



Deliverable: Translation of the source attribution methodological approach to the One-Health context for surveillance of Salmonella, Campylobacter, VTEC, and antimicrobial resistance: options, needs

D-JRPFBZ-1-WP5.3

Workpackage 5

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# Preamble - Rationale

The source attribution (SA) activities carried out by the DiSCoVeR project aimed at evaluating through different methodological approaches how much different animal reservoirs, food production chains and transmission routes contribute to the burden of disease in humans caused by *Salmonella*, *Campylobacter*, Shigatoxin-producing *E.coli* (STEC) and antimicrobial resistance (AMR) with particular focus on Extended-Spectrum Beta-lactamase *E. coli*.

SA has been defined as the partitioning of a human disease burden to specific animal sources (Pires et al. 2009). However, the application of SA studies to animals and to livestock may be sub-optimal for zoonotic foodborne pathogens with complex epidemiology that can also spread to humans by non-foodborne transmission routes such as direct animal contact or contaminated environmental sources (drinking water, wastewater, soil, top-soil improvers, etc.) or non-food producing animal reservoirs (wildlife, pets including horses, etc.) or by inter-human transmission. This is the case of the hazards targeted by the DiSCoVeR, which are considered top-priority zoonotic foodborne hazards in the EU, although it is well known that they are non-100% foodborne pathogens. They can in fact be transmitted to humans from multiple known and unknown reservoirs and transmission routes, which go beyond the foodborne route.

SA is important to guide intervention to control the spread of zoonotic pathogens from animal reservoirs and non-human sources to humans and support decision making.

In this document, the main components and activities of SA (figure 1) within the DiSCoVeR are summarised and critically evaluated to provide suggestions to optimise and ease the translation of the SA DiSCoVeR approach into the field, support capacity-building for SA and translate the results into actions in the field of the DiSCoVeR project.

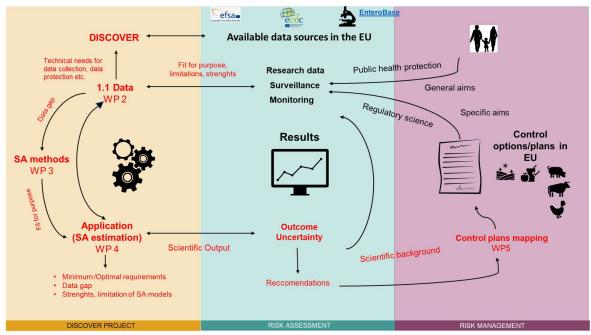


Figure 1. Main components and activities of the SA process within the DiSCoVeR (left panel) and the connections with the sectors of risk assessment (middle panel) and risk management (right panel).





# Setting the DiSCoVeR source attribution process

The DiSCoVeR project applied multiple methodological approaches to maximise the exploitation of existing data and data collected during the project (DiSCoVeR partners' data, data obtained through partners' network, as well as public available data) and to encompass various levels of the epidemiological transmission pathways, i.e. the reservoir level and the exposure level. This allowed to comprehensively explore the fit-for purpose of the different methods in relation to the quality and the amount of data and conversely to highlight the current data gaps to be covered to optimise future SA exercises. SA is a data driven process, where the quality of data (e.g. information on the hazard) and metadata (information on the sample and the sampling context), the criteria for sampling and source selection and prioritization, are key elements.

#### Core components of the DiSCoVeR SA are:

- Methods
- Data
- Skills

#### Methods

SA analyses in the DiSCoVeR were carried out using various methodological approaches to obtain comparable results. The objective was to maximise the exploitation of available data, typically collected for many different purposes using a wide range of study designs, in order to better characterise and quantify sources and transmission routes through the comparison or combination of outcomes from different approaches. An exhaustive description of existing attribution methods and principles is available in recent article by Mughini-Gras et al. (2019). As part of DiSCoVeR, quantitative SA methods have been catalogued and analysed in the D-JRPFBZ-1WP3.2 (*Report on the critical and quantitative evaluation of existing and novel source attribution methods*).

Briefly, methods applied in DiSCoVeR for SA, by pathogen are summarised in table 1 and comprised the following approaches:

- SA methods at the reservoir level: these methods identify the primary source of human infection but do not fully clarify the transmission route(s) of zoonotic infection (e.g. foodborne, waterborne, direct contact etc.). Methods focusing on reservoirs are based on microbial typing/subtyping and aim to compare the distribution of hazard subtypes in the different animal reservoirs and environmental sources and in humans to quantify the relative contribution of the sources to human infection. Typing/subtyping-based methods for SA have evolved in parallel to the technical evolution and have moved towards large-scale application of more discriminatory characterisation methods, mainly due to reduction of costs. Phenotypic-based methods (i.e. using serotyping, phagetyping, ribotyping, AMR profiling etc.) have been progressively replaced by genotype-based methods (MLVA, MLST, virulotyping) and Whole-genome-sequence (WGS) based methods.
- Frequency matching SA methods based on phenotyping and also genotyping (e.g. MLVA)
  data has been widely used by for many years to guide control actions e.g. for Salmonella.
  Information on serotypes, phagetype and antimicrobial resistance etc. is/has been collected
  in case-based surveillance making this approach suitable for the application to 'large' amount
  of data collected in surveillance. This microbial subtyping approach requires the minimum





- set of data, at current. SA based on frequency matching has been applied to *Salmonella* and STEC (table 1). For further details, see Deliverables D-JRPFBZ-1-WP4.1/2/3/4.
- The application of SA methods making use of WGS data (machine-learning, population structure) allowed to reduce remarkably the uncertainty compared to previous studies since the use of sequence based characterisation methods have a high discriminatory power. In DiSCoVeR this approach was used for all hazards (Table 1).
   Methods based on microbial subtyping approaches were used to perform the analysis at different geographical level, mainly nationally, regionally and/or more generally for the EU if possible and appropriate
- At the exposure/near exposure level, methods applied for SA in the DiSCoVeR were (Table 1):
  - Metanalysis of case-control studies from the scientific literature.
  - Methods based on aggregation of data from outbreak investigation, in collaboration with EFSA.
  - Exposure assessment based approaches which looks into exposure (e.g. animal contact) and hazard prevalences at the exposure level
  - Infectious disease modelling based on compartmental multidirectional dynamic models for AMR

A critical assessment of the existing quantitative model for SA and how to improve existing models including those that account for multidirectional transmission mode of agent between reservoir and humans have been carried out under WP3. Results are summarised in Deliverable D-JRPFBZ-1-WP3.2

Table 1: Summary of the methodological approaches used for the source attribution analysis, by pathogen (DiSCoVeR, 2022)

	Methods						
	Microbia	al Subtyping a	pproach	Meta- analysis of	Analysis of	Comparative	Multi- directional
	Frequency- Matching	Machine Learning	Population genetic	case control studies	outbreak data	exposure assessment	dynamic model
Salmonella	yes	yes	yes	yes	yes	yes	
Campylobacter		yes	yes	yes		yes	
STEC	yes	yes	yes			yes	
AMR (ESBL E.coli)	yes	yes				yes	yes

Data need, data availability at the EU level and critical evaluation of the DiSCoVeR datasets

## **Optimal data requirements**

Data needs for SA according to the different methodological approaches are shown in figure 2 and include information on infectious agents, the source, the sampling context, the production systems, human hosts including information on the clinical condition, demographic and animal population data (human population, animal population, food products, animal and products volumes,





#### companion animals population, wildlife numbers and distribution).

		Point of	attribution		Pathoge	n data		Human data			Source da	ata
Method	Reservoir*	Exposure	Transmission route	Risk factor	Subtyping	Fitness	Subtyping	Sporadic or epidemic status	Exposure & risk factors**	Subtyping	Prevalence and exposure***	Need to consider all potential sources
Epidemiological (case-control study)	No	Yes	Yes	Yes	No	No	No	Sporadic	Yes	No	No	High
Epidemiological (outbreak investigation)	No	Yes	Yes	Yes	No	No	No‡	Epidemic	Yes	No	No	High
Microbiological (frequency-matching models)	Yes	Yes	Yes	No	Yes	No±	Yes	Sporadic and epidemic	No‡	Yes	Yes	Medium
Microbiological (population genetics models)	Yes	Yes	Yes	No	Yes	No	Yes	Sporadic or epidemic	No	Yes	No	High
Quantitative exposure/risk assessment	Yes	Yes	Yes	Yes	No	Yes	No	N/A	Yes	No	Yes	Medium
Expert elicitations	Yes	Yes	Yes	Yes	No	No	No	Sporadic or Epidemic	No	No	No	Low

N/A, not applicable. \*Or amplifying hosts. \*\*Including travel status. \*\*\*In this case, food consumption weights. †But can be used if available.

The use of different categories of data type depend on the models and approaches. DiSCoVeR got data from many different sources. Many were from surveillance and monitoring of the hazards in individual countries, survey and studies from published and un-published, national and international statistics including industry annual report, and expert opinions.

### **Data quality**

Good quality data underpinning SA analyses is key for good quality SA estimates. In particular data should be accurate and representative (i.e. they should ideally be obtained from, but at least be representative of the population under investigation). Data and isolates collected under surveillance and monitoring programs in human and non-human reservoirs would ideally represent the best option because they are usually associated with high quality attributes/metadata (i.e. detailed information on source/sample/risk factors) on the domain of provenance of the samples/ isolates and are collected within specific sampling framework. These characteristics are important to both optimise the quality of input data (i.e. making it possible to select the more appropriate isolates according to the SA method being applied) and support the inference process of SA outputs, which is also important for the evaluation of the uncertainty around SA results.

#### Data availability at the EU level

Information on typing/subtyping of *Salmonella*, *Campylobacter*, Shigatoxin-producing *E. coli* and Extended-Spectrum Beta-lactamase *E. coli*, and the associated metadata (i.e. data describing the sampling context and the characteristic of the host/source/reservoir) were needed for DiSCoVeR SA.

At the EU-level, data on zoonotic agents typing/subtyping in humans and in animals, food and feed are collected from Member States within the case-based surveillance in humans (ECDC-Atlas, 2021) and the EFSA monitoring zoonotic agents (EFSA, 2022). Typing/subtyping Information focuses on species, serogroup/serotypes, phagetypes, AMR profiles and virulence genes types and subtypes (for STEC) despite descriptive metadata are generally highly detailed. However, the reporting of microbial typing/subtyping is not mandatory (except for data collected within Reg. (CE) 2160/2003 on Salmonella National Control Programs and Reg. (CE) 2073/2005 on Salmonella and STEC microbiological food-safety criteria).

WGS-based data collection, which for SA methods is the best option due to the high discriminatory power, is not well established within the surveillance and monitoring framework for foodborne pathogens and AMR at the EU level, although the main EU bodies have recently made important progresses. WGS-data are widely used at country level or EU level to support cluster detection and outbreak investigation under the framework of event-based surveillance (Epipulse). EFSA and ECDC have recently developed a One-Health interoperable WGS system for *Salmonella*, *Listeria* and STEC which could be queried in order to discover matching profiles to mainly support cluster detection,





outbreak investigation and source tracking (EFSA, 2022b). Data are stored according to specific legal framework and the interoperability of the two systems is subjected to specific business rules.

#### The DiSCoVeR datasets

The DiSCoVeR SA activities allowed us to use a wealth of data that were either originally produced for the DiSCoVeR needs or collected for other purposes by the DiSCoVeR partners (i.e. data from surveillance & monitoring, survey, official control, research, literature) or available in public repositories (e.g. EFSA dataset for foodborne outbreaks). This approach together with the production of secondary evidence studies (e.g. meta-analysis, analysis of foodborne outbreaks) matches well with international policies recommending re-use of data such as FAIR (Findability, Accessibility, Interoperability, and Reuse) (<a href="https://www.go-fair.org/fair-principles/">https://www.go-fair.org/fair-principles/</a>). At the same time, the re-use of data was remarkably challenged by barriers and/or complexity hampering accessibility to existing data in either the public or private domain or due to the lack of harmonisation.

Barriers for data sharing with specific focus on the use of WGS data in the context of foodborne outbreak detection and investigation were discussed with EFSA and ECDC with main focus on:

- **Technical**: e.g. lack of harmonization and standardization of methodological processes (e.g. pipelines for WGS analysis)
- Practical: lack of open platforms for sharing genomic sequences associated with metadata
- Formal: barriers associated with restrictive rules and policies for data protection and data usability that also go beyond the legitimate legal requirements (i.e. ownership of data and data usability subjected to specific restrictive terms of reference etc.). Overcoming these barriers would be greatly beneficial for future translation of SA into the field and call into action also actors not traditionally involved in the One-Health paradigm such as lawyers, bioethicists, food companies etc. to commit on how to do this in the future, as part of the societal engagement and responsibility in protecting public health

Inventory of available data for *Salmonella*, *Campylobacter*, STEC, and ESBL *E.coli* and metadata was done by the DiSCoVeR WP2. All the DiSCoVeR datasets were single structured stand-alone data collections, which received active contribution from the OHEJP partners. As such, datasets have been implemented for the purpose at the best of partners' ability in order to comply with the specific 'optimal' data requirements according to different SA methods implemented under the DiSCoVeR (DiSCoVeR D-JRPFBZ-1-WP2.6-data description). The DiSCoVeR datasets include typing/subtyping information as well as descriptive metadata (description of the samples and sampling context, information on area of sampling and date of sampling etc.):

- Inventories of available data and dataset for each hazard targeted by the DiSCoVeR are described in deliverables D-JRPFBZ-1-WP2.6 and are available in Zenodo (<u>link</u>)
- Genomic sequences of isolates collected for SA within the DiSCoVeR are/will be made public available in a DiSCoVeR dedicated project in the European Nucleotide Archive (ENA)

<sup>&</sup>lt;sup>1</sup> Several OH-EJP projects have been investigating barriers for data sharing on surveillance of zoonoses: Cohesive, NOVA, Orion. In addition, in 2020 a Workshop was funded by the Med-Vet-Net Association with EFSA, ECDC and other stakeholders focused on this topic in WGS data use foodborne zoonoses. <a href="https://www.foodsafetynews.com/2020/10/experts-assess-use-of-genome-sequencing-in-multi-country-outbreaks/">https://www.foodsafetynews.com/2020/10/experts-assess-use-of-genome-sequencing-in-multi-country-outbreaks/</a>





- Metadata associated with genomic sequences are/will be made public available in Zenodo.
   Reference key are/will be the accession number.
- The datasets will serve the DiSCoVeR aims and will not be updated in the future, which raises the question of the obsolescence of the datasets. This represents a major limitation and raises the question for the future sustainability of the DiSCoVeR experience for the translation of the SA into the One-Health 'surveillance & monitoring' in the EU.

#### **Evaluation of the DiSCoVeR datasets**

SA is a data driven exercise and covering with accurate, detailed and representative data the different sources, is very important. Precision of the estimates depends on both the quality of data and the amount of information collected.

Compared to previous studies the DiSCoVeR aimed at increasing the data coverage in different sources also targeting non livestock reservoirs and non-food transmission routes. Data have been obtained, by collecting and collating data from DiSCoVeR partners where possible, also on non-livestock reservoirs (i.e. wildlife) and environment to fill out knowledge gap that the DiSCoVeR had identified.

Figure 3 describes the final size of the DiSCoVeR data inventory for *Salmonella*, *Campoylobacter* and STEC in years 2015-2019 (for STEC 2010-2020) and coverage in non-human sources, by sector and type of source. The distribution was rather unbalanced with the vast amount of data available for few animal species, namely the species whose epidemiological role of reservoir is well known for the given pathogen:

- S. Enteritidis in Layers and broilers
- S. Typhimurium (including monophasic variant) in pigs
- S. Infantis in broilers
- S. Derby in pigs
- Campylobacter in broilers
- STEC in cattle

Altogether, the hazard/source combinations listed above accounted for more than half of total data (54.7%) collected in years 2015-2019. Data from livestock species (pigs, cattle, sheep and goats, *Gallus gallus*, turkeys, ducks and goose) accounted for the 84.2 % of all data. Pets and cats were the sources most frequently reported among non-food producing animals (6.2%) followed by wild birds (4.2%), horses (1.7%), reptiles (1.3%) and other wild animals (0.5%). However, data from non-livestock sources were contributed with by only a few countries. Environmental samples were scarcely represented for all hazards except for *Campylobacter* (17.2%) with only two countries contributing to this source and hazard. Distribution of data on positive samples for *Salmonella*, *Campylobacter* and STEC collected by EFSA in the same period (years 2015-2019²) is shown in figure 4 for comparison with the DiSCoVeR datasets and get information on the representativeness of the DiSCoVeR data collection against monitoring data of the same pathogens in animals, in the EU.

<sup>&</sup>lt;sup>2</sup> Data are available in the EFSA on-line resources and in the EFSA Knowledge Junction in Zenodo: 2015 (https://efsa.onlinelibrary.wiley.com/doi/epdf/10.2903/j.efsa.2016.4634); 2016 (https://doi.org/10.5281/zenodo.1044742); 2017 (https://doi.org/10.5281/zenodo.1475841); 2018 (https://doi.org/10.5281/zenodo.3527706); 2019 (https://doi.org/10.5281/zenodo.4298993).





Figure 3 - Distribution of positive samples by zoonotic hazard and source in year 2015-2019 (data source DiSCoVeR inventory). For STEC the sampling period is 2010-2020.

5 years sum of positive samples		S. Enteritidis		S.Typhimurium+ Monoph	hdo	S.	S. Infantis	S. Newport	oort	S. Derby	rby	Campylobacter	bacter	STEC	O
	N positive	N positive N country	N positive	N country		N positive	N positive N country	N positive N country		N positive N country	l country	N positive N country		N positive N country	country
Pigs	2	2	494	6		12	æ	20	2	92	8	146	4	119	2
Cattle *	9	2	57	9		m	8	15	1	9	П	221	c	773	10
Sheep/goats	1	1	19	2		1	1	1	1	0	0	107	1	209	7
Chicken (broilers)	248	9	77	5		98	2	65	2	15	2	755	9		
Layers	369	9	70	5		6	4	15	2	7	2	55	1		
Gallus gallus (other)															
Turkeys *	10	8	62	4				36	2	30	2	35	2		
Ducks/goose *	34	2	35	8		13	1	7	1	1	1				
Pets (dogs/cats)	15	2	94	4		17	2	21	2	7	П	161	ĸ	2	2
Pets (reptiles)	20	2	7	2		17	2	21	2						
Wild animals *	12	4	7	8		e	1			1	1	7	1		
Wild birds	18	8	б	2		1	1			0	0	190	co		
Horses *	13	1	49	2		2	1	21	1	0	0				
Other animals *														86	7
Zoo animals															
Shellfish															
Environment (Water)	6	ю	4	2		0	0	-	ч	9	н	287	7	31	4
Total	748	6	086	6	6	164	6	222	6	159	6	1672	9	1204	10

For STEC the sampling period refers to years 2010-2020.





Figure 4 - Distribution of positive samples by zoonotic hazard and source in year 2015-2019 (data source EFSA).

5 years sum of positive		S. Enteritidis	S.T	S.Typhimurium	nmihd.	phimurium, monoph		S. Infantis	S. Newport	vport	S. D	S. Derby	Campylobacter	obacter	STEC	, SI
samples		EFSA		EFSA		EFSA	8	EFSA	EFSA	Ä	Ħ	EFSA	EFSA	SA	EFSA	Ą
	N positive	Country	N positive	Country	N posi	N positive Country	N positive Country	Country	N positive Country	Country	N positive Country	Country	N positive Country	Country	N positive Country	Country
Pigs	53	6	1952	20	287	7 14	75	<b>6</b>	24	5	773	16	136	10	299	9
Cattle *	424	14	3627	16	114	1 12	20	6	147	2	14	9	2282	12	3723	16
Sheep/goats	34	2	83	6	34	2	•	ı	m	1	4	2	454	10	1036	13
Chicken (broilers)	2905	25	1578	27	62	7	22871	23	269	10	279	14	8789	17		
Layers	2593	28	711	29	12	9	787	21	54	6	23	10				
Gallus gallus (other)													140	ıc		
Turkeys *	258	12	557	17	6	9	1119	11	710	6	1981	9	786	9		
Ducks/goose *	221	2	117	თ	_	'	15	2	15	2	9	33	1	1		
Pets (dogs/cats)	43	7	792	7	6	3	44	2	m	2	8	2	412	6	190	9
Pets (reptiles)	8	3	'		_		•	,	,	,			,	,		
Wild animals *	68	14	840	14	52	4	42	9	24	9	24	9				
Wild birds													227	2		
Horses *	14	4	113	2	27	2	2	2	37	ю	-	,	9	е		
Other animals *															1573	∞
Zoo animals															2	1
Shellfish	,	ı	,	ı	'	ı	,	'	•	ı	,	ı	,	,		
Environment (Water)	,	•	,		•	•	,		•	ı	•	,	,		44	11





As expected, data on environmental sources were absent from the EFSA dataset but present in the DiSCoVeR. Main differences between the DiSCoVeR dataset and EFSA data summary statistics = were a much higher proportion of samples from non-food producing animals (Dogs/Cats, Reptiles, Wild animals, Wild birds, Other animals, Zoo animals) in the DiSCoVeR dataset for all pathogens with few exception. Proportion of *S*. Typhimurium including monophasic and *S*. Derby in pigs were higher in the DiSCoVeR and the same was for STEC in cattle. *Campylobacter* and *S*. Infantis in broiler chicken were reported in lower proportion in the DiSCoVeR than in the EFSA dataset, and the same was observed for *S*. Newport in turkey.

Comparison on data on *Salmonella*, *Campylobacter*, and STEC in human cases reported to DiSCoVeR and ECDC as well as coverage of non-human sources by agent and by countries are reported in the Appendix.

## Optimisation of dataset for SA

In summary, both DiSCoVeR and EFSA datasets are characterised by unbalanced distribution of positive samples among the different sources. This is probably due to a most intensive sampling activities carried out in specific sources according with the targeted zoonotic hazards. It also reflects a higher probability to obtain isolates for further analysis in those sources (e.g. *S.* Enteritidis in layers). Environmental samples (water) are only available in DiSCoVeR dataset in particular for *Campylobacter*.

Datasets characterised by unbalanced distribution with few oversampled sources (e.g. *Campylobacter* in poultry) and others missing information or with few data available may be a problem, because they can lead to biased attribution estimates.

For this reason efforts should be made to balance the availability of data with active and supplementary sampling in specific sources, to improve the quality of SA estimated based on typing/subtyping method.

A sustainable way to optimise the availability of data could be to promote sampling through synergies with other already existing sampling programs (e.g. pigs or turkeys monitoring survey for AMR could be used to get samples to be tested for *Campylobacter*). Synergies with regular sampling from baseline surveys would allow to get comparable data from all countries.

Optimisation of SA means finding a compromise between the need to enlarge the dataset and the coverage of the different sources and the quality of data being introduced in the model. Increase of data to enlarge source coverage may result in increased uncertainty, if data are not as accurate and representative as needed (e.g. data collected outside sampling survey or surveillance and monitoring). For this reason the quality of data especially for reservoir and sources less frequently populated in terms of data should be carefully evaluated in advance. Optimal data requirements for the DiSCoVeR has been specified in deliverable D-JRPFBZ-1-WP2.6.

## Skill

#### The DiSCoVeR scientific community

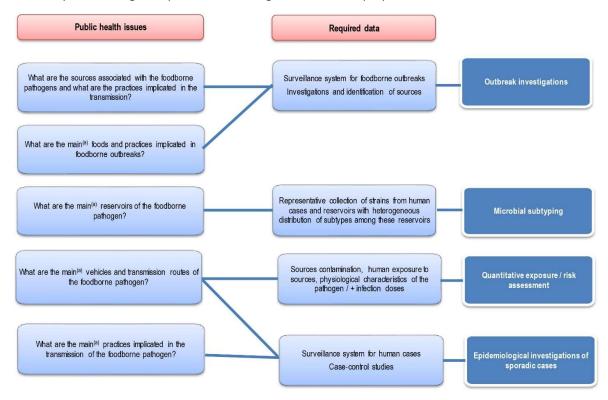
Under the umbrella of the DiSCoVeR project, expertise and skills from multiple sectors and disciplines have been brought together into a small scientific One-Health community. Collaboration, integration, networking as well as exchange of knowledge and skills among partners' experts (19 partners from 13 countries) allowed to comprehensively approach 'SA science' including its main components (methods, data, knowledge, operational framework). This was an added value of the DiSCoVeR and an important achievement for future translation of the capacity building into the EU context.





## Conclusions

- Although past experience were reported of SA sequence-based studies, the DiSCoVeR project allowed for the first time to conduct large-scale SA exercise at the EU-level thanks to the availability of large high-quality (fit for purpose) data also including non-traditional tested sources (pets, wildlife, environment). This is an important achievement of the DiSCoVeR and OHEJP consortium in terms of scientific evidence outcomes and uncertainty that have been made available to end-users, i.e. primarily public health authorities, and other stakeholders including farmers, industry etc. to be taken into account for future development of policy of intervention to protect public, animal and environmental health and support sustainability of primary production.
- The release of the DiSCoVeR comprehensive large One-Health datasets in the public domain
  has been an important initiative for ease the translation of the SA into the field and
  represents an important achievement of the project, which, however, raises in the near
  future the question of the obsolescence and future sustainability of these important
  resources, as resources for updating the datasets are not available.
- At the same time, the DiSCoVeR highlighted the importance of building public and accessible repositories of WGS data and metadata not only to support investigation of outbreaks. It also provided a proof of concept for the long-term use of WGS-based data from and the structural integration of WGS data into monitoring and surveillance activities pursuing One-Health objectives. This would enable performing SA studies on a regular basis in order to support the monitoring of the effect of intervention programs and the development of new policies to protect public and animal health against priority foodborne zoonoses and AMR.
- Optimisation of SA approach is a complex task and selection of best appropriate approach
  can be guided by the aims, the quality of available data, the skills and the public health
  questions. Figure 5 provides a useful guidance for this purpose.







# Figure 5 Preferential choice of source attribution methods based on public health issues. (a)Ranking and/or quantifying the relative importance (from Mughini-Gras et. al 2019)

- SA methods based on genomic data represent nowadays the gold standard for SA, since they
  have the highest discriminatory power. However, the choice of the best methodological
  approach for SA should take into account that genomic-based data requires high quality data
  and skills to be reliably implemented.
- On the other hand, frequency-matching SA models making use of traditional pheno/genotypic characterisation data for *Salmonella* and STEC continue to represent an interesting option at country level and EU level, since these data can be regularly collected within surveillance and monitoring of foodborne zoonoses in human and non-human sources.
- Training and dissemination to improve skills and capacity-building in SA is definitely a crucial resource for SA translation into the field.

## References:

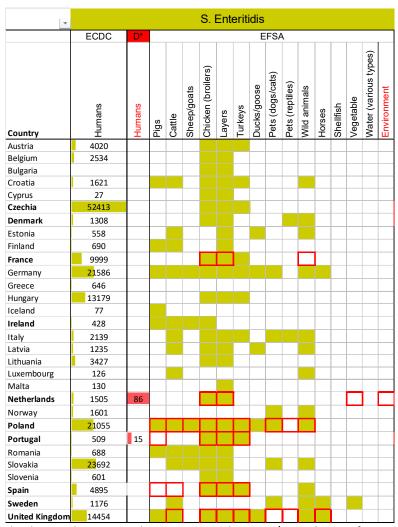
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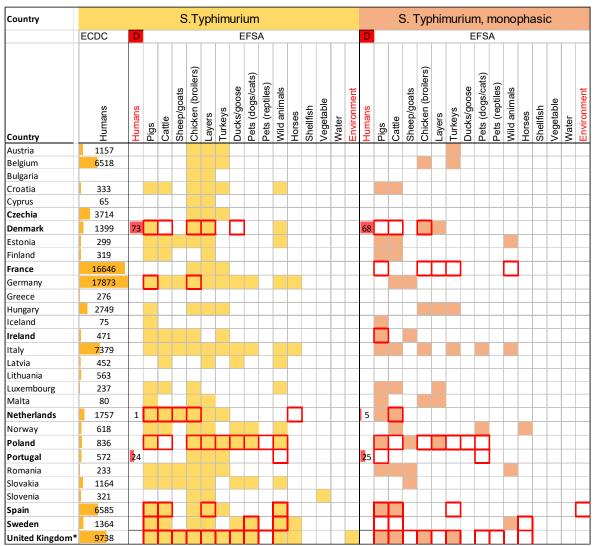
# **Appendix**

Comparison on data on *Salmonella*, *Campylobacter*, and STEC for human cases reported to DiSCoVeR and ECDC (data Source ECDC Atlas) and data coverage for the same agents in different animal species and other non-human sources within the DiSCoVeR and the EFSA datasets, in years 2015-2019









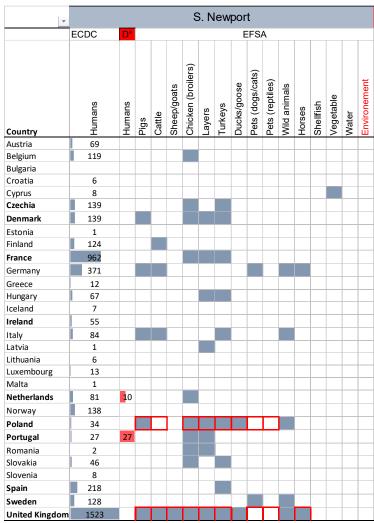




¥						S	. In	fan	tis							
	ECDC	D*							EF	SA						
Country	Humans	Humans	Pigs	Cattle	Sheep/goats	Chicken (broilers)	Layers	Turkeys	Ducks/goose	Pets (dogs/cats)	Pets (reptiles)	Wild animals	Horses	Shellfish	Vegetable	Water
Austria	375															
Belgium	288															
Bulgaria																
Croatia	59															
Cyprus	9															
Czechia	525															
Denmark	96															
Estonia	150															
Finland	54															
France	1240															
Germany	1421															
Greece	15															
Hungary	1224															
Iceland	1															
Ireland	37															
Italy	452															
Latvia	62															
Lithuania	115															
Luxembourg	17															
Malta	56															
Netherlands	161	25														
Norway	57															
Poland	345															
Portugal	5	14														
Romania	20	Г														
Slovakia	524															
Slovenia	55															
Spain	334				Ī								ĺ			
Sweden	50															
United Kingdom	1176															











	S. Derk	ΟV															
	ECDC	D*							_	FSA							
	ECDC	_								FOF	`						
Country	Humans	Humans	Pigs	Cattle	Sheep/goats	Chicken (broilers)	Layers	Turkeys	Ducks/goose	Pets (dogs/cats)	Pets (reptiles)	Wild animals	Horses	Shellfish	Vegetable	Water	Environment
Austria	28																
Belgium	217																
Bulgaria																	
Croatia	28																
Cyprus	2																
Czechia	87																
Denmark	70																
Estonia	14																
Finland	7																
France	727																
Germany	699																
Greece	3																
Hungary	263																
Iceland																	
Ireland	5																
Italy	536																
Latvia	24																
Lithuania	92																
Luxembourg	13																
Malta																	
Netherlands	62	10															
Norway	17																
Poland	54																
Portugal	1	9															
Romania	14																
Slovakia	71																
Slovenia	9																
Spain	142			П					Ì				Ì				
Sweden	15			Т													
United Kingdom				Т													





	Camyloba	cter j	eju	ni												
	ECDC	D					E	EFS/	4							
Country	Humans	Humans	Pigs	Cattle	Sheep/goats	Chicken (broilers)	other Gallus gallus	Turkeys	Ducks/goose	Pets (dogs/cats)	Wild birds	Horses	Horses	Shellfish	Vegetable	Water
Austria	40522															
Belgium	48886															
Bulgaria	1171															
Cyprus	135															
Czechia	132676															
Germany	388148															
Denmark	26997	1230														
Estonia	1924															
Greece	1741															
Spain	82588															
Finland	25069															
France	42474					Г				П		Г				
Croatia	9344															
Hungary	42683															
Ireland	15982					П							Ī			
Iceland	742															
Italy	7538															
Lithuania	6225															
Luxembourg	3010															
Latvia	547															
Malta	1508															
Netherlands	19106	262														
Norway	18762															
Poland	4148	7														
Portugal	3513	299											Ī			
Romania	2973															
Sweden	49069															
Slovenia	7579															
Slovakia	42467															
United Kingdom	305929															





		S	STE	С				
	ECDC	D			EF	SA		
Country	Humans	Humans	Pigs	Cattle	Sheep/goats	Pets (dogs/cats)	Other/zoo animals	Environment
Austria	2010							
Belgium	1160							
Bulgaria	2							
Cyprus	0							
Czechia	245							
Germany	22454							
Denmark	3245							
Estonia	65							
Greece	20							
Spain	858							
Finland	1273							
France	2651							
Croatia	60							
Hungary	136							
Ireland	6648							
Iceland	52							
Italy	675	364						
Lithuania	17	_						
Luxembourg	71							
Latvia	59							
Malta	165							
Netherlands	7660	2937						
Norway	2605	557						
Poland	53	_						
Portugal	9	6						
Romania	123							
Sweden	6138							
Slovenia	295							
Slovakia	55							
United Kingdom	13551	468						