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この博士論文全文の閲覧方法については、以下のサイトをご参照ください。
Studies on risk assessment of transboundary swine diseases in Japan

日本における越境性豚伝染病のリスク評価に関する研究

Satoshi Ito
Notes

The contents of chapter I have been published in Pathogens (https://www.mdpi.com/journal/pathogens).


The contents of chapter II have been published in Transboundary and Emerging Diseases (https://onlinelibrary.wiley.com/journal/18651682).

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Abbreviations

ASF: African swine fever
ASFV: African swine fever virus
CI: Confidence interval
CSF: Classical swine fever
CSFV: Classical swine fever virus
DP: Domestic pig
EU: European Union
FMD: Foot-and-mouth disease
GLOBCOVER: Global land cover vegetation
IPPAP: Illegal pork products being brought in international air passengers’ luggage
JAQS: Japan Animal Quarantine Service
JPPA: Japan Pork Producers Association
LSL: Labor Standard Law
MAFF: Ministry of Agriculture, Forestry and Fisheries
MLIT: Ministry of Land, Infrastructure, Transport and Tourism
OIE: World Organisation for Animal Health
PPAP: Pork products being brought in international air passengers’ luggage
QAH: Quality of available habitat
SDE: Standard Deviational Ellipse
VLA: Veterinary Laboratories Agency
WAHIS: World Animal Health Information database interface
WB: Wild boar
Preface

Classical swine fever (CSF) is a contagious viral swine disease which can cause enormous economic loss in the pig industry worldwide [1]. The disease is caused by infection with CSF virus (CSFV), which belongs to the genus Pestivirus (family Flaviviridae). Infected animals exhibit various symptoms and lesions, which vary in severity and may result in high mortality depending on the acute and chronic forms of the disease [2]. In the last decade, several outbreaks of CSF have been reported worldwide: 14 in Asia, 5 in Europe, 4 in Latin America, 3 in the Caribbean, and 1 in Africa [3].

African swine fever (ASF), which is caused by infection with ASF virus (ASFV) belongs to the Asfarviridae family, is a highly contagious hemorrhagic viral disease that infects domestic and wild pigs; like CSF, it can result in substantial economic losses in pig production [4]. At present, an approved vaccine with high efficacy does not exist for ASF; therefore, where the introduction of ASF is possible, detection and culling practices must be implemented even more thoroughly than in CSF cases. Clinical signs and mortality rates vary according to the virulence of ASF and the types/species of pig infected [4]. In recent years, the prevalence of ASF has increased globally, with outbreaks reported in sub-Saharan Africa, Central and Eastern Europe, and Asia. In 2007, ASF was transmitted to Georgia, potentially via contaminated food waste used as pig feed [5], from where it spread to the European Union (EU) in 2014, with outbreaks reported in nine EU territories. After the first confirmed ASF outbreak in China in 2018, the disease spread rapidly inside and outside of the territory; currently, in Asia, ASF outbreaks have been reported in Cambodia, China, Democratic People’s Republic of Korea, Hong Kong, Indonesia, Lao People’s Democratic Republic, Myanmar, the Philippines, Republic of
Korea, Timor-Leste, and Vietnam [6,7].

Territories with land borders tend to suffer from transboundary animal diseases, whereas Japan, an island territory with well-established quarantine measures, has seen a limited number of disease outbreaks. However, foot-and-mouth disease (FMD) emerged in Miyazaki prefecture in 2010 and severely damaged livestock production. The results of phylogenetic analyses indicated that the pathogenic virus isolated in Miyazaki prefecture was closely related to viruses isolated in nearby territories [8]; thus, the disease apparently entered Japan through the movement of people or infected materials from these affected territories. In Japan, the outbreaks were contained in 3 months using a preventive vaccination and the disposal of approximately 290,000 animals [8].

Eight years after the last outbreak of FMD, CSF reemerged in Japan in September 2018; this was the first CSF outbreak since the World Organization for Animal Health gave Japan CSF-free status in 2007 [10]. The first CSF outbreak ever reported in Japan occurred in 1888 and subsequent outbreaks occurred repeatedly for more than a century [9]. Following the use of a live vaccine in Japan, the number of CSF infections decreased substantially, and the last CSF outbreak (before 2018) occurred in 1992. The 2018 spread of CSF in Japan had not been eradicated even one year after the initial disease notification due to the disease being spread mainly by wild boar; the infection continued despite an oral vaccination of wild boar was applied in the first 6 months after the initial notification [11]. During the outbreak, more than 150,000 pigs were slaughtered, and the preventive vaccination for domestic pigs was applied one year after the initial disease notification [11]. The results of phylogenetic analyses revealed that the strain of CSFV isolated in Japan in 2018 showed the highest homology with strains isolated in China and Mongolia in 2014 and 2015 [10], respectively. Therefore, similar to the outbreak of FMD
in 2010, the CSF outbreak in Japan in 2018 was likely to have been introduced from nearby affected territories.

The number of foreign tourists entering Japan increases every year, with tourists from Asia accounting for a large proportion of visitors to Japan [12]. Concurrent with the increase in foreign visitors to Japan, there has been a proportional increase in the number of confiscated animal products that may potentially contain infectious pathogens. In 2018, more than 85% of the animal products confiscated in Japan originated from other Asian territories [13]. Given that ASF has spread rapidly across the world, it is possible that ASFV-contaminated products could have been brought into Japan in large numbers. To prevent the introduction of contagious diseases from other territories, the Japanese Animal Quarantine Service has strengthened border controls by imposing penalty charges, increasing the number of detection dogs, and increasing the number of flights investigated; nevertheless, the risk of contamination remains because incoming travelers and imported commodities cannot all be investigated with 100% confidence.

The present thesis consists of three chapters. In Chapter I, a spatio-temporal analysis was conducted to obtain epidemiological information on the current epidemic of CSF in Japan. Our findings will help to assess the potential risk of disease spread among pig species, including domestic pigs and wild boar, and assist in implementing control measures effectively. In Chapter II, a quantitative assessment of the risk of ASF virus (ASFV) being introduced into Japan via pork products brought in air passengers’ luggage was conducted. Identification of ASFV risks such as origin territories and destination airports in Japan will be useful for improving the risk management schemes and monitoring systems that aim to prevent the introduction of ASFV into Japan. In Chapter III, the risk of domestic pigs and wild boars in prefectures of Japan being exposed to
ASFV via pork products brought in air passengers’ luggage was assessed. The results could be useful for reviewing the current risk reduction activities applicable for ASFV exposure in susceptible animals. The foci of this thesis are the current control measures used in Japan for contagious diseases in animals, which are investigated by epidemiological analysis of the CSF outbreak, and the efficacy of risk assessment for these diseases, which was investigated using a risk assessment for ASF.
Chapter I

Role of wild boar in the spread of Classical swine fever in Japan
Introduction

CSF is caused by infection with the CSFV, which belongs to the genus Pestivirus, family Flaviviridae. CSF is described by the World Organisation for Animal Health as a highly contagious febrile disease with potential for high mortality that causes enormous economic loss in the pig industry worldwide [9]. CSFV is a positive-sense, single-stranded RNA virus with a genome of approximately 12.3 kb, comprising one large open reading frame that encodes a polyprotein and flanked by 5’-untranslated region (5’-UTR) and 3’-untranslated region [14]. During virus replication, the polyprotein is processed by cellular and viral proteases into four structural and nine nonstructural proteins [15]. Outbreaks of CSF have been reported over the past decade in Asia (Bhutan, Cambodia, China, India, Indonesia, Republic of Korea, Lao PDR, Mongolia, Myanmar, Nepal, the Philippines, Thailand, Timor-Leste, and Vietnam), Europe (Latvia, Lithuania, the Russian Federation, Serbia, and Ukraine), Africa (Madagascar), the Caribbean (the Dominican Republic, Guatemala, and Haiti), and Latin America (Bolivia, Colombia, Ecuador, and Peru) [16]. Based on the amino acid sequence of the 5’-UTR and E2 which is one of the structural regions of the protein, CSFVs are classified into three genotypes (1, 2, and 3), and several subgenotypes (1.1–1.4, 2.1–2.3, and 3.1–3, and 3.1-3.4) [17,18]. The virulence of CSFV is categorized via a clinical score into highly virulent, moderately virulent, low virulent, and avirulent [19,20]. Although the CSFV genotype 2.1b isolated from Republic of Korea was highly virulent, the same genotype isolated in Mongolia was moderately virulent [21,22]. Moreover, the recently classified CSFV genotype 2.1d from China was moderately virulent compared to different variants and antigenicity from field strains identified in China in the past [23].
No notifications of CSF were reported in Japan since 1992, and the territory had an 11-year stretch of CSF-free status defined by the World Organisation for Animal Health (OIE) Terrestrial Animal Health Code since 2007. However, CSF reemerged in Japan in September 2018 in Gifu prefecture, which is located in the central part of the main island of Japan. Phylogenetic analysis revealed that the CSFV strain isolated in Japan in 2018 showed the highest identity in the complete E2 gene sequence with Chinese strains isolated between 2011 and 2015, and in the partial 5’-UTR sequence with strains isolated in China and Mongolia in 2014 and 2015 [10].

By the end of August 2019, a total of 39 CSFV outbreaks on pig farms in four prefectures and 1,071 cases in wild boar in seven prefectures have been reported [24]. Despite the implementation of intensive responses, including movement bans of domestic pigs, surveillance, and oral immunization of wild boar, new notifications of CSF cases in both wild boar and domestic pigs were being reported continually [25]. This might indicate that the pathogenic viruses were widely prevalent and persisted in wildlife around the affected area. As the Eurasian wild boar is also susceptible to CSFV, the circulation and persistence of CSFV among food animals and wildlife makes it difficult to carry out effective control measures for eradicating it in affected areas. Due to contact with infected animals and feeds contaminated with contagious pathogens in garbage dumped on the human sphere, naïve wild boar populations are often infected with CSFV [21,26-36]. Before the 1990s, CSF cases in wild boar were rare concerns as infection was detected rapidly due to the high virulence of circulating strains. However, disease detection appears delayed in the current epidemic due to infection with more moderately virulent strains [37]. As a consequence, there have been serious outbreaks of CSF in the wild boar population in Germany. During an outbreak of CSF in Germany from 1993 to
1998, an epidemiological field investigation confirmed that 59% of the primary cases in domestic pigs could be attributed to either direct or indirect contact with infected wild boar [29]. Virus characteristics and population size both can be considered critical factors for the persistence of CSFV, especially in wild boar populations [37]. It has been suggested that CSFV would be self-limiting within one year in populations of 2,000 wild boar, whereas it will persist and become endemic in a larger population [38]. In addition, the population density of wild boar has also been suggested as being a potential factor for the persistence of CSF, because more frequent turnover occurs in dense populations, which provides faster renewal of susceptible piglets that increases the chance that the virus will persist in the population [37]. Once the contagious viruses are transmitted to wildlife, specific control measures for wild boar will be needed to eradicate CSF in the affected area and to contain it more effectively.

The present study conducted a spatio-temporal analysis to obtain epidemiological information on current epidemics of CSF in Japan. Based on the official CSF reports on domestic pig farms and wild boar, notified in Gifu prefecture from September 2018 to June 2019, we assessed the direction of the spread of the disease and identified areas with high densities of notifications. In addition, to identify spatio-temporal aggregation of notifications and to characterize land cover vegetation in areas of disease aggregation, a clustering analysis was conducted, and obtained clusters were then overlapped with quality habitat map. The obtained information can be used to develop more effective disease control measures for application in both domestic pigs and wild boar.
Material and Methods

Data and data sources

Epidemiological data for the periods from September 9, 2018, to June 25, 2019, were provided by the Gifu Prefectural Government, which provided the dates and coordinates (latitude and longitude) of the notifications of CSF in domestic pigs and wild boar. A total of 743 CSF notifications, 16 outbreaks on domestic pig farms, and 727 cases in wild boar were confirmed by RT-PCR and/or ELISA tests in the laboratory [25]. As we focused on local transmission of CSFV, notifications of CSF in slaughterhouses or in facilities through which CSF-affected pigs had been transported were removed from the current study. Notifications of CSF in wild boar reported on the same day and location were regarded as one case.

Standard deviational ellipse analysis

Standard deviational ellipse (SDE) analysis is a tool that provides the orientation and shape of a distribution, as well as its location, and dispersion or concentration of the data [39]. It requires a single point that is used to define the standard deviational ellipse. The analysis was conducted to describe the trend and spatial characteristics of CSF notifications in the study area in ArcGIS 10.6.1 software (ESRI Inc., Redlands, CA, USA) following an approach similar to Fonseca et al. and Lu et al. [40,41]. The ratio (R) of the long and short axes was used to identify the degree of clustering (R>1) or dispersion (R = 1) [40,41]. To analyze temporal changes of CSF notifications, the study period was divided into three stages: (i) September to December 2018 (4 months), (ii) January to March 2019 (3 months), and (iii) April to June 2019 (3 months).

Multi-distance spatial cluster analysis
A multi-distance spatial cluster analysis tool in ArcGIS software version 10.6.1 was used to identify the maximum distance of the relationships between CSF notifications according to the guide on the manufacture’s website [42]. In brief, the tool uses a common transformation of Ripley's k function, wherein the expected result with a random set of events is equal to the input distance. The transformation $L(d)$ is given by the following formula:

$$L(d) = \sqrt{\frac{A \sum_{i=1}^{N} \sum_{j=1, j \neq i}^{N} k(i, j)}{\pi N (N - 1)}}$$

where $A$ is the area, $N$ is the number of events, $d$ is the distance, and $k(i, j)$ is the weight, in which it is 1 when the distance between $i$ and $j$ is less than or equal to $d$ and it is 0 when the distance between $i$ and $j$ is greater than $d$. To analyze the spatial pattern of CSF notifications, Observed K values were compared to the Expected K values of a completely random spatial distribution of CSF notifications with 999 simulations, which is equal to confidence levels of 99.9%.

The Diff K values contain the Observed K values minus the Expected K values. In the present analysis, the Expected K values that yield the highest Diff K values were applied as the maximum distance for relationships between notifications of CSF outbreaks in Gifu prefecture.

**Kernel density estimation analysis**

Kernel density estimation is a non-parametric estimator for describing the spatial extent of a series of events [43]. In the current study, the kernel density tool was applied to explore the influence of the CSF notifications in the study area by calculating the density of CSF notifications in ArcGIS 10.6.1. A radius of 23 km based on results obtained from Ripley’s k function, was applied as the maximum distance for significant
spatial association between CSF notifications. Kernel density estimation was divided into five categories according to the equal interval method.

**Space-time cluster analysis**

A space-time permutation technique was applied to examine the presence of space-time clusters in Gifu prefecture. The upper limit on the geographical size of the cluster was set as 23 km, the minimum time aggregation as 7 days, and the maximum temporal cluster size as 50% of the total study period (default setting) [44]. Monte Carlo process was implemented using 999 replications to test for the presence of candidate clusters ($P < 0.05$). Analyses were conducted in SaTScan software v9.6 (Kulldorff, Boston, MA, USA) [45].

**Quality of available habitat (QAH) within space-time cluster area**

CSF notifications within significant space-time clusters were overlaid on a QAH map to characterize land cover vegetation in areas of disease aggregation. The QAH map developed by Bosch et al. [46] is a cartographic tool previously suggested as a potential tool for managing ASF. Briefly, it is a standardized distribution map based on global land cover vegetation (GLOBCOVER) that quantifies QAH for wild boar [47]. The QAH map provides seven levels of QAH, namely (i) 0, “absent”; (ii) 0.1, “unsuitable”; (iii) 0.5, “worst suitable area”; (iv) 1, “suitable areas for food or shelter”; (v) 1.5, “suitable areas for food and shelter, but used mainly for one or the other”; (vi) 1.75, “suitable areas for food and shelter, but mainly used for food”; and (vii) 2, “suitable areas for both food and shelter.” In addition, the QAH map also differentiates between landscapes such as natural (mainly QAHs 2 and 1.5) and agricultural landscapes (QAHs 1.75 and 1), among others.
Results

Standard deviational ellipse analysis

A standard deviational ellipse analysis was applied to describe the directional trend and dispersion of CSF notifications in the study area throughout the study period. The study covered the period between September 2018 and June 2019, which was divided into three stages (September–December, January–March, and April–June). Figure 1 illustrates standard deviational ellipses and CSF notifications between September 2018 and June 2019 (Figure 1). To indicate the potential explanation for the directional trend of the CSF outbreaks, the ellipses were overlaid on a map of snowfall area in Gifu prefecture obtained from the National Land Information Division, Ministry of Land, Infrastructure, Transport and Tourism [48]. The findings showed that CSF notifications appeared to move northeast while spreading along the border of the snowfall area.

Multi-distance spatial cluster analysis

The multi-distance spatial cluster analysis was applied to explore the maximum distance between cases of CSF notifications. The results indicated that 23 km was the maximum distance of the significant spatial association between CSF notifications in Gifu prefecture. The obtained maximum distance was used in the subsequent analyses.

Kernel density estimation analysis

The kernel density estimation analysis was applied to describe the spatial distribution of the CSF notifications. The analysis showed that the highest density of CSF notifications was located in the southern part of Gifu prefecture (Figure 2) with further expansion to the east. Among the 16 CSF-positive farms, 37.5% were located in areas with very high or high density of notifications, 31.25% in areas of medium density and
31.25% in areas of low density. Moreover, most of the non-affected domestic farms were located in areas with very low density of notifications (80%), followed by areas with low density (20%). The analysis revealed that CSF-positive farms were located in areas with higher density of notifications, whereas the non-affected farms tended to locate in areas with low density.
**Figure 1.** Directional distribution of CSF notifications from September 2018 to June 2019. Standard deviational ellipses (SDEs) identified between September and December 2018, between January and March 2019, and between April and June 2019. Ellipses were overlaid with CSF notifications distinguishing domestic pig (DP) (square) and wild boar (WB) (circle). Ellipses with centroids were combined to indicate the directional trend of the CSF outbreaks.
Figure 2. Density of CSF notifications in Gifu prefecture. The heat map illustrates the estimated kernel density of CSF notifications (notifications/km²) from very high (red) to very low (transparent). Each coloured area indicates the density of CSF notifications per square kilometer: very high (>0.400), high (0.300–0.399), medium (0.200–0.299), low (0.100–0.199), and very low (<0.100). The highest density of CSF notifications was located in the southern part of Gifu prefecture. A very low density of CSF notifications was located in other areas of the prefecture. Locations of CSF-negative pig farms are represented by crosses.
**Space-time cluster analysis**

The space-time permutation analysis was applied to analyze the space-time patterns of the CSF notifications. The analysis identified two significant space-time clusters (P < 0.05) in Gifu prefecture during the study period. Cluster 1, which had a radius of 12.12 km, covered September 9, 2018, to January 13, 2019, and contained 83 notifications, including 4 outbreaks on domestic pig farms. Cluster 2 had a radius of 19.79 km, spanning the period from February 11, 2019, to May 19, 2019, and contained 198 notifications, including three outbreaks in domestic pigs (Table 1).

**QAH within space-time cluster area**

In order to characterize the land cover vegetation within two significant space-time clusters, the clusters were overlaid with a QAH map. The results showed different patterns between cluster 1 and cluster 2 (Figure 3). In cluster 1, 50.6% of CSF notifications were reported in areas at QAH 1, while 31.3% were reported in areas at QAH 1.5 and 18.1% were reported in areas at QAH 2 (Table 2). In cluster 2, 22.7% of CSF notifications were reported in areas at QAH 1, 52.5% were reported in areas at QAH 1.5, 2.5% were reported in areas at QAH 1.75, and 22.2% were reported in areas at QAH 2 (Table 2).

The CSF notifications within clusters 1 and 2 occurred within habitats that included rainfed croplands (QAH 1), a closed (>40%) needle-leaved evergreen forest (>5 m) (QAH 1.5), a mosaic of cropland (50%–70%) and vegetation (grassland/shrubland/forest) (20%–50%) (QAH 1.75), a mosaic of vegetation (grassland/shrubland/forest) (50%–70%) and cropland (20%–50%) (QAH 2), closed (>40%) broadleaved deciduous forest (>5 m) (QAH 2), and closed to open (>15%) mixed broadleaved and needle-leaved forest (>5 m) (QAH 2).
Although different patterns of land cover vegetation were observed between clusters 1 and 2, nearly 50% of CSF notifications within cluster 1 and more than 75% within cluster 2 were notified in QAH 1.5–2, which provides the greatest opportunities for food and shelter for wild boar.
Table 1. Observed and expected notifications, duration, start and end dates, and radius of each space-time cluster detected ($P < 0.05$) in CSF notifications in Gifu prefecture.

<table>
<thead>
<tr>
<th>Cluster</th>
<th>Observed notifications</th>
<th>Expected notifications</th>
<th>Duration (Days)</th>
<th>Start date</th>
<th>End date</th>
<th>Radius (km)</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>83</td>
<td>17.34</td>
<td>124</td>
<td>2018/9/9</td>
<td>2019/1/13</td>
<td>12.12</td>
</tr>
<tr>
<td>2</td>
<td>198</td>
<td>131.87</td>
<td>98</td>
<td>2019/2/11</td>
<td>2019/5/19</td>
<td>19.79</td>
</tr>
</tbody>
</table>

Table 2. Quality of availability habitats (QAH) of CSF notifications within the two identified space-time clusters.

<table>
<thead>
<tr>
<th>QAH Category</th>
<th>Land Cover</th>
<th>Cluster 1</th>
<th>Cluster 2</th>
<th>Cluster 1</th>
<th>Cluster 2</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>DP (n)</td>
<td>WB (n)</td>
<td>Total (n) (%)</td>
<td>DP (n)</td>
</tr>
<tr>
<td>1.0</td>
<td>Rainfed croplands</td>
<td>4</td>
<td>38</td>
<td>42 (50.6)</td>
<td>0</td>
</tr>
<tr>
<td>1.5</td>
<td>Closed (&gt;40%) needleleaved evergreen forest (&gt;5m)</td>
<td>0</td>
<td>26</td>
<td>26 (31.3)</td>
<td>3</td>
</tr>
<tr>
<td></td>
<td>Mosaic cropland (50-70%)</td>
<td>0</td>
<td>0</td>
<td>0 (0.0)</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td>/ vegetation (grassland/shrubland/forest) (20-50%)</td>
<td>0</td>
<td>0</td>
<td>0 (0.0)</td>
<td>0</td>
</tr>
<tr>
<td>2.0</td>
<td>Mosaic vegetation (grassland/shrubland/forest) (50-70%)</td>
<td>0</td>
<td>0</td>
<td>0 (0.0)</td>
<td>0</td>
</tr>
<tr>
<td></td>
<td>/ cropland (20-50%)</td>
<td>0</td>
<td>0</td>
<td>0 (0.0)</td>
<td>0</td>
</tr>
<tr>
<td>2.0</td>
<td>Closed (&gt;40%) broadleaved deciduous forest (&gt;5m)</td>
<td>0</td>
<td>1</td>
<td>1 (1.2)</td>
<td>0</td>
</tr>
<tr>
<td>2.0</td>
<td>Closed to open (&gt;15%) mixed broadleaved and needleleaved forest (&gt;5m)</td>
<td>0</td>
<td>14</td>
<td>14 (16.9)</td>
<td>0</td>
</tr>
<tr>
<td>Total</td>
<td></td>
<td>4</td>
<td>79</td>
<td>83 (100.0)</td>
<td>3</td>
</tr>
</tbody>
</table>

DP: domestic pig. WB: wild boar. $n$: the number of notifications
Figure 3. Locations of the significant space-time clusters of CSF notifications ($P < 0.05$) in Gifu prefecture overlaid on a map of the quality of available habitat (QAH) levels for wild boar. Graduated colors indicate the quality of habitat availability from darker colors (areas with better quality of habitat availability) to lighter colors (areas with worse-quality of habitat availability).
Discussion

From 2018 until August 2019, all notifications of CSF outbreaks in Japan have been made in Gifu prefecture as well as in the surrounding four prefectures. A total of 1,110 notifications had been reported, so far, with 1,071 affecting wild boar and 39 affecting domestic pig farms. The continuous notification of CSF in the area might have been attributed to wide spread of the virus within wild boar populations favored by free animal movements, as well as to the emergence of epidemiologically related domestic pig farms. To prevent the disease spreading in wild boar, control measures including (i) fencing to restrict animal movements, (ii) hunting activities for active monitoring and to reduce susceptible populations, and (iii) disseminating baits for oral immunization, were implemented. However, the efficacy of these strategies has not been confirmed. Therefore, we conducted a spatio-temporal analysis to obtain epidemiological information of the spread of CSF in Gifu prefecture. Results from this analysis could help to increase our understanding of the current CSF epidemic and to contribute strategies for the containment of the disease in domestic pigs and wild boar.

Japan is an island territory which has achieved the status of freedom from several contagious animal diseases by implementing adequate control measures that took advantage of the territory’s geography. Nevertheless, Japan has imported outbreaks of contagious animal diseases from neighboring territories. In 2010, there was an outbreak of FMD in Miyazaki prefecture in the southern part of Japan, which caused extensive losses in animal husbandry [8]. According to the high degree of sequence homology between an original virus isolated in Japan and viruses that were circulating widely in East Asia, it was suspected that the FMD virus might have been introduced via movement
of people or commodities from East Asia [8]. The high homology of genetic sequences between the CSFV isolated in Japan and viruses prevailing in China suggests that the infectious CSFV may have been introduced from China [10]. Potential factors that could have contributed to disease introduction include easy access from the international airport to the affected area, which has regular and direct flights from China, and the relatively high population density of Chinese people in the affected area.

In the present study, standard deviational ellipse analysis was conducted to measure the standard distance of CSF notifications. Shifting the centroids of identified ellipses indicated that the disease notification has spread in a northeast direction. Overlaying the three identified ellipses with a map of snowfall area in Gifu prefecture revealed that the disease spread along the border of the snowfall area. In the south of Gifu prefecture, there is a widespread area of flat land with field crops or animal farms, residential areas, and forests surrounded by mountains to the north. As suggested by other authors [49,50], wild boar do not move to the snowfall or high mountain areas. Therefore, mountains could have acted as an effective geographical barrier to limit wild boar movements and guide the direction of the spread of CSF.

Another concern regarding the spread of the disease is the potential for it to jump to remote areas. During the epidemic, CSFV infections were confirmed on 7 farms that were geographically distant from, but epidemiologically linked, to the farms affected by CSFV (i.e., run by the same owner, supported by the same husbandry company, etc.) [25]. Given the potential for transmission of the virus between pigs on any farms or from wild boar near that farm, the epidemiologically related farms may further expand the spread of disease. This “disordered” spread of disease could affect the accuracy of spatio-temporal analysis by overestimating the maximum distance of significant spatial
association between notifications. During the FMD epidemic in Miyazaki, the disease was confirmed 70 km away from the zone of movement restriction, which could have been caused by vehicle transportation [8]. Unexpected occurrences of disease in epidemiologically related farms would require reviewing farm biosecurity measures, as well as disease monitoring protocols.

In the present study, the results of the multi-distance spatial cluster analysis revealed that the maximum distance of relationship between CSF notifications was 23 km. Because of the small number of CSF outbreaks on domestic pig farms, we estimated the maximum distance of the relationship between notifications of domestic pigs and wild boar. This assumption could have influenced our estimated distance resulting in overestimation due to long distance spread observed on domestic pig farms. Nevertheless, similar approaches have studied another transboundary animal disease, ASF, which shares hosts and most of the transmission mechanisms with CSF [44,51,52]. When comparing our results with other studies, the estimated distance (23 km) was similar to that obtained for notifications of ASF in domestic pigs (15 km) and wild boar (25 km) in Sardinia [44]. This finding may be useful for setting the range of effective surveillance and control zones in the affected area.

The application of cluster analysis to identify areas with significant spatio-temporal aggregation of the ASF outbreaks in Sardinia from 2004 to 2013 indicated four clusters, the largest of which had a radius of 30 km [52]. This does not correspond with the results of another report that identified one cluster with a radius of 3 km in the same area [44]. As discussed in Iglesias et al. methodological differences could have led to the discrepancy [44]. In present study, because of the small number of CSF outbreaks in pig farms, we could not identify the maximum distance for the relationship between
notifications of CSF in pigs alone, but we were able to do it by considering pigs and wild boar together. The discordance between the findings of the two spatio-temporal analyses in Sardinia may suggest that by using mixed data for two species in the present study, we may have overestimated the distance of the spread of disease compared to true distance of transmission in each of the two species. However, we believe that this uncertainty would be acceptable for setting the monitoring area with high efficacy. Thus, these findings may be useful for setting the range of an effective surveillance and control zone.

Data on wild boar cases consisted of animals found dead and/or captured during surveillance activities. Many of wild boar were captured during active surveillance activities by setting traps and conducting hunting activities. Considering that most of the reported wild boar cases were located close to human habitats, the wild boar capture area may have been biased. Therefore, the disease could be wider spread in the area than what has been reported in official notifications, and the identified clusters could have had a shorter radius. Ideally, active virologic surveys should be intensively implemented to decrease the reporting biases by providing more samples to detect low levels of prevalence [53,54]. The Gifu Animal Health Administration has authorized hunting activities to reduce the number of susceptible, as well as potentially infected, individuals. Hunters are a critical group for implementing population control and proper disposal of wild boar carcasses.

According to the investigative report of the affected farms, there were some factors that might have increased the risk of CSFV introduction into affected farms, including (i) improper preparedness against invasion of wild or small animals into farms; (ii) imperfect clothing and boot changes in farms and pig pens, or disinfection of those materials; and (iii) inadequate vehicle disinfection [25]. To prevent contact among each
of the hosts, in addition to raising awareness of disease among farmers and hunters, it is important to improve biosecurity measures in pig farms against CSFV as well as other infectious diseases.

Finally, we analyzed the QAH level of areas within the two identified clusters to characterize land cover vegetation in areas of disease aggregation. According to Bosch et al, a QAH 1 level corresponded to suitable areas for food or shelter for wild boar (mainly agricultural landscapes) [46]. In cluster 1, 50.6% of CSF notifications were reported in areas at QAH 1, whereas in cluster 2, 22.7% of CSF notifications were reported in areas at QAH 1. Considering that frequent direct and indirect contact is likely to occur between both hosts, contagious viruses in wild boar could be transmitted to pigs in the farms due to insufficient biosecurity in the affected farms since wild boar was the suspected source of infection on 80% of affected domestic pig farms in Gifu prefecture during the studied epidemic [25,54]. On the other hand, almost 50% of CSF notifications within cluster 1 and over 75% within cluster 2 were associated with QAH 1.5–2, which mainly corresponded to natural landscapes. These natural areas provided the greatest opportunities for food and shelter for wild boar. In the case of ASF, it has been reported that wild boar can transmit the disease efficiently at local levels within their own population [44,46]. Furthermore, De la Torre et al. suggested that the spread of ASF in Europe was driven by contact between animals from different populations that moved short distances [55]. Although ASF is caused by another virus, given that wild boar play an important role in both diseases, it is plausible to assume that CSF also could have expanded through contact between individual wild boar. Therefore, it would be critical to control wild boar populations and manage wild boar carcasses adequately from the environment to reduce habitat contamination.
Interestingly, the QAH map could also identify routes of CSF introduction or spread, mediated by wild boar, through vegetation or travel corridors. Travel corridors are either unbroken vegetation corridors or patches of habitat that enable animals to travel securely from one habitat to another [46]. These patches of habitat and vegetation corridors could be used as strategic points of vaccination where oral baits could be placed. In Gifu prefecture, the vegetation is composed mainly of broadleaved evergreen and broadleaved deciduous forests, which provide suitable habitat for wild boar [56,57]. Given that the composition of the vegetation in Gifu prefecture is common throughout Japan, it is likely that the disease could spread similarly to other prefectures.

It should be noted that vegetation types and wild boar behavior could vary among geographical features. For example, mountains usually have gentle slopes in Germany, whereas Japanese mountains tend to have precipitous slopes [40]. These topographical differences may require different approaches for control of wild boar populations.

Almost one year has passed since the first notification of the CSF outbreak in Japan, and the spread of the disease has been confirmed mainly in wild boar. Fortunately, CSF outbreaks on domestic pig farms have been limited. Nevertheless, the potential risk of CSF introduction on farms could be high due to limited biosecurity, high number of wild boar cases in the area, and difficulties in implementing disease control measures in wildlife [25]. The results from this study provide information on the current epidemic, which may help improve current approaches for controlling CSF in Japan. Information on the direction and distance of disease spread could help with the implementation of control measures by modifying the area for control and surveillance zones or identifying specific locations for increasing efforts of oral immunization.

Given the potential risk of the ASF introduction from neighboring territories, we
should summarize and disseminate the lessons learned from the current CSF outbreak to achieve the protection of ASF invasion or rapid containment of its occurrence even if it occurred.
Summary

Since September 2018, nearly 900 notifications of CSF have been reported in Gifu prefecture (Japan) affecting domestic pig and wild boar by the end of August 2019. To determine the epidemiological characteristics of its spread, a spatio-temporal analysis was performed using actual field data on the current epidemic. The spatial study, based on standard deviational ellipses of official CSF notifications, showed that the disease likely spread to the northeast part of the prefecture. A maximum significant spatial association estimated between CSF notifications was 23 km by the multi-distance spatial cluster analysis. A space-time permutation analysis identified two significant clusters with an approximate radius of 12 and 20 km and 124 and 98 days of duration, respectively. When the area of the identified clusters was overlaid on a map of habitat quality, approximately 82% and 75% of CSF notifications, respectively, were found in areas with potential contact between domestic pigs and wild boar. The obtained results provide information on the current CSF epidemic which is mainly driven by wild boar cases with sporadic outbreaks on pig farms. These findings will help implement control measures in Gifu prefecture.
Chapter II

Quantitative risk assessment of African swine fever virus introduction to Japan via pork products brought in air passengers’ luggage
Introduction

ASF is one of the most important transmissible infectious diseases in animal health and is caused by infection with the ASFV. ASF is classified as a List A disease in the Old Classification of Diseases Notifiable to the OIE, indicating that it can potentially have serious socio-economic and public health consequences and is of major importance in the international trade of animals and animal products [58].

Direct contact with contaminated fluids or excretion is the main route of ASF spread on a farm, on the other hand, indirect contact with contaminated fomites and feed on infective pork products are other important routes of disease spread over long distances [55,59]. There is currently no effective vaccine against ASFV and it is known to be highly resistant to various environmental conditions—for instance, the viral infectivity can persist for longer than 1,000 days in frozen meat, 110 days in chilled meat, and 1 month in contaminated pig pens [59]. Therefore, early detection and culling and the restriction of pig movements in infected areas are critical to controlling this disease [60].

In recent years, ASF has become prevalent in many parts of the world, including sub-Saharan Africa, Central and Eastern Europe, and Asia. In 2007, ASF was transmitted to Georgia through the Poti Sea Port, potentially via contaminated food waste that was used as pig feed [5], from where it spread to neighboring territories, such as Azerbaijan, Armenia, and the Russian Federation. Since then, wild boar and domestic pig populations in these territories have been infected with ASFV, as well as in Belarus and Ukraine. In 2014, the disease spread to the north and west, reaching nine different EU territories: Lithuania (2014), Poland (2014), Latvia (2014), Estonia (2014), Czech Republic (2017), Romania (2017), Hungary (2018), Bulgaria (2018), and Belgium (2018) [61].
In August 2018, the first notification of an ASF outbreak in East Asia was reported on a domestic pig farm in Liaoning Province, China. Genetic analysis based on the partial sequence of the ASFV p72 gene indicated that the outbreak isolate, ASFV-SY18, belonged to the genotype II group and had an identical nucleotide sequence to the Georgian (Georgia 2007/1), Russian (Krasnodar 2012 and Irkutsk 2017), and Estonian (Estonia 2014) isolates [62], suggesting that it was most likely introduced from epidemic areas in Europe. Since then, more than 130 outbreaks have been reported in China and approximately 1,000,000 domestic pigs have been slaughtered as at May 23, 2019 [63]. However, disease control has not yet been achieved, so the biggest concern to neighboring territories is the potential spread of ASF by human movement and trading with China, which have caused recent outbreaks in Vietnam [64]. Likewise, 30 provinces in Vietnam have reported the presence of ASF and more than 1,300,000 domestic pigs have been culled. Moreover, a positive sample from a domestic pig collected in a slaughterhouse in Hong Kong was recently found [65,66].

There are various potential ASFV introduction pathways into disease-free territories. For disease-free territories, legal or illegal importation of live pigs or pork products by vehicles are considered as important pathways [67]. Furthermore, territories sharing border with ASF-affected territories have potential risk that movement of ASF affected wild boar introduces the disease into their territories [55]. Considering that Japan is an island territory which does not share border with any territories and quarantine measures for live animal import at ports of entry are established, there are a negligible risk of ASFV introduction into Japan via vehicle or wild boar movements or via live pig imports, but might be a potential risk via pork products. The number of foreign tourists to Japan increases each year, with tourists from the Asian region accounting for a large
The proportion of these visitors [68]. The number of confiscated animal products at airports is proportional to the number of foreign visitors, and in 2016, more than 80% of the confiscated animal products originated from Asian region, with 42% coming from China, 11.4% from Taiwan, and 10.1% from Vietnam [69]. Pork products contaminated with ASFV that were being illegally carried in international passengers’ luggage have been detected at border controls in several ASF-free territories, including Republic of Korea, Taiwan, Thailand, Australia, and Japan [70-74]. As of April 2, 2019, two isolations of ASFV from pork products brought from mainland China into Japan were reported [74], indicating that passengers’ luggage could be a potential route of ASFV introduction to this territory. The Japan Animal Quarantine Service (JAQS) plays an important role in preventing the incursion of animal diseases [75]. However, it is no easy task to confiscate all animal products that are illegally brought into airports. Therefore, limitations of the current quarantine systems, as well as a failure to detected illegally transported animal products could exacerbate the potential risk of foreign animal diseases being introduced to Japan.

Japanese authorities are concerned about the economic losses and animal health impacts that would result from the arrival of animal transboundary infectious diseases. In 2010, there was an outbreak of foot and mouth disease in Miyazaki prefecture, Japan, which affected a total of 292 farms and resulted in nearly 290,000 animals being culled (including vaccinated animals) [76]. This outbreak took 3 months to be contained and had an estimated economic impact of approximately 235 billion JPY [77]. Considering that there are 9,189,000 head of pigs in Japan [78], an outbreak of ASF would cause serious losses to the swine industry. Furthermore, it may also affect wild boar populations, which are considered to have played an important role in the spread of this disease in Europe.
The Biodiversity Center of Japan estimated that there are approximately 220,000 to 1,200,000 wild boars living across a broad area of Japan [80], indicating that it would require tremendous amounts of money and time to eradicate the disease from Japan should it spread to these populations.

The aim of the present study was to quantitatively evaluate the risk of ASFV introduction to Japan via pork products being brought in international air passengers’ luggage (PPAP). The results from the evaluation and spatial characterization of the risk of ASFV introduction could be used to improve the effectiveness of preventive measures already in place. The Japanese quarantine system relies heavily on border controls at ports of entry. Therefore, this study will help to focus available resources on specific airports and flights according to the estimated risk of ASFV entry.
Materials and Methods

A quantitative stochastic risk assessment model was developed to estimate the annual probability of ASFV introduction to each international airport in Japan via PPAP based on information obtained from relevant websites, public databases, published literature, and expert opinion. The risk model was developed using @RISK 7.6 (Palisade Corporation, Newfield, NY, USA) in Microsoft Excel 2013® and 10,000 iterations were run using the Latin hypercube sampling method. In the current analysis, 214 international commercial flights traveling from 47 origin territories to 31 destination airports in Japan were considered as potential routes of ASFV introduction to Japan via PPAP, information on which was obtained for the period August 2015 to October 2017 from the Ministry of Land, Infrastructure, Transport and Tourism (MLIT) database [81]. As PPAP include both legal and illegal pork products, the legal importation of pork products to Japan requires specific certification of inspection provided by the Government of the territory of origin [82], we assumed that legal pork products would not be contaminated with ASFV. Therefore, we focused only on the risk of illegal PPAP (IPPAP) in the current analysis. A dataset on confiscated items was obtained from the JAQS database [69], and information about ASF disease status and number of susceptible pigs in each territory of origin was obtained from the OIE-WAHIS (World Animal Health Information database interface) as at April 5, 2019 [61] and FAOSTAT 2018 [83], respectively.

Model structure

The current risk model was developed following an approach similar to [84], who assessed the risk of ASFV and CSFV introduction to the U.S. via swine products carried in air passengers’ luggage. In this quantitative model, the risk of introduction
followed a binomial process according to the formula:

\[ P(x \geq 1) = \sum 1 - (1 - P_{i,o})^{V_{od}} \]

where \( V_{od} \) is the assumed weight (kg) of PPAP brought into each destination Japanese airport \((d)\) by international commercial flights from each origin territory \((o)\), \(P_{i,o}\) is the estimated probability of at least 1 kg of PPAP being contaminated with ASFV from each origin territory, and \(\sum\) refers to the probability of entry considering all of the different international commercial flights. Thus, the model consists of two main components: (a) the quantity of PPAP brought into Japan in kg and (b) the probability of these PPAP being contaminated with ASFV. All input values, parameterizations, and references are presented in Tables 3 and 4 and are explained in the following sections. The structure of model is described in Figure 4.
Figure 4. The structure of the model to estimate the quantity of pork products brought in air passengers’ luggage (PPAP) in kg which is introduced into Japan. The origin territories are classified as three categories of high, medium and low risk considering the situation of ASF.
Table 3. Description and parameterization of model inputs for estimation of the quantity (kg) of illegal pork products brought in air passenger's luggage (IPPAP) introduce into Japan per origin territory and destination airport

<table>
<thead>
<tr>
<th>Notation</th>
<th>Definition</th>
<th>Parameterization</th>
<th>Values</th>
<th>Reference/ Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>$N_{f-ann,od}$</td>
<td>Annual number of international commercial flights coming to Japan per origin territory ($o$), per destination airport ($d$)</td>
<td>Normal ($\mu, \alpha$)</td>
<td></td>
<td>[85]</td>
</tr>
<tr>
<td>$C_{p,a}$</td>
<td>Capacity of air passengers per aircraft type</td>
<td>(min, max)</td>
<td></td>
<td>[86]</td>
</tr>
<tr>
<td>$A_{o}$</td>
<td>Type of aircraft used per $o$</td>
<td></td>
<td>[87]</td>
<td></td>
</tr>
<tr>
<td>$C_{f,o}$</td>
<td>Flight capacity coming to Japan per $o$</td>
<td>Normal ($\mu, \alpha$)</td>
<td></td>
<td>Assumption based on fourteen airline carriers [88-102]</td>
</tr>
<tr>
<td>LF</td>
<td>Average international passenger load factor</td>
<td></td>
<td>[103]</td>
<td></td>
</tr>
<tr>
<td>$N_{p,of}$</td>
<td>Number of passengers coming to Japan per $o$, per flight ($f$)</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>$N_{p-ann,od}$</td>
<td>Annual number of passengers coming to Japan per $o$, per $d$</td>
<td>$N_{f-ann,od}N_{p,of}$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$V_{lal,po}$</td>
<td>Weight (kg) of luggage allowance per air passenger, per $o$</td>
<td>Pert (min, most likely, max)</td>
<td></td>
<td>Checked luggage allowance of 14 most important airline carriers listed above</td>
</tr>
<tr>
<td>$V_{l,ann,od}$</td>
<td>Annual weight (kg) of luggage coming to Japan per $o$, per $d$</td>
<td>$N_{p-ann,od}V_{lal,po}$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$V_{l,ann,o}$</td>
<td>Annual weight (kg) of luggage coming from $o$ to Japan</td>
<td>$N_{p-ann,o}V_{lal,po}$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Pro$_{vl,od}$</td>
<td>Proportion of weight (kg) of luggage from each $o$ to each $d$</td>
<td>$V_{l,ann,od}/V_{l,ann,o}$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$V_{c}$</td>
<td>Individual weight of each confiscated animal product</td>
<td>Mean</td>
<td>[104]</td>
<td></td>
</tr>
<tr>
<td>Symbol</td>
<td>Description</td>
<td>Formula</td>
<td></td>
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<td>-----------------------------------------------------------------------------</td>
<td>------------------------------------------</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$N_{c\text{-ann,o}}$</td>
<td>Annual number of confiscated pork products brought in international passengers' luggage per $o$</td>
<td>Mean $V_{c\text{-ann,o}}$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$V_{c\text{-ann,o}}$</td>
<td>Annual weight (kg) of confiscated PPAP per $o$</td>
<td>$V_{c\text{-ann,o}}N_{c\text{-ann,o}}$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$V_{c\text{-ann,od}}$</td>
<td>Annual weight (kg) of IPPAP per $o$, per $d$</td>
<td>$V_{c\text{-ann,od}}\text{Pr}_\text{od}$</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$H_{\text{non}}$</td>
<td>Probability of non-detection of ASFV contaminated IPPAP at airport quarantine</td>
<td>Triang (min, most likely, max)</td>
<td></td>
<td></td>
</tr>
<tr>
<td>$V_{\text{od}}$</td>
<td>Annual weight (kg) of IPPAP introduction into Japan without detection at quarantine per $o$, per $d$</td>
<td>$V_{c\text{-ann,od}}H_{\text{non}}(1-H_{\text{non}})$</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Table 4. Description and parameterization of model inputs for estimation of the probability of PPAP being contaminated with ASFV per origin territories ($o$), per month ($m$)

<table>
<thead>
<tr>
<th>Notation</th>
<th>Definition</th>
<th>Parameterization</th>
<th>Values</th>
<th>Reference/Source</th>
</tr>
</thead>
<tbody>
<tr>
<td>$N_{dp,o}$</td>
<td>Domestic pig population in $o$</td>
<td>NA</td>
<td>Data on 2017</td>
<td>[106]</td>
</tr>
<tr>
<td>$Pro_{slt,m}$</td>
<td>Proportion of annual domestic pig population slaughtered for pork product per $m$</td>
<td>Pert (min, most likely, max)</td>
<td>Pert (0.1, 0.18, 0.25)</td>
<td>[106]</td>
</tr>
<tr>
<td>$N_{slt,om}$</td>
<td>Number of domestic pigs slaughtered for pork products in $o$ per $m$</td>
<td>Pert (min, most likely, max)</td>
<td>$N_{dp,o} \times Pro_{slt,m}$</td>
<td></td>
</tr>
<tr>
<td>$Pre_{v,o}$</td>
<td>Prevalence rate of ASF on affected farms in $o$</td>
<td></td>
<td>Mean prevalence rate of ASF on affected farms in $o$</td>
<td>[6]</td>
</tr>
<tr>
<td>$N_{af,o}$</td>
<td>Average number of pigs on affected farm in $o$</td>
<td></td>
<td></td>
<td>[6]</td>
</tr>
<tr>
<td>$N_{asf,om}$</td>
<td>Monthly number of ASF outbreaks in $o$</td>
<td></td>
<td></td>
<td>[6]</td>
</tr>
<tr>
<td>$H_{under}$</td>
<td>Probability of notification underreporting</td>
<td>Pert (min, most likely, max)</td>
<td>Pert (0.2, 0.4, 0.6)</td>
<td>[107]</td>
</tr>
<tr>
<td>$T_{inf}$</td>
<td>Duration of ASF infection in months</td>
<td>Pert (min, most likely, max)</td>
<td>Pert (3, 11, 30)</td>
<td>Modified [107]</td>
</tr>
<tr>
<td>$N_{nasf,m}$</td>
<td>Potential number of nonreported ASFV infected domestic pigs per $m$</td>
<td></td>
<td>$\left( Pre_{v,o} \times N_{af,o} \times N_{asf,om} \times T_{inf} \right)/H_{under}$</td>
<td>[6]</td>
</tr>
<tr>
<td>$Pro_{imp,m}$</td>
<td>Proportion of pigs improperly slaughtered for pork product per $m$</td>
<td>Uniform (min, max)</td>
<td>min = $Pro_{imp,m}$, max = 1</td>
<td>Modified [106]</td>
</tr>
<tr>
<td>$H_{slt-high,m}$</td>
<td>Probability of ASFV infected domestic pigs slaughtered in $m$ in origin territory of &quot;High&quot; risk</td>
<td>Beta ($\alpha_1, \alpha_2$)</td>
<td>$\alpha_1 = N_{nasf} \times Pro_{imp,m} + 1$, $\alpha_2 = N_{slt,om} - (N_{nasf} \times Pro_{imp,m}) + 1$</td>
<td></td>
</tr>
<tr>
<td>$Pro_{asf}$</td>
<td>Proportion of domestic pig population affected by ASF outbreak</td>
<td>Pert (min, most likely, max)</td>
<td>Pert (0, 0.002, 0.02)</td>
<td></td>
</tr>
<tr>
<td>Symbol</td>
<td>Description</td>
<td>Probability Distribution</td>
<td>Parameters</td>
<td>Notes</td>
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<td>------------------------------------------------------------------------------</td>
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</tr>
<tr>
<td>$H_{med}$</td>
<td>Probability of occurrence of an outbreak of ASF in origin territory of &quot;Medium&quot; risk</td>
<td>Pert (min, most likely, max)</td>
<td>Pert (0, 0.0022, 0.022)</td>
<td>Modified [107]</td>
</tr>
<tr>
<td>$H_{n_{med-m}}$</td>
<td>Probability of ASFV infected domestic pigs slaughtered in $m$ in origin territory of &quot;Medium&quot; risk</td>
<td>Pert (min, most likely, max)</td>
<td>$\text{Pro}<em>{\text{asf}} \times H</em>{med} \times T_{inf} \times H_{under}$</td>
<td></td>
</tr>
<tr>
<td>$H_{low}$</td>
<td>Probability of occurrence of an outbreak of ASF in origin territory of &quot;Low&quot; risk</td>
<td>Pert (min, most likely, max)</td>
<td>Pert (0, 0.00022, 0.0022)</td>
<td>Modified [107]</td>
</tr>
<tr>
<td>$H_{n_{low-m}}$</td>
<td>Probability of ASFV infected domestic pigs slaughtered in $m$ in origin territory of &quot;Low&quot; risk</td>
<td>Pert (min, most likely, max)</td>
<td>$\text{Pro}<em>{\text{asf}} \times H</em>{low} \times T_{inf} \times H_{under}$</td>
<td></td>
</tr>
</tbody>
</table>
Annual number of passengers coming to Japan ($N_{\text{p-ann\_od}}$)

The annual number of passengers from each origin territory and at each destination airport ($N_{\text{p-ann\_od}}$) was estimated by multiplying the annual number of international commercial flights coming to Japan per origin territory and destination airport ($N_{\text{f-ann\_od}}$) by the number of passengers coming to Japan per origin territory and flight ($f$) ($N_{\text{p\_of}}$). First, the value of $N_{\text{f-ann\_od}}$ during the period from August 2015 to October 2017 was parameterized using a normal distribution. The value of $N_{\text{p\_of}}$ was then estimated as the product of the flight capacity per origin territory ($C_{\text{f\_o}}$) and the average international passenger load factor (i.e., seat occupancy) in 2017, which was obtained from the MLIT database [108]. The value of $C_{\text{f\_o}}$ was estimated by obtaining information on the aircraft types that traveled between each origin territory and the seven most important Japanese airports (Narita International Airport, Kansai International Airport, Tokyo International Airport, Chubu Centrair International Airport, Fukuoka Airport, Naha Airport, and New Chitose Airport), which represented 94.7% of the total international commercial flights, from the travel booking website Expedia.com [87]. Information on the flight capacity for each aircraft type and airline carrier was then obtained from the SeatGuru website [86]. In particular, data were obtained for the 14 most important airline carriers, which represented 63% of the total international commercial flights to Japan from August 2015 to October 2017 and included Japan Airlines [99], All Nippon Airways [89], Cathay Pacific Airways [91], Eva Air [98], Air China [88], China Airlines [92], China Eastern Airlines [93], China Southern Airlines [94], Korean Air [101], Asiana Airlines [90], Delta Air Lines [95], United Airlines [102], Jeju Air [100], and Peach Aviation [109]. The value of $C_{\text{f\_o}}$ was then parameterized using the normal distribution based on the estimated minimum, mean, and maximum flight capacities per
flight, which were estimated from the aircraft type and flight capacity data.

**Weight of luggage allowance (V<sub>l-ann_od</sub>)**

The annual weight (kg) of luggage arriving in Japan per origin territory and destination airport (V<sub>l-ann_od</sub>) was estimated as the product of the weight (kg) of luggage allowance per air passenger (p) per origin territory (V<sub>lal_po</sub>) and the annual number of passengers to Japan per origin territory and destination airport (N<sub>p-ann_od</sub>). The value of V<sub>lal_po</sub> was estimated at the regional level based on the free checked luggage allowance in economy class for the 14 most important airline carriers listed above. Since flights between the African continent and Japan were operated by either Ethiopian Airlines or EgyptAir, the weight of air passengers’ luggage from the African region was estimated based on the free checked luggage allowance in economy class for both companies [96,97]. This resulted in the estimation of a minimum value, which corresponded to the weight of hand luggage alone, and a maximum value, which corresponded to the total weight of the hand luggage and the maximum free luggage allowance for each origin territory. The median values of these estimated free luggage allowances were then used as the most probable values. The value of V<sub>lal_po</sub> was parameterized using a pert distribution with minimum, most likely, and maximum values. The obtained data were used to estimate the proportion of the weight of air passengers’ luggage that came from each origin territory to each destination airport (Pro<sub>vl_od</sub>).

**Annual weight of IPPAP (V<sub>c-ann_od</sub>)**

Under the current JAQS regulation, pork products that have specific certification of inspection by the Government of the territory of origin are allowed to be imported into Japan. Therefore, it was assumed that only IPPAP were potentially contaminated with ASFV. The annual weight of confiscated PPAP (i.e., IPPAP) coming from each origin
territory to each destination airport \((V_{c-ann_od})\) was estimated from information obtained from the JAQS database [69]. Annual information on the mean weight confiscation was available by referring the JAQS database. Information on weight was aggregated as animal products without pointing out any differences in weight between them. Therefore, it was assumed that the individual weight of PPAP was equivalent to the individual weight of confiscated animal products. Moreover, data on the number of confiscated pork products that were brought in by airline passengers and ship passengers per year (PPAP and PPSP, respectively) was available and public [110]. However, there was not distinction between PPAP and PPSP. So, the annual number of confiscated pork products brought in by international passengers per origin territory \((N_{c-ann_o})\) was considered as the annual number of confiscated PPAP per origin territory in the current analysis. Thus, the annual weight of confiscated PPAP per origin territory \((V_{c-ann_o})\) was estimated as the product of the individual weight of each confiscated animal product \((V_c)\) and \(N_{c-ann_o}\) [69,110]. \(N_{c-ann_od}\) was estimated by multiplying the value of \(Provl_{od}\) by the value of \(V_{c-ann_o}\). 

**Weight of IPPAP being missed at border controls \((V_{od})\)**

In the current analysis, we assumed that IPPAP could only enter airports in Japan when the products were missed at border controls at the destination airport. Unfortunately, we were unable to obtain any data or expert opinion that would allow us to investigate the probability of the non-detection of IPPAP at border controls in Japan \((H_{non})\), so this was modeled with a triangular distribution that had minimum, most likely, and maximum values of 0.2, 0.5, and 0.9, respectively, which were obtained from [84]. The weight of IPPAP that had been missed at border controls per origin territory and destination airport \((V_{od})\) was then estimated from the obtained probability.
Probability of PPAP being contaminated with ASFV

To estimate the probability that PPAP could be contaminated with ASFV, an approach was developed following the methods of [84] and Veterinary Laboratories Agency (VLA) [107]. As mentioned above, although PPAP includes both legal and illegal pork products, only IPPAP was considered to have the potential of being contaminated with ASFV.

A total of 47 origin territories that have direct or indirect flights to Japan were classified into three risk categories based on information on their ASF disease status contained in the OIE-WAHIS as at April 5, 2019 [6]: “High” where the disease is present or suspected in pig species; “Medium” where a territory shares a national border with at least one territory in the “High” category; and “Low” for all other territories that officially have no ASF outbreaks and do not share borders with territories in the “High” category. For each of these categories, it was assumed that one kg of PPAP that was contaminated with ASFV indicated that at least one domestic pig was infected in the origin territory.

For each territory in the “High” category, the probability of ASFV infection was modeled using a beta distribution with consideration of the potential number of non-reported infected domestic pigs in each origin territory per month ($N_{nasf,m}$) and the number of slaughtered domestic pigs in each origin territory per month ($N_{slt,om}$). $N_{nasf,m}$ was estimated from the prevalence rate of ASF on affected farms in the origin territory ($Prev_{o}$), the size of ASF-affected farms in the origin territory ($N_{af,o}$), the average number of outbreaks per month in the origin territory ($N_{asf,om}$), the duration of the infection ($T_{inf}$), and the probability of notification underreporting ($H_{under}$). The number of non-reported infected domestic pigs slaughtered per month in each origin territory was estimated as the product of $N_{nasf,m}$ and the proportion of pigs that were improperly slaughtered per month.
(Pro_{imp,m}). Since the probability of slaughtering pigs may be greatly influenced by the farmer’s attitude, economic condition, and other social factors, a conservative approach was applied whereby Pro_{imp,m} was modeled using a uniform distribution with the mean value of the Pro_{slt,m} distribution as a minimum and 100% as the maximum (indicating the worst-case scenario where all pigs would be slaughtered if farmers suspected the disease). The number of pigs slaughtered per month (N_{slt,om}) was estimated by considering the annual pig census in each origin territory (N_{dp,o}) and the proportion of pigs slaughtered per month (Pro_{slt,m}).

The probability of ASFV infection in the “Medium” and “Low” risk categories (H_{slt-med,m} and H_{slt-low,m}, respectively) was estimated by multiplying the probability of an outbreak occurring in the origin territory (H_{med} and H_{low}, respectively) by the average size of an ASF outbreak (Pro_{asf}), the duration of the infection (T_{inf}), the probability of outbreaks being undetected, and the proportion of pigs going to the slaughterhouse that month. The probability of outbreak occurrence was set 10 times higher for territories in the “Medium” category than for those in the “Low” category based on expert opinion considering the geographical closeness to territories in the “High” category.

Specific standards were applied to some origin territories due to their unique situations. In the case of Ethiopia, no data on ASF outbreaks were available even though the disease is suspected there [111], so the median value of the probability of infection for the other affected African territories was used. In the case of Italy, the risk was set at “Medium” even though ASF outbreaks have been reported there for a long time because the disease is confined to the island of Sardinia and has only escaped from there once in the last 40 years. Taiwan, Hong Kong, Macau, and mainland China (China) were recognized as origin territories based on the territory classification that is used in the OIE-
WAHIS [111]. Finally, in the case of official international commercial indirect flights [81], the probability of infection was estimated as the combination of the probability of ASFV infection in the origin territory and the connecting territory.

**Sensitivity analysis**

In order to identify the most influential parameters to model outputs, Spearman correlation coefficients ($\rho_i$) between each input and the annual probability of ASFV entry into Japan were calculated. Inputs with $\rho_i \geq 0.4$ that contributed $> 9\%$ to the variance of the output were identified as the most influential parameters for the model and were analyzed in detail using the advanced sensitivity analysis tool @RISK 7.6, with 10,000 iterations for each scenario. A total of 10 scenarios were assessed for each selected parameter by changing the base values in 10 consecutive steps from a minimum of 50% reduction to a maximum of 50% increase.
Results

The overall mean annual probability of ASFV introduction to Japan via PPAP was found to be 0.941 with a 95% confidence interval (CI) of 0.661–1.000, which approximately corresponds to one introduction every 1.06 years.

At the flight level, the highest risk of ASFV entry was the direct flight between Mongolia and Narita International Airport (0.864; 95% CI, 0.434–1.000), followed by Vietnam and Narita (0.485; 0.104–0.955), China and Kansai (0.413; 0.08–0.903), and China and Narita (0.313; 0.052–0.779) (Table 5).

In particular, the risk was clustered around flights from several territories in the “High” category (Mongolia, China, and Vietnam) that were traveling to the four largest airports (Narita, Kansai, Chubu, and Tokyo). Among the indirect flights, the flight from France to Narita via the Russian Federation was found to have the highest risk (0.014; 95% CI, 0.001–0.043), but none of the indirect flights were in the top 10 highest risk flights.

At the origin territory level, the top 10 highest risk territories included six Asian territories, three European territories, and one African territory. Mongolia was the highest risk territory, with a risk of 0.864 (95% CI, 0.434–1.000), followed by China (0.697; 0.223–0.999), Vietnam (0.662; 0.196–0.998), the Russian Federation (0.136; 0.018–0.401), Cambodia (0.073; 0.009–0.226), and Ethiopia (0.002; 0.0002–0.0050). Poland was also among the top 10 highest risk territories, where ASF is currently present. However, non-affected territories were also in this list, including Taiwan, Germany, and Thailand. Results are depicted in Figure 5 and presented in Table 6.

At the destination airport level, a high probability of ASFV introduction to Japan
via PPAP was clustered within the four largest airports in Japan, with Narita having the highest risk (0.905; 95% CI, 0.537–1.000), followed by Kansai (0.496, 0.109–0.961), Tokyo (0.389; 0.072–0.879), and Chubu (0.338; 0.058–0.816). Results are depicted in Figure 6 and presented in Table 7.
Table 5. The risk result of ASFV introduction to Japan via pork product brought in air passenger's luggage per international flight

<table>
<thead>
<tr>
<th>Flight: Origin / Airport</th>
<th>Risk value of ASFV introduction into Japan [95% confidence interval]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mongolia / Narita</td>
<td>0.864 [0.434 - 1.000]</td>
</tr>
<tr>
<td>Vietnam / Narita</td>
<td>0.485 [0.104 - 0.955]</td>
</tr>
<tr>
<td>China / Kansai</td>
<td>0.413 [0.079 - 0.903]</td>
</tr>
<tr>
<td>China / Narita</td>
<td>0.313 [0.052 - 0.779]</td>
</tr>
<tr>
<td>Vietnam / Tokyo</td>
<td>0.272 [0.043 - 0.708]</td>
</tr>
<tr>
<td>China / Chubu</td>
<td>0.236 [0.035 - 0.637]</td>
</tr>
<tr>
<td>Vietnam / Kansai</td>
<td>0.217 [0.032 - 0.596]</td>
</tr>
<tr>
<td>China / Tokyo</td>
<td>0.211 [0.030 - 0.583]</td>
</tr>
<tr>
<td>Vietnam / Chubu</td>
<td>0.164 [0.022 - 0.471]</td>
</tr>
<tr>
<td>the Russian Federation / Narita</td>
<td>0.120 [0.015 - 0.356]</td>
</tr>
<tr>
<td>China / Fukuoka</td>
<td>0.120 [0.015 - 0.355]</td>
</tr>
<tr>
<td>Vietnam / Fukuoka</td>
<td>0.087 [0.011 - 0.262]</td>
</tr>
<tr>
<td>Cambodia / Narita</td>
<td>0.073 [0.009 - 0.226]</td>
</tr>
<tr>
<td>China / Naha</td>
<td>0.063 [0.007 - 0.192]</td>
</tr>
<tr>
<td>China / Shizuoka</td>
<td>0.048 [0.006 - 0.147]</td>
</tr>
<tr>
<td>China / New Chitose</td>
<td>0.039 [0.005 - 0.121]</td>
</tr>
</tbody>
</table>
Figure 5. The risk results of ASFV introduction into Japan via pork products brought in air passengers’ luggage at origin territory level. The graduated color map represents the annual risk from the highest (darker) to the lowest (lighter).
**Table 6.** The risk result of ASFV introduction from origin territories into Japan via pork product brought in air passenger's luggage

<table>
<thead>
<tr>
<th>Territory of origin</th>
<th>Risk value of ASFV introduction into Japan [95% confidence interval]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Mongolia</td>
<td>0.864 [0.434 - 1.000]</td>
</tr>
<tr>
<td>China</td>
<td>0.697 [0.223 - 0.999]</td>
</tr>
<tr>
<td>Vietnam</td>
<td>0.662 [0.196 - 0.998]</td>
</tr>
<tr>
<td>the Russian Federation</td>
<td>0.136 [0.018 - 0.401]</td>
</tr>
<tr>
<td>Cambodia</td>
<td>0.073 [0.009 - 0.226]</td>
</tr>
<tr>
<td>Ethiopia</td>
<td>0.002 [2 x 10^{-4} - 0.005]</td>
</tr>
<tr>
<td>Taiwan</td>
<td>0.001 [2 x 10^{-5} - 0.006]</td>
</tr>
<tr>
<td>Poland</td>
<td>0.001 [1 x 10^{-4} - 0.004]</td>
</tr>
<tr>
<td>Germany</td>
<td>0.001 [1 x 10^{-5} - 0.003]</td>
</tr>
<tr>
<td>Thailand</td>
<td>0.001 [1 x 10^{-5} - 0.003]</td>
</tr>
</tbody>
</table>
Figure 6. The risk results of ASFV introduction into Japan via pork products brought in air passengers’ luggage at airport level. Airports were represented by colored circles from the highest (larger) to the lowest (smaller) at airport level.
Table 7. The risk result of ASFV introduction into Japan via pork product brought in air passenger's luggage per airport

<table>
<thead>
<tr>
<th>Airport</th>
<th>Risk value of ASFV introduction into Japan [95% confidence interval]</th>
</tr>
</thead>
<tbody>
<tr>
<td>Narita</td>
<td>0.905 [0.537 - 1.000]</td>
</tr>
<tr>
<td>Kansai</td>
<td>0.496 [0.109 - 0.961]</td>
</tr>
<tr>
<td>Tokyo</td>
<td>0.389 [0.072 - 0.879]</td>
</tr>
<tr>
<td>Chubu</td>
<td>0.338 [0.058 - 0.816]</td>
</tr>
<tr>
<td>Fukuoka</td>
<td>0.185 [0.026 - 0.524]</td>
</tr>
<tr>
<td>Naha</td>
<td>0.063 [0.008 - 0.192]</td>
</tr>
<tr>
<td>New Chitose</td>
<td>0.057 [0.007 - 0.175]</td>
</tr>
<tr>
<td>Shizuoka</td>
<td>0.048 [0.006 - 0.147]</td>
</tr>
</tbody>
</table>
Sensitivity analysis

Examination of the Spearman correlation coefficients indicated that the probability of PPAP not being detected at border controls, the probability of ASF infection in Mongolia, and the volume of PPAP coming into Japan were highly correlated input parameters in the model. The advanced sensitivity analysis revealed that the probability of PPAP not being detected at border controls was the most influential parameter in the model, followed by the volume of PPAP coming into Japan and the probability of ASF infection in Mongolia (Figure 7).

Figure 7. Results of the advanced sensitivity analysis for the risk of ASFV introduction via into Japan via pork products brought in air passengers’ luggage. Horizontal axis shows the percentage change of the selected three input parameters against the annual risk in the vertical axis.
Discussion

Several studies have assessed the import risk of diseases such as rabies and bovine spongiform encephalopathy being introduced to Japan [112,113]. However, this is the first study to quantify the risk of ASFV introduction to Japan via PPAP. This information is urgently needed as although Japan has never been affected by ASF, outbreaks have now been reported in the neighboring territories of China, Mongolia, Vietnam, Cambodia, and Hong Kong [6]. A similar study has also been carried out in Europe and the U.S. following the introduction of this disease into the Caucasus region in 2007 [67,84,114-116].

The number of detected cases of ASFV-contaminated pork products in Asian airports has increased in proportion to the number of ASF outbreaks in this region, and these products have mainly been brought in by air passengers traveling from epidemic areas. Therefore, considering the risk of ASFV entry to Japan from neighboring epidemic areas, routine quarantine measures will definitely play a critical role in preventing the introduction of this disease. Passengers who are boarding international flights are usually informed in advance about the ban on bringing animal products to Japan without official certification, and once they arrive at their destination port, passengers are obliged to fill out a questionnaire that includes their flight number and the contents of their luggage. Voluntary luggage inspections are conducted by quarantine officers based on this questionnaire as well as their impression of the passengers. Detection dogs are also engaged in finding potential contrabands, including drugs and prohibited animal products in passengers’ luggage at major international airports [117]. However, it is almost impossible to find all animal products that have been illegally brought into airports, so
limitations on current quarantine capacities as well as the failure to detect illegally transported animal products will have a large effect on the risk of animal disease introduction. Consequently, the identification of flights, origin territories, and destination airports that are at high risk of ASFV introduction will help to improve the current surveillance system and allow a monitoring system to be developed for conducting risk-based surveillance, such as preparing a specific quarantine system for particular territories of origin.

The availability and quality of data are always major limitations for any risk assessment. Ideally, information on the number of air passengers arriving in Japan at each airport from each origin territory would be used to accurately assess the volume of air passengers’ luggage. However, the data used in this study were not categorized in great detail. Consequently, the number of passengers per origin territory and aircraft type were estimated using an approximation algorithm based on data that were available for 14 airline carriers. Since information on the size of the passengers’ hand luggage and the proportion of pork products in the passengers’ hand luggage were not available for this risk assessment, we assumed the total volume of IPPAP based on the number of confiscated pork products per origin territory [110]. This assumption also considered the range of the free luggage allowance for passengers at a regional level because the acceptable size of free luggage varies between different regions within origin territories. The data on confiscated pork products, which are accounted for both data of PPAP and one of PPSP and are not separated. However, since 97.5% of foreign visitors travel to Japan by airplane [118], it seemed safe to assume that most of these pork products were brought in by air passengers.

The Ministry of Agriculture, Forestry and Fisheries reported 36 incidences of
ASFV detection from pork products in Japanese airports during the period from August 2018 to May 2019 [74]. Based on these data, our finding that Mongolia, China, and Vietnam are the highest risk territories for ASFV introduction to Japan appears to be appropriate. However, our model may have underestimated the risk of the annual probability of ASFV introduction at the level of the origin territory, as 32 cases of ASFV detection from PPAP were reported from China in 9 months, whereas our model indicated that the risk for China was 0.697 (95% CI, 0.223–0.999), which approximately corresponds to one ASFV introduction into Japanese airports every 1.43 years. While our analysis considered the underreporting probability of ASF in the origin territory, we were unable to fully account for inadequate behavior by farmers due to a lack of reliable data sources. Some pig farmers tend to put pigs that are infected with ASFV on sale as they are afraid of suffering losses, which may greatly increase the risk of pork products being contaminated with ASFV in epidemic areas, as evidenced by the case of Hong Kong [65]. Therefore, additional data are required to fill these gaps and conduct a more accurate risk assessment.

In the current analysis, Mongolia was identified as the highest risk origin territory for ASFV introduction to Japan. This is likely due to the large proportion of domestic pigs in Mongolia that have been affected by ASF since the first outbreak was notified [119]. However, there may be a huge gap in the pig census data for Mongolia, as the Food and Agriculture Organization of the United Nations database reported 31,870 head of pigs in Mongolia in 2017 [83], whereas Enkhbold et al. [21] assumed approximately 150,000 head. Therefore, the final risk posed by Mongolia could have been overestimated. However, as Mongolia is geographically close to Japan and has the highest risk, more accurate data are required to accurately assess this risk.
China had the second highest risk despite having the greatest number of ASFV-contaminated PPAP detections. This difference between the results of our assessment and the reported ASFV detections in Japanese airports may have been caused by the underestimation of several factors, such as the probability of PPAP not being detected at border controls, the number of confiscated pork products, and the number of ASF outbreaks and affected animals. Therefore, given that detected ASFV-contaminated pork products were brought in by air passengers from various parts of China, the risk for this territory should be reassessed at a regional level.

Vietnam had the third highest risk of introducing ASFV to Japan. This result may have been greatly influenced by the fact that Vietnam has the largest number of ASF outbreaks per month and the highest prevalence rate of ASF in affected farms among the ASF-affected territories, as well as the large number of PPAP that originated from this territory.

The Russian Federation was the fourth highest risk origin territory, despite the disease having been present there for a long period of time and having affected a large number of domestic pigs. This might be because the average number of ASF outbreaks per month and the prevalence of ASF in affected farms were lower than in China, Vietnam, and Mongolia. Following this was Cambodia, as the fifth highest risk territory, which reflected the fact that although only one ASF outbreak has been reported in Cambodia to the OIE as at April 5, 2019, it is geographically close to Vietnam, so the potential risk of a further outbreak should not be overlooked. Ethiopia was the sixth highest risk territory of origin. While no official notification of ASF has been reported to the OIE from Ethiopia, ASFV was detected in tissue samples collected from domestic pigs in this territory from 2011 to 2014 [120], so it was classified as an ASF-affected territory and its risk was
assessed based on the disease situation of other ASF-affected African territories. Therefore, further information is required to more accurately assess the risk from Ethiopia.

Remarkably, Taiwan was found to pose a higher risk than Poland, which is an ASF-affected territory, and similarly Germany was considered a higher risk than ASF-affected Belgium. Categorization of these ASF-free territories as higher risk territories of ASFV introduction to Japan would be owing to three potential parameters: the high number of PPAP originating from these territories, the number of visitors from these territories to Japan, and the geographical distance between these territories and ASF-affected territories. On the other hand, the risk from Belgium may have been lower due to no ASF outbreak having occurred in the domestic pig population there (it has only been detected among wild boars and our model did not include products originating from wildlife) and the low number of visitors to Japan.

In terms of the destination airport, the risk was clustered around the four largest airports in Japan: Narita, Kansai, Tokyo, and Chubu Centrair. This result may have been highly influenced by the volume of air passengers’ hand luggage arriving from high risk territories—indeed, Narita International Airport has the largest volume of air passengers’ hand luggage, followed by Kansai, Tokyo, and Chubu Centrair, which corresponds with the order of risk for these airports.

While limitations to the quality and accessibility of data always pose challenges in any risk assessment, our study revealed the potential risk of ASFV introduction at the flight, origin territory, and destination airport levels, which will be useful for improving risk management activities and monitoring systems to prevent the introduction of ASFV into Japan. However, it should also be noted that half of the pork that is consumed in Japan is imported [121], and these products are mainly transported by ship freight or air
cargo for commercial purposes. In addition, live pigs are imported to Japan from overseas, mainly for breeding purposes [122]. Therefore, these also represent potential ASFV introduction pathways into Japan. These pathways were not evaluated in the present study due to a lack of reliable information, so further assessment is required to estimate these risks more accurately.
Summary

The spread of ASF has reached pandemic levels over the last decade, and outbreaks of this disease in China, Mongolia, Vietnam, and Cambodia in 2018 and 2019 could accelerate its transmission to neighboring Asian territories. Thus, the risk that the ASFV will be introduced to disease-free territories increases each year. Since Japan is an island nation, the most likely way in which ASFV would be introduced is via PPAP. Therefore, in the present study, we assessed the risk of ASFV introduction to Japan via PPAP. For the purposes of this analysis, we considered 214 international commercial flights traveling from 47 origin territories to 31 destination airports as potential routes of ASFV introduction via PPAP. The risk was estimated quantitatively through a stochastic model that considered the volume of air passengers’ luggage, the amount of confiscated pork products that were carried in air passengers’ luggage, and the disease status of the origin territory. The overall mean annual probability of ASFV introduction to Japan via PPAP was found to be 0.941 [95% CI, 0.661–1.000], which approximately corresponds to one introduction every 1.06 years. At the origin territory level, Mongolia was led as the highest risk territory, with a risk of 0.864 (95% CI, 0.434–1.000), followed by China (0.697; 0.223–0.999), Vietnam (0.662; 0.196–0.998), and the Russian Federation (0.136; 0.018–0.401). At the destination airport level, Narita International Airport had the highest risk (0.905; 0.537–1.000), followed by Kansai International Airport (0.496; 0.109–0.961), Tokyo International Airport (0.389; 0.072–0.879), and Chubu Centrair International Airport (0.338; 0.058–0.816). This information will help improve risk management activities and monitoring systems to prevent the introduction of ASFV to Japan.
Conclusion

CSF is caused by infection with the CSFV, which belongs to the genus *Pestivirus*, family *Flaviviridae*. CSF is a highly contagious febrile disease with potential for high mortality that causes enormous economic loss in the pig industry worldwide [9]. After re-emergence of the CSF in Japan 2018, the diseases have transmitted continuously in both domestic pig and wild boar population, nevertheless of the vigorous effort to address the disease control.

The results of the spatio-temporal analysis showed that the disease likely spread to the northeast part of the prefecture. A maximum significant spatial association estimated between CSF notifications was 23 km by the multi-distance spatial cluster analysis. A space-time permutation analysis identified two significant clusters with an approximate radius of 12 and 20 km and 124 and 98 days of duration, respectively. When the area of the identified clusters was overlaid on a map of habitat quality, approximately 82% and 75% of CSF notifications, respectively, were found in areas with potential contact between domestic pigs and wild boar. These findings will help implement control measures in the affected area.

It has already well known that there are effective vaccines against CSFV but no approved vaccines against ASFV is available. The histories of transboundary animal disease outbreaks in Japan and the current ASF epidemic situation in Asia have warned Japan against the considerable threat of ASF spread from affected territories. Given that wild boar play an important role in both CSF and ASF, it is plausible to assume that similar or even worth epidemic situation could be caused once the ASF outbreak occurred in Japan.
In the present study, we assessed the risk of ASFV introduction to Japan via PPAP, which is one of the most likely risk pathways for the introduction of ASFV into Japan. For the purposes of this analysis, we considered 214 international commercial flights traveling from 47 origin territories to 31 destination airports as potential routes of ASFV introduction via PPAP. The risk was estimated quantitatively through a stochastic model that considered the volume of air passengers’ luggage, the amount of confiscated pork products that were carried in air passengers’ luggage, and the disease status of the origin territory. The overall mean annual probability of ASFV introduction to Japan via PPAP was found to be 0.941 [95% CI, 0.661–1.000], which approximately corresponds to one introduction every 1.06 years. At the origin territory level, Mongolia was led as the highest risk territory, with a risk of 0.864 (95% CI, 0.434–1.000), followed by China (0.697; 0.223–0.999), Vietnam (0.662; 0.196–0.998), and the Russian Federation (0.136; 0.018–0.401). At the destination airport level, Narita International Airport had the highest risk (0.905; 0.537–1.000), followed by Kansai International Airport (0.496; 0.109–0.961), Tokyo International Airport (0.389; 0.072–0.879), and Chubu Centrair International Airport (0.338; 0.058–0.816). This information will help improve risk management activities and monitoring systems to prevent the introduction of ASFV to Japan.

The obtained results could be useful for launching or improving effective control measures to prevent the introduction of ASFV, and for reviewing the current risk reduction activities in the field.
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CSF（豚コレラ）はフラビウイルス科ペスチウイルス属 CSF ウイルスによって引き起こされる家畜伝染病で、感受性動物である豚及びイノシシに高い致死率と強い伝染力をもたらす。2018年9月に日本国内で26年ぶりとなるCSFの発生が確認されたが、迅速な終息には至らず、発生から1年3ヶ月経過しても家畜豚・野生イノシシへの感染が報告されている。本研究では、日本国内におけるCSF発生拡大に関する疫学的知見を得るために発生報告情報を用いて時空間疫学解析を実施した。分布指向性分析の結果、本疾病の北東方向への拡大が示唆された。また、RipleyのK関数法を用いた解析では、各発生間での最大関連距離は23kmと算出された。推定イノシシ最適生息域と時空間置換モデルクラスター解析の結果を重ね合わせたところ、クラスターは野生イノシシと家畜豚の接触が想定される地域に形成された。本解析で得られた情報は今後のCSF感染拡大防止策を検討するうえで有益であると考えられる。

CSFと同様に越境性動物感染症に分類され、疫学的にもCSFと類似しているASF（アフリカ豚コレラ）は、アスファウイルス科アスフィウイルス属ASFウイルス（ASFV）によって引き起こされる豚及びイノシシに感染する家畜伝染病である。2018年より流行地域であるアフリカ・欧州諸国からアジア地域へと感染が急速に拡大していることから、日本への侵入が危惧されている疾病であり、ASFに対して有効なワクチンが開発されていないことから、一度我が国への侵入を許させば畜産産業に甚大な被害が及ぶことが想定される。そこで本研究では、ASFVの侵入経路の1つとして監視すべき重要な経路である国際航空旅客手荷物由来の豚肉製品（PPAP）を介したASFVの日本への侵入リスクと、国内侵入後の感受性動物への曝露リスクを発症国別かつ都道府県別に評価した。侵入リスク評価について、本研究では47の国と地域から日本国内31の国際空港に就航している214航空路線を対象とした確率論的モデルを用い、旅客の手荷物重量、各出発国由来の豚肉製品押収品量および出
発国の ASF 流行状況を考慮して定量的にリスクを評価した。定量的リスク評価の結果、
PPAP を介したASFVの日本への侵入リスクは0.941（95%信頼区間 0.661, 1.000）と導かれ、
この結果はASFVがPPAPを介して1.06年に1回日本へ侵入する可能性があることを示
している。また感受性分析の結果、検疫での違法持ち込み豚肉製品の見逃し率、PPAPの重
量が結果に大きな影響を及ぼしていることがわかった。本研究で得られた結果は、リスク分
析に基づく高リスク地域におけるASFV侵入対策の実施に有用であると考えられる。