Original Research Article

Distracted multinomial model for corona screening at entry ports

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ABSTRACT

Background: On 24 January 2020, 1287 corona cases were noticed in Wuhan, China, causing 41 deaths. Its incubation period is at least 14 days. Now, this deadly virus has spread to other foreign countries. The prevalence of corona cases is changing daily. See www.who.org for daily reports. The corona cases are mystic and nightmare to the public, health professionals, and governing agencies globally.

Methods: The Center for Disease Control (CDC) compiled in their webpage (www.cdc.org) the number of confirmed, number of healthy, and the number of pending cases at the port of entries in United States of America (USA). These numbers are perhaps under-estimates because of inappropriate diagnostics and imprecise incubation period. To resolve the under estimating, this article, introduces a distracted multinomial model to refine the imprecise corona screening process and interpret the probability of detecting a corona case in US entry gates.

Results: An alternate expression (2) for the correlation between the corona ill and corona free cases at the USA ports of entry reveals that it was rising since 31st January 2020, reached its maximum on 5th February 2020, then declined to hit a bottom on 7th February 2020 only to rise again.

Conclusions: Most desirable is an accurate predictability of a traveller with the corona virus at the portal entry to minimize its spread. To make such prediction, a regression is necessary with involvement of covariates like age, body’s immunity level, comorbidity, and precise understanding of its incubation period. The model in this article is the starting point for further future research work.

Keywords: Conditional prediction, Inappropriate diagnostic, Multinomial frequency trend, Model, Probability indices, Prevalence of corona cases, Under-estimate

INTRODUCTION

What is corona virus? It is witnessed to be an extremely acute respiratory deadly airborne disease whose cause or treatment remains unknown even at this time. The genesis of the name corona is traced to a Latin word corona, meaning crown or halo. The corona type illness can be traced to an earlier time in 1960 according to Habibzadeh and Stoneman.1 Recently, during December 2019, the corona illness was first identified by health authorities in Wuhan, capital of Hubei province in China.2,3 As of March 25, 2020, a total of 81,285 confirmed corona cases including 3,287 deaths in China. Among them, a total of 67,801 confirmed cases in Hubei Province (50006 cases in Wuhan) including 3169 death cases (2531 cases in Wuhan).4 The corona virus (also recognized as COVID-19) originally spread to humans from animals but subsequently spread from a human to other humans.5 The time between exposure and disease onset is typically 2 to 14 days.6 The Symptoms may include fever, cough, and shortness of breath.7 The Complications may include pneumonia and acute respiratory distress syndrome. There is no vaccine or specific antiviral treatment, with efforts typically aiming at managing symptoms and supportive therapy.8 Hand washing is recommended to prevent the disease.9 Anyone who is suspected of carrying the virus is advised to stay at home and monitor their health for two weeks, wear a
mask, and seek medical advice by calling a doctor before visiting a clinic.10

Because the illness is seriously contagious, the travellers between the source locations in China and different parts of world are carriers of the corona virus, the corona incidences are noticed in several countries though in less numbers compared to the incidences within China. Since the first incidence of coronavirus (COVID-19) happened on Jan 20, 2020, in South Korea, the number of cases rapidly increased, resulting in 6284 cases including 42 deaths as of Mar 6, 2020. An empirical model on the reporting delay was speculated. The fatality rate was higher among males compared to females and it increased with older age. The implementation of social distancing was highly recommended.11

The World Health Organization (WHO) declared this outbreak of corona virus is a global emergency and recommended travel restrictions to and from Wuhan, Hubei, China. The train stations and airports within Wuhan, Hubei have installed body temperature checking booths to identify carriers and quarantine them. Many countries evacuated their citizens and diplomats from Hubei, China by chartered flights to their home nation. The USA is one of those countries. Around the world, especially in the United States of America (USA), traveller’s from China are checked at the (sea, air) portal entries for the presence of corona virus. Neither an appropriate diagnostic test nor an effective treatment is available at the USA entry ports.11,12

On 30 January 2020, Center for Disease Control (CDC) witnessed the first USA citizen in Chicago, Illinois with corona virus as he returned home from Wuhan, China on 13 January 2020. The virus was confirmed in five other states in USA. As of 31 January 2020, the USA signed an executive order to deny entry to foreign nationals who travelled to China in the past two weeks. On the same day, the CDC declared a mandatory quarantine for 14 days for the 195 American evacuees from Wuhan, China. Even during the incubation periods (3 to 24 days, according to WHO), the virus is infectious from a human to another.7

It is in this scenario of uncertainties, this article introduces an apt model, name it distracted multinomial and exercise it to understand the implications of the corona data (https://www.worldometers.info/usa-coronavirus for details).

As show in Table 1, a summary of the dates, the time, t since 31 January 2020, the numbers $x_1$, $x_2$, $x_3$ of positively, negatively, and pending cases respectively among those, $n = x_1 + x_2 + x_3$ tested cases in USA entry portals. The Figure 1 and Figure 2 depict respectively the chronologic (that is as of the time, t since the beginning 31 January 2020) proportions and index of dispersion (ID) in the three (that is, positive, negative, pending) categories. The ID value is the ratio of the variance to the mean. In the scenario with multinomial frequencies, the ID values are ought to be in the open interval [0,1] as it happened in this data.

### Table 1: Corona at an entry in US (on 31 Jan, 2020, CDC announced that corona is a serious epidemic) (https://www.worldometers.info/usa-coronavirus).

| Date   | $t$ | $x_1$ | $x_2$ | $x_3$ | $n = x_1 + x_2 + x_3$ |
|--------|-----|-------|-------|-------|----------------------|
| 29/1/2020 | -2  | 5     | 68    | 92    | 165                  |
| 31/1/2020 | 0   | 6     | 114   | 121   | 241                  |
| 3/2/2020  | 3   | 11    | 116   | 82    | 260                  |
| 5/2/2020  | 5   | 11    | 706   | 76    | 793                  |
| 7/2/2020  | 7   | 12    | 225   | 100   | 337                  |
| 10/2/2020 | 10  | 12    | 318   | 68    | 398                  |

$x_1$= Positive, $x_2$= negative, $x_3$= pending cases

### Figure 1: Chronic prevalence of corona cases.

### Figure 2: Index of dispersion of corona cases.

The outline of the articles is as follows. In Section 2, author derive and assess properties of the distracted multinomial model in different sequential days since 31 January 2020, construct estimates of the model parameters, and hypothesis testing procedures. In Section 3, author illustrate the results for the USA data over the dates and make comparative judgements on the scrutiny of potential corona cases among the incoming travellers from Hubei, China. In Section 4, author make some
conclusive remarks from data analytic points of view and their relevance for future dealings with respect to any other outbreak of infectious global diseases.

**METHODS**

In an incidence of infectious disease in which a proven diagnostic test is available, the probability for a randomly chosen suspect to get confirmed with the presence or absence of the disease is respectively \( \pi \) and \( 1 - \pi \) because of the dichotomy. The corona virus turns the situation to a trichotomy because the practiced diagnostic(s) create a third category called pending category because of applying repeatedly several diagnostics. It is so because neither the cause nor the real incubation day(s) is mysterious. This reality defines the prevalences in the three categories as follows.

Let \( \pi_1, \pi_2, \) and \( (1 - \pi - \pi_2) \) be the prevalences of corona ill, healthy from corona, and pending with respect to corona virus. Given that \( n \geq 1 \) travelers are subjected to a diagnostic test on a day at a port of entry to USA, then the probability for noticing the numbers \( x_1, x_2, \) and \( x_3 = n - x_1 - x_2 \) in the corona ill, healthy from corona, and pending categories is a multinomial type as in (1).

That is,

\[
\Pr(x_1, x_2, n - x_1 - x_2) = \frac{n!}{x_1!x_2!(n-x_1-x_2)!} \pi_1^{x_1} \pi_2^{x_2} (1-\pi_1-\pi_2)^{n-x_1-x_2};
\]

\( \pi_1, \pi_2 < 1; 0 < \pi_1 + \pi_2 < 1; \)

\( x_1, x_2 = 0, 1, 2, \ldots, n; x_1 + x_2 \leq n. \)

Because neither the causes of corona virus are not clear and the healthcare researchers have not formulated an effective/appropriate diagnostic test for corona infection, the health professionals at screening center of USA entry portals probably try one or another diagnostic in their list for detecting infectious corona virus. In the midst of uncertainty about the aptness of the diagnostic, let \( 0 \leq \phi < 1 \) be the probability that a selected diagnostic test to try is *inappropriate*, then the prevalences get distorted to \( \frac{\pi_1 - \phi}{1 - \phi} \) \( \frac{\pi_2}{1 - \phi} \) \( 1 - \frac{\pi_1 + \pi_2}{1 - \phi} \) with added restrictions,

\( 0 < \pi_1 + \phi < 1, 0 < \pi_2 + \phi < 1; \)

\( 0 < \pi_1 + \pi_2 + \phi < 1 \)

The underlying probability model for the outcomes now changes to,

\[
\Pr(x_1, x_2, n - x_1 - x_2) = \frac{n!}{x_1!x_2!(n-x_1-x_2)!} \left( \frac{\pi_1}{1 - \phi} \right)^{x_1} \left( \frac{\pi_2}{1 - \phi} \right)^{x_2} \left( 1 - \frac{\pi_1 + \pi_2}{1 - \phi} \right)^{n-x_1-x_2};
\]

\( (1 - \frac{\pi_1}{1 - \phi} - \frac{\pi_2}{1 - \phi}) \frac{\phi + \pi_1}{1 - \phi} \) for \( i = 1, 2, 3 \) with an understanding that \( x_i = n - x_1 - x_2 \). When the diagnostic for the classification is appropriate (that is with an appropriate diagnostic, \( \phi = 0 \)), the variance-mean relationship becomes \( \text{Var}(X_i) = \frac{(1-\pi)}{1-\phi} E(X_i) \), a well known identity in binomial/multinomial for \( i = 1, 2, 3 \). The relationship between the variance and expected value reveals that the index of dispersion (that is, the ratio of the variance over the expected value) is

\[
ID_{x_i} = \left\{ \frac{1}{1-\phi} \right\} \left( \frac{\phi + \pi_i}{1 - \phi} \right) \text{ for } i = 1, 2, 3.
\]

Authors designate the quantity \( CDI_i = \left\{ \frac{1}{1-\phi} \right\} \text{ for } i = 1, 2, 3 \) as the aptness level (AL) of the diagnostic(s), where \( \phi = \Pr[\text{diagnostic is inappropriate}] \). When the diagnostic is quite appropriate, note that the parameter, \( \phi = 0 \) and \( CDI_{\phi=0} = 1 \). Analogously, authors designate the quantity \( DC_{x_i} = \left\{ \frac{\phi + \pi_i}{1 - \phi} \right\} \) as the distortive classification (DC) index for category for \( i = 1, 2, 3 \).

When the diagnostic is quite appropriate (that is, \( \phi = 0 \)), the classification index is for \( i = 1, 2, 3 \) is \( DC_{x_i} = \pi_i \). Otherwise, when an inappropriate diagnostic is exercised, an amount of distortiveness occurs and the deviation amount from an ideal classification index \( \pi_i \) is

\[
\left( \frac{1-\phi}{1-\phi} \right) \pi_i - \phi. \]

An interpretation of \( \phi \) in the deviation in the distortiveness is the difference \( CDI_{\phi=0} - CDI_{\phi} \). Notice that this difference is not only the initial amount but also the rate of change in the deviation.
Also, let us focus on the pending category. Note
\[
\text{Var}(X_i) = \text{Var}(n - X_i - X_j) = \text{Var}(X_i + X_j) \text{ which is equal to,}
\]
\[
\text{Var}(X_i) + \text{Var}(X_j) + 2\rho_{X_i, X_j}\sqrt{\text{Var}(X_i)\text{Var}(X_j)},
\]
where \(\rho_{X_i, X_j}\) is the correlation between the number of corona ill and healthy (corona ill free) cases. An easily recognizable is the expected value of the pending category is
\[
E(X_i) = n - E(X_i) - E(X_j) = n(1 - \frac{\pi_i + \pi_j}{1 - \phi}).
\]

Hence, the index of dispersion for the pending category is a weighted average of the indices of dispersion for the corona ill category and the corona virus free healthy category with their expected values as weights. That is
\[
ID_{X_i} = \frac{E(X_i) + ID_{X_j} E(X_j)}{E(X_i) - E(X_j)}. (2)
\]

An implication of the above relationship is summarized in theorem 1 below for the correlation between the corona ill and corona virus free healthy categories.

**Theorem 1:** The correlation, \(\rho_{X_i, X_j}\) between the corona ill cases, \(X_i\) and the corona virus free healthy cases, \(X_j\) is
\[
\rho_{X_i, X_j} = \frac{n - E(X_i) - E(X_j)}{n}ID_{X_i},
\]
where
\[
ID_{X_i} = 1 - \frac{\pi_i}{1 - \phi} = (\frac{\pi_i + \pi_j}{1 - \phi}) \cdot \text{Var}(X_i) \text{ and } E(X_i) \text{ are indices of dispersion of pending category, variance and expected values for } i = 1, 2.
\]

An implication of the theorem is that the correlation \(\rho_{X_i, X_j}\) increases proportionally as the index of dispersion \(ID_{X_i}\) increases with the proportionality equal to \(n - E(X_i) - E(X_j)\). Next, author have to estimate the model parameters. It is quite clear that the maximum likelihood estimate (MLE) is most efficient.\(^{15}\) The MLE has a virtue that the MLE of a function of the parameters is the function of the MLE of the parameters.\(^{14}\) To be specific, the MLEs of \(ID_{X_i} = (1 - \frac{\pi_i}{1 - \phi}) = \frac{n - \pi_i}{n} \cdot \text{Var}(X_i) = s^2_x\), and \(E(X_i) = \bar{x}_i\), where \(s^2_x\) and \(\bar{x}_i\) denote respectively sample variance and mean for category \(i = 1, 2\). The MLE of the inappropriate level of the diagnostic is \(\hat{\phi} = 1 - (\frac{\hat{\pi}_i}{\hat{\pi}_2})\). Substituting these MLEs in the alternate expression for the correlation \(\rho_{X_i, X_j}\) in Theorem 1 and that is
\[
\hat{\rho}_{X_i, X_j} = \frac{n - \bar{x}_i - \bar{x}_j}{n}ID_{X_i}.
\]

| Closing day | Total cases | \(\bar{x}_i\) | \(\bar{x}_j\) | \(s^2_x\) | \(s^2_{xj}\) | \(\frac{\hat{\pi}_i}{1 - \phi}\) | \(\frac{\hat{\pi}_j}{1 - \phi}\) | \(\hat{\phi}\) | \(\hat{ID}_{X_i}\) | \(\hat{\rho}_{X_i, X_j}\) |
|-------------|-------------|---------------|---------------|-----------|-----------|-----------------|-----------------|----------|--------------|-------------------|
| 31/1/2020   | 241         | 5.5           | 91            | 0.5       | 1058      | 0.02            | 0.38            | 0.94     | 0.56         | 0.33              |
| 3/2/2020    | 260         | 7.3           | 116.3         | 10.3      | 2454.3    | 0.03            | 0.45            | 0.93     | 0.62         | 0.33              |
| 5/2/2020    | 793         | 8.2           | 263.7         | 10.2      | 88562.9   | 0.01            | 0.33            | 0.97     | 0.88         | 0.58              |
| 7/2/2020    | 337         | 9             | 256           | 10.5      | 66722.5   | 0.04            | 0.76            | 0.96     | 0.72         | 0.15              |
| 10/2/2020   | 398         | 9.5           | 266.3         | 9.9       | 54018.6   | 0.02            | 0.67            | 0.96     | 0.77         | 0.24              |

**RESULTS**

In this section, author illustrate the alternate expression (2) for the correlation between the corona ill and corona ill free cases under the distorted multinomial model for the USA data in Table 1. The average number of corona cases, \(\bar{x}_i\) is the observed number of corona cases over the total cases.

The average number of healthy cases, \(\bar{x}_j\) is the observed number of healthy cases over the total cases, \(n\). The sample variances \(s^2_x\) and \(s^2_{xj}\) are based on the observed corona and healthy cases as of the chosen closing day.

The mle, \(\frac{\hat{\pi}_i}{1 - \phi}\) are the ratio \(\frac{\bar{x}_i}{n}\) for \(i = 1, 2\). The mle, \(\frac{\hat{\phi}}{1 - \phi}\) portrays the estimated level of inappropriate level of the employed diagnostic to identify the corona ill cases. The mle of the index of dispersion is done using \(\hat{ID}_{X_i} = (1 - \frac{\hat{\pi}_i}{1 - \phi}) = \frac{n - \bar{x}_i}{n}\). The mle of the correlation is
done employing the core result, \( \hat{\rho}_{x_1,x_2} = \frac{n - \overline{x}_1 - \overline{x}_2}{\sqrt{n}} \hat{D}_{x_1} \) of the theorem 1 in this article. The proportionality between the estimated correlation, \( \hat{\rho}_{x_1,x_2} \) and the index of dispersion, \( \hat{D}_{x_1} \) of the pending cases is displayed in figure.

**Figure 3: Proportionality of \( \hat{\rho}_{x_1,x_2} \) versus and \( \hat{D}_{x_1} \).**

The ASC was more in males bilaterally than in females. The difference was statistically significant on right side (p=0.00). The ASS was more in females than in males on right and vice versa on the left side.

**DISCUSSION**

The predictability of a traveler into a category of corona ill is desirable in an entry portal. To make such prediction, a regression would help.\(^{13}\) To construct such a regression, data on covariates are necessary and the data are not in a public domain for several reasons of privacy and practicality. However, the developments in this article about the underlying probability model for the chance mechanism for the outcomes with reference to travelers with corona virus remain the foundation for further research work.

A multivariate logistic curve for the number of deaths in several countries due to corona virus is feasible based on the distorted multinomial model of this article. These require a comprehensive data collection apparatus and it can be implemented perhaps by the WHO in future pandemics. The knowledge saves life.

It will be of great practical importance to understand the cross-country spread of the corona pandemic for the sake of complete global prevention. Also, the cross-regional connections among cities as well as between urban and rural sections within a country are worthwhile in the pursuit of stopping the deadly corona virus.

A space-time correlation analysis of the distance from the epicenter (that is Wuhan in China) and the other global locations the number of corona cases will reveal a lot more the role played by the travellers and the gravity law in the social dynamics sense.

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