Catch me if you can: control disease in 40 days

G.A. Puspitarani^{1,2}, H. Schuster³, E. Coleman⁴, R. Fuchs⁵, A. Desvars-Larrive^{1,2}

¹University of Veterinary Medicine Vienna, Vienna, Austria; ²Complexity Science Hub, Vienna, Austria;

³Vienna University of Economic and Business, Vienna, Austria; ⁴Bristol Medical School, United Kingdom; ⁵Department for Data, Statistic and Risk Assessment, Austrian Agency for Health and Food Safety (AGES), Graz, Austria; ⁶Institute of System Science, University of Graz, Graz, Austria.

Background

The Austrian swine industry plays a crucial role in the national economy, with an average pork consumption of 43.6 kg per capita annually. With a self-sufficiency level of 103%, the industry may face critical losses in the event of an exotic disease introduction. Using pig movement data from 2021, this study investigates the potential spread of an African swine fever-like disease within Austria following its introduction.

Methods

Data

Daily pig movements in Austria (2021), consisting of 23,722 holdings with 250,136 movements.

source	target	date	n	types
Holding of origin	Receiving holding	Daily record	Number of pigs	Purpose of movement: domestic, slaughter, import/export, import/export for slaughter

2.2. Between holdings:

We considered two transmission paths:

- **Direct transmission** via time-respective trade network edges
- Localized spread within 5-km radius using spatial kernel

We defined holding status as infected if at least one pig in *E* or *I* compartments was

Objectives

- **Explore pig mobility flows and** identify key hubs.
- Model the spread of an introduced **ASF-like disease** within the pig movement network to simulate outbreak scenarios and assess potential transmission pathways.
- Inform targeted mitigation strategies.

Results

We identified Straß in Steiermark as the highest-import municipality (m^*) , a key node for disease introduction. Trade network analysis revealed that 89% of Austrian municipalities could be reached within six trade steps from m^* . The mean trade distance was 46.8 km (SD:57.3 km), with long-distance trades defined as those exceeding 161.4 km.

Tab 1. Overview of pig movements records of between pairs of holdings in Austria.

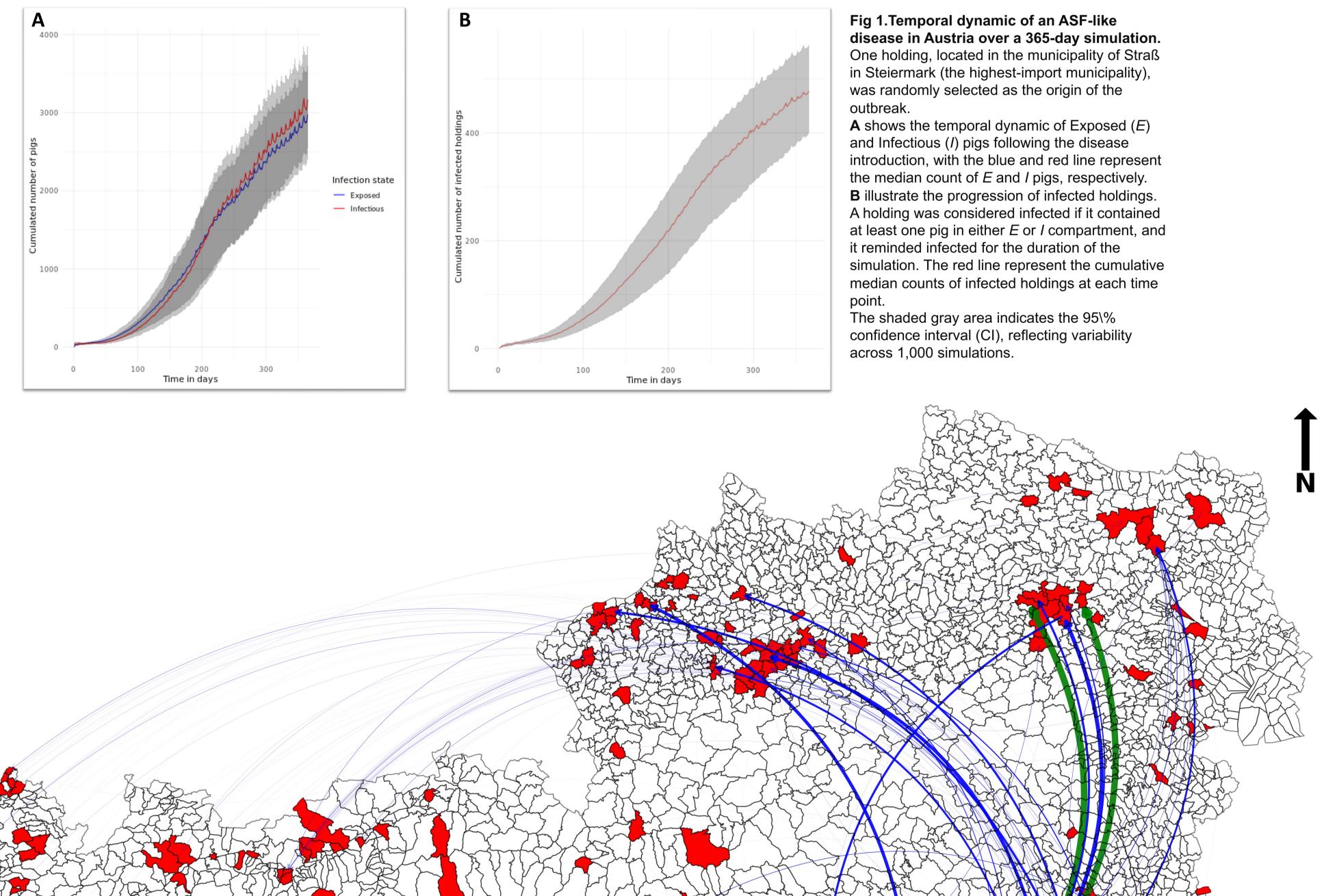
1. Identify municipality with high import

We aggregated nodes by municipality and analyzed trade flows, identifying the highest-import municipality (*m**). Trade distribution from *m*^{*} was examined, with Euclidean distance. We defined for longdistance trade as mean +2SD.

2. Simulate the spread of ASF-like seeded from *m**

2.1: Within holding transmission:

We seeded infections in a randomly selected holdings within m*. Using ASF-like parameters, we developed SEIR model.



present:

We introduced the disease at two different times: January (low trade activity) and April (heightened trade activity).

3. Evaluate the impact

1. Estimate epidemic size.

- Count the number infected pigs, infected holdings, and municipalities.
- **1.** Analyzed long-distance transmission. Identified infection jumps, including timing and contributing municipalities.

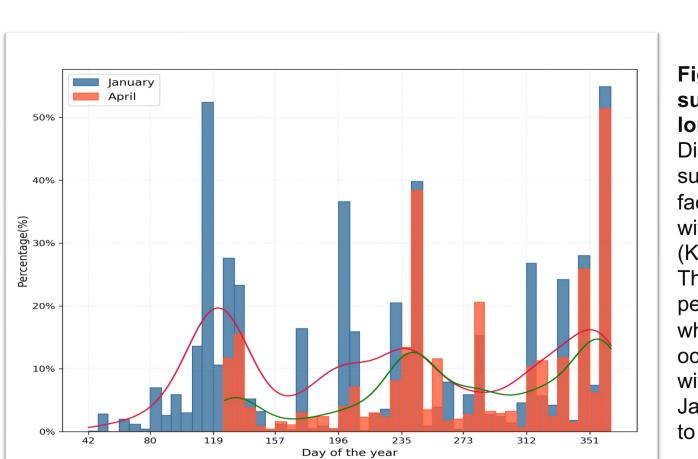
The SEIR model estimated a peak of 3,183 infectious pigs and 477 infected holdings (Fig 1), reaching 10.2% of municipalities. On average, 5.6 longdistance infection jumps occurred per simulation, with high-risk municipalities clustered in Styria and Carinthia (Fig 2). The first infection jump occurred within 40 days (days 42-48, P: 0.7%). When seeded during high-trade period reduced to 20 days (P: 11.7%) (Fig 3). Infection jump entered their first peak phase within 100 days post-introduction.

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University of Veterinary Medicine, Vienna

Conclusion

High-import regions act as potential entry points for exotic disease introduction. In Austria, control measures within the first 40 days after introduction are crucial for limiting large scale outbreak. However, when introduced during a period of increased trade activity this window shortens. If the outbreak circulates more than 100 days, effective containment requires intervention both locally, in the region of introduction, and distant regions.



50.0 km

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Fig 3. Temporal distribution of successful infection jump via long-distance trade. Distribution of the percentage of successful infection jumps facilitated by long-distance trade with an overlaid red KDE (Kernel Density Estimate) curve. The histogram represents the percentage of simulations in which an infection jump occurred within a given time bin, with blue bars corresponding to January seeding and orange bar to April seeding.

Straß in Steiermark (m^*)

Fig 2. Likelihood of long-distance infection jumps. Municipalities colored in red participated in at least one infection jump via long-distance trade (as source or recipient). Directed blue arcs indicate longdistance infection jumps, with arrows pointing from the source municipality to the recipient. The thickness of each arc is proportional to the likelihood of the corresponding infection pathway, with thicker arcs representing higher probabilities of transmission. Green arcs shows the highest-likelihood pathways from Großklein to Karlstetten(P: 0.73) and Großklein to Würmla(P: 0.72).

Contact:

Gavrila A. Puspitarani: gavrila.puspitarani@vetmeduni.ac.at Amélie Desvars-Larrive: amelie.desvars@vetmeduni.ac.at