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# Early-Life Hepatitis E Infection in Pigs: The Importance of Maternally-Derived Antibodies



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# Abstract

Passive immunity (PI), acquired through colostrum intake, is essential for piglet protection against pathogens. Maternallyderived antibodies (MDAs) can decrease the transmission of pathogens between individuals by reducing shedding from infected animals and/or susceptibility of naïve animals. Only a limited number of studies, however, have been carried out to quantify the level of protection conferred by PI in terms of transmission. In the present study, an original modeling framework was designed to estimate parameters governing the transmission of infectious agents in the presence and absence of PI. This epidemiological model accounts for the distribution of PI duration and two different forces of infection depending on the serological status of animals after colostrum intake. A Bayesian approach (Metropolis-Hastings algorithm) was used for parameter estimation. The impact of PI on hepatitis E virus transmission in piglets was investigated using longitudinal serological data from six pig farms. A strong impact of PI was highlighted, the efficiency of transmission being on average 13 times lower in piglets with maternally-derived antibodies than in fully susceptible animals (range: 5-21). Median infection-free survival ages, based on herd-specific estimates, ranged between 8.7 and 13.8 weeks in all but one herd. Indeed, this herd exhibited a different profile with a relatively low prevalence of infected pigs (50% at slaughter age) despite the similar proportions of passively immune individuals after colostrum intake. These results suggest that the age at HEV infection is not strictly dependent upon the proportion of piglets with PI but is also linked to farm-specific husbandry (mingling of piglets after weaning) and hygiene practices. The original methodology developed here, using populationbased longitudinal serological data, was able to demonstrate the relative impact of MDAs on the transmission of infectious agents.

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# Introduction

Passive immunity (PI) is the primary protection against infections in early life for several species, including humans [1–4]. However, this protection is only partial, rarely totally preventing from infection, but slows down the transmission process among individuals and/or reduces clinical consequences whenever infections occur. Moreover, this protection is only temporary due to maternal antibodies waning and, in the absence of vaccination, individuals become fully susceptible to infection. Although vaccination could overcome this issue, several studies evidenced antagonistic effects between vaccine-induced immunity and maternally derived antibodies through inhibition of vaccine protection, which could potentially worsen the dynamics of infection [5–9]. Hence, timely vaccination in regards to passive immunity waning appears crucial to optimize vaccine efficacy.

In this context, we developed a methodological framework to analyze the main characteristics of agent-specific passive immunity, in terms of duration and protection, based on longitudinal data. Our model explicitly takes into account the distribution of passive immunity duration to define the transition rate between passively immune and susceptible states. The protective impact of passive immunity is assessed through the estimation of differential forces of infection (FOI), defined as the *per capita* rate of infection per time unit, in regards with the early life serological status (*i.e.* with or without maternally derived antibodies). A Bayesian approach, based on Monte-Carlo Markov Chains (MCMC), was used to estimate the protection conferred by maternally derived antibodies on the transmission process. Bayesian methodology is of particular interest for the estimations of parameters related to infectious disease dynamics [10–18], since such data are subject to uncertainties due, for example, to under-reporting of cases, partially observed processes, or time-aggregated data.

As passive immunity was shown to be partially protective in the context of Hepatitis E virus (HEV) infection in pigs [19], the developed methodology was applied to quantify the protection level conferred to new-born piglets by maternally derived antibodies (MDAs). Hepatitis E virus is a recognized zoonotic agent, which can cause enterically-transmitted hepatitis in humans, and for which domestic pigs are considered as the main

reservoir [20]. Understanding the impact of passive immunity on infection dynamics is therefore of pivotal importance to identify strategies to decrease the prevalence of infected animals at slaughter time. To our knowledge, few studies have focused on quantifying the protection conferred by passive immunity (PI) in terms of pathogen transmission in pigs. In 1997, Bouma et al. calculated a reproduction number of 0.2 for pseudorabies virus in PI animals (vs. 6.3 in absence of MDAs) [5]. Another study on PCV-2 transmission demonstrated a two-fold reduction in pigs with MDAs and an estimated reproduction number of 1.5 [21]. Allerson et al. recently studied the transmission potential of swine influenza virus in the presence of homologous and heterologous MDAs (obtained from vaccinated dams) with regard to the challenge strain [22]. The results showed high protection with homologous MDAs but weaker protection in piglets born to sows vaccinated with a heterologous strain. The main risk regarding hepatitis E infection in humans in industrialized countries is the introduction of viremic pigs into the food chain, which is strongly related to the age at infection. Data from a cohort study were first used to characterize MDAs (IgG) kinetics in the early life of piglets (up to 14 weeks of age) and to estimate the distribution of the duration of passive immunity. A second dataset derived from a longitudinal study in 6 Spanish pig-herds [23] was then analyzed, accounting for previous estimates of MDAs waning time, to estimate forces of infection from herd-specific HEV prevalences.

# **Materials and Methods**

#### Distribution of passive immunity duration

A non-linear mixed effect model was used to represent the decay of MDAs titer. The antibody titers A in piglets which received maternally-derived antibodies (MDAs) were assumed to decrease exponentially with age according to the equation dA/dt = -rA. The antibody titers would thus depend on the initial level of antibodies delivered at birth  $A_0$  and the antibody decay rate r. As the initial antibody titer would likely be related to the dam's serological status, dam antibody titers were consequently considered as a covariate related with the initial level of antibodies  $A_0$  in offspring. Thus, the model describing serological titer of individual i at observation time  $t_{ij}$  (with a constant residual error model) is given by:

$$A_{ii} = A_0^{(i)} \exp\left(-r_i t_{ij}\right) + a\varepsilon_{ij}$$

where  $A_0^{(i)}$  and  $r_i$  are the individual parameters and  $\varepsilon_i$  is a vector of standardized random variables. As described in [24], individual parameters were assumed to be log-normally distributed. The serological titer of the dam  $A_{Dam}^{(i)}$  one week after parturition was considered as covariate for the initial serological titer of each individual  $(A_0^{(i)})$ . The model for individual parameters is given by:

 $A_0^{(i)} = A_0^{pop} \left( A_{Dam}^{(i)} \right)^{\beta_{Dam}} e^{\eta_A^{(i)}}$ 

and

$$r_i = r_{pop} e^{\eta_r^{(i)}},$$

where *i* denotes the individual and  $r_{pop}$  the median decay rate at the population level.  $A_0^{pop}$  represents the typical initial serological titer in the population, the predicted initial serological titer for individual *i* being given by  $A_0^{pop} \left(A_{Dam}^{(i)}\right)^{\beta_{Dam}}$ .  $\eta_A^{(i)}$  and  $\eta_r^{(i)}$  are vectors of random effects assumed as independent centered Gaussian vectors with variance  $\Omega_A$  and  $\Omega_r$ , representing interindividual variability. Parameters were estimated using MLE and the SAEM algorithm for the analysis of hierarchical nonlinear mixed-effects models [25].

Individual antibody kinetics were derived from the individual empirical parameter estimates. Subjects with antibody levels below the cut-off value of the ELISA test were considered seronegative (HEV ELISA 4.0v, MP Diagnostics, Illkirch, France, [26]). Different distributions were tested to model the time to immunity waning (Weibull, lognormal and gamma distributions). Distribution parameters were estimated using the maximum likelihood method and model selection was based on the AIC values. A nonparametric bootstrap procedure was used to determine the 95% confidence intervals for the parameter estimates and thus measure their accuracy. One thousand bootstrap replicates were generated by resampling individual profiles for each dataset. For each bootstrap replicate, selected model was refitted to get an estimate of the distribution parameters. The 95% confidence interval was constructed from the 2.5<sup>th</sup> and 97.5<sup>th</sup> percentiles of distribution parameters [27].

#### Impact of passive immunity on transmission

PI is known to be a protective factor against infections in early life. However, this protection is rarely fully efficient. We developed a general model framework to represent the evolution of passive immunity with age of the animals. For this purpose, the transition from passively immune to susceptible states was considered as agedependent according to the estimated distribution described in the previous subsection. The following procedure was adopted: Let g(t)and G(t) be the probability density and cumulative probability function of this distribution, respectively. Let  $T_i$  be the interval duration between two observation times  $[a_i,a_{i+1}]$ . In the absence of infection, animals will then either be passively immune or susceptible and the transition between these two states is governed by:

$$M_{i+1} = M_0(1 - G(a_{i+1})) = M_i \frac{(1 - G(a_{i+1}))}{1 - G(a_i)}$$
(1)

and

$$S_{i+1} = S_i + (M_i - M_{i+1}) = S_i + \frac{M_i}{1 - G(a_i)} (G(a_{i+1}) - G(a_i)) \quad (2)$$

The second term in equation (2) corresponds to the proportion of individuals losing passive immunity between ages  $a_i$  and  $a_{i+1}$ .  $M_i$  (resp.  $S_i$ ) represents the proportion of passively immune (resp. susceptible) animals of age  $a_i$ .

**Infectious process.** The present study was designed to quantify the protective impact MDAs from longitudinal serological data. For this purpose, and based on equations (1 - 2), the following epidemiological model was developed to account for passive immunity duration and differential forces of infection according to the serological status of individuals at birth.

Let  $\lambda_M$  be the force of infection exerted on individuals with passive immunity and  $\lambda$  the FOI in the absence of passive immunity (Full model). The probability that an animal with maternal antibodies will escape infection between  $a_i$  and  $a_{i+1}$  is therefore expressed by  $\exp(-\lambda_M T_i)$  and the proportion of individuals remaining passively immune is given by:

$$M_{i+1} = M_i \frac{(1 - G(a_{i+1}))}{1 - G(a_i)} \exp(-\lambda_M T_i)$$
(3)

Individuals for which immunity waning occurs at time  $\tau \in [a_i, a_{i+1}]$  are susceptible at age  $a_{i+1}$  if they were not infected between  $a_i$  and  $\tau$  (*i.e.* when they were still passively immune), or between  $\tau$  and  $a_{i+1}$ , when they were susceptible. Moreover, individuals that were susceptible at age  $a_i$  remain so if they escaped infection during the time interval  $[a_i, a_{i+1}]$  with a probability of  $\exp(-\lambda T_i)$ . The proportion of susceptible individuals is therefore represented by:

$$S_{i+1} = \frac{M_i}{1 - G(a_i)} \int_0^{T_i} g(a_i + \tau) \exp(-\lambda_M \tau) \exp(-\lambda(T_i - \tau)) d\tau + S_i \exp(-\lambda T_i)$$
(4)

To clearly identify the impact of PI, we compared this model structure to a simplified version assuming that passive immunity played no role in the transmission process ( $\lambda_M = \lambda$ ). In this sub-model, hereafter referred to as the single force of infection model (SFOI model), the proportion of individuals with passive immunity and susceptible individuals are governed by:

$$M_{i+1} = M_i \frac{(1 - G(a_{i+1}))}{1 - G(a_i)} \exp(-\lambda T_i)$$
(5)

$$S_{i+1} = \left(\frac{M_i}{1 - G(a_i)} (G(a_{i+1}) - G(a_i)) + S_i\right) \exp(-\lambda T_i) \quad (6)$$

Whatever the assumption concerning the impact of passive immunity, the proportion of infected individuals  $C_i$  is given by:

$$C_i = 1 - M_i - S_i.$$
 (7)

Model (3-4-7), which considers two forces of infection, is hereafter termed the "full model", in contrast to model (5-6-7) which considers a single force of infection.

**Parameter inference.** Let *n* be the number of sampled animals from a total number of Nanimals per herd. As described by Backer *et al.* [18], the number of infected individuals is not directly observed so the true number of infected individuals of age  $a_i$  in herd *h*, and denoted by  $J_i^{(h)}$  follows the distribution:

$$f\left(J_{i}^{(h)}|N,n,k_{i}^{(h)}\right) = \frac{\left(J_{i}^{(h)}\right)\left(N-J_{i}^{(h)}\right)}{\binom{N}{n-k_{i}^{(h)}}}\frac{n+1}{N+1},$$

with  $k_i^{(h)}$  the observed prevalence at age  $a_i$  in herd h.

Considering that the proportion of infected animals at each sampling time is given by equation (7), we can calculate the herd-specific contribution to the likelihood of the observations if there were  $J^{(h)}$  infected animals and  $N-J^{(h)}$  non infected pigs in herd h:

$$L\left(\mathbf{k}^{(\mathbf{h})}, n, N, \mathbf{a}, \mathbf{T} \middle| \lambda_{M}^{(h)}, \lambda^{(h)}, \mathbf{J}^{(\mathbf{h})} \right) = \Pi_{i} f\left(J_{i}^{(h)} \middle| N, n, k_{i}^{(h)}\right) {N \choose I_{i}^{(h)}} \left(C_{i}^{(h)}\right)^{J_{i}^{(h)}} \left(1 - C_{i}^{(h)}\right)^{N - J_{i}^{(h)}}$$

The global likelihood is thus expressed as the product of the herd-specific contributions:

$$L(\mathbf{k}, n, N, \mathbf{a}, \mathbf{T} | \boldsymbol{\lambda}_{\mathbf{M}}, \boldsymbol{\lambda}, \mathbf{J}) = \Pi L(\mathbf{k}^{(\mathbf{h})}, n, N, \mathbf{a}, \mathbf{T} | \boldsymbol{\lambda}_{\mathbf{M}}, \boldsymbol{\lambda}, \mathbf{J}^{(\mathbf{h})})$$

A Bayesian framework was used to infer the following parameters: the herd-specific forces of infection  $(\lambda_M^{(h)}, \lambda^{(h)})$ , the actual number of infected individuals for each observation time  $(\mathbf{J}^{(\mathbf{h})})$  and the initial proportion of passively immune animals  $(M_0^{(h)})$ . Flat uninformative priors were assumed for all parameters but  $M_0^{(h)}$  for which a beta distribution was selected in line with the average percentage of seropositive sows in the sampled herds (59%, sd: 10; [23]). Parameters were updated using the Metropolis-Hastings algorithm with gamma proposal distributions for the forces of infection and normal proposal distributions for J for which the mean value corresponded to the last accepted value and using the variance as tuning parameter. Uniform proposal distributions, also centered on current value (Y ~  $U(M_{0:t} - \delta, M_{0:t})$  $(+ \delta)$ , Y denoting the proposal value and delta being the tuning parameter) were used to estimate  $M_0$ . Three chains were run with random initial conditions, 50000 steps per chain, a burn-in of 1000 steps and a thinning parameter of 10. Convergence was assessed by visual inspection and diagnostic tests (autocorrelation plots, Geweke and Gelman-Rubin diagnostics).

#### Application to Hepatitis E virus transmission in pigs

No data were specifically collected for the present analysis. Data were available from previous field studies which were described in references [23] and [28].

**Passive immunity duration.** Serological data, used to estimate the duration of PI, were derived from a longitudinal epidemiological study on HEV prevalence in France [28] by monitoring 120 piglets, from 3 consecutive batches in one herd known to be infected by HEV, from birth to 14 weeks of age. Ten sows were chosen at random from this herd and from each followed batch. One week post-farrowing, 4 piglets born to the selected sows were chosen at random and ear-tagged. At the same time, blood samples were collected from the sows to determine their serological status with regard to HEV infection. Blood samples were also taken from piglets at the ages of 1, 6, 10 and 14 weeks for serological analysis.

A commercial test validated for veterinary analyses (HEV ELISA 4.0v (MP Diagnostics, Illkirch, France)) was used to detect anti-HEV antibodies in pigs as described in [26]. This test is based on a double sandwich ELISA that allows the detection of all classes of immunoglobulins (IgG, IgA and IgM) regardless of the animal species. The HEV ELISA 4.0v utilizes a proprietary recombinant antigen, which is highly conserved between different HEV strains.

Table 1. Parameter estimates of HEV maternal antibodies           kinetics.						
	Parameter estimate	SE	p-value			
Ao	0.47	0.07				
r	0.4	0.02				
$\beta_{Dam}$	0.66	0.04	<10 <sup>-3</sup>			

A<sub>0</sub>: initial antibodies concentration (ELISA test Optic density (OD)).

*r*: antibody decay rate ( $week^{-1}$ ).

 $\beta_{sow}$ : sow's serological titer covariate coefficient on piglets' initial antibody titer. doi:10.1371/journal.pone.0105527.t001

Analyses were performed according to the manufacturer's instructions except that 10  $\mu$ l of sera were used. Samples were considered positive when the OD450 value of the sample was above the cut-off value (Co = mean of the negative control + 0.300).

Impact of passive immunity on transmission. The forces of infection in the presence and absence of PI were estimated from data provided by Casas et al. [23] because these data had shown evidence of early infections occurring in young piglets in the presence of residual maternal antibodies. Briefly, 6 farrow-to-finish Spanish swine herds of similar size (100-300 productive sows) and known to be infected by HEV were included in the study. In each farm, 20 piglets from 6 previously selected sows, were randomly selected and ear-tagged and serial blood samples were collected at 3, 7, 13, and 18 weeks. Blood was also collected at slaughter (25 weeks of age approximately) at bleeding. We assumed that this number of sampled animals corresponded to 10% of the total number of pigs in the batch (N = 200). In this study, the sera were tested for specific IgG and also IgM which allowed differentiation between passively derived antibodies and a post-infectious response following a recent infection (IgM) [29]. Based on individual serological results, the course of infection was determined from the cumulated number of IgM positive piglets among the 20 piglets examined at each time point.

#### Results

### Antibody decay rate

The serological investigations in sows revealed a mean HEV seroprevalence of 77% at farrowing with variations (from 70 to 90%) between batches. Seventy percent of the selected piglets were positive for anti-HEV antibodies at one week of age. The MDA titers of piglets were correlated with the titer of their dam ( $\beta_{Sow}$  = 0.66, SE = 0.04, p<0.001; Table 1). The estimated antibody decay rate was 0.4 per week (SE: 0.02, Table 1) which corresponded to a half life of 12.1 days. Individual fits of serological profiles are shown in Figure S1. The individual duration of PI, based on the individual parameter estimates, was determined by long term projection of the antibody kinetics (assuming a threshold value of 0.38 Optic Density (OD)). Different distributions were fitted to the individual durations of PI (Weibull, lognormal and gamma distributions). The mean durations of PI were similar for all distributions but the gamma distribution was selected, based on the Akaike Information Criterion (AIC). Thus, the duration of PI was fitted to a gamma distribution with a mean duration of 6.5 weeks (95% CI: [6.0; 7.2]; Figure 1). This distribution was then used to model the transition between the PI state and the fully susceptible state in the epidemiological model (Equations (3-7)).



**Figure 1. Distribution of the duration of HEV passive immunity** (Weeks). We assumed gamma distributions for the duration of PI and used the maximum likelihood method to estimate the shape and scale parameters (red line). 95% confidence bands (black lines) were obtained by the bootstrapping method (see text for details). doi:10.1371/journal.pone.0105527.q001

#### HEV Infectious process and passive immunity

Convergence of the MCMC Algorithm is shown in figure S2. A graphical representation of the posterior distribution of herd-specific prevalences, *i.e.* the proportion of infected individuals  $(\mathbf{J}^{(\mathbf{h})}/N)$ , according to the model structure is provided in Figure 2. As previously highlighted by Casas *et al.* [23], all but one herd (herd 2) exhibited similar profiles with a huge increase in prevalence between 7 and 13 weeks of age, leading to asymptotic behavior close to 1. This increase in prevalence coincided with the estimated duration of PI, suggesting an impact on transmission dynamics. Visually, the estimated prevalence values based on the full model (blue boxplots) were better adapted to reproduce this behavior than the SFOI approach (black boxplots).

The force of infection estimates, obtained by using the SFOI model, ranged between 0.019 and 0.136 week<sup>-1</sup> (Table 2). They tended to overestimate the predicted prevalence of infected animals before 7 weeks of age and led to underestimations thereafter. The full model structure was able to disentangle the transmission process in the presence and absence of PI. Both visual inspection (Figure 2) and the DIC values (Table 2) led us to conclude that the model with two different forces of infection better captured the dynamics of infection observed in field conditions. With this full model, the forces of infection exerted on fully susceptible individuals were estimated on average 1.8 times higher than with the SFOI model (range: 1.3 - 2.0; Table 2). In contrast, transmission in piglets with PI was found, on average, 13 times lower than in fully susceptible animals, with a relatively large variability between herds (range: 5-21).

The impact of PI on time to infection was assessed by modeling the infection-free survival functions in piglets with MDAs (Figure 3; full line) and fully susceptible animals (Figure 3; dashed lines) for each herd. In the herds with high prevalence (all but herd 2), fully susceptible individuals escaped infection according to exponential distributions with median infection-free survival ages ranging between 2.7 and 7.6 weeks. In contrast, in passively immune animals, the combined gamma-distributed duration of PI



**Figure 2. Estimated age-specific HEV seroprevalences.** The distributions of estimated seroprevalences are shown according to model structure (SFOI model: black boxplots; Full model: blue boxplots). Observed prevalence data, derived from the longitudinal data in six pig herds in Spain [23], are represented by red dots. doi:10.1371/journal.pone.0105527.q002

and the differential forces of infection produced sigmoid survival curves with a slight probability of infection in the presence of MDAs (before 7 weeks of age), followed by an increased probability of infection thereafter. In consequence, the infection-free survival age in piglets which had received maternal antibodies was much higher than in fully susceptible animals, ranging between 8.7 and 13.8 weeks. In herd 2, the infection-free survival curves were similar in shape for both fully susceptible and passively immune animals (median: 27.5, 95% CI: [17.0; 53.4] and 32.0 weeks, 95% CI: [22.3; 51.5], respectively), but are not shown in Figure 3 since they extended beyond the normal slaughter age.

# Discussion

Piglets do not receive any maternal antibodies during gestation due to the epitheliochorial placentation in sows [30]. Neonate piglets are thus susceptible to the majority of infectious agents to which they are exposed in early life. It is therefore essential to provide new-born piglets with MDAs via colostrum to protect them against viral and bacterial infections before their own immune system develops [31]. However, the protection conferred by PI has been shown to be partial for several pathogens, reducing the clinical expression and/or within–host viral replication in infected piglets [21,32–35]. To our knowledge, the extent to which MDAs affect the transmission process has rarely been quantified to date [5,21,22]. We present an original approach to investigate and model the impact of PI on the transmission of an infectious agent in pigs based on batch-level prevalence data using two nested modeling structures: (i) differential forces of infection dependent on the serological status of piglets following colostrum intake (full model); and (ii) an equal force of infection exerted on fully susceptible and passively immune animals (SFOI model). This modeling approach, which combines a realistic distribution of the duration of PI and differential forces of infection, with and without maternal antibodies, provides a better understanding of early-life transmission of infectious agents. This approach was used to analyze serological data from a longitudinal study on hepatitis E prevalence in 6 pig herds.

The model only provides a simplified representation of the actual process of hepatitis E virus transmission with a main shortcut relying on the assumption of constant forces of infection. Indeed, the role of the environment on the course of infection was highlighted in a previous study, indicating that the force of infection represents a combination of transmission by direct contact between animals and ingestion of viral particles present in the environment [36]. Moreover, it is likely that the susceptibility to infectious agent in passively immune individuals varies as the level of maternally derived antibodies decreases with age. Thus,



Figure 3. HEV Infection-free survival function. Median (red curves) and 95% credibility interval (black curves) of age specific probability to escape infection according to piglet's serological status after colostrum intake (dashed lines: fully susceptible animal; full lines: pig with HEV passive immunity).

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	Herd N°	λ <sup>1</sup>	$\lambda_{M}^{2}$	DIC	
SFOI model <sup>3</sup>	1	0.105 (0.079;0.140)	-	90	
	2	0.019 (0.012;0.028)	-	81	
	3	0.099 (0.074;0.129)	-	101	
	4	0.055 (0.038;0.077)	-	94	
	5	0.106 (0.079;0.140)	-	94	
	6	0.136 (0.102;0.181)	-	88	
Full model <sup>4</sup>	1	0.181 (0.109;0.290)	0.017 (9E-04;0.082)	78	
	2	0.025 (0.015;0.042)	0.006 (5E-04;0.032)	78	
	3	0.190 (0.122;0.301)	0.009 (7E-04;0.048)	83	
	4	0.092 (0.056;0.140)	0.008 (7E-04;0.042)	84	
	5	0.215 (0.129;0.351)	0.010 (5E-04;0.054)	75	
	6	0.274 (0.148;0.615)	0.023 (0.001;0.112)	75	

 $^{1}\lambda$ : Posterior median estimate with 95% credibility interval of the force of infection without PI (week $^{-1}$ ).

 $^{2}\lambda_{M}$ : Posterior median estimate with 95% credibility interval of the force of infection in the presence of PI (week<sup>-1</sup>).

<sup>3</sup>SFOI: Single force of infection model ( $\lambda_M = \lambda$ ).

<sup>4</sup>Full model: differential forces of infection in the presence and absence of MDAs ( $\lambda_M \neq \lambda$ ).

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the estimated FOIs should be considered as the average infection pressure exerted on individuals to assess and quantify the protection conferred by PI. As all pigs within a specific herd were reared under the same conditions, the transmission process was summarized by forces of infection, by considering that a similar infection pressure was exerted on each individual independently of their serological status after colostrum intake. In consequence, the protection induced by PI is clearly related to a decrease in susceptibility.

The aim of the present study was to quantify the protective impact conferred by PI against Hepatitis E infection. Data from a longitudinal survey in France were used to characterize the kinetics of maternally-derived antibodies in 120 piglets. The serological titers of piglets at 1 week of age were highly correlated with those of the dam, thus supporting previous observations [37,38]. The duration of PI was modeled by a gamma distribution with a mean duration of 45.6 days [41.4; 50.4]. Although the duration was slightly shorter than in several epidemiological studies (8-9 weeks of age) [19,37,38], because of the gamma distribution, the probability of remaining passively immune fell below 0.1 by 60 days of age, as in these previous studies. In a recent study, the transmission parameters for hepatitis E virus in pigs were estimated from field data and the effectiveness of different hypothetical control strategies (early life vaccination and delayed vaccination around 10 weeks of age) was evaluated [18]. The author concluded that vaccination should lead to a 75% reduction of the main parameters governing the transmission process (host susceptibility, transmission rate or infectious period duration) to significantly decrease the expected prevalence at slaughter age. However, several studies showed interference between the maternally-derived antibodies and the immune response after vaccination [5,6,32,39,40]. Such interactions could have dramatic consequences on transmission. For example, the study by Bouma et al. revealed an increased reproduction number in piglets vaccinated against pseudorabies virus in the presence of PI as compared to vaccinated piglets without MDAs [5]. Further work should therefore be carried out to assess possible interactions between MDAs and future candidate vaccines before considering their usefulness in HEV control. However, a delayed vaccination strategy, as studied in [18], would certainly avoid such interactions as only 10% of piglets who received MDAs remained passively immune at 10 weeks of age.

Based on individual serological data for 120 pigs from 6 herds published by Casas et al. [23], we estimated the herd specific cumulative incidence by assuming that the first-positive IgM detection reflected a recent infection. Two nested model structures were tested assuming either an equal force of infection in the presence and absence of MDAs (SFOI) or two different forces of infection (full model). However, neither model fitted the null prevalence observed at 3 weeks of age for two main reasons: (i) the estimated observed initial proportion of passively immune animals was about 60% in all herds so that 40% of the animals were deemed fully susceptible at birth; (ii) even in the presence of PI, piglets are not fully protected against infection as shown by the non negligible FOIs estimated in the presence of maternal antibodies. Moreover, only 20 piglets in each herd were sampled to establish the herd-specific seroprevalence profiles, which might have led to considerable uncertainty, especially if prevalence was low. These considerations are supported by the good concordance between the estimates and the data from 7 weeks of age onwards. In herd 2, the prevalence increased slowly, continuously and exponentially. Indeed the cumulative prevalence observed at slaughter age was about 50%. As the level of PI was similar to that of the other herds, this difference reflects a low infection pressure

which might be due to differences in husbandry practices and herd structure.

In models with a single force of infection, the inverse of the force of infection corresponds to the average age at infection [41]. However, this relationship cannot be applied to the full model due to the interplay between piglet serological status and the differential forces of infection. The impact of PI on time-toinfection was therefore assessed by modeling the infection-free survival functions, *i.e.* the probability of escaping infection according to age of the animal, in piglets with and without MDAs, for each herd. In a fully susceptible animal, transmission occurs much earlier than in pigs with MDAs. Several epidemiological studies have shown that peak HEV prevalence occurs after the nursery stage, at about 13 weeks of age, which corresponds to the estimated median age at infection in passively immune individuals [19,38,42]. However, the predicted average age-atinfection in herd 2 was considerably higher and beyond the usual slaughter age (25 weeks). As the level of PI in all herds was similar, this difference could presumably only be explained by differences in herd structures and husbandry practices which might help to reduce the infection pressure due to piglet mixing and the environmental viral load. These results suggest that the age at HEV infection does not strictly depend on the proportion of passive immune piglets but also relies on later events which condition the force of infection after PI waning [43].

The origins of most autochthonous HEV cases remain undetermined, although some cases have been related to the consumption of uncooked meat, liver or offal from domestic pigs or wild animals (e.g. wild boar or deer) [44-46]. In consequence, control measures in pig farms are aimed at preventing the presence of viremic animals at slaughter age. The modeling framework developed in this study was successfully used to evaluate the age at infection according to the serological status of piglets after colostrum intake in different pig farms. In herd 2, a large proportion of pigs remained susceptible before slaughter-age, mainly due to a lower force of infection after the wane of PI. These conditions could be associated with an increased risk as regards public health especially if later events, shortly before slaughter, enhance the infectious process and lead to massive infection of pigs. Further work needs to be carried out to fully understand which factors are involved in the resulting variations in forces of infection between herds.

We developed an original modeling framework to analyze the role of passive immunity in the transmission of infectious agents in animal populations based on batch-level incidence data. The protective characteristics of MDAs, acquired either through colostrum intake or transplacental passage, play an important part in several mammalian species [47–49]. In consequence, the methodology developed in this study could be extended to quantify the level of protection conferred by maternally-derived antibodies for different pathogens and species.

## **Supporting Information**

Figure S1 Individual fits of decay of maternally derived antibodies (12 individual profiles are shown). (TIFF)

Figure S2 Convergence plot of Monte Carlo Makov Chains for Herd 1 parameters (lines correspond to smoothing averages for each chain). (TIFF)

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# References

- Brinkhof MWG, Mayorga O, Bock J, Heininger U, Herzog C (2013) Kinetics of maternally acquired anti-hepatitis A antibodies: Prediction of waning based on maternal or cord blood antibody levels. Vaccine 31: 1490–1495.
- Elbers ARW, Stockhofe-Zurwieden N, van der Poel WHM (2014) Schmallenberg virus antibody persistence in adult cattle after natural infection and decay of maternal antibodies in calves. BMC Veterinary Research 10.
- Pravieux JJ, Poulet H, Charreyre C, Juillard V (2007) Protection of Newborn Animals through Maternal Immunization. J Comp Pathol 137: S32–S34.
- Swain P, Nayak SK (2009) Role of maternally derived immunity in fish. Fish and Shellfish Immunology 27: 89–99.
- Bouma A, De Jong MCM, Kimman TG (1997) The influence of maternal immunity on the transmission of pseudorabies virus and on the effectiveness of vaccination. Vaccine 15: 287–294.
- Kitikoon P, Nilubol D, Erickson BJ, Janke BH, Hoover TC, et al. (2006) The immune response and maternal antibody interference to a heterologous H1N1 swine influenza virus infection following vaccination. Vet Immunol Immunopathol 112: 117–128.
- Pastoret PP (2007) Challenges and Issues of Early Life Vaccination in Animals and Humans. J Comp Pathol 137, Supplement 1: S2–S3.
- Siegrist CA (2007) The Challenges of Vaccine Responses in Early Life: Selected Examples. J Comp Pathol 137, Supplement 1: S4–S9.
- Siegrist CA, Barrios C, Martinez X, Brandt C, Berney M, et al. (1998) Influence of maternal antibodies on vaccine responses: inhibition of antibody but not T cell responses allows successful early prime-boost strategies in mice. Eur J Immunol 28: 4138–4148.
- Britton T, Kypraios T, O'Neill PD (2011) Inference for Epidemics with Three Levels of Mixing: Methodology and Application to a Measles Outbreak. Scandinavian Journal of Statistics 38: 578–599.
- Cauchemez S, Ferguson NM (2008) Likelihood-based estimation of continuoustime epidemic models from time-series data: application to measles transmission in London. Journal of the Royal Society, Interface/the Royal Society 5: 885– 897.
- Courcoul A, Vergu E, Denis JB, Beaudeau F (2010) Spread of Q fever within dairy cattle herds: key parameters inferred using a Bayesian approach. Proceedings Biological sciences/The Royal Society 277: 2857–2865.
- Elderd BD, Dwyer G, Dukic V (2013) Population-level differences in disease transmission: A Bayesian analysis of multiple smallpox epidemics. Epidemics 5: 146–156.
- Jewell CP, Kypraios T, Neal P, Roberts GO (2009) Bayesian analysis for emerging infectious diseases. 10.1214/09-ba417: 465–496.
- McBryde ES, Gibson G, Pettitt AN, Zhang Y, Zhao B, et al. (2006) Bayesian modelling of an epidemic of severe acute respiratory syndrome. Bulletin of mathematical biology 68: 889–917.
- O'Neill PD (2002) A tutorial introduction to Bayesian inference for stochastic epidemic models using Markov chain Monte Carlo methods. Mathematical Biosciences 180: 103–114.
- Streftaris G, Gibson GJ (2004) Bayesian analysis of experimental epidemics of foot-and-mouth disease. Proceedings of the Royal Society B: Biological Sciences 271: 1111–1117.
- Backer JA, Berto A, McCreary C, Martelli F, van der Poel WHM (2012) Transmission dynamics of hepatitis E virus in pigs: Estimation from field data and effect of vaccination. Epidemics 4: 86–92.
- Kanai Y, Tsujikawa M, Yunoki M, Nishiyama S, Ikuta K, et al. (2010) Longterm shedding of hepatitis E virus in the feces of pigs infected naturally, born to sows with and without maternal antibodies. Journal of Medical Virology 82: 69– 76.
- Pavio N, Meng XJ, Renou C (2010) Zoonotic hepatitis E: Animal reservoirs and emerging risks. Veterinary Research 41.
- Rose N, Blanchard P, Cariolet R, Grasland B, Amenna N, et al. (2007) Vaccination of porcine circovirus type 2 (PCV2)-infected sows against porcine Parvovirus (PPV) and Erysipelas: effect on post-weaning multisystemic wasting syndrome (PMWS) and on PCV2 genome load in the offspring. J Comp Pathol 136: 133–144.
- Allerson M, Deen J, Detmer SE, Gramer MR, Joo HS, et al. (2013) The impact of maternally derived immunity on influenza A virus transmission in neonatal pig populations. Vaccine 31: 500–505.
- Casas M, Cortes R, Pina S, Peralta B, Allepuz A, et al. (2011) Longitudinal study of hepatitis E virus infection in Spanish farrow-to-finish swine herds. Vet Microbiol 148: 27–34.
- Snoeck E, Chanu P, Lavielle M, Jacqmin P, Jonsson EN, et al. (2010) A comprehensive hepatitis C viral kinetic model explaining cure. Clinical pharmacology and therapeutics 87: 706–713.

## **Author Contributions**

Conceived and designed the experiments: NR MA NP. Performed the experiments: MA NR NP. Analyzed the data: MA MC. Contributed reagents/materials/analysis tools: MC NP. Wrote the paper: MA NR NP MC.

- Lavielle M, Mentre F (2007) Estimation of population pharmacokinetic parameters of saquinavir in HIV patients with the MONOLIX software. J Pharmacokinet Pharmacodyn 34: 229–249.
- Barnaud E, Rogee S, Garry P, Rose N, Pavio N (2012) Thermal inactivation of infectious hepatitis E virus in experimentally contaminated food. Appl Environ Microbiol 78: 5153–5159.
- 27. Efron B, Tibshirani J (1993) An introduction to the bootstrap. New York: Chapman & Hall, Inc.
- Rose N, Lunazzi A, Dorenlor V, Merbah T, Eono F, et al. (2011) High prevalence of Hepatitis E virus in French domestic pigs. Comparative Immunology, Microbiology and Infectious Diseases 34: 419–427.
- Peralta B, Casas M, de Deus N, Martín M, Ortuño A, et al. (2009) Anti-HEV antibodies in domestic animal species and rodents from Spain using a genotype 3-based ELISA. Vet Microbiol 137: 66–73.
- Pluske JR, Le Dividich J (2003) Weaning the pig: concepts and consequences: Wageningen Academic Publishers.
- Rooke JA, Bland IM (2002) The acquisition of passive immunity in the new-born piglet. Livest Prod Sci 78: 13–23.
- Kitching RP, Salt JS (1995) The interference by maternally-derived antibody with active immunization of farm animals against foot-and-mouth disease. British Veterinary Journal 151: 379–389.
- 33. Loeffen WLA, Heinen PP, Bianchi ATJ, Hunneman WA, Verheijden JHM (2003) Effect of maternally derived antibodies on the clinical signs and immune response in pigs after primary and secondary infection with an influenza H1N1 virus. Vet Immunol Immunopathol 92: 23–35.
- Vandeputte J, Too HL, Ng FK, Chen C, Chai KK, et al. (2001) Adsorption of colostral antibodies against classical swine fever, persistence of maternal antibodies, and effect on response to vaccination in baby pigs. Am J Vet Res 62: 1805–1811.
- Ward LA, Rich ED, Besser TE (1996) Role of Maternally Derived Circulating Antibodies in Protection of Neonatal Swine against Porcine Group A Rotavirus. J Infect Dis 174: 276–282.
- Andraud M, Dumarest M, Cariolet R, Aylaj B, Barnaud E, et al. (2013) Direct contact and environmental contaminations are responsible for HEV transmission in pigs. Vet Res 44: 102.
- Meng XJ, Purcell RH, Halbur PG, Lehman JR, Webb DM, et al. (1997) A novel virus in swine is closely related to the human hepatitis E virus. Proc Natl Acad Sci U S A 94: 9860–9865.
- de Deus N, Casas M, Peralta B, Nofrarias M, Pina S, et al. (2008) Hepatitis E virus infection dynamics and organic distribution in naturally infected pigs in a farrow-to-finish farm. Veterinary Microbiology 132: 19–28.
- Francis MJ, Black L (1986) Response of young pigs to foot-and-mouth disease oil emulsion vaccination in the presence and absence of maternally derived neutralising antibodies. Res Vet Sci 41: 33–39.
- Klinkenberg D, Moormann RJ, de Smit AJ, Bouma A, de Jong MC (2002) Influence of maternal antibodies on efficacy of a subunit vaccine: transmission of classical swine fever virus between pigs vaccinated at 2 weeks of age. Vaccine 20: 3005–3013.
- Anderson RM, May RM (2004) Infectious diseases of humans: dynamics and control: Oxford University Press.
- 42. Fernandez-Barredo S, Galiana C, Garcia A, Vega S, Gomez MT, et al. (2006) Detection of hepatitis E virus shedding in feces of pigs at different stages of production using reverse transcription-polymerase chain reaction. Journal of Veterinary Diagnostic Investigation 18: 462–465.
- 43. Walachowski S, Dorenlor V, Lefevre J, Lunazzi A, Eono F, et al. (2013) Risk factors associated with the presence of hepatitis E virus in livers and seroprevalence in slaughter-age pigs: a retrospective study of 90 swine farms in France. Epidemiol Infect 10.1017/S0950268813003063: 1–11.
- Colson P, Borentain P, Queyriaux B, Kaba M, Moal Vr, et al. (2010) Pig Liver Sausage as a Source of Hepatitis E Virus Transmission to Humans. J Infect Dis 202: 825–834.
- Masuda J-I, Yano K, Tamada Y, Takii Y, Ito M, et al. (2005) Acute hepatitis E of a man who consumed wild boar meat prior to the onset of illness in Nagasaki, Japan. Hepatol Res 31: 178–183.
- Tei S, Kitajima N, Takahashi K, Mishiro S (2003) Zoonotic transmission of hepatitis E virus from deer to human beings. Lancet 362: 371–373.
- Hassig M, Stadler T, Lutz H (2007) Transition from maternal to endogenous antibodies in newborn calves. Veterinary Record 160: 234–235.
- Tomlinson A, Chambers M, Delahay R (2012) Mycobacterium bovis infection in badger cubs: Re-assessing the evidence for maternally derived immunological protection from advanced disease. Vet Immunol Immunopathol 148: 326–330.
- Van De Perre P (2003) Transfer of antibody via mother's milk. Vaccine 21: 3374–3376.