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de Freitas Costa, Eduardo; Corbellini, Luis Gustavo; da Silva, Ana Paula Serafini Poeta; Nauta, Maarten

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- A stochastic model to assess the effect of meat inspection practices on the contamination of
 the pig carcasses
- 3
- 4 Eduardo de Freitas Costa^{*1}; Luis Gustavo Corbellini¹; Ana Paula Serafini Poeta da Silva¹;

5 Maarten Nauta²

- 6 1: Laboratory of Veterinary Epidemiology (Epilab), Department of Preventive Veterinary
- 7 Medicine, Federal University of Rio Grande do Sul, Brazil
- 8 Address: Avenida Bento Gonçalves, n° 9090, zip code 91540-000, Porto Alegre, RS, Brazil;
- 9 phone: +55 51 3308 8025
- 10 2: Technical University of Denmark National Food Institute
- 11 Mørkhøj Bygade, 19, Building G, DK-2860 Søborg, Denmark

12 Abstract

The objective of meat inspection is to promote animal and public health, by preventing, detecting 13 and controlling hazards originating from animals. With the improvements of sanitary level in pig 14 herds the hazards profile has shifting and the inspection procedures have no longer targeting 15 major foodborne pathogens (i.e., not risk-based). Additionally carcass manipulations performed 16 when searching for macroscopic lesions can lead to cross-contamination. We therefore 17 18 developed a stochastic model to quantitatively describe cross-contamination when consecutive carcasses are submitted to classic inspection procedures. The microbial hazard used to illustrate 19 the model was *Salmonella*, the data set was obtained from Brazilian slaughterhouses and some 20 21 simplifying assumptions were made. The model predicted that, due to cross-contamination during inspection, the prevalence of contaminated carcass surfaces increased from 1.2% to 22 95.7%, whereas the mean contamination on contaminated surfaces decreased from 1 to -0.8723 24 logCFU/cm², and the standard deviations decreased from 0.65 to 0.19. These results are explained by the fact that, due to carcass manipulations with hands, knives and hooks, including 25 the cutting of contaminated lymph nodes, *Salmonella* is transferred to previously uncontaminated 26 carcasses, but in small quantities. These small quantities can easily go undetected during 27 28 sampling. Sensitivity analyses gave insight in the model performance and showed that the 29 touching and cutting of lymph nodes during inspection can be an important source of carcass contamination. The model can serve as a tool to support discussions on the modernization of pig 30 carcass inspection. 31

32 Keywords: Mathematical modeling, cross-contamination, carcass inspection

33

34 1. INTRODUCTION

The main objective of meat inspection is to promote both animal and public health, by 35 preventing, detecting and controlling microbial hazards originating from animals.⁽¹⁾ Although no 36 precise definition about meat inspection procedures has been proposed, the *Codex Alimentarius* 37 refers to two types of inspection. Ante-mortem examination consists of a clinical examination 38 aimed at detection of non-healthy animals. Post-mortem examination consists of a pathological 39 examination to identify potential hazards for human or animal health.⁽²⁾ Classically the 40 inspection of pigs is done at all carcasses and the procedures are based on physical examinations, 41 42 like incisions, palpation and observation of the carcass, organs and lymph nodes, searching for macroscopic lesions, typical for classical zoonotic diseases.⁽³⁾ 43 Although the recognition of animals as a source of pathogens to humans dates from prehistoric 44 times, the current procedures were developed in Europe by Robert von Ostertag in 1900.⁽⁴⁾ They 45 have an important role in controlling zoonotic diseases, mainly in places, where the production is 46 not done in an intensive production chain⁽⁵⁾ and, consequently, classic zoonosis are endemic. 47 The global livestock production systems have undergone an industrial revolution and the 48 production has shifted increasingly from smallholders to large-scale, industrial production 49 chains. An increasing share of production comes from pigs and chickens that are more easily 50 adapted to large-scale industrial production than ruminants.⁽⁶⁾ In 2010, even in developing 51 countries, at least 50% of the herds in pork production are processed in integrated productions 52 systems.⁽⁷⁾ 53 Nowadays, farms adhere to specific management requirements like all-in-all-out production, 54 55 controlled feed sources, indoor production, and a traceability system from the farm to the

56 slaughterhouse.⁽⁸⁾ As a consequence, hazards like parasites are getting rare in the industrial pork

production chain.⁽⁸⁻¹⁰⁾ On the other hand, the intensification of the production brings changes in
the epidemiology and other microbial pathogens are emerging.⁽¹¹⁾ Salmonella spp., Yersinia *enterocolitica*, Toxoplasma gondii and Trichinella spp. are identified as the most important
hazards to be covered by the meat inspection of swine carcasses.⁽⁵⁾

The interaction of these pathogens with the host and the environment raises some concerns about the suitability of the classic inspection procedures. It demands structured control using all food chain information to reach a risk-based inspection system.⁽¹²⁾ The modernization of meat inspection has been extensively studied in Europe and since 2014, according to EC Regulation 219/2014, the inspection of pig carcasses is visual-only for pig herds that have been reared in integrated farm systems, doing palpation and incision when a lesion has been found after visualonly inspection. ⁽¹³⁾

In 2011 the European Food Safety Authority (EFSA) discussed the limitations of the meat

69 inspection system procedures, such as lymph node incision, in terms of consumer health

70 protection and stated that the classic procedures could increase the level of cross-contamination,

also for zoonotic pathogens. However, quantitative data on the impact of the inspection

72 procedures on cross-contamination are lacking and Hill *et al.*⁽¹⁴⁾ highlighted the need of studies

regarding the cross-contamination during the inspection procedures to support a risk-based

approach to meat inspection, which could improve the efficiency in dealing with public healthissues related to animal slaughter.

In this paper we describe a modelling approach to study the impact of meat inspection practices on cross-contamination between pig carcasses and to provide insight in the potential effect of these practices on the prevalence and concentration of pathogens on pig carcasses. Using methods applied in quantitative microbiological risk assessment (e.g. Nauta *et al.*)⁽¹⁵⁾, we aim to

quantify the cross-contamination during meat inspection of pig carcasses via specific transfer 80 routes and to assess their relevance for the contamination of the carcasses. The model is set up as 81 a generic model for cross-contamination during meat inspection and is applied to Salmonella 82 transfer during inspection of pig carcasses in Brazil, because there is relevant data available from 83 some large slaughterhouses in Brazil, and transfer of this pathogen from lymph nodes to the 84 carcass surface has been considered a potential hazard.^(16,17) To illustrate the model, we focus on 85 the point of the meat inspection identified as "CARCASS" by the Food and Agriculture 86 Organization of the United Nations (FAO).⁽¹⁸⁾ This point of inspection is not the same in all 87 88 countries. In Brazil this inspection occurs after the carcass splitting and refers to inspection of specific parts of the pig carcass by looking, cutting and touching the skin, musculature, exposed 89 bones, joints, tendon sheaths and serous membrane. It also includes several cuttings and 90 palpation of the following lymph nodes: superficial inguinal, supramammary, external and 91 internal iliac, according to ordinance 711/1995.⁽¹⁹⁾ 92

93 2. MATERIALS AND METHODS

94 **2.1 Conceptual model**

The basic structure of the model and the transfer routes considered are shown in Fig. 1. The 95 model has a similar structure as the one developed by Nauta et al.⁽¹⁵⁾ for broiler processing and is 96 based on classic meat inspection procedures, where a series of consecutively slaughtered 97 98 carcasses are submitted to several manipulations, and cross-contamination between carcasses 99 may occur via equipment (like cutting knives and hooks used to hang up the carcasses) and hands. Therefore, the *knife*, *hands and hook* were considered as the relevant components of the 100 101 slaughter environment. As both the surface and organs of the pig may get in contact with hands 102 and equipment, the carcass was separated in two components: the carcass surface and the

possible *organs* evaluated during the inspection. Contact between the carcass (carcass surface
and organs) and the environmental components occurs on specific areas of the carcass surface
and the organs. The transfer of bacteria can happen from the environmental components to
carcass and from the carcass to environmental components.



107

Fig. 1. Schematic representation of the pig carcass inspection. Consecutive carcasses pass
through the point of inspection and get in contact with the environmental components, which can
lead to cross-contamination via bacterial transfer from the environment to the carcass (arrow a)
or from the carcass to the environment (arrow b). The arrows d represent the reduction in the
concentration of the bacteria due to inactivation or removal.

113

The model only considers the carcass and the predefined environmental components as sources of bacteria, the influence of the air, carcass to carcass contact or other external factors are ignored. Also, bacterial growth during the inspection is excluded from the model. Removal of the bacteria from the carcass (surface or organs) can only occur by the inspection activities that are included in the model. Bacteria on the knife are frequently inactivated by putting the knife in

hot water (i. e. 83 °C). Washing of hands and cleaning of the hooks are unusual or don't follow a
clear rule during meat inspection and have therefore not been considered.

121 **2.2 Mathematical model**

127

122 The model can be written as a system of five difference equations as given below (1). It

describes the changes in the concentrations in the five components for consecutively slaughtered carcasses *i*, before inspection (stage S-1) and after inspection (stage S). Variables are listed in Table 1. The upper cases letters represent variables, and lower case letters represent model parameters. *Ae* and *Ac* are used as index and refer to the different areas on carcass (Ae =knife

(K), hand (H), hook (G)) and different compartments of carcass (Ac = surface (C) or organs (O)).

$$C_{i,S} = \sum_{Ae \in \{K,H,G\}} (1 - d_{C})(1 - b_{C,Ae})^{J_{i,C,Ae}} C_{i,(S-1),Ae} + E_{(i-1),Ae} (1 - (1 - a_{Ae,C})^{J_{i,C,Ae}})$$

$$O_{i,S} = \sum_{Ae \in \{K,H,G\}} (1 - d_{O})(1 - b_{O,Ae})^{J_{i,O,Ae}} O_{i,(S-1),Ae} + E_{(i-1),Ae} (1 - (1 - a_{Ae,O})^{J_{i,O,Ae}})$$

$$K_{i} = K_{(i-1)} \prod_{Ac \in \{C,O\}} (1 - a_{K,Ac})^{J_{(i-1),Ac,K}} + \sum_{Ac \in \{C,O\}} N_{i,Ac} (1 - d_{Ac}) (1 - (1 - b_{Ac,K})^{J_{i,Ac,K}})$$

$$H_{i} = H_{(i-1)} \prod_{Ac \in \{C,O\}} (1 - a_{H,Ac})^{J_{(i-1),Ac,H}} + \sum_{Ac \in \{C,O\}} N_{i,Ac} (1 - d_{Ac}) (1 - (1 - b_{Ac,H})^{J_{i,Ac,H}})$$

$$G_{i} = G_{(i-1)} \prod_{Ac \in \{C,O\}} (1 - a_{G,Ac})^{J_{(i-1),Ac,G}} + \sum_{Ac \in \{C,O\}} N_{i,Ac} (1 - d_{Ac}) (1 - (1 - b_{Ac,G})^{J_{i,Ac,G}})$$

The variable $E_{i,Ae}$ is the generic term to refer to the environment and the value of Ae for knife, 133 hands or hook will be used according to the area modeled. Similarly, the variable $N_{i,Ac}$ is a 134 135 generic term to refer to the carcass surface or organs, according to the component Ac modeled. 136 Organs will be referred to from here onward, as lymph nodes, because that is the most relevant 137 organ evaluated during this specific inspection point. The variables are explained in the Table I. 138 The numbers of contacts between the environmental components and the carcasses compartments are represented generically by $J_{i,Ae,Ac}$. When Ae = K, H, G it refers to the number of 139 contacts between the carcass and the knife, hand and hook respectively. These values are 140 sampled from empirical distributions (see appendix A) and are assumed to affect either the 141

142	surface or lymph nodes ($Ac=C$, O) with equal probability (50%). Also, the three areas on carcass
143	or lymph nodes are considered mutually exclusive: the worker does not touch the same carcass
144	area with his hands as the worker cuts with a knife or holds the carcass with the hook.
145	

Table I. Overview of model variables (eq. 1): Each variable describes a quantity that is changing

147 for consecutive carcasses (with rank number i) and over the process steps S-1 (before) or S (after

148 inspection). Values before inspection are sampled from the indicated distributions and values

149 after inspection are calculated by the model

Variable	Description	Distribution/function	Unit	Source
Ji,Ac,Ae	Number of cuts, touches or hooking (Ae) in the surface or organs (Ac) of the carcass <i>i</i>	Empirical [#]	Count	Appendix A
$log_{10}[C_{i,(S-1)}]$	<i>Salmonella</i> concentration on the carcass surface <i>i</i> before inspection (S-1) on contaminated carcass.	Normal(-5.4;2.2) ^{\delta #}	log10CFU/cm2	(20) Appendix B
PrevC _{i(S-1)}	Status of carcass surface contamination on the carcass <i>i</i> before inspection (S-1)	100%	Positive/Negative	(20) Appendix B
$C_{i,(S-1)}$	Salmonella counts on the carcass surface <i>i</i> before inspection (S-1) on contaminated carcass.	$Poisson([C_{i(S-1)}]ea_{Ae}J_{i,C,Ae})$	CFU ^{##}	Calculation, see Table 2
$[O_{i,(S-1)}]$	<i>Salmonella</i> concentration in organs (lymph nodes) <i>i</i> before inspection (<i>S-1</i>) in contaminated lymph nodes.	Triangular(0.1;1;100) #	CFU/cm ²	Assumption
$O_{i,(S-1)}$	<i>Salmonella</i> counts in organs (lymph nodes) <i>i</i> before inspection (<i>S-1</i>) in contaminated lymph nodes.	$Poisson([O_{i(S-1)}]ea_{Ae}J_{i,0,Ae})$	CFU ^{##}	Calculation, see Table 2
PrevO _{i,(S-1)}	Status of organs contamination in the carcass <i>i</i> before inspection (S-1) (i.e., carrying Salmonella in lymph nodes)	Bernoulli(14.1%)	Positive/Negative	(11)
K _i	Amount of Salmonella on knife by the carcass i after inspection	Model	CFU	Calculation
H _i	Amount of Salmonella on hands by the carcass i	Model	CFU	Calculation

	after inspection			
	Amount of <i>Salmonella</i> on hook by the carcass i G_i	Model	CFU	Calculation
450	after inspection	and is Operation	(u = and p	
150	Distribution expressing variability between carcas	sses <i>i</i> ; 'Parameters	$(\mu, \sigma \text{ and } Prev$	$C_{i(S-1)}$) were
151	fitted according a zero inflated normal distribution b	oy Maximum Log	likelihood est	imation
152	method (Appendix B); ##CFU per inspected area.			
153				
154	The model was implemented as a Monte Carlo simu	llation model. Tra	nsfers were de	scribed as
155	binomial processes taking into account the successive	ve contacts betwee	en environmer	it and carcass,
156	as explained in appendix C. For example, in the first	t term in the equat	ions consideri	ing the
157	carcass, $(1 - d_c)(1 - b_{C,Ae})^{J_{i,C,Ae}}$ is the fraction of the n	umber of Salmone	ella that are no	ot lost by
158	removal (d) and not transferred from the carcass, to	the environment of	on different are	eas, indicated
159	by the index Ae (knife area, hand area, and hook are	ea). The second ter	m is $(1 - (1 - $	$a_{Ae,C})^{J_{i,C,Ae}}),$
160	the fraction of the number of Salmonella received fr	rom the environme	ent indicated b	y the index
161	Ae (knife, hand and hook) to the carcass and can be	derived as explain	ned in appendi	x C.
162	In the last three equations, modeling the environment	ntal components, u	ising the knife	as example,
163	the first term: $K_{(i-1)} \prod_{Ac \in \{C, O\}} (1 - a_{K, Ac})^{J_{(i-1), Ac, K}}$ conce	rns the Salmonelle	a that are not t	ransferred
164	from the knife to the carcass $(i-1)$ on different comp	artments indexed	by <i>Ac</i> (surface	and lymph
165	nodes). The second term $(1 - d_{Ac})(1 - (1 - b_{Ac,K})^{J_{i,Ac,K}})$	indicates the Saln	<i>nonella</i> receiv	ed from the
166	carcass indicated by the index Ac (surface or lymph	nodes). The varia	bles $C_{i,(S-1)}$	and
167	$O_{i,(S-1)}$ represent the counts of <i>Salmonella</i> before t	he inspection and	describe the v	ariability
168	between inspected carcass surfaces and organs respe	ectively. To account	nt for the rand	om spatial
169	distribution of cells over the inspected area, a Poisso	on distribution wa	s used. In orde	er to assess
170	the true prevalence, the variables $C_{(i-1),S}$ and $O_{(i-1),S}$	1), <i>s</i> were multiplie	ed by the posit	ive/negative

171 status (1 or 0) of carcass surface contamination $PrevC_{i,(S-1)}$ and lymph nodes $PrevO_{i,(S-1)}$, 172 both sampled from Bernoulli distributions.

The transfer parameters a, b are used in combination with the index Ae or Ac according to the area on the carcass or the environmental components modeled. For instance when the parameter a is used with index Ae, it refers to the probability of transfer of a CFU from the environment according index Ae used (knife, hand or hook) to the carcass (C) or lymph nodes (O). The removal parameter d, is indexed by Ac because detection or reduction are accounted only on carcass surface and lymph nodes. Table II provides an overview of the parameters used in the model.

180 Counts of *Salmonella* were expressed in CFU and the outputs were calculated for the inspected

181 areas (CFU/cm²) and then transformed to natural logarithm (presented here as "log"),

182 considering only the contaminated carcasses (because $\log (0)$ is not defined). When a carcass has not been submitted to any contact with the environment by hands, knife or hook, the carcass was 183 considered as not inspected and, consequently, the concentration on inspected area is assumed to 184 be the same as before (S-1). Also the probability of inactivation or removal on carcass or in 185 lymph nodes $(d_c \text{ or } d_0)$ are underlying assumed to be zero. The analyses were done using 186 @Risk 6.2.1 (Palisade) for Excel with 10000 iterations using 500 and 100 consecutively 187 processed carcasses in two separate simulations. These numbers were chosen to approximate 188 realistic numbers of pigs slaughtered in a slaughter line per shift of two hours (i.e. 350 189 carcasses/hour), whilst keeping the model manageable and restricting the running time. 190

191

Table II. Parameters used to illustrate the dynamics of the model. The indices *Ae* and *Ac* are
given by the initials of environment and carcass compartments respectively

Parameters	Description	Unit	Value	Source
$a_{K,C}$	Transfer probability knife-carcass	%	0.17	(21)
$a_{K,O}$	Transfer probability knife-lymph nodes	%	0.17	(21)
$a_{H,C}$	Transfer probability hand-carcass	%	0. 21	(22)
<i>a_{H,0}</i>	Transfer probability hand-lymph nodes	%	0. 21	(22)
$a_{G,C}$	Transfer probability hook-carcass	%	0. 17	(21)
$a_{G,O}$	Transfer probability hook-lymph nodes	%	0.17	(21)
$b_{C,K}$	Transfer probability carcass-knife	%	0.17	(21)
b _{C,H}	Transfer probability carcass-hand	%	3.1	(22)
b _{C,G}	Transfer probability carcass-hook	%	0. 17	(21)
<i>b</i> _{0,K}	Transfer probability lymph nodes-knife	%	0.17	(21)
<i>b</i> _{0,H}	Transfer probability lymph nodes-hand	%	0. 21	(22)
<i>b</i> _{0,G}	Transfer probability lymph nodes-hook	%	0. 17	(21)
ea _{Ae}	Environmental components area			
Ae=H	Area of touch (cm ²)	cm ²	150	Assumption*
Ae=G	Area of hook (cm ²)	cm ²	1	Assumption*
Ae=K	Area of cut (cm ²)	cm ²	10	Assumption*
ala	Drobability of shane's - the legits	0/	00	Assumption (based on
СК	Provability of changing the knile	%0	90	observations)

¹⁹⁴

* Estimates for the medium size of these areas, author's best guess.

195 **2.3 Sensitivity analysis**

196 First, the baseline model was built with the parameter values indicated in Tables 1 and 2. Next,

197 two types of sensitivity analyses were performed. First, several *univariate* analyses were done to

198 assess the impact of parameters on the model outputs. To avoid unrealistic values we used a

199 range of values between each parameter baseline value (*y*) and realistic minimum and maximum

values of the parameter considered (y^-) and (y^+) respectively (Appendix D). To assess the impact of ranges of input values, above and below the baseline we applied:

202
$$f(x; y, y^-, y^+) = \begin{cases} y + (y - y^-)x, & \text{for } x < 0 \\ y - (y - y^+)x, & \text{for } x \ge 0 \end{cases}$$
 (2)

witch runs from minimum to maximum when x runs from -1 to 1 and meets the baseline when x=0. The univariate analyses were ran with 10000 iterations using 100 carcasses. Based on the *univariate* results, nine scenarios were submitted to *multivariate* analyses (Appendix E) and simulated with 10000 iterations using 500 carcasses.

207 **2.4 Data sources**

The data on the carcass surface contamination were obtained from da Silva et al.⁽²⁰⁾ These 208 authors collected carcass surface swabs in three Brazilian commercial slaughterhouses. Data 209 210 regarding the lymph nodes prevalence where obtained from 12 cohorts representing finishing herds located in the state of Santa Catarina, Brazil.⁽¹¹⁾ These herds belong to an integrated system 211 responsible for approximately 7% of all Brazilian pork production in 2007. Manipulation data 212 213 were observed during two weeks in March 2015 in a large Brazilian pig slaughterhouse dedicated to exportation. 778 inspection procedures were counted during this period, of which 214 290 in the inspection point "CARCASS". The numbers of manipulations were recorded in a 215 database. Although no data regarding transfer probability in slaughterhouse environment could 216 be found, results from Kim et al.⁽²¹⁾ and Hong and Bahk⁽²²⁾, provide transfer probabilities 217 between hands and pork and knife and pork, respectively. We have not found suitable 218 concentration data for Salmonella in lymph nodes, the values used were based on estimates of 219 the authors. 220

221 **3. RESULTS**

222 **3.1 Baseline and distributions**

223 In the baseline model, the mean of the mean concentrations (μ) and the mean prevalence were determined for two independent simulations with 500 and 100 consecutive carcasses, over 10000 224 iterations. As the mean concentration is the mean of logs, only contaminated carcasses are 225 included in the calculations. The results are summarized and given in Table III. The mean 226 concentration on inspected areas of the contaminated carcass surfaces is decreasing after 227 inspection procedures, from 1 to -0.87 logCFU/cm². Standard deviations of the means decrease 228 as well, from 0.65 to 0.2. The reason is that many more carcass surfaces are getting contaminated 229 by cross-contamination, resulting in a large number of carcasses (i. e. a prevalence difference of 230 231 94.6 percentage points) contaminated with lower concentrations. Consequently the variability is decreasing (See Fig. 2). 232

Table III. Outputs for inspected areas of carcass surface and lymph nodes before and after inspection procedures in model simulations with 500 and 100 carcasses. The mean (μ) and standard deviation (σ) are those of the mean values for the contaminated inspected carcass areas among the 500 or 100 simulated carcasses; the prevalence is the mean prevalence after 10000 iterations

	500 carcasses			10	00 Carcasses	
	μ (σ) logCFU/cm ²	Prevalence % [#]	CFU in the system##	$\mu \left(\sigma \right) logCFU/cm^{2}$	Prevalence %#	CFU in the system ^{##}
Surface before	1 (0.65)	1.2	$2.6*10^5$	1 (1.22)	1.2	$4.2*10^{4}$
Surface after	-0.87 (0.2)	95.8	3.7*10 ⁵	-1.6 (0.47)	92	5.2*10 ⁴
LN before	3.17 (0.13)	22.2	2.3*10 ⁶	3.17 (0.3)	23.8	4.6*10 ⁵
LN after	0.08 (0.2)	96.7	2.1*10 ⁶	-0.41 (0.51)	93.9	4.1*10 ⁵

LN=Lymph node; [#] The prevalence refers to at least one cell on inspected area; ^{##} arithmetic
mean (over 10000 iterations) of the total number of *Sallmonella* on carcasses surface and lymph
nodes (i. e. whole carcass) in all simulated areas per iteration.

241



Fig. 2. Distribution of mean carcasses surface contamination (logCFU/cm²) considering only
contaminated carcasses before and after inspection in two separated simulations with 100 (a) and
500 (b) carcasses.

246

242

For the lymph nodes the effects of the inspection procedures on the mean and prevalence are similar to those observed in the carcass surface (i.e. decrease and increase respectively), but the standard deviation increases after inspection. Although the results suggest a reduction in mean carcass surface contamination, after inspection, the sum of the numbers of *Salmonella* ("CFU in the system") is increasing on the carcass surface and decreasing in the lymph nodes. The reason

is that the geometric mean (mean of logCFU), which only can includes values larger than zero
(i.e. contaminated carcasses), should not be interpreted as an arithmetic mean. As the model does
not assume any growth, the only sources of contamination are the carcasses entering into the
slaughterhouse and therefore the results indicate a flow of contamination from the lymphatic
tissue to the surface by inspection procedures.

Results differ depending on the number of simulated carcasses. With a lower number of 257 carcasses, the variation in mean concentrations sampled from the zero inflated Poisson 258 Lognormal is larger. For example, with the prevalence 1.2%, the probability that the 259 concentrations in all 100 carcass surfaces are zero is $(1-0.012)^{100} = 30\%$, for 500 carcass surfaces 260 it is $(1-0.012)^{500} = 0.24\%$ (compare Fig 2a and 2b). The peaks in figure 2a before inspection 261 reflect the sampling of 1 and 2 positive carcasses, with the variability in concentrations around it. 262 Differences between distributions are smaller when considered after inspection, because more 263 carcass surfaces are contaminated. Hence, the number of carcasses used in the analysis is 264 relevant and the number of carcasses used to run the model should be realistic. Still, very large 265 numbers of carcasses slow down the calculations considerably. 266

267 **3.2 Univariate sensitivity analyses**

Of the 23 parameters analyzed (see appendix D) seven had a significant impact on the output mean of the means (μ) and four on the mean prevalence. Here, the impact is considered significant if the mean output values in the sensitivity analysis fall out of the range correspondent to 2.5th and 97.5th percentiles of the distribution describing the variability between model iterations in the mean μ from -2.57 to -0.83 logCFU/cm² and prevalence 81-98%. As shown in Fig. 3 the mean concentration of carcass surface contamination before inspection [$C_{i,(S-1)}$] and its standard deviation had an important effect on the surface contamination after inspection

procedures. The mean after inspection increased from -1.6 to 9.07 logCFU/cm² on the inspection 275 area when the load of Salmonella on carcass surfaces before inspection approaches the maximum 276 value $(2 \log_{10} CFU/cm^2 \text{ compared to } -5.4 \text{ in the baseline})$. The same effect cannot be seen when 277 mean and standard deviation are decreased below the baseline. Changes in lymph nodes 278 contamination, by changing the maximum value of the triangular distribution used in $[O_{i,(S-1)}]$ 279 from 100 to 1000 CFU/cm², increased the mean to 0.54 logCFU/cm², and decreases it to -3.6 280 logCFU/cm² when the parameter is reduced to 10 CFU/cm². Also the prevalence of animals 281 carrying Salmonella in lymph nodes had an important effect by reducing the mean contamination 282 to -4.3 logCFU/cm² and increasing it to -0.7 logCFU/cm² compared to the baseline (-1.6 283 logCFU/cm²). 284



285

Fig. 3. Mean of the means (μ) logCFU/cm² in in function of different *x* values regarding the variables mean [$C_{i,(S-1)}$], standard deviation of [$C_{i,(S-1)}$], Prev O_{i,(S-1)} and maximum [O_{i, (S-1)}].

288

Fig. 4 shows how changes in transfer probabilities affect the mean contamination on the carcass surface. If the transfer probability from hands to lymph nodes $(a_{H,O})$ decreases to 0% using x = -1, the lack of bacterial transfer from hands to lymph nodes leads to an increase of the amount on the hands and a subsequent increase of transfer to the carcass surface, leading to a small increase of the mean to approximately -1.45 logCFU/cm². But when the same parameter is increased, the mean decreases because the cells transferred to the lymph nodes can no longer be transferred to the carcass surface.



Fig. 4. Mean of the means (μ) logCFU/cm² as a function of different *x* values regarding the parameters ($b_{0,H}$), ($a_{H,O}$) and ($a_{H,C}$).

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Both transfer probabilities from the lymph node to hand $(b_{O,H})$ and to carcass by the hand 300 $(a_{H,C})$, show similar results below the baseline, but $b_{O,H}$ keeps increasing the mean until x 301 approaches 1 ($b_{0,H}$ = 100%). On the other hand, $a_{H,C}$ has a peak when x is close to 0.05. There is 302 a peak because, at some point, the transfer from hand to carcass gets so large that the 303 concentration on the hands gets too low. Once a large number of bacteria are transferred to the 304 first carcasses only a few bacteria are transferred to the subsequent carcasses, reducing the mean 305 concentration without relevant effects on prevalence (Fig. 5). As the $a_{H,C}$ keeps increasing, 306 bacteria get even more concentrated on the first carcasses after hands contamination, reducing 307

- 308 prevalence compared to the situation with a lower $a_{H,C}$ (Fig. 5). As the mean log can be
- 309 calculated only for contaminated carcasses (i. e. one or more CFU), the reduction of
- 310 contaminated carcasses leads to increases of the mean ($\log CFU/cm^2$) when the x increases for the



311 variables $a_{H,C}$ and $a_{H,O}$.

Fig. 5. Prevalence of carcass surface contamination as a function of different *x* values regarding the variable Prev $O_{,i}(S-1)$ and parameters $(b_{0,H})$, $(a_{H,0})$ and $(a_{H,C})$.

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Fig. 5 shows the effects of tested parameters on prevalence after inspection. The effect of reduction of $PrevO_{i,(S-1)}$, $b_{O,H}$ and $a_{H,C}$ to zero (minimum values, when x=-1) leads to a reduction of the surface prevalence to approximately 40%, 29% and 12% respectively, whereas reductions in $a_{H,O}$ do not seem to affect the surface prevalence. When the values of the transfer parameters $a_{H,O}$ and $a_{H,C}$ are increased, a reduction of the prevalence is observed. The reduction in the number of positive carcasses leads to an increase of the mean log surface contamination as observed in Fig. 4, as this can be calculated for positive carcasses only.

323 **3.3 Multivariate sensitivity analyses**

Table IV shows the mean of the means (μ) logCFU/cm², its standard deviation (σ) and mean prevalence on carcass surface 'before' and 'after' inspection in the multivariate sensitivity analyses. The first scenario shows the baseline for comparison proposes. The second and third scenarios present a stress analysis to verify the model performance. As expected, when transfer probabilities are set to zero, the outputs 'before' and 'after' were the same. Also, the absence of sources of contamination results in a completely uncontaminated scenario after inspection, meeting the null contamination set by the parameters.

Table IV. Scenarios used in multivariate analyses to test the effect of different variables

combination on mean of the means (μ) logCFU/cm² its standard deviation (σ) and mean

333	prevalence of	contaminated	carcass	surface	before	and	after	inspection	
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	μ (σ) log	gCFU/cm ²	Prevalen	nce %
Scenario	before	after	before	after
Baseline	1 (0.65)	-0.87 (0.2)	1.2	95.8
No transfer	1 (0.65)	1 (0.65)	1.2	1.2
No contamination (S-1)	-	-	0	0
Only carcass (S-1)	1 (0.65)	-5.08 (0.89)	1.2	43.9
Only LN (S-1)	-	-0.93 (0.18)	0	95.7
Hand influence high mean on $log_{10}[C_{i,(S-1)}]$	1.8 (0.28)	0.64 (0.82)	11.3	96.7
Hand influence high standard deviation	2.6 (0.52)	0.12 (1.56)	6.2	96.4

Hand Influence high $[O_{i,(S-1)}]$	1 (0.65)	1.31 (0.18)	1.2	97.3
Hand Influence high $PrevO_{i,(S-1)}$	1 (1.65)	0.5 (0.11)	1.2	97.4

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335 When only the carcass surface was included as the source of *Salmonella*, an important difference could be found as both the level of surface contamination and prevalence after inspection were 336 drastically reduced compared to the baseline. The influence of high transfer probability involving 337 the hands and carcass, by increasing the parameters $a_{H,C}$, $b_{C,H}$ (appendix E) tested together with a 338 higher initial concentration on contaminated carcass surfaces ($log_{10}[C_{i,(s-1)}] = -3 \log_{10} \text{CFU/cm}^2$), 339 increased the mean from -1.6 to $0.64 \log CFU/cm^2$ and the mean prevalence to 96.7%. The 340 influence of high transfer probability involving the hands and carcass was also tested with an 341 342 increase of variability of contamination on carcass surface. It resulted in an increase of carcass 343 surface contamination because surface contamination (S-1) is entered in the model as \log_{10} , so increases in variability affect the expected value, as the arithmetic mean of C_{iS} equals 344 $10^{(\mu+(\frac{1}{2}\log(10)\sigma^2))}$ and the transfer of bacteria acts as a factor of quantity and not of the log-345 quantity. 346

Also, the influence of high transfer probability involving hands and carcass was tested with a
higher concentration of *Salmonella* in lymph nodes (mean=337 CFU/cm²) and a high frequency
of animals carrying *Salmonella* in lymph nodes (100%). The increase in lymph nodes
contamination O_i, (S-1) had an important effect on the mean, changing it from -0.88 to 1.31
logCFU/cm². Changings in Prev O_i, (S-1) also increased the surface contamination and prevalence
after inspection. The mean on surface contamination increased to 0.55 logCFU/cm² and the
prevalence to 97.4%.

354 4. DISCUSSION

We developed a generic mechanistic model to assess the effect of cross-contamination during pig 355 carcass inspection, which can be applied to different hazards for different inspection practices. 356 Its performance has been studied for one inspection step, using a Brazilian data set on 357 Salmonella contamination and some parameters assumptions. The results allow us to draw 358 conclusions on the potential impact of the cross-contamination during meat inspection, but are 359 360 not necessarily considered representative for the impact of the whole inspection process of pig carcasses and the related policies in Brazil, since it deals with only one point of inspection. To do 361 so, all the three points in Brazilian inspection of pig carcasses should be included and data about 362 the contamination in lymph nodes should be also used as an input. Corbellini et al.⁽²³⁾ have 363 reported the importance of variability between different days and slaughterhouses on Salmonella 364 contamination in Brazil and this information is essential for a realistic assessment of the impact 365 366 of meat inspections practices in the country.

367 With the inputs used, the model showed that the meat inspection leads to a redistribution of Salmonella over the carcasses, which implies that many more carcasses become contaminated, 368 369 but with (very) low numbers of bacteria. In terms of prevalence and concentrations we found an 370 increase in the surface contamination prevalence with more than 90 percentage points through the inspection process and, due to the increase in the number of contaminated carcasses, a 371 372 decrease in the mean of the mean log concentrations in contaminated carcasses. The cutting of 373 the lymph nodes during inspection plays an important role, as it adds Salmonella to the carcass surface areas that were not present on carcass surfaces before inspection. Overall, the model 374 shows that the conduction of meat inspection can lead to a spread of Salmonella from the 375 lymphatic tissue to carcass surface, decreasing the differences between the surface contamination 376

377 of different carcasses.

Note that the baseline depicts a scenario of high lymph node contamination, and although we 378 have no data about lymph nodes contamination, prevalence studies have shown that this is not 379 always realistic.^(24–26) The phenomena described here meet results from previous research on the 380 effect of carcass manipulation on carcass surface contamination by Salmonella, where the 381 importance of lymphatic tissue manipulation has been observed in herds with a high number of 382 pigs harboring the bacteria in lymph nodes.^(17,24) If the model would be applied to obtain realistic 383 estimates, the user should adjust parameter values and distributions to their observations. For 384 example, variation in the prevalence of contaminated pigs or contaminated lymph nodes entering 385 in slaughterhouse can be found as a function of season, slaughterhouse and slaughter day.⁽²⁷⁻²⁹⁾ 386 387 In the sensitivity analyses, equation (2), used to standardize the domain, can result in a sudden

changing on the value of the parameters (i.e. f(x)) when x=0 (e.g. Fig 4.). It occurs because the 388 derivative f'(x) is $(y - y^{-})$ for x<0 and $(y - y^{+})$ for x≥0, so when $(y - y^{-}) \neq (y - y^{+})$ the 389 equation has two different slopes below and above the baseline x = 0. As the distances in the 390 image (i.e. $\Delta f(x)$) are not the same, this can give the impression that the effect is stronger on one 391 side than the other (Fig. 4 and 5). As an example when the bacterial transfer baseline is 0.17%, 392 (i. e. far from the 50%, center of this domain) with minimum and maximum values as 0 and 1 393 respectively, applying equation (2) we obtain a transfer value of approximately 11% when x=0.1394 395 and 0.1% when x=-0.1.

The multivariate sensitivity analysis (table IV) showed that, when the lymph nodes were considered to be uncontaminated (i.e only the carcass surface was a source of contamination), the surface contamination after inspection was much lower than in the baseline. Also, the mean (SD) logCFU/cm² decreased to -5 (0.89) and in such a scenario a large number of positive carcasses would be below the limit of detection (-4 logCFU/cm²)⁽²⁰⁾, so the observable prevalence would
only be 5% instead of 44%. On the other hand, when only the lymph nodes are considered as
source of bacteria, the results were kept similar to the baseline, indicating that the effect of
lymph nodes inspection dominated the surface carcass contamination.

When only considering the prevalence, the results obtained here may seem to be alarming and 404 unrealistic, because the increase in more than 90 percentage points is very large and it is not 405 observed in prevalence studies, which give values like 24%⁽¹¹⁾ and 14%.⁽²⁰⁾ Although these 406 prevalence results were obtained in Brazilian slaughterhouses before chilling, no inactivation 407 step is used in Brazil between the carcass inspection and the chilling. A reason that observed 408 409 prevalences are so much lower than predicted by the model may be the localization of the contaminating bacteria, which is restricted to areas manipulated by the inspection workers. These 410 411 may not correspond with areas sampled when these prevalence studies were performed. According to Jongenburger et al.⁽³⁰⁾ batches with localized bacterial concentration reduce the 412 observed prevalence with a factor *l*, derived from the relative size of the contaminated areas 413 compared to the whole surface (see Appendix F). Another issue could be related with the 414 difference between the measured prevalence (observed frequency of carcasses positive for 415 416 Salmonella in the microbial test) and the modeled prevalence, which refers to the true 417 prevalence, that is carcasses with one or more CFU. Although this difference must to be taken into account, in our simulations the mean (SD) contamination on carcass surfaces after 418 inspection was -0.8 (0.2) logCFU/cm² (Figure 2) which assuming a normal distribution, means 419 420 that none carcasses will have a level of contamination below the limit of detection (LOD=-4 logCFU/cm²)⁽²⁰⁾ (calculations not shown). Hence, it is expected that the modeled and measured 421 prevalences are expected to be the same in our simulations. 422

423 The carcass inspection is one of the many activities in the whole pork production and the model does not allow us to access the impact of the inspection procedures when compared to more 424 extensive dressing activities. In this sense studies as conducted by Swart et al. (31) describing 425 Salmonella concentrations at different stages of the slaughterhouse process should be conducted. 426 Also no direct conclusion can be drawn regarding the impact of inspection procedures on the 427 428 number of human salmonellosis cases. A quantitative microbial risk assessment (QMRA) could help to answer such question, since the outputs of this model can be applied to assess the impact 429 of the cross-contamination on human exposure.⁽³²⁾ 430

The present model is a tool to account for cross-contamination during the carcass inspection. The 431 432 purpose of the model is to capture the essential dynamics, and therefore the right balance between reality/complexity and simplicity is important.⁽³³⁾ Simplifying assumptions about the 433 inspection process are for example that we choose to give an equal probability of handling the 434 435 carcass surface and lymph nodes and that the inspection workers are equally capable to run the inspections and treat each carcass randomly. Only knife, hand and hook are modelled, because 436 procedures regarding the use of these three components are more standardized and easier to 437 quantify. In general, the effect of direct contact (hand, knife) is more important to carcass 438 contamination than other potential sources of contamination.^(34,35) 439

440 Also, in this model, only carcass surface and lymph nodes are considered as sources of

441 contamination. Regarding the meat inspection procedures, this can be considered a realistic

442 approach because viscera, like intestines, are cut and manipulated, usually, in another step of the

slaughter process⁽²⁾. Although some adaptations of the model may be necessary, the model is

444 generic enough to deal with different hazards and inspection processes.

445 Several authors have reported different approaches to deal with cross-contamination and transfer

446 in food products⁽³⁶⁻⁴¹⁾, and a particularly interesting approach is proposed by Smid *et al.*⁽⁴²⁾,
447 taking into account the uncertainty generated in transfer experiments. Here, we preferred to use

a binomial process, assuming a mechanistic approach regarding the transfer of cells⁽³⁸⁾, but the
model can be updated in order to consider new evidence about the transfer of different hazards in
pork.

The model applied to *Salmonella* has shown that the manipulation parameters and the initial contamination of the carcass surfaces and the lymph nodes are the most important for the surface contamination after inspection. Although some studies have reported bacterial quantification on carcass surfaces^(43,44), these studies are scarce and they do not always account for quantification immediately before the inspection point. To our knowledge, no data are available on *Salmonella* concentrations in the lymph nodes of pigs, whereas the model shows that this information is essential to assess the impact of cross-contamination during meat inspection.

458 Although the traditional meat inspection procedures aim to protect human health, our results 459 show that the cutting and handling of the carcasses and organs during inspection may also have the opposite effect. According to Hill et al.⁽¹⁴⁾ modernization of meat inspection towards to 460 461 visual-only approach does not seem to be a threat to public health. However, these authors also 462 identify a lack of knowledge regarding cross-contamination during the traditional pig carcass inspection and indicate that this information is needed. Furthermore, Ravel et al.⁽⁴⁵⁾ discuss that 463 during the traditional inspection system, cross-contamination can occur between the lymph nodes 464 and other parts of the same carcass or even between consecutive carcasses, but the cross-465 contamination level has not been described so far.⁽⁴⁵⁾ 466

467 The results, also, highlight the importance of bacterial transfer between carcass surface, hands468 and lymph nodes if a high number of animals carrying *Salmonella* in lymph nodes are expected.

469 Furthermore it sheds some light on the potential inadequacy of classic pig carcass inspection and

470 therefore it can be considered as a tool to quantify the effects of cross-contamination and answer

471 questions about the modernization of the classic carcass inspection system for the

472 implementation of risk-based approaches in meat-inspection.⁽⁴⁶⁾

473 5. CONCLUSIONS

In the classic veterinary meat inspection of pig carcasses, the effect of cross-contamination may not be negligible. The model presented in this paper offers a tool to quantify these effects. Our analyses how that, especially when animals that carry high concentrations of *Salmonella* in lymph nodes are entering the slaughterhouse, bacteria will be spread to many previously uncontaminated carcasses. The model had not been validated, so far, and this step is important to figure out the suitability of this model in describe cross-contamination during classic inspection procedures and support the modernization of inspection of pig carcass.

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633

Appendix A. Values used in the empirical distribution that is applied to sample the number of
interactions between carcass/organ with hands, knife and hook using the index Ae= (H, K,
G). Probabilities in columns sum up to 1

Counts	J _{i,Ac,H}	J _{i,Ac,G}	J _{i,Ac,K}
0	0.009	0	0
1	0.009	0.017	0.008
2	0.018	0.483	0.051

3	0.027	0.433	0.034
4	0.152	0.033	0.144
5	0.116	0.017	0.161
6	0.143	0.017	0.169
7	0.134	0	0.161
8	0.161	0	0.059
9	0.045	0	0.068
10	0.098	0	0.051
11	0.009	0	0.000
12	0.018	0	0.034
13	0	0	0
14	0.018	0	0.017
15	0.009	0	0.025
16	0.018	0	0
17	0	0	0.008
18	0	0	0

19	0.009	0	0.008
20	0.009	0	0
21	0	0	0
23	0	0	0

637

638 Appendix B

It is assumed that the concentrations on the carcass surfaces before inspection $[C_{i (S-1)}]$ can be described by a zero inflated lognormal distribution with prevalence *p* and thus a probability of an uncontaminated carcasses 1-*p*. Let $u = Log_{10}(x)$ and the probability density function f(u) for concentration $[C_{i (S-1)}]$, can be defined by:

$$643 f(u) = \begin{cases} p * \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{u-\mu}{\sigma}\right)^2} & u > LOQ \\ p * \int_{LOD}^{LOQ} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{u-\mu}{\sigma}\right)^2} du & LOD < u < LOQ \\ (1-p) + \left(p * \int_0^{LOD} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{u-\mu}{\sigma}\right)^2} du\right) & u < LOD \end{cases}$$

644 Where *LOQ* and *LOD* are limits of quantification and detection of *Salmonella*, respectively. 645 Applying this probability density function to each sample x_i for carcass surfaces *i* in the study 646 from da Silva *et al.*⁽²⁰⁾, the parameters (μ , σ and *p*) were assessed by the Maximum Log 647 likelihood estimation using the Solver function in Excel, i.e. maximizing

648 $\sum_{i=1}^{n} \operatorname{Log} f(x_i; \mu, \sigma, p)$

649 Where, i is the carcass number and n is the last carcass sampled.

650

651 Appendix C

- 652 Consider a system with two compartments *E* and *C* in which a sequence of events (in this case
- 653 the manipulations: cuts, touches or hookings) is defined by the index J, with $J \in N$ indicating the
- number of manipulations. Also consider no loss and no increase of units, only transfer is
- 655 modeled with a given initial condition:

656 $E_0 = 100 \text{ CFU}$

- The transfer probability from the compartment E to the C is called q, and it is assumed to be constant through the sequential steps J.
- 659 So, the amount on the *E* compartment after *J* steps is the amount before the step *J* minus a 660 fraction *q*:
- 661 $E_I = E_{I-1} (E_{I-1} * q)$
- 662 so

663 $E_J = E_{J-1} * (1 - q)$

664 This applies to *J* subsequent steps, so.

665 $\begin{cases} E_1 = E_0 * (1 - q) \\ E_2 = E_0 * (1 - q) * (1 - q) \\ E_3 = E_0 * (1 - q) * (1 - q) * (1 - q) ... \\ E_J = E_0 * (1 - q)^J \end{cases}$

666 In the stochastic model this is interpreted as a Binomial process, with N = E_0 and p = $(1-q)^J$. For 667 example in equation (1), considering the term $(1 - b_{C,Ae})^{J_{i,C,Ae}}$. It was implemented by sampling

668	values from a Binomial distribution as: ~ Binomial $(C_{i,(S-1),Ae}; (1 - b_{C,Ae})^{J_{i,C,Ae}})$ describing the
669	number of cells not transferred to from carcass to environment after J interactions.
670	Using the previous equation (Appendix C) we can derive the amount of Salmonella in
671	compartment C . Considering the fact that only transfer between two compartments is possible
672	(no die off or growth), the amount in C is the difference between the E_0 and the E_J . Applying it
673	in the previous equations:
674	$C_J = E_0 - (E_0 * (1 - q)^J)$
675	SO
676	$C_J = E_0 * (1 - (1 - q)^J)$
677	
678	
679	

Appendix D. Parameters, baseline, minimum and maximum values used during the univariate
sensitivity analyses. See tables I and II for an explanation of the parameter symbols and indices.

Parameters	Baseline (y)	Maximum (y ⁺)	Minimum (y ⁻)	Unit
$a_{K,C}$	0.17	100	0	%
$a_{K,O}$	0.17	100	0	%
<i>b_{C,K}</i>	0.17	100	0	%
<i>b</i> _{0,K}	0.17	100	0	%
$a_{H,C}$	0.21	100	0	%

<i>a</i> _{<i>H,O</i>}	0.21	100	0	%
b _{C,H}	3.1	100	0	%
<i>b</i> _{0,H}	3.1	100	0	%
a _{G,C}	0.17	100	0	%
$a_{G,O}$	0.17	100	0	%
b _{C,G}	0.17	100	0	%
<i>b</i> _{0,G}	0.17	100	0	%
d_{C}	0	100	0	%
d_O	0	100	0	%
ka	10	15	5	cm^2
ha	150	200	100	cm^2
ga	1	3	0.5	cm^2
$\mu \log_{10}[C_{i(S-1)}]$	-5.2	2	-15	$log_{10}CFU/cm^2$
$\sigma \log_{10}[C_{i(S-1)}]$	2.2	4	0	log10CFU/cm ²
$PrevO_{i(S-1)}$	14.1	100	0	%
$PrevC_{i(S-1)}$	100	100	0	%
$[O_{i(S-1)}]$ Min	0.01	0.1	0.0001	CFU/cm ²
$[O_{i(S-1)}]\mathrm{MP}$	1	100	0.01	CFU/cm ²
$[0_{i(S-1)}]$ Max	100	1000	10	CFU/cm ²

 μ = mean; σ =standard deviation; Min= minimum; MP=most likely; Max=maximum values in

683 triangular distribution

685 Appendix E. Parameter values for scenarios used in multivariate sensitivity analyses. The

	Scenarios						
				Hand influence	Hand influence		
				high	high	Hand Influence	Hand Influence high
Parameters	Baseline	Only carcass (S-1)	Only LN (S-1)	$log_{10}[C_{i(S-1)}]$	$\sigma \log_{10}[C_{i(S-1)}]$	high $[O_{i(S-1)}]$	$PrevO_{i(S-1)}$
PrevC _{i (S-1)}	100%	100%	0%	100%	100%	100%	100%
PrevO _{i (S-1)}	14.4%	0%	14.1%	14.1%	14.1%	14.1%	100%
a _{H,C}	0.21%	0.21%	0.21%	10%	10%	10%	10%
$b_{C,H}$	3.1%	3.1%	3.1%	10%	10%	10%	10%
$\mu \log_{10}[C_{i(S-1)}]$	-5.4	-5.4	0	-3	-5.4	-5.4	-5.4
$\sigma \log_{10}[C_{i(S-1)}]$	2.2	2.2	0	2.2	3.3	2.2	2.2
$[O_{i(S-1)}]$ Min	0.01	0	0.01	0.01	0.01	0.1	0.01
$[O_{i(S-1)}]\mathrm{MP}$	1	0	1	1	1	10	1
$[O_{i(S-1)}]$ Max	100	0	100	100	100	1000	100

686 scenarios: no transfer and no contamination were omitted

687

 μ = mean; σ =standard deviation; LN=Lymph nodes; Min=minimum; MP= most probable; Max=

688 maximum values in triangular distribution

689 Appendix F

Adopting Jongenburger's approach⁽³⁰⁾ the observed prevalence will be a product of the factor *l* and the modeled prevalence. The value of *l* can be considered as a probability that at least one sample unit (CFU) is drawn from an inspected area: $l = P(x \ge 0)$. This probability follows as a hypergeometric distribution. Consider four swabs taken with a sponge in fixed and mutually exclusive areas of 100cm² on carcass surface (*n*=4). Next, consider the whole carcass area as 14000 cm²⁽⁴⁷⁾ represented as a rectangle composed by N = 14000/100 = 140 different *N* areas that can potentially be sampled.

According to the model the total area inspected per carcass during meat inspection is on average

- 698 700 cm² (data not shown), and the number of possible inspected areas sampled is K=700/100=7
- 699 different areas. The probability that at least one sample is drawn from the inspected areas is (1-

700 P(*x*=0)) has hypergeometric distribution according:

701
$$P(x=0) = \frac{\binom{K}{x}\binom{N-K}{n-x}}{\binom{N}{n}} = \frac{\binom{7}{0}\binom{140-7}{4-0}}{\binom{140}{4}} \approx 81\%$$

702 So, 1-P(x=0) = 19% is the *l* factor.