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 trans-
missions 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 mod-
els to reflect the pathogen transmission in wildlife [1].

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

$$\begin{align*}
\dot{S}_i &= A_i - \sum_{j=1}^{n} \beta_{ji} I_j S_i - \mu_i S_i, \\
\dot{I}_i &= \sum_{j=1}^{n} \beta_{ji} I_j S_i - \mu_i I_i - \gamma_i I_i - \alpha_i I_i, \\
\dot{R}_i &= \gamma_i I_i - \mu_i R_i.
\end{align*}$$

(1)

$S_i$, $I_i$ and $R_i$ represent the number of susceptibles, infectives, and recovered
individuals for species $i$, $i=1,2,\ldots,n$. $A_i$ is the birth or immigration rate for species
$i$. $\mu_i$ is the natural mortality rate. $\alpha_i$ is the disease-induced mortality rate. $\gamma_i$ is
the recovery rate. And $\beta_{ji}$ 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, domes-
tic animals and humans into account to limit the value of $\beta_{ji}$ 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 anthropo-
genic 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.

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 $i = LH$. $A_i$ is the birth or immigration rate for species $i = W, D, HH$ or $LH$. $\mu_i$, $\gamma_i$, and $\beta_{ji}$ 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.

\[
\begin{align*}
S_W &= A_W - \beta_{WW} I_W S_W - \mu_W S_W, \\
I_W &= \beta_{WW} I_W S_W - (\mu_W + \gamma_W + \alpha_W) I_W, \\
R_W &= \gamma_W I_W - \mu_W R_W, \\
S_D &= \Lambda_D - (\beta_{WD} I_W + \beta_{DD} D) S_D - \mu_D S_D, \\
I_D &= (\beta_{WD} I_W + \beta_{DD} D) S_D - (\gamma_D + \alpha_D + \mu_D) I_D, \\
R_D &= \gamma_D I_D - \mu_D R_D, \\
S_{HH} &= \Lambda_{HH} - [\beta_{WH} I_W + \beta_{DH} D + \beta_{HH}(I_{HH} + I_{LH})] S_{HH} - \mu_H S_{HH}, \\
I_{HH} &= [\beta_{WH} I_W + \beta_{DH} D + \beta_{HH}(I_{HH} + I_{LH})] S_{HH} - (\gamma_H + \alpha_H + \mu_H) I_{HH}, \\
R_{HH} &= \gamma_H I_{HH} - \mu_H R_{HH}, \\
S_{LH} &= \Lambda_{LH} - \beta_{HH}(I_{HH} + I_{LH}) S_{LH} - \mu_H S_{LH}, \\
I_{LH} &= \beta_{HH}(I_{HH} + I_{LH}) S_{LH} - (\gamma_H + \alpha_H + \mu_H) I_{LH}, \\
R_{LH} &= \gamma_H I_{LH} - \mu_H R_{LH}.
\end{align*}
\] (2)

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.

\[
\begin{align*}
S_W &= A_W - \beta_{WW} I_W S_W - (\mu_W + \delta_S) S_W, \\
I_W &= \beta_{WW} I_W S_W - (\mu_W + \gamma_W + \alpha_W + \delta_I) I_W, \\
R_W &= \gamma_W I_W - (\mu_W + \delta_R) R_W, \\
S_D &= \Lambda_D - (1 - \theta_D) \beta_{WD} I_W + \beta_{DD} D) S_D - \mu_D S_D, \\
I_D &= ((1 - \theta_D) \beta_{WD} I_W + \beta_{DD} D) S_D - (\gamma_D + \alpha_D + \mu_D) I_D, \\
R_D &= \gamma_D I_D - \mu_D R_D, \\
S_{HH} &= \Lambda_{HH} - [(1 - \theta_H) \beta_{WH} I_W + \beta_{DH} D + \beta_{HH}(I_{HH} + I_{LH})] S_{HH} - \mu_H S_{HH}, \\
I_{HH} &= [(1 - \theta_H) \beta_{WH} I_W + \beta_{DH} D + \beta_{HH}(I_{HH} + I_{LH})] S_{HH} - (\gamma_H + \alpha_H + \mu_H) I_{HH}, \\
R_{HH} &= \gamma_H I_{HH} - \mu_H R_{HH}, \\
S_{LH} &= \Lambda_{LH} - \beta_{HH}(I_{HH} + I_{LH}) S_{LH} - \mu_H S_{LH}, \\
I_{LH} &= \beta_{HH}(I_{HH} + I_{LH}) S_{LH} - (\gamma_H + \alpha_H + \mu_H) I_{LH}, \\
R_{LH} &= \gamma_H I_{LH} - \mu_H R_{LH}.
\end{align*}
\] (3)

\(\delta_s, \delta_I\) and \(\delta_R\) represent lethal control or slaughter rate of susceptibles, infectives, and recovered individuals in wildlife. \(\theta_D, \theta_H\) represent effectiveness of fencing (physical barriers), \(\theta_D, \theta_H \in [0, 1]\). 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.

If \( \Theta_H \) represents the per capita incidence rate from animals to humans, \( \beta_I \) represents the rate of getting infected, to take the initiative and get away from susceptible animals. So taking isolation strategies in animals, it is the susceptible humans, who are afraid of quarantine and isolation strategies in humans are different from animals. For method to limit the pathogen transmission except for vaccination. But the effect of isolation, \( \Theta_H \) 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, \( \beta_{W_H} \) and \( \beta_{D_H} \), is decreased by \( \theta_H \) and \( \Theta_H \). But in humans, we quarantine and isolate the infected people to cut off pathogen transmission way. \( \beta_{H_H} \) would not change at this time, but there is a new compartment \( O \) produced, which denotes the isolation compartment [25].

\[
\begin{align*}
\dot{S}_H &= \alpha_{HH} - [\beta_{W_H}I_W + \beta_{D_H}I_D + \beta_{H}(I_{HH} + I_{LH})]S_{HH} - \mu_H S_{HH} - \varphi(I)S_{HH} + \gamma_{H1}O_{HH1}, \\
\dot{O}_{HH1} &= \varphi(I)S_{HH} - \gamma_{H1}O_{HH1} - \mu_H O_{HH1}, \\
\dot{I}_{HH} &= [\beta_{W_H}I_W + \beta_{D_H}I_D + \beta_{H}(I_{HH} + I_{LH})]S_{HH} - \gamma_{H} + \alpha_{H} + \mu_{H} + \sigma I_{HH}, \\
\dot{O}_{HH2} &= \sigma I_{HH} - \gamma_{H2}O_{HH2} - \mu_{H}O_{HH2}, \\
\dot{R}_{HH} &= \gamma_{H}I_{HH} + \gamma_{H2}O_{HH2} - \mu_{H}R_{HH}, \\
\dot{S}_L &= \alpha_{LH} - \beta_{H}(I_{HH} + I_{LH})S_{LH} - \mu_H S_{LH} - \varphi(I)S_{LH} + \gamma_{H1}O_{LH1}, \\
\dot{O}_{LH1} &= \varphi(I)S_{LH} - \gamma_{H1}O_{LH1} - \mu_H O_{LH1}, \\
\dot{I}_{LH} &= \beta_{H}(I_{HH} + I_{LH})S_{LH} - (\gamma_{H} + \alpha_{H} + \mu_{H} + \sigma)I_{LH}, \\
\dot{O}_{LH2} &= \sigma I_{LH} - \gamma_{H2}O_{LH2} - \mu_{H}O_{LH2}, \\
\dot{R}_{LH} &= \gamma_{H}I_{LH} + \gamma_{H2}O_{LH2} - \mu_{H}R_{LH}.
\end{align*}
\]
\( O_{HH1} \) 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 infectives in high risk group respectively. 

by (3), (4) and (5) in wildlife, domestic animals and humans as the form:

\[
\begin{align*}
\dot{S}_W &= Aw - \beta_{WW}I_WS_W - (\mu_W + \varepsilon_W\delta_S)S_W,
\dot{I}_W &= \beta_{WW}I_WS_W - (\mu_W + \gamma_W + \alpha_W + \varepsilon_W\delta_I)I_W,
\dot{R}_W &= \gamma_WI_W - (\mu_W + \varepsilon_W\delta_R)R_W,
\dot{S}_D &= \Lambda_D - (1 - \varepsilon_W\theta_D)\beta_{WD}I_W + \beta_{DD}I_D)S_D - \mu_DS_D,
\dot{I}_D &= ((1 - \varepsilon_W\theta_D)\beta_{WD}I_W + \beta_{DD}I_D)S_D - (\gamma_D + \alpha_D + \mu_D + \varepsilon_D\Delta_I)I_D,
\dot{R}_D &= \gamma_DI_D - \mu_DR_D,
\dot{S}_{HH} &= A_{HH} - [(1 - \varepsilon_W\theta_H)\beta_{WH}I_W + (1 - \varepsilon_D\Theta_H)\beta_{DH}I_D + \beta_{HH}(I_{HH} + I_{LH})]S_{HH} - \mu_HS_{HH} - \varepsilon_H\phi(I)S_{HH} + \gamma_{H1}O_{HH1},
\dot{I}_{HH} &= [(1 - \varepsilon_W\theta_H)\beta_{WH}I_W + (1 - \varepsilon_D\Theta_H)\beta_{DH}I_D + \beta_{HH}(I_{HH} + I_{LH})]S_{HH} - (\gamma_H + \alpha_H + \mu_H + \varepsilon_H\sigma)I_{HH},
\dot{O}_{HH1} &= \varepsilon_H\sigma I_{HH} - \gamma_{H2}O_{HH1} - \mu_HO_{HH1},
\dot{R}_{HH} &= \gamma_HI_{HH} + \gamma_2O_{HH2} - \mu_RF_{HH},
\dot{S}_{LH} &= A_{LH} - \beta_{HH}(I_{HH} + I_{LH})S_{LH} - \mu_HS_{LH} - \varepsilon_H\phi(I)S_{LH} + \gamma_{H1}O_{LH1},
\dot{I}_{LH} &= \beta_{HH}(I_{HH} + I_{LH})S_{LH} - (\gamma_H + \alpha_H + \mu_H + \varepsilon_H\sigma)I_{LH},
\dot{O}_{LH2} &= \varepsilon_H\sigma I_{LH} - \gamma_{H2}O_{LH2} - \mu_HO_{LH2},
\dot{R}_{LH} &= \gamma_HI_{LH} + \gamma_2O_{LH2} - \mu_FR_{LH}.
\end{align*}
\]

with Strategy 1,

\[
\begin{align*}
\varepsilon_W &= 0, I_{LH} + I_{HH} < I_{WC}, \\
\varepsilon_W &= 1, I_{LH} + I_{HH} \geq I_{WC}
\end{align*}
\]

Strategy 2,

\[
\begin{align*}
\varepsilon_D &= 0, I_{LH} + I_{HH} < I_{DC}, \\
\varepsilon_D &= 1, I_{LH} + I_{HH} \geq I_{DC}
\end{align*}
\]

Strategy 3,

\[
\begin{align*}
\varepsilon_H &= 0, I_{LH} + I_{HH} < I_{HC}, \\
\varepsilon_H &= 1, I_{LH} + I_{HH} \geq I_{HC}
\end{align*}
\]

The feasible set \( \Omega = \{S_W(t), I_W(t), R_W(t), S_D(t), I_D(t), R_D(t), S_{HH}(t), O_{HH1}(t), I_{HH}(t), O_{HH2}(t), R_{HH}(t), S_{LH}(t), O_{LH1}(t), I_{LH}(t), O_{LH2}(t), R_{LH}(t) \mid S_i(t), I_i(t), R_i(t), O_j(t) \geq 0, 0 < N \leq \frac{A_w}{\mu_w} + \frac{A_D}{\mu_D} + \frac{A_{HH}}{\mu_H} + \frac{A_{LH}}{\mu_H}, i = \} \)
Proof.

It is difficult for us to take any strategies to control emerging zoonoses in first time. Only the infected 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) \cite{5, 14, 19, 16}.

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

$$
\begin{align*}
\dot{S}_W &= A_W - \beta_{WW} I_W S_W - \mu_W S_W, \\
\dot{I}_W &= \beta_{WW} I_W S_W - \mu_W I_W - \gamma_W I_W - \alpha_W I_W.
\end{align*}
$$

We can get the basic reproductive number in wildlife $R_0(W) = \frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)}$.

The disease-free equilibrium is \(E_{0(W)} = (\dot{S}_W, \dot{I}_W) = (\frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)}, \frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} - 1))\). 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 $R_0(W) < 1$, the disease-free equilibrium $E_{0(W)}$ in wildlife is stable. If $R_0(W) > 1$, 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

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

If $R_0(W) = \frac{A_W \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} < 1$, 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

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

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

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

In (11), the epidemic equilibrium $E_{(W)}^*$ is stable.

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 $E_{0(WD)} = (S_W, I_W, S_D, I_D) = (\frac{A_w}{\beta_{WW}}, 0, \frac{A_D}{\beta_{DD}}, 0)$, the epidemic equilibrium is $E_{(W)}^* = (\hat{S}_W, \hat{I}_W, \hat{S}_D, \hat{I}_D) = (\frac{\mu_W + \gamma_W + \alpha_W}{\beta_{WW}}, \frac{\mu_W}{\beta_{WW}}(A_w \beta_{WW} - 1), \hat{S}_D, \hat{I}_D)$.

**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 only one unique positive epidemic equilibrium $E_{(W)}^*$, and $E_{(W)}^*$ is stable.

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

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

with

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

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 $E_{(W)}^*$ satisfies

$$\begin{align*}
A_W - \beta_{WW} I_W \hat{S}_W - \mu_W \hat{S}_W &= 0 \\
\beta_{WW} \hat{I}_W \hat{S}_W - \mu_W \hat{I}_W - \gamma_W \hat{I}_W - \alpha_W \hat{I}_W &= 0 \\
A_D - \beta_{WD} I_W \hat{S}_D - \beta_{DD} I_D \hat{S}_D - \mu_D \hat{S}_D &= 0 \\
\beta_{WD} I_W \hat{S}_D + \beta_{DD} I_D \hat{S}_D - \gamma_D \hat{I}_D - \alpha_D \hat{I}_D &= 0
\end{align*}$$

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

$$\begin{align*}
\hat{S}_W &= \frac{\mu_W + \gamma_W + \alpha_W}{\beta_{WW}} \\
\hat{I}_W &= \frac{\mu_W}{\beta_{WW}}(A_w \beta_{WW} - 1) \\
\hat{I}_D &= \frac{\beta_{WD} \hat{I}_W \hat{S}_D}{\mu_D + \gamma_D + \alpha_D - \beta_{DD} \hat{S}_D} \\
&= \frac{\beta_{WD} \hat{S}_D}{\mu_D + \gamma_D + \alpha_D - \beta_{DD} \hat{S}_D} \times \frac{\mu_W}{\beta_{WW}}(A_w \beta_{WW} - 1)
\end{align*}$$
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 \(E^{*}_{(W_D)}\).

In fact, for \(\hat{S}_D\), \(\hat{S}_D\) satisfies
\[
g_1(\hat{S}_D) = \mu_D\beta_{DD}\hat{S}_D - \left[A_D\beta_{DD} + (\mu_D + \gamma_D + \alpha_D)(\mu_D + \beta_{WD}\hat{I}_W)\right]\hat{S}_D \\
+ A_D(\mu_D + \gamma_D + \alpha_D) = 0
\]

If \(g_1(\hat{S}_D) = 0\), \(\hat{S}_D\) always has two positive roots because of \(-A_D\beta_{DD} + (\mu_D + \gamma_D + \alpha_D)(\mu_D + \beta_{WD}\hat{I}_W) < 0\), \(A_D(\mu_D + \gamma_D + \alpha_D) > 0\), 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 \(E^{*}_{(W_D)}\) is
\[
J_{WD}(E^{*}_{(W_D)}) = \left(\begin{array}{cc}
J_W(E^{*}_{(W)}) & 0 \\
* & J_D(E^{*}_{(D)})
\end{array}\right)
\]
with
\[
J_D(E^{*}_{(D)}) = \left(\begin{array}{cc}
-\beta_{WD}\hat{I}_W - \beta_{DD}\hat{I}_D - \mu_D & -\beta_{DD}\hat{S}_D \\
\beta_{WD}\hat{I}_W + \beta_{DD}\hat{I}_D & \beta_{DD}\hat{S}_D - \mu_D - \gamma_D - \alpha_D
\end{array}\right)
\]

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

If \(E^{*}_{(D)}\) exists, all of the real parts of eigenvalues of \(J_D(E^{*}_{(D)})\) are negative for \(\frac{A_D}{\mu_D} + \frac{\beta_{WD}\hat{I}_W}{I_D} > 0\) and \(\frac{\beta_{WD}\beta_{DD}\hat{S}_D\hat{I}_W + \beta_{DD}^2\hat{S}_D\hat{I}_D}{I_D} > 0\).

In conclusion, if \(R_{0(WW)} = \frac{A_D\beta_{WD}\hat{I}_W}{\mu_D(\mu_D + \gamma_D + \alpha_D)} > 1\), the real parts of eigenvalues of \(J_D(E^{*}_{(D)})\) are negative and \(E^{*}_{(W_D)}\) is stable.

The human class in (2) can be separated as the form:
If equilibrium $E$ of system (2) is stable (Theorem 2.1, Theorem 2.2).

**Proof.**

$$
\begin{aligned}
S_{HH} &= A_{HH} - \beta_{WH}I_W S_{HH} - \beta_{DH}I_D S_{HH} - \beta_{HH}(I_{HH} + I_{LL}) S_{HH} - \mu_H S_{HH}, \\
I_{HH} &= \beta_{WH}I_W S_{HH} + \beta_{DH}I_D S_{HH} + \beta_{HH}(I_{HH} + I_{LL}) S_{HH} - \gamma_H I_{HH} \\
S_{LL} &= A_{LL} - \beta_{HH}(I_{HH} + I_{LL}) S_{LL} - \mu_H S_{LL}, \\
I_{LL} &= \beta_{HH}(I_{HH} + I_{LL}) S_{LL} - \gamma_H I_{LL} - \alpha_H I_{HH} - \mu_H I_{HH}.
\end{aligned}
$$

(16)

There always exists disease-free equilibrium $E_0(WDH) = (S_W, I_W, S_D, I_D, S_{HH}, T_{HH}, S_{LL}, I_{LL}) = (\frac{\lambda_W}{\mu_W}, 0, \frac{\lambda_D}{\mu_D}, 0, \frac{\lambda_H}{\mu_H})$ in (2).

The next generation matrix at $E_0(WDH)$ is

$$
J_{WDH}(E_0(WDH)) = \begin{pmatrix}
J_W(E_0(W)) & 0 & 0 \\
0 & J_D(E_0(D)) & 0 \\
* & * & J_H(E_0(H))
\end{pmatrix}
$$

with $J_H(E_0(H)) =

$$
\begin{pmatrix}
-\mu_H & -\beta_{HH} \frac{A_{HH}}{\mu_H} & 0 & -\beta_{HH} \frac{A_{HH}}{\mu_H} \\
0 & -\beta_{HH} \frac{A_{HH}}{\mu_H} - \mu_H - \gamma_H - \alpha_H & 0 & -\beta_{HH} \frac{A_{HH}}{\mu_H} \\
0 & -\beta_{HH} \frac{A_{HH}}{\mu_H} & -\mu_H & -\beta_{HH} \frac{A_{HH}}{\mu_H} \\
0 & 0 & 0 & -\beta_{HH} \frac{A_{HH}}{\mu_H} - \mu_H - \gamma_H - \alpha_H
\end{pmatrix}
$$

The characteristic equation of $J_H(E_0(H))$ is

$$
f_H(\lambda) = (\lambda + \mu_H)^2(\lambda^2 - (\beta_{HH} \frac{A_{HH} + A_{LL}}{\mu_H} - 2\mu_H - 2\gamma_H - 2\alpha_H)\lambda - (\mu_H + \gamma_H + \alpha_H)(\beta_{HH} \frac{A_{HH} + A_{LL}}{\mu_H} - \mu_H - \gamma_H - \alpha_H)) = 0
$$

If there is $R_0(H) = (\frac{A_{HH} + A_{LL}}{\mu_H(\mu_H + \gamma_H + \alpha_H)}) < 1$, we have $\beta_{HH} \frac{A_{HH} + A_{LL}}{\mu_H} - 2\mu_H - 2\gamma_H - 2\alpha_H < 0$ and $(\mu_H + \gamma_H + \alpha_H)(\beta_{HH} \frac{A_{HH} + A_{LL}}{\mu_H} - \mu_H - \gamma_H - \alpha_H) < 0$. Then we get all the real parts of eigenvalues of $J_H(E_0(H))$ are negative, $E_0(WDH)$ is stable.

At the same time, the spectral radius of $\begin{pmatrix} R_0(\HHHH) & R_0(\HHLL) \\
R_0(\HLLL) & R_0(\LLHH) \end{pmatrix}$ is $R_0(H) = \frac{(A_{HH} + A_{LL})\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H)}$, with $R_0(\HHHH) = R_0(\HHLL) = \frac{A_{HH}\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H)}$ and $R_0(\HLLL) = R_0(\LLHH) = \frac{A_{LL}\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H)}$.

**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

$$
J_{WDH}(E_0(WDH)) = \begin{pmatrix}
J_W(E_0(W)) & 0 & 0 \\
0 & J_D(E_0(D)) & 0 \\
* & * & J_H(E_0(H))
\end{pmatrix}
$$

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 $E^*_{(W,D,H)} = (\hat{S}_W, \hat{I}_W, \hat{S}_D, \hat{I}_D, \hat{S}_{HH}, \hat{I}_{HH}, \hat{S}_{LH}, \hat{I}_{LH})$ satisfies

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

(17)

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

(18)

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

(19)

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

(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

$$\hat{S}_H = \frac{1}{2\mu_H\beta_{HH}}[(A_{HH} + A_{LH})\beta_{HH} + (\gamma_H + \alpha_H + \mu_H)(\mu_H + \eta_S\beta_{WH}\hat{I}_W$$

$$+ \eta_S\beta_{DH}\hat{I}_D)] - \frac{1}{2\mu_H\beta_{HH}}[(A_{HH} + A_{LH})^2\beta_{HH}^2 + 2(A_{HH} + A_{LH})\beta_{HH}$$

$$(\gamma_H + \alpha_H + \mu_H)(\eta_S\beta_{WH}\hat{I}_W + \eta_S\beta_{DH}\hat{I}_D - \mu_H) + (\gamma_H + \alpha_H + \mu_H)^2$$

$$(\mu_H + \eta_S\beta_{WH}\hat{I}_W + \eta_S\beta_{DH}\hat{I}_D)]^2$$

Similarly to the calculation of Theorem 2.2, we have

$$g_2(\hat{S}_H) = \mu_H\beta_{HH}\hat{S}_H + (A_{HH} + A_{LH})(\mu_H + \gamma_H + \alpha_H) - [(A_{HH} + A_{LH})\beta_{HH}$$

$$+ (\mu_H + \gamma_H + \alpha_H)(\mu_H + \eta_S\beta_{WH}\hat{I}_W + \eta_S\beta_{DH}\hat{I}_D)]\hat{S}_H$$

$$= 0$$

So if $R_0(W,D) > 1$, there exists epidemic equilibrium $E^*_{(W,D,H)}$ in (2). The next generation matrix at $E^*_{(W,D,H)}$ is

$$J_{WDH}(E^*_{(W,D,H)}) = \left( \begin{array}{ccc} J_W(E^*_{(W)}) & 0 & 0 \\ * & J_D(E^*(D)) & 0 \\ * & * & J_H(E^*_{(H)}) \end{array} \right)$$
with

$$J_H(E_{1h}) = \begin{pmatrix} J_{11} & -\beta_{HH}\hat{S}_{HH} & 0 & -\beta_{HH}\hat{S}_{HH} \\ J_{21} & J_{22} & 0 & 0 \\ 0 & -\beta_{HH}\hat{S}_{HH} & J_{33} & -\beta_{HH}\hat{S}_{HH} \\ 0 & \beta_{HH}\hat{S}_{HH} & \beta_{HH}(\hat{I}_{HH} + \hat{I}_{LH}) & J_{44} \end{pmatrix}$$

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

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

$$J_{22} = \beta_{HH}\hat{S}_{HH} - \gamma_H - \alpha_H - \mu_H$$

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

$$J_{44} = \beta_{HH}\hat{S}_{LH} - \gamma_H - \alpha_H - \mu_H$$

The characteristic equation of $J_H(E_{1h}^*)$ is

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

with

$$a_1 = \frac{A_{HH}}{S_{HH}} + \frac{A_{LH}}{S_{LH}} - \beta_{HH}(S_{HH} + S_{LH}) + 2(\gamma_H + \alpha_H + \mu_H),$$

$$a_2 = (\frac{A_{HH}}{S_{HH}} + \frac{A_{LH}}{S_{LH}})(\gamma_H + \alpha_H + \mu_H) - \mu_H\beta_{HH}(\hat{S}_{HH} + \hat{S}_{LH})$$

$$- \beta_{HH}\hat{S}_{HH}\hat{S}_{LH} + (\frac{A_{HH}}{S_{HH}} + \beta_{WH}\frac{\hat{I}_{W}}{I_{HH}} - \beta_{HH}\hat{S}_{HH} + \hat{S}_{LH})$$

$$+ \beta_{HH}\frac{\hat{I}_{LH}}{I_{HH}}\hat{S}_{HH}(\frac{A_{LH}}{S_{LH}} + \beta_{HH}\frac{\hat{I}_{HH}}{I_{LH}}\hat{S}_{LH}),$$

$$a_3 = \frac{A_{LH}}{S_{LH}}(\frac{A_{HH}}{S_{HH}} + \gamma_H + \alpha_H + \mu_H)(\gamma_H + \alpha_H + \mu_H)$$

$$- \mu_H\beta_{HH}\hat{S}_{LH}(\frac{A_{HH}}{S_{HH}} + \gamma_H + \alpha_H + \mu_H) - \beta_{HH}\hat{S}_{HH}\frac{A_{LH}}{S_{LH}}(\gamma_H + \alpha_H + \mu_H)$$

$$+ \alpha_H + \mu_H) + \frac{A_{HH}}{S_{HH}}(\frac{A_{LH}}{S_{LH}} + \gamma_H + \alpha_H + \mu_H)(\gamma_H + \alpha_H + \mu_H)$$

$$- \mu_H\beta_{HH}\hat{S}_{HH}\frac{A_{LH}}{S_{LH}}(\gamma_H + \alpha_H + \mu_H) - \beta_{HH}\hat{S}_{LH}\frac{A_{HH}}{S_{HH}}(\gamma_H + \alpha_H + \mu_H)$$

$$a_4 = \frac{A_{HH}}{S_{HH}}\frac{A_{LH}}{S_{LH}}(\gamma_H + \alpha_H + \mu_H)^2 - \mu_H\beta_{HH}(\hat{S}_{HH}\frac{A_{LH}}{S_{LH}} + \hat{S}_{LH}\frac{A_{HH}}{S_{HH}})(\gamma_H + \alpha_H + \mu_H)$$

It is assumed that $\mu_H \approx 0$ for $\mu_H$ is much smaller than other parameters. Then we get $a_1 > 0$ $a_2 > 0$ $a_3 > 0$ $a_4 > 0$ $b_1 = \frac{\mu_{aw} - \alpha_{aw}}{\alpha_{aw}} > 0$ and $c_1 = \frac{b_{aw} - \alpha_{aw}}{\alpha_{aw}} > 0$. So if $E_{WH}^*$ exists, all of the real parts of eigenvalues of $J_H(E_{1h}^*)$ are negative according to Routh–Hurwitz stability criterion.

In conclusion, if $R_0(WW) = \frac{A_{WW}}{\mu_W(\mu_W + \gamma_W + \alpha_W)} > 1$, 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 $R_0(WW) = \frac{A_{WW} \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)} > 1$, 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).

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

In (25), we get the control reproductive number in wildlife is $R_{1(WW)} = \frac{A_{WW} \beta_{WW}}{\mu_W (\mu_W + \delta)}$, 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 $R_{1(H)} = \frac{A_{HH} + A_{LL}}{\gamma_H + \alpha_H}$.

The epidemic equilibrium of $I_W$, $I_D$ and $I_H$ are

\[
\begin{align*}
\dot{I}_W &= \frac{1}{\beta_{WW} - \mu_W} \left( \frac{1 - \theta_D}{\beta_{WD} \mu_D + \gamma_D + \alpha_D} \right) \dot{I}_D = \frac{1 - \theta_H}{\beta_{WH} \mu_W + \gamma_W + \alpha_W} \dot{I}_D = \frac{1 - \theta_H}{\beta_{WH} \mu_W + \gamma_W + \alpha_W} \left( \frac{\gamma_H I_{IH} + \alpha_H I_{IH} - \mu_H I_{IH}}{\gamma_H + \alpha_H} \right),
\end{align*}
\]

Strategy 1.

For the epidemic equilibrium of $S_W$, $S_D$ and $S_H$, we get $\dot{S}_W = \mu_W S_W - \beta_{WW} I_W S_W > \dot{S}_W$, $\dot{S}_D > \dot{S}_D$ and $\dot{S}_H > \dot{S}_H$ for $g_1(S_D^1) < 0$ and $g_2(S_H^1) < 0$.

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

**Strategy 2.**

In (4), we get the control reproductive number in wildlife is $R_{2(WW)} = \frac{A_{WW} \beta_{WW}}{\mu_W (\mu_W + \gamma_W + \alpha_W)}$, the control reproductive number in domestic animals is $R_{2(DD)} = \frac{A_D \beta_{DD}}{\mu_D (\mu_D + \gamma_D + \alpha_D + \Delta)}$, the control reproductive number in humans is $R_{2(H)} = \frac{A_{HH}}{\mu_H}$. 
If Theorem 2.5.

Strategy 3.

The epidemic equilibrium of $I_W$, $I_D$ and $I_H$ are $\hat{I}_W^2 = \frac{1}{\beta_{WW}} (-\mu_W + \frac{A_{W}W}{\mu_W+\gamma_W+\alpha_W})$, $\hat{I}_D^2 = \frac{\beta_{WD}W^2\hat{S}_D^2}{\mu_D+\gamma_D+\alpha_D+\Delta_D} - \frac{\beta_{WW}W^2\hat{S}_D^2}{\mu_W+\gamma_W+\alpha_W}$ and $\hat{I}_H^2 = \frac{1}{\gamma_H+\alpha_H+\mu_H-\beta_{HH}\hat{S}_H}

(\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)$ in strategy 2.

For the epidemic equilibrium of $S_W$, $S_D$ and $S_H$, we get $\hat{S}_W^2 = \frac{\mu_W+\alpha_W}{\beta_{WW}} = \hat{S}_D$, $\hat{S}_D^2 > \hat{S}_D$ and $\hat{S}_H > \hat{S}_H$ for $g_2(\hat{S}_H^2) < 0$.

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

$$g_3(\hat{S}_D^2) = \mu_D\beta_{DD}(\hat{S}_D^2)^2 - \left[A_D\beta_{DD} + (\mu_D + \gamma_D + \alpha_D + \Delta_D)(\mu_D + \beta_{WD}\hat{I}_W)\right]\hat{S}_D^2 + A_D(\mu_D + \gamma_D + \alpha_D + \Delta_D) = 0$$

So we have

$$g_3(\hat{S}_D) = \mu_D\beta_{DD}\hat{S}_D^2 - \left[A_D\beta_{DD} + (\mu_D + \gamma_D + \alpha_D)(\mu_D + \beta_{WD}\hat{I}_W)\right]\hat{S}_D + A_D(\mu_D + \gamma_D + \alpha_D) = 0$$

If

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

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

$$g_3(\hat{S}_D) = \left[A_D\beta_{DD} + (\mu_D + \gamma_D + \alpha_D)(\mu_D + \beta_{WD}\hat{I}_W)\right]\hat{S}_D - A_D(\mu_D + \gamma_D + \alpha_D) - \left[A_D\beta_{DD} + (\mu_D + \gamma_D + \alpha_D + \Delta_D)(\mu_D + \beta_{WD}\hat{I}_W)\right]\hat{S}_D + A_D(\mu_D + \gamma_D + \alpha_D + \Delta_D) = -\Delta_D(\mu_D + \beta_{WD}\hat{I}_W)\hat{S}_D + \Delta_D A_D = \Delta_D\beta_{DD}\hat{I}_D\hat{S}_D > 0.$$

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 $E_{2(\beta_{WH}\hat{I}_W^2\hat{S}_H^2 + (1 - \Theta_H)\eta_S\beta_{DH}\hat{I}_D^2\hat{S}_H^2)}^2$ in system (4) is stable. If $R_{2(WW)} > 1$, there exists epidemic equilibrium $E_{2(\beta_{WH}\hat{I}_W^2\hat{S}_H^2 + (1 - \Theta_H)\eta_S\beta_{DH}\hat{I}_D^2\hat{S}_H^2)}^2$ and $E_{2(\beta_{WH}\hat{I}_W^2\hat{S}_H^2 + (1 - \Theta_H)\eta_S\beta_{DH}\hat{I}_D^2\hat{S}_H^2)}^2$ 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 $\varphi(1) = \rho(1_H + 1_{LH})$ to simplify the model.
Equilibrium \( E \) and \( L \) are stable if the following conditions are satisfied:

\[
\begin{align*}
\dot{S}_H &= A_{HH} - \beta_{HH}I_HS_H - \beta_{DH}I_DS_H - \beta_{HH}(I_{HH} + I_{LL})S_H - \mu_HS_H - \rho(I_{HH} + I_{LL})S_H + \gamma_{HH}O_{HH}, \\
\dot{O}_{HH1} &= \rho(I_{HH} + I_{LL})S_H - \gamma_{HH}O_{HH1} - \mu_HO_{HH1}, \\
\dot{I}_{HH} &= \beta_{WH}I_WS_H + \beta_{DH}I_DS_H + \beta_{HH}(I_{HH} + I_{LL})S_H \\
&\quad - \mu_{IHH} - \alpha_{HHHH} - \mu_{HH} - \sigma_{IHH}, \\
\dot{O}_{HH2} &= \sigma_{IHH} - \gamma_{HHH}O_{HH2} - \mu_HO_{HH2}, \\
\dot{S}_L &= A_{HH} - \beta_{HH}(I_{HH} + I_{LL})S_L - \mu_HS_L - \rho(I_{HH} + I_{LL})S_L \\
&\quad + \gamma_{HH}O_{HL1}, \\
\dot{O}_{LH1} &= \rho(I_{HH} + I_{LL})S_L - \gamma_{HH}O_{LH1} - \mu_HO_{LH1}, \\
\dot{I}_L &= \beta_{HH}(I_{HH} + I_{LL})S_L - \mu_I_{IHH} - \alpha_{HH}I_{LL} - \mu_HI_{LL} - \sigma_{IHL}, \\
\dot{O}_{LH2} &= \sigma_{IHL} - \gamma_{HHH}O_{LH2} - \mu_HO_{LH2}.
\end{align*}
\]

We get the control reproductive number in humans as:

\[
R_{3(H)} = \frac{(A_{HH} + A_{HL})\beta_{HH}}{\mu_H(\mu_H + \gamma_H + \alpha_H + \sigma)}
\]

**Theorem 2.6.** If \( R_{0(WW)} < 1 \), \( R_{0(DD)} < 1 \) and \( R_{3(H)} < 1 \), the disease-free equilibrium \( E_{0(WDH)}^3 \) 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 | \( R_{0(WW)} = \frac{R}{\mu_H} \) | \( R_{1(WW)} = \frac{R}{\mu_H} \) | \( R_{2(WW)} = \frac{R}{R_{0(WW)}} \) | \( R_{3(WW)} = \frac{R}{R_{0(WW)}} \) |
| Reproductive number in domestic animals | \( R_{0(DD)} = \frac{R}{\mu_E} \) | \( R_{1(DD)} = R_{0(DD)} \) | \( R_{2(DD)} = \frac{R}{R_{0(DD)}} \) | \( R_{3(DD)} = \frac{R}{R_{0(DD)}} \) |
| Reproductive number in humans | \( R_A = \frac{R}{\mu_H} \) | \( R_1(H) = R_0(H) \) | \( R_2(H) = R_0(H) \) | \( R_3(H) = \frac{R}{R_{1(H)}} \) |

\[
\begin{align*}
R_{1} &= A_{WW}B_{WW}, \\
R_{2} &= B_{HH}(\mu_W + \gamma_W + \alpha_W), \\
R_{3} &= (\mu_W + \delta)(\mu_W + \gamma_W + \alpha_W + \delta), \\
R_{4} &= A_{DD}B_{DD}, \\
R_{5} &= \mu_D(\mu_D + \gamma_D + \alpha_D), \\
R_{6} &= (A_{HH} + A_{HL})\beta_{HH}, \\
R_{7} &= \mu_H(\mu_H + \gamma_H + \alpha_H), \\
R_{8} &= \mu_H(\mu_H + \gamma_H + \alpha_H + \sigma).
\end{align*}
\]

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 $R_0(W,D) = 0.1 \times R_0(D,D)$, $R_0(W,H) = 0.1 \times R_0(H)$ and $R_0(D,H) = 0.1 \times R_0(H)$.

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 $\delta, \Delta_I, \sigma$ in Strategy 1, Strategy 2 and Strategy 3 on control reproductive numbers have been shown in FIGURE 4. The existence of parameters $\delta, \Delta_I, \sigma$ would decrease the value of $R_1(W,W), R_2(D,D)$ and $R_3(H)$. If $\delta < 0.142 \times 10^{-3}$, $\Delta_I < 0.258$ and $\sigma < 0.066$, $R_1(W,W), R_2(D,D)$ 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 $\delta, \theta_D, \theta_H, \Delta_I, \Theta_H, \sigma$ and $\rho$ 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 Strategy 3 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 $\delta$ on $R_{1(WW)}$ in a. $R_{1(WW)} = 1$, when $\delta = 0.142 \times 10^3$. The effect of $\Delta_I$ on $R_{2(DD)}$ in b. $R_{2(DD)} = 1$, when $\Delta_I=0.258$. The effect of $\delta$ 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 $I_{LN} + I_{HH} < I_{HC}$, $\varepsilon_W = 0$; when $I_{LN} + I_{HH} \geq I_{HC}$, $\varepsilon_W = 1$. $\varepsilon_D = 0$, $\varepsilon_H = 0$. Strategy 2: when $I_{LN} + I_{HH} < I_{DC}$, $\varepsilon_D = 0$; when $I_{LN} + I_{HH} \geq I_{DC}$, $\varepsilon_D = 1$. $\varepsilon_W = 0$, $\varepsilon_H = 0$. Strategy 3: when $I_{LN} + I_{HH} < I_{HC}$, $\varepsilon_H = 0$; when $I_{LN} + I_{HH} \geq I_{HC}$, $\varepsilon_H = 1$. $\varepsilon_W = 0$, $\varepsilon_D = 0$. ($\delta=0.1$, $\theta_D=0.1$, $\theta_H = 0.1$, $\Delta_H = 1$, $\Theta_H = 0.1$, $\sigma = 0.01$ 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 26th day in a, 17th 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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Figure 6. The effects of $\delta, \theta_D, \theta_H, \Delta_I, \Theta_H, \sigma$ and $\rho$ on the number of infected humans ($I_H$) in the first 90 days (a in a, $\theta_D$ in b, $\theta_H$ in c, $\Delta_I$ in d, $\Theta_H$ in e, $\sigma$ in f and $\rho$ in g) (high risk group: low risk group=1:9)
Table 2. Parameter definitions and their values for avian influenza in China.

| Parameter         | Definitions                                                                 | Values               | Sources |
|-------------------|------------------------------------------------------------------------------|----------------------|---------|
| $A_W$             | birth or immigration rate of wild aquatic birds                              | 0.137 birds/day      | Est.    |
| $\mu_W$           | natural mortality rate of wild aquatic birds                                 | 0.000137/day         | [33]    |
| $\gamma_W$        | recovery rate of wild aquatic birds                                          | 0.25/day             | Est.    |
| $\alpha_W$        | disease-induced mortality rate of wild aquatic birds                         | 0.0025/day           | Est.    |
| $A_D$             | birth or immigration rate of domestic birds                                  | 48.72 birds/day      | [33]    |
| $\mu_D$           | natural mortality rate of domestic birds                                     | 0.0058/day           | [33]    |
| $\gamma_D$        | recovery rate of domestic birds                                              | 0.25/day             | [26]    |
| $\alpha_D$        | disease-induced mortality rate of domestic birds                             | 0.0025/day           | Est.    |
| $A_{HH} + A_{LH}$ | birth or immigration rate of humans                                          | 0.07 people/day      | [23]    |
| $\mu_H$           | natural mortality rate of humans                                             | 0.000035/day         | [23]    |
| $\gamma_H$        | recovery rate of humans                                                      | 0.33/day             | [26, 31]|
| $\gamma_H1$       | remove rate from isolation compartment to susceptible compartment.           | 0.5/day              | Est.    |
| $\gamma_H2$       | remove rate from isolation compartment to recovery individual compartment.   | 0.5/day              | Est.    |
| $\alpha_H$        | disease-induced mortality rate of humans                                     | 0.0033/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]    |
| $\beta_{WD}$      | per capita incidence rate from wild aquatic birds to domestic birds          | $6.15 \times 10^{-6}$| Est.    |
| $\beta_{WH}$      | per capita incidence rate from wild aquatic birds to humans                 | $2 \times 10^{-5}$   | Est.    |
| $\beta_{DH}$      | per capita incidence rate from domestic birds to humans                     | $2 \times 10^{-5}$   | Est.    |
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