

Problem Chosen

A

**2020
MCM/ICM
Summary Sheet**

Team Control Number

2014906

**Novel Coronavirus Pneumonia: Aloha and to Say
Aloha**

Summary

Begin Abstract...

Keywords: Begin Keywords...

Novel Coronavirus Pneumonia: Aloha and to Say Aloha

February 15, 2020

Contents

1	Introduction	3
1.1	Problem Background	3
1.2	Our Work	3
2	Assumptions and Justifications	4
3	Notations and Definitions	4
4	Model Analysis	5
4.1	Sea Surface Temperature Forecasting Model	5
4.1.1	SST Based on Gray Forecasting	6
4.1.2	Influence of Global Warming	7
4.2	Fish Immigration	7
5	title	9
5.1	Background Illustration	9
5.2	subsection1	9
5.3	subsection2	9
6	S6	9
6.1	Simulation Results	9
7	Model Analysis	9
7.1	Sensitivity Analysis	9
7.2	Strengths and Weaknesses	9
7.2.1	Strengths	9
7.2.2	Weaknesses	9

8 Conclusion	9
Appendices	11
Appendix A Solve and Plot Ordinary Differential Equation Systems	11

1 Introduction

1.1 Problem Background

The ocean is covering approximately 1.3 billion cubic kilometers on Earth, which is the equivalent of 97 percent of total water. Playing a vital role in oxygen producing, carbon sequestering and food providing for billions of people, the ocean's abilities are being eroded by plenty of stressors.

Since the 1950s, the oceans have absorbed more than 93 percent of all the heat that produced by human, which is beginning to show its drawbacks at a price. Marine organisms, as an important part in the ocean, can be affected when there are changes in their natural habitat as well as changes in ocean chemistry. As a primary producer of the food chain in the ocean, phytoplankton, a kind of marine plants, is facing an incrementally decrease if the water becomes warmer, which results in the reduction of the amount of nutrients in the ocean. The scarcity of nutrients will limit the growth of marine organisms in the food chain. In addition, temperature plays an important role in controlling biological rhythm of marine plants and animals involving birth, growth, feeding, production and death. The changes in ocean temperature will cause a disruption of biological rhythm making marine organisms become vulnerable to illness. Moreover, sequence lowering of the pH of seawater is unfavorable to the hatching of eggs and the growth of fry of some sea fish.

The anticipated increase in ocean temperature is predicted to stimulate the migration of marine organisms based on their temperature tolerance, with heat-tolerant species expanding their range northward and those less tolerant species retreating. Just as the lobster of Maine, USA, thousands of aquatic species is forced to migrating north to find a better habitat. As an impact, relative companies need to move their base to survive.

1.2 Our Work

Based on our understanding of the problem, we make Gray Prediction Model as well as Differential Equations to predict the location of fish. Sequenced analysis are based on our prediction.

The remainder of this article is organized as follows. In Section 2 and Section 3, we put forward main assumptions and notations in this paper. Section 4 shows our predicting process. Section 5In Section 6, In Section 7, we discuss the strengths and weaknesses of our models. And finally, we conclude the paper in Section 8. The major logic flow is depicted as follows, in Figure 1:

Figure 1: Major Logic Flow of our Model - ALOHA

2 Assumptions and Justifications

We make some general assumptions to simplify our model. We will list all of the assumptions together with corresponding justification below:

- 1) **All of human beings are susceptible to NCP.** NCP was recently discovered and has gene specificity, which make it difficult to be immunized by the body. Thus, we can infer that human beings are susceptible to this virus without previous contact.
- 2) **People will not get susceptible to secondary infection of NCP.** Depending on the nature of the body's immune system, after recovery most people will have better immune reaction to virus. We ignore the possibility of secondary infection to simplify our model.
- 3) **Assume that incubation period to different individuals are the same.** Since incubation period of different host individual is different and also hard to predict, we assume that incubation periods are the same.
- 4) **NCP will only complete the process of human-to-human transmission via close contact such as talking face-to-face.** We simplify our model here since people suffer a rather low level of risk if they do not have close contact with the infected when it comes to NCP.
- 5) **The number of infected people around the world is almost the number of infected people in China.** Although countries except China accounts for 71.57% of the world population (till August 28th, 2019, by United Nations [?]), only less than 0.77% of confirmed cases occurred outside China (till February 3rd, 2020, by various organizations [?][?][?]; with detailed information listed in Appendix). Meanwhile, after NCP was discovered in China, all the other countries took immediate action against the spread. Therefore, what deserves our attention is more the circumstances in China.

3 Notations and Definitions

For convenience, we define a series of notations for mathematical usage, which are listed in Table 1 to describe our model.

Table 1: Notations mentioned in this paper

Symbol	Definition
$S(t)$	Number of susceptible person at time t
$E(t)$	Number of enfective person at time t
$I(t)$	Number of infectious person at time t
$R(t)$	Number of removed person at time t
t	Time
N	Gross population in this model
α	Morbidity of NCP
β	Infection rate
γ	Removal rate
d	Mortality of NCP
h	Recovery rate of NCP
k	Average number of people exposed to infections every day
b	Probability of infection by contact
R_0	Basic reproduction number
m_i	Affecting parameters of i th measure by government
ND	Number of death
NR	Number of recovery
NSP	Number of suspected people
NDP	Number of diagnosed people

4 Model Analysis

In this section, we present the **Gray Forecasting** method to simulate the change of the average sea surface temperature of a region on the ocean. Also, we use **Differential Equation** to introduce the influence of global warming to the temperature. The framework of the model is shown in Figure 1. Referencing the temperature data onto recent years, we calculate the factors of our model using Gray Forecasting, which can identify the degree of variation in development trends among system factors. After conducting the solution by Gray Forecasting, we additional add the global warming factor of the heavier of carbon emission. Finally, we analyze the immigration of these two fish species as well as the most likely location for these two fish species over the next 50 years.

4.1 Sea Surface Temperature Forecasting Model

The temperature on the ocean always gets affected by a huge number of factors, including solar radiation, ocean currents, carbon emission, Nino phenomenon, etc. To avoid the complex analysis and simplify the problem, we use a gray forecasting model to calculate the correlation degree of factors. Besides, the acceleration of global warming, extra differential equations are needed as correction of our model.

4.1.1 SST Based on Gray Forecasting

We first denote the symbols and terms used in this part.

Table 2: Symbols and Terms

Symbol	Definition
<i>AGO</i>	Accumulated Generating Operation
<i>GM(1,1)</i>	Gray Model
<i>IAGO</i>	Inverse Accumulated Generating Operation
...	...

According to the already existing SST data, we initialize our forecasting model using monthly statistics from January, 2006 as the first month to January, 2020 .

Assuming that the SST in the original k -th month is $x^{(0)}(k)$, set the original data as follows:

$$x^{(0)} = x^{(0)}(1), x^{(0)}(2), \dots, x^{(0)}(n)$$

We denote

$$x^{(1)}(k) = \sum_{i=1}^k x^{(0)}(i) \quad , where \quad k = 1, 2, \dots, n$$

Then, we calculate the class ratio λk

$$\lambda(k) = \frac{x^{(0)}(k-1)}{x^{(0)}(k)}$$

Due to all $\lambda k \in [0.982, 1.0098]$, we can reach a fatisfying model using GM(1,1). By *AGO*,

$$x^{(1)}(k) = \sum_{i=1}^k x^{(0)}(i) \quad , where \quad k = 1, 2, \dots, n$$

Take differential, i.e.

$$d(k) = x^{(0)}(k) = x^{(1)}(k) - x^{(1)}(k-1)$$

Denote $z^{(1)}(k)$ as the adjacent value generated sequence, i.e.

$$z^{(1)}(k) = \alpha x^{(1)}(k) + (1 - \alpha)x^{(1)}(k-1)$$

So, we get the model as

$$d(k) = \alpha z^{(1)}(k) = b$$

We then construct Data Matrix B and Data Vector Y

$$B = \begin{bmatrix} -\frac{1}{2}(x^{(1)}(1) + x^{(1)}(2)) & 1 \\ -\frac{1}{2}(x^{(1)}(2) + x^{(1)}(3)) & 1 \\ \dots & \dots \\ -\frac{1}{2}(x^{(1)}(k) + x^{(1)}(k+1)) & 1 \end{bmatrix}, Y = \begin{bmatrix} x^{(0)}(2) \\ x^{(0)}(3) \\ \dots \\ x^{(0)}(k) \end{bmatrix} \quad (1)$$

$$\hat{\mu} = (a, b)^T = (B^T B)^{-1} B^T Y = ???$$

$$\frac{dx^{(1)}}{dt} + ??x^{(1)} = ??$$

Solution is,

$$x^{(1)}(k+1) = (x^{(1)}(k) - \frac{b}{a})e^{-ak} + \frac{b}{a} = ???$$

Until now, we get the forecasting value of future sea surface temperature.

4.1.2 Influence of Global Warming

Equation 4.1.2 shows the mathematical foundation of the influence of global warming. The parameters are defined as follows:

- ρ indicate the density of the ocean. Since there are only delicate distinction between different region, we let ρ approximately equal to 1.
- h shows the mixed layer depth, which we take

$$h = 70m$$

- C_p shows the specific heat of a part of ocean, which is assumed as 300000000J.
- γ is a parameter defined in terms of the fraction of land. We assume $\gamma \in [0.72, 0.75]$
- λ the climatic feedback factor, which is defined as

$$\lambda = 3.58$$

- δF express the flux lost from the mixed layer, which is ignored in our model.
- δQ shows the net surface flux, which is defined as

$$\delta Q = 34.6534 \times 10^{-4} t e^{8.686 \times 10^{-3} t}$$

- t indicates the number of the year, such as 2018, 2019, etc.

Wait sx...

$$\gamma \rho C_p h \frac{d\delta t}{dt} = \delta Q - \lambda \delta T - \delta F$$

Wait sx...

4.2 Fish Immigration

Based on the water temperature predictions we made in Section 4.1, we can get the main position where these two fish species located over 50 years. In order to simplify our model, we make the following assumptions:

- 1) **A school of fish will not scatter during immigration.** The randomness of scattering will profoundly increase the difficulty.
- 2) **The sea surface temperature can represent the temperature at which fishes live.** Since the ocean is too deep for us to analyze the temperature in different depth, we simply take the surface temperature to present the temperature where fishes live.

Due to the initial place of fish school is unknowable, we choose five largest fishing companies as the most possible region as the starting region. To iteration, we decide to choose eight direction in the map for moving. Each iteration contains a movement towards on e certain direction where the tempture is favorable to the fish school. The algorithm is shown as Algorithm 1.

Algorithm 1: Fish Movement Simulation

input : Fish starting Position (x, y)

output: Fish's Location over 50 Years

Initialize fish starting position (x, y);

for year $\leftarrow 0$ **to** 50 **do**

for *eight direction*(x_i, y_i) **do**

if (x_i, y_i) *is land* **then**

break;

end

if $\|SST(x_i, y_i, \text{year} + 1) - SST(x, y, \text{year})\|$ *reaches minimum* **then**

 position = (x_i, y_i, year)

end

end

end

Thus, after 50 rounds of iteration, a solution figure can easily get.

5 title

5.1 Background Illustration

5.2 subsection1

5.3 subsection2

6 S6

6.1 Simulation Results

7 Model Analysis

7.1 Sensitivity Analysis

7.2 Strengths and Weaknesses

7.2.1 Strengths

- Our model achieves a good simulation in fitting the current infections.
- Our model takes into account the effects of the change of pedestrian flow during the Spring Festival travel rush as well as virus mutations
- The model is a well-organized epidemic model which can be widely applied to other virus disease.

7.2.2 Weaknesses

- The prediction simulation result can't be verified due to the unknown data in future.
- A large number of assumptions reduce the accuracy of the models simulation results.

8 Conclusion

References

Brief Assessment Report

Appendices

Appendix A Solve and Plot Ordinary Differential Equation Systems