_{HH} + A_{LH})\beta_{HH},$					
$R_H = \mu_H(\mu_H + \gamma_H + \alpha_H), R_I = \mu_H(\mu_H + \gamma_H + \alpha_H + \sigma).$					

3. Numerical simulation. In this section we take avian influenza epidemic in China as an example to analyze the effects of different strategies on emerging zoonoses. Avian influenza is a kind of zoonoses, which have been prevalent in humans since 150 years ago. Avian influenza virus originated from aquatic birds, and it infected domestic birds by sharing watersheds. Humans can be infected by avian influenza virus via infected domestic birds[11, 30, 6, 9]. But for birds, we

cannot get the exact parameters to reflect the virus transmission clearly. So we take some similar data to estimate the process of avian influenza virus transmission approximately (TABLE 2).

The number of domestic birds is 4.2 times more than the number of humans in China [7], so we assume that the number of domestic birds is 8400 and the number of humans is 2000 to simplify the calculation. And it is assumed that there are about 1000 wild aquatic birds for no exact data found. And it is assumed that $R_{0(WD)} = 0.1 * R_{0(DD)}, R_{0(WH)} = 0.1 * R_{0(H)}$ and $R_{0(DH)} = 0.1 * R_{0(H)}$.

The avian influenza virus transmission has been shown in model (2), which included wildlife, domestic animals, high risk group and low risk group [10, 21, 13]. For high risk group and low risk group in humans, there may be shown in different proportion in different areas. Less people are needed to take care of live animals in modern farming than tradition. Few people have opportunities to contact with live animals in some areas, which are the potential hosts of some pathogens in emerging zoonoses. But in some other areas, stock raising is the main economy origin of the residents. More people have to look after live animals to help support the family. The proportion of high risk group and low risk group is higher in these areas than others. Here we choose different proportions of high risk group and low risk group, such as 1:9, 1:3, 1:1, 3:1 and 9:1, to reflect emerging avian influenza prevalence in different areas (FIGURE 2).

From a to e in FIGURE 2, we get that more and more high proportion of humans are infected in the first 90 days. More people would be infected with higher proportion of them having the opportunity to contact with susceptible animals. From FIGURE 3, we get that the incidence rate on epidemic equilibrium is increasing with higher proportion of high risk group in humans. Although the proportion of high risk group in humans would never change the basic reproductive number, it could impact the final prevalence in humans.

The effects of parameters δ , Δ_I , σ in Strategy 1, Strategy 2 and Strategy 3 on control reproductive numbers have been shown in FIGURE 4. The existence of parameters δ , Δ_I , and σ would decrease the value of $R_{1(WW)}$, $R_{2(DD)}$ and $R_{3(H)}$. If $\delta < 0.142 * 10^{-3}$, $\Delta_I < 0.258$ and $\sigma < 0.066$, $R_{1(WW)}$, $R_{2(DD)}$ and $R_{3(H)}$ would get the value below threshold to control the zoonoses in wildlife, domestic animals and humans respectively. The effects of Strategy 1, Strategy 2 and Strategy 3 in different areas are shown in FIGURE 5, when $I_{DC} = I_{WC} = I_{HC} = 15$. The effects of δ , θ_D , θ_H , Δ_I , Θ_H , σ and ρ on the number of infected humans in the first 90 days are shown in FIGURE 6.

4. **Discussion.** From Ebola, Hendra, Marburg, SARS to H1N1, H7N9, more and more zoonotic pathogens come into humans. Tens of thousands of people have dead of these zoonoses in the last hundreds of years. Some public health policies have to be established to answer emerging or remerging zoonoses. For different species participating in an emerging zoonosis, different strategies should been taken for controlling. In this paper, we established model (3), model (4) and model (5) to reflect the effects of Strategy 1, Strategy 2 and Strategy3 about isolation and slaughter in emerging zoonoses respectively. Strategy 1 is the controlling measure for wildlife. Strategy 2 is the controlling measure for domestic animals. And Strategy 3 is the controlling measure for humans.

All of the three strategies would change the basic reproductive number to their own control reproductive number. The involvement of Strategy 1, Strategy 2 and Strategy 3 would change the conditions, which determine the zoonoses prevalence or



FIGURE 2. Avian influenza prevalence in wildlife, domestic animals and humans with high risk group: low risk group=1:9 in a, 1:3 in b, 1:1 in c, 3:1 in d 9:1 in e.

not. At the same time, we conclude that the extinction of zoonoses must satisfy the conditions ensuring all of basic (control) reproductive numbers in different species are less than 1, whether it is taken controlling strategy or not. But if and only if basic (control) reproductive numbers in wildlife is more than 1, the zoonoses might be prevalent in all of the susceptible species.



FIGURE 3. Incidence rate on epidemic equilibrium change in different proportion of high risk group in humans



FIGURE 4. The effect of δ on $R_{1(WW)}$ in a. $R_{1(WW)} = 1$, when $\delta = 0.142 * 10^3$. The effect of Δ_I on $R_{2(DD)}$ in b. $R_{2(DD)} = 1$, when $\Delta_I = 0.258$. The effect of δ on $R_{3(H)}$ in c. $R_{3(H)} = 1$, when $\delta = 0.066$.



FIGURE 5. Phase portrait of S_H and I_H in system (6) with no strategy: $\varepsilon_W = 0$, $\varepsilon_D = 0$, $\varepsilon_H = 0$. Strategy 1: when $I_{LH} + I_{HH} < I_{WC}$, $\varepsilon_W = 0$; when $I_{LH} + I_{HH} \ge I_{WC}$, $\varepsilon_W = 1$. $\varepsilon_D = 0$, $\varepsilon_H = 0$. Strategy 2: when $I_{LH} + I_{HH} < I_{DC}$, $\varepsilon_D = 0$; when $I_{LH} + I_{HH} < I_{DC}$, $\varepsilon_D = 0$; when $I_{LH} + I_{HH} < I_{DC}$, $\varepsilon_D = 0$; when $I_{LH} + I_{HH} < I_{HC}$, $\varepsilon_H = 0$; when $I_{LH} + I_{HH} \ge I_{HC}$, $\varepsilon_H = 1$. $\varepsilon_W = 0$, $\varepsilon_D = 0$. ($\delta = 0.1$, $\theta_D = 0.1$, $\theta_H = 0.1$, $\Delta_I = 1$, $\Theta_H = 0.1$, $\sigma = 0.01$ and $\rho = 0.001$; high risk group: low risk group=1:9 in a, 9:1 in b; $I_{DC} = I_{WC} = I_{HC} = 15$, at 26^{th} day in a, 17^{th} day in b)

The stability analysis on models in section 2 reflects the effects of three strategies on control reproductive numbers and equilibriums. In section 3, some numerical simulations show the effects of the three strategies on avian influenza epidemic in different areas in China at beginning. In this paper, we take isolation and slaughter strategies into consideration to study their effects on emerging zoonoses. But the other effective strategies like vaccination are neglected, which could be proposed in a forthcoming paper.

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REFERENCES

- L. J. S. Allen, Mathematical Modeling of Viral Zoonoses in Wildlife, Natural Resource Modeling, 25 (2012), 5–51.
- [2] J. Arino, R. Jordan and P. V. D. Driessche, Quarantine in a multi-species epidemic model with spatial dynamics, *Mathematical Biosciences*, **206** (2007), 46–60.
- [3] R. M. Atlas and S. Maloy, One Health: People, Animals, and the Environment, ASM Press, 2014.
- [4] R. G. Bengis, F. A. Leighton and J. R. Fischer, The role of wildlife in emerging and reemerging zoonoses, *Revue Scientifique Et Technique*, 23 (2004), 497–511.
- [5] F. Brauer and C. Chavez, Mathematical Models in Population Biology and Epidemiology, 2^{nd} edition, Springer, 2001.
- [6] N. Busquets, J. Segals and L. Crdoba, Experimental infection with H1N1 European swine influenza virus protects pigs from an infection with the 2009 pandemic H1N1 human influenza virus, Veterinary Research, 41 (2010), 571–584.
- [7] China Agricultural Yearbook Editing Committee, ChinaAgriculture Yearbook, China Agriculture Press, China, 2012.



FIGURE 6. The effects of δ , θ_D , θ_H , Δ_I , Θ_H , σ and ρ on the number of infected humans (I_H) in the first 90 days (δ in a, θ_D in b, θ_H in c, Δ_I in d, Θ_H in e, σ in f and ρ in g) (high risk group: low risk group=1:9)

Parameter	Definitions	Values	Sources
A_W	birth or immigration rate of wild aquatic birds	0.137 birds/day	Est.
μ_W	natural mortality rate of wild aquatic birds	$0.000137/\mathrm{day}$	[33]
γ_W	recovery rate of wild aquatic birds	$0.25/\mathrm{day}$	Est.
$lpha_W$	disease-induced mortality rate of wild aquatic birds	$0.0025/\mathrm{day}$	Est.
A_D	birth or immigration rate of domestic birds	48.72 birds/day	[33]
μ_D	natural mortality rate of domestic birds	$0.0058/\mathrm{day}$	[33]
γ_D	recovery rate of domestic birds	$0.25/\mathrm{day}$	[26]
α_D	disease-induced mortality rate of domestic birds	$0.0025/\mathrm{day}$	Est.
$A_{HH} + A_{LH}$	birth or immigration rate of humans	0.07 people/day	[23]
μ_H	natural mortality rate of humans	$0.000035/\mathrm{day}$	[23]
γ_H	recovery rate of humans	$0.33/\mathrm{day}$	[26, 31]
γ_{H1}	remove rate from isolation compartment to susceptible compartment.	$0.5/\mathrm{day}$	Est.
γ_{H2}	remove rate from isolation compartment to recovery individual compartment.	$0.5/\mathrm{day}$	Est.
$lpha_H$	disease-induced mortality rate of humans	$0.0033/\mathrm{day}$	Est.
$R_{0(WW)}$	basic reproductive number of wild aquatic birds	2	Est.
$R_{0(DD)}$	basic reproductive number of domestic birds	2	Est.
$R_{0(H)}$	basic reproductive number of humans	1.2	[26]
β_{WD}	per capita incidence rate from wild aquatic birds to domestic birds	$6.15 * 10^{-6}$	Est.
β_{WH}	per capita incidence rate from wild aquatic birds to humans	$2 * 10^{-5}$	Est.
β_{DH}	per capita incidence rate from domestic birds to humans	$2 * 10^{-5}$	Est.

TABLE 2.Parameter definitions and their values for avian in-fluenza in China.

- [8] G. Chowell, Model parameters and outbreak control for SARS, Emerging Infectious Diseases, 10 (2004), 1258–1263.
- [9] B. J. Coburn, B. G. Wagner and S. Blower, Modeling influenza epidemics and pandemics: Insights into the future of swine flu (H1N1), *Bmc Medicine*, **7** (2009), p30.
- [10] B. J. Coburn, C. Cosne and S. Ruan, Emergence and dynamics of influenza super-strains, Bmc Public Health, 11 (2011), 597–615.
- [11] R. W. Compans and M. B. A. Oldstone, *Influenza Pathogenesis and Control Volume I*, Current Topics in Microbiology & Immunology, 2014.
- [12] M. R. Conover, *Human Diseases from Wildlife*, Boca Raton : CRC Press, Taylor & Francis Group, 2014.
- [13] M. Derouich and A. Boutayeb, An avian influenza mathematical model, Applied Mathematical Sciences, 2 (2008), 1749–1760.
- [14] K. Dietz, W. H. Wernsdorfer and I. Mcgregor, Mathematical Models for Transmission and Control of Malaria, Malaria, 1988.
- [15] A. Dobson, Population dynamics of pathogens with multiple host species, American Naturalist, 164 (2004), 64–78.
- [16] P. Van Den Driessche and J. Watmough, Reproduction numbers and sub-threshold endemic equilibria for compartmental models of disease transmission, *Mathematical Biosciences*, 180 (2002), 29–48.
- [17] X. Fang, The Role of Mammals in Epidemiology, Acta Theriologica Sinica, 2 (1981), 219–224.
- [18] Z. Feng, Final and peak epidemic sizes for SEIR models with quarantine and isolation, Mathematical Biosciences & Engineering, 4 (2007), 675–686.
- [19] Z. Feng, Applications of Epidemiological Models to Public Health Policymaking, World Scientific, 2014.
- [20] A. Fritsche, R. Engel and D. Buhl, Mycobacterium bovis tuberculosis: From animal to man and back, International Journal of Tuberculosis & Lung Disease the Official Journal of the International Union Against Tuberculosis & Lung Disease, 8 (2004), 903–904.
- [21] S. Iwami, Y. Takeuchi and X. Liu, Avian-Chuman influenza epidemic model, Mathematical Biosciences, 207 (2007), 1–25.
- [22] A. M. Kilpatrick and S. E. Randolph, Drivers, dynamics, and control of emerging vector-borne zoonotic diseases, *Lancet*, **380** (2012), 1946–1955.
- [23] J. Lee, J. Kim and H. D. Kwon, Optimal control of an influenza model with seasonal forcing and age-dependent transmission rates, *Journal of Theoretical Biology*, **317** (2013), 310–320.
- [24] J. S. Mackenzie, One Health: The Human-Animal-Environment Interfaces in Emerging Infectious Diseases, Springer, Berlin, 2013.
- [25] A. Mubayi, A cost-based comparison of quarantine strategies for new emerging diseases, Mathematical Biosciences & Engineering, 7 (2010), 687–717.
- [26] R. A. Saenz, H. W. Hethcote and G. C. Gray, Confined animal feeding operations as amplifiers of influenza, Vector Borne & Zoonotic Diseases, 6 (2006), 338–346.
- [27] P. M. Sharp and B. H. Hahn, Cross-species transmission and recombination of 'AIDS' viruses, Philosophical Transactions of the Royal Society B Biological Sciences, 349 (1995), 41–47.
- [28] A. Sing, Zoonoses Infections Affecting Humans and Animals, Springer Netherlands, Berlin, 2015.
- [29] S. Towers and Z. Feng, Pandemic H1N1 influenza: predicting the course of a pandemic and assessing the efficacy of the planned vaccination programme in the United States, *European Communicable Disease Bulletin*, 14 (2009), 6–8.
- [30] Q. Xian, L. Cui and Y. Jiao, Antigenic and genetic characterization of a European avian-like H1N1 swine influenza virus from a boy in China in 2011, Archives of Virology, 158 (2013), 39–53.
- [31] W. D. Zhang, Optimized strategy for the control and prevention of newly emerging influenza revealed by the spread dynamics model, *Plos One*, **91** (2014), 5–51.
- [32] J. Zhang, Z. Jin and G. Q. Sun, Modeling seasonal rabies epidemics in China, Bulletin of Mathematical Biology, 74 (2012), 1226–1251.

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