



# Inflammatory Responses Induced by the Monophasic Variant of *Salmonella* Typhimurium in Pigs Play a Role in the High Shedder Phenotype and Fecal Microbiota Composition

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**ABSTRACT** Pigs infected with *Salmonella* may excrete large amounts of *Salmonella*, increasing the risk of spread of this pathogen in the food chain. Identifying *Salmonella* high shedder pigs is therefore required to mitigate this risk. We analyzed immune-associated markers and composition of the gut microbiota in specific-pathogen-free pigs presenting different shedding levels after an oral infection with *Salmonella*. Immune response was studied through total blood cell counts, production of anti-*Salmonella* antibodies and cytokines, and gene expression quantification. Total *Salmonella* shedding for each pig was estimated and hierarchical clustering was used to cluster pigs into high, intermediate, and low shedders. Gut microbiota compositions were assessed using 16S rRNA microbial community profiling. Comparisons were made between control and inoculated pigs, then between high and low shedders pigs. Prior to infection, high shedders had similar immunological profiles compared to low shedders. As soon as 1 day postinoculation (dpi), significant differences on the cytokine production level and on the expression level of several host genes related to a proinflammatory response were observed between high and low shedders. Infection with *Salmonella* induced an early and profound remodeling of the immune response in all pigs, but the intensity of the response was stronger in high shedders. In contrast, low shedders seroconverted earlier than high shedders. Just after induction of the proinflammatory response (at 2 dpi), some taxa of the fecal microbiota were specific to the shedding phenotypes. This was related to the enrichment of several functional pathways related to anaerobic respiration in high shedders. In conclusion, our data show that the immune response to *Salmonella* modifies the fecal microbiota and subsequently could be responsible for shedding phenotypes. Influencing the gut microbiota and reducing intestinal inflammation could be a strategy for preventing *Salmonella* high shedding in livestock.

**IMPORTANCE** Salmonellosis remains the most frequent human foodborne zoonosis after campylobacteriosis and pork meat is considered one of the major sources of human foodborne infections. At the farm, host heterogeneity in pig infection is problematic. High *Salmonella* shedders contribute more significantly to the spread of this foodborne pathogen in the food chain. The identification of predictive biomarkers for high shedders could help to control *Salmonella* in pigs. The purpose of the present study was to investigate why some pigs become super shedders and others low shedders. We thus investigated the differences in the fecal microbial composition and the immune response in orally infected pigs presenting different *Salmonella* shedding patterns. Our data show that the proinflammatory response

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induced by *S. Typhimurium* at 1 dpi could be responsible for the modification of the fecal microbiota composition and functions observed mainly at 2 and 3 dpi and to the low and super shedder phenotypes.

**KEYWORDS** *Salmonella*, pig, gut microbiota, immunity, high shedder, inflammation

Host heterogeneity in animal infection is problematic and can contribute to the emergence and spread of foodborne pathogens. A minority of infected animals, the high shedders, also known as super shedders, are considered responsible for the majority of the infections (1, 2). Heterogeneity in *Salmonella* infection has been described in mice (2, 3), chickens (4), and pigs (5). *Salmonella* control on farms is extremely important, as *Salmonella* is a major cause of zoonotic diseases. Indeed, salmonellosis remains the most frequent human foodborne zoonosis after campylobacteriosis in the European Union (1). Pork meat is considered one of the major sources of human foodborne infections (6).

However, controlling *Salmonella* in the food chain is challenging because of the ability of *Salmonella* to colonize livestock without causing clinical disease. Pigs are typically asymptomatic healthy carriers of nonhost-restricted *Salmonella* serotypes (7) with varied levels and durations of fecal *Salmonella* shedding postinfection in both experimental and on-farm infections (8). High levels of *Salmonella* detected on carcasses have been positively correlated with both *Salmonella* shedding by carrier swine and *Salmonella* prevalence on a farm (9). In addition, the load of *Salmonella* can increase after transport, confirming that this phase of the production is also a critical point for the control of *Salmonella* contamination (10). Among serotypes infecting pigs, the monophasic variant of *Salmonella* Typhimurium is now one of the most prevalent (11–13) with *S. Derby* or *S. Typhimurium*, and its zoonotic importance in human salmonellosis is also highlighted (1, 14).

It has been reported that within a few hours of ingestion, *S. Typhimurium* can be detected in high numbers in the feces (15). Shortly after infection, epithelial barrier damage and inflammation are observed, mainly in the ileal mucosa, as well as a migration of the pathogen to the gut-associated lymphoid tissue via monocyte-derived cells (16). Moreover, 2 days after infection, *Salmonella* infection has been shown to impact the microbiome diversity at the mucosa (17). These changes in commensal gut microbiota could contribute to the pathogen's ability to colonize the gut successfully (18) and suggest that 16s rRNA microbial community analysis could be used to shape intervention strategies to mitigate the effects of *Salmonella* infection and transmission by a prompt identification of high shedders. These results suggest that the manipulation of certain taxa within the porcine intestinal microbial community could increase disease resistance against *Salmonella* in pigs (18).

Heterogeneity of *Salmonella* colonization and excretion is a multifactorial process involving the dose, the *Salmonella* serotype, the feed, and the host genetics (19, 20). However, the host immune status and gut microbiota composition appear to play a crucial role in the variation of *Salmonella* shedding in the high and low shedder phenotypes (5).

*In vivo* and *in vitro* gene expression studies (21, 22) as well as proteomic studies (23) have shown that an immune response is stimulated when pigs are infected with *Salmonella* Typhimurium. Detection of *Salmonella* by immune cell receptors triggers the innate immune response and induces the production of proinflammatory mediators. The release of chemotactic cytokines attracts other immune cells like dendritic or T cells and initiate a Th1 response (24). Different studies have explored the interaction between immune response and the shedding phenotypes. Pigs that shed elevated *Salmonella* in their feces have higher IFN- $\gamma$  levels in serum (25). However, it seems difficult to identify immunological markers that could be useful in predicting shedding status. Classical analyses of differentially expressed (DE) genes (19, 26, 27), comparing future high shedders with future low shedders before infection, have difficulty in finding predictive markers. By using gene coexpression analyses, Kommadath et al. (27) were able to identify modules of coexpressed genes that were associated with the

shedding phenotypes before infection. Of four of these modules, two were associated with immune response.

To date, very few studies have analyzed both the immune response and the gut microbiota composition in the same experiment (28, 29). Yet, understanding the changes occurring both in the host gut microbiota and the host immune response due to *Salmonella* infection is of paramount importance with respect to developing a better understanding of *Salmonella* pathogenesis and disease progression. Moreover, the analysis of these parameters before infection could lead to the identification of predictive biomarkers, which may provide tools to help prevent *Salmonella* carriage in pigs. The purpose of the present study was to investigate the dynamics between the pig gut microbiota, the immune response, and the levels of *Salmonella* shedding in orally infected pigs. Depending on the shedding levels, we determined three groups of pigs, high (HS), intermediate (IS) and low (LS) shedders, where zootechnical observations (temperature, growth, and food intake), immune parameters in the blood (total cell counts, level of anti-*Salmonella* antibodies, level of cytokines, and expression of immune genes), and gut microbiota (feces, ileum and cecum) composition were analyzed.

## RESULTS

**Salmonella shedding and carriage resulting from experimental infection.** All control pigs remained negative for *Salmonella* throughout the experiment while all the inoculated pigs shed *Salmonella* from 1 day postinfection (dpi) to the end of the experiment (21 dpi). *Salmonella* shedding ranged from 1.48 to 9.09 log<sub>10</sub>CFU/g of feces depending on the day and pig. The excretion peak was observed at 2 dpi, with 7.05 ± 1.77 log<sub>10</sub>CFU/g (median). Although all pigs were inoculated with the same dose of *Salmonella*, we observed different *Salmonella* excretion patterns.

The hierarchical clustering on the area under the log curve (AULC) values permits clustering the pigs into three groups. The three classes (Fig. 1) gathered 13, 16, and 11, high (HS), intermediate (IS), and low shedder (LS) pigs, respectively, with a median AULC of 107.70 ± 12.13, 88.90 ± 5.11, and 69.81 ± 4.69, respectively. These three classes were significantly different (ANOVA, *P* value = 9E-15). For all classes, excretion peak was observed at 2 dpi for the HS and IS class and at 3 dpi for the LS class (Fig. 2).

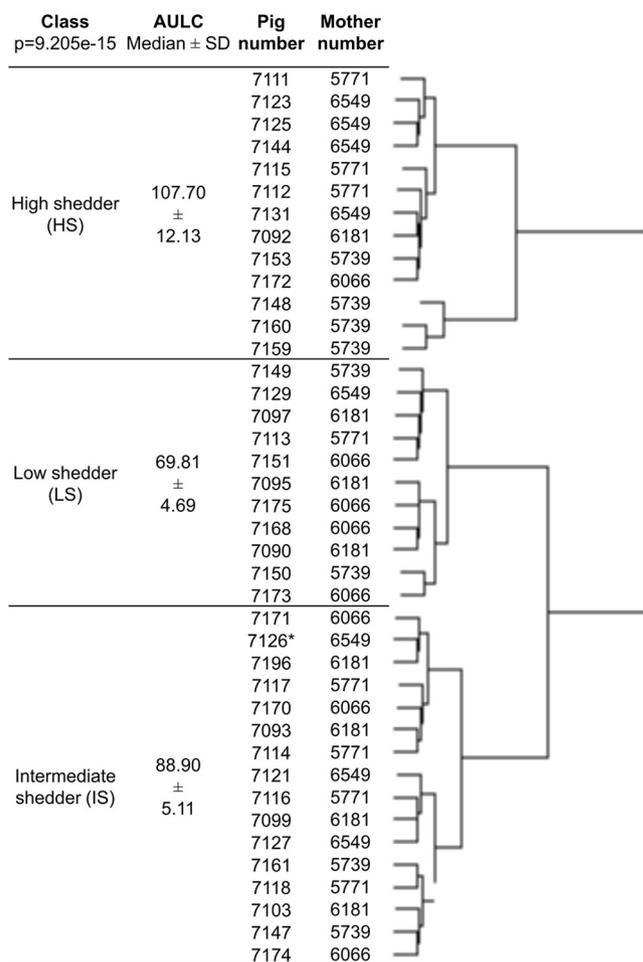
Pigs from the same sow were found in the three classes, LS, IS, and HS. Furthermore, based on AULC value, there was no difference between pigs according to the sow (*P* value = 0.424) (data not shown), indicating that this parameter has no detectable effect on the excretion level of *Salmonella*.

At the end of the experiment, tonsils, cecum, and ileum contents of the pigs were highly contaminated with *Salmonella* (mean, 5.60 ± 0.37; 3.71 ± 1.06; and 3.46 ± 1.66 log<sub>10</sub>CFU/g, respectively), in contrast to mesenteric lymph nodes (MLN) (mean, 0.86 ± 0.68 log<sub>10</sub>CFU/g). However, for the HS class, levels of *Salmonella* were significantly higher (*P* value < 0.01) for MLN, ileum, and cecum contents than for the LS group (1.26 ± 0.60; 4.56 ± 1.47; and 4.60 ± 0.96 log<sub>10</sub>CFU/g, respectively, versus 0.35 ± 0.39; 2.45 ± 1.66; and 3.10 ± 0.77 log<sub>10</sub>CFU/g, respectively) (Fig. 3).

**Zootechnical data.** A temperature increase was observed for all inoculated pigs. At 1 dpi, the difference was significant (*P* value < 0.01) between the control group (39.3°C) and inoculated group (39.9°C) but not according to shedding classes HS, IS, and LS.

Growth of the piglets was significantly slower (*P* value < 0.005) for HS pigs from 4, 11, and 18 dpi to the end compared to control and LS pigs (Table 1).

**Variation in markers of immunity during the experimental infection. (i) Total blood cell counts.** The impact of *Salmonella* was analyzed on different blood cell populations and leukocyte subpopulations at different time points before infection (data not shown) and postinfection (dpi). Statistical difference in cell numbers between inoculated and control pigs and between HS and LS pigs was evaluated by the Kruskal-Wallis test. The comparison of inoculated pigs to control ones showed that, after inoculation, at some time points, the number of cells between the blood cell populations was different. When compared, the HS and LS showed very few differences. Instead, at



**FIG 1** Hierarchical tree produced using AULC data. The analysis classifies the 40 inoculated pigs into three classes (ANOVA, *P* value = 9E-15).

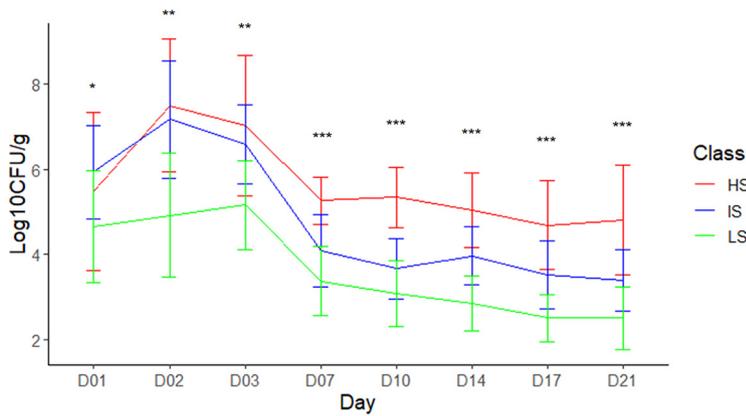
1 and 10 dpi, there was a difference in the number of lymphocytes and in the percentage of granulocytes (Table 2).

**(ii) Antibody response to *Salmonella*.** The IDEXX test was used to detect the level of antibodies against *Salmonella*; an optical density (OD) percentage over 15% indicates that pigs had seroconverted.

No seroconversion occurred in control pigs. Inoculated pigs seroconverted from 7 dpi in 12 pigs, from 10 dpi in 7 pigs, from 14 dpi in 4 pigs, and from 21 dpi in 4 pigs. Thirteen pigs did not seroconvert before the end of the experiment, at 21 dpi; 30.80% were HS pigs and 15.40% LS pigs. LS pigs seroconverted earlier than HS pigs in spite of a nonsignificant difference in the level of excretion at 1 dpi between low and super shedders. Seven of the HS pigs (53.80%) seroconverted only at 21 dpi (*n* = 3) or after (*n* = 4) (Fig. 4).

**(iii) Cytokine levels.** A peak of production was observed at 1 dpi for interleukins IL-1β, IL-6, and interferon IFN-γ, and at 3 dpi for TNF-α. A significant difference was observed on the median production level between HS and LS for IL-1β and IL-6 at 1 dpi (*P* value = 0.007 and 0.019, respectively) (Fig. 5).

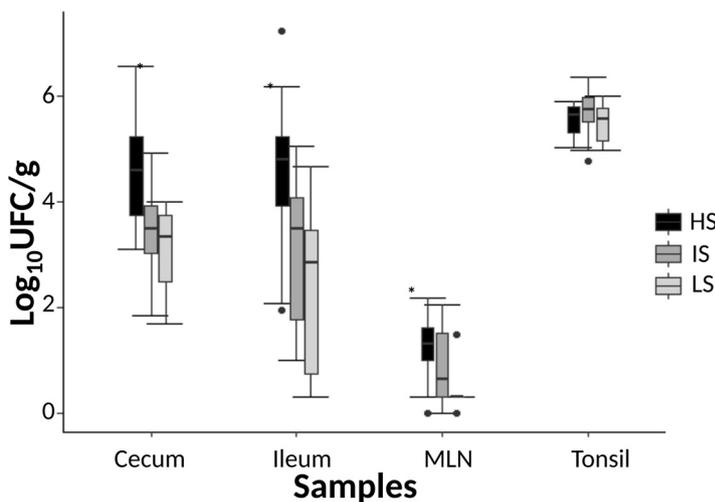
**(iv) Immune gene expression in HS and LS pigs.** To identify biomarkers that could predict the shedding phenotype, we compared immune gene expression between HS and LS pigs before infection. Very few differences were found (Table 3). Twenty days before inoculation, *IL-10* and *CLEC7A* were more expressed in HS pigs while *GNLY*, encoding an antimicrobial protein, a member of the saposin-like protein



**FIG 2** *Salmonella* enumeration in feces at each sampling day postinfection for each class (median excretion in log<sub>10</sub> CFU/g ± median absolute deviation according to days postinfection [D01 to D21]). Significant differences found after Kruskal-Wallis comparisons were represented by asterisks (\*): the *P* values were *P* = 0.029, 0.001, 0.004, <0.001, <0.001, <0.001, <0.001, and <0.001 at D1, 2, 3, 7, 10, 14, 17, and 21, respectively. Nevertheless, when comparing extreme categories (i.e., HS and LS, Student *t*-tests), no difference was observed at day 1 (*P* = 0.214), whereas significant differences were observed over the rest of the time series (*P* < 0.001, *P* < 0.001, *P* = 0.012, *P* < 0.001, *P* < 0.001, *P* < 0.001, and *P* < 0.001 at day 2, 3, 7, 10, 14, 17, and 21).

family that kills various pathogens, including *Salmonella*, was less expressed. At 7 days before inoculation, *IL-10* was still more expressed in HS pigs, and *CSF3* was less expressed in HS pigs the day of inoculation.

After inoculation, when HS or LS were compared to noninoculated pigs, out of 70 expressed genes, 46 differentially expressed (DE) genes were found in infected pigs (Table 4). The peak of gene expression modulation was found at 1 dpi for the HS pigs, while the difference in the level of excretion at 1 dpi between low and high shedders was not significant. In this regard, it is interesting to note that the peak production of *IL-1β* and *IL-6*, two master cytokines, in the innate immune response, was also found at 1 dpi. In most of cases, the fold changes were higher for HS than for LS pigs; for example, at 1 dpi, *S100A12* and *S100A9* were expressed 49.60 and 12.00 times more in HS than in controls, respectively, but this ratio was only 6.20 and 4.80 in LS. The corresponding proteins bind calcium and zinc and, by regulating several pathways, have proinflammatory properties; this suggests that HS could develop a stronger inflamma-



**FIG 3** *Salmonella* enumeration from different sample types, according to excretion classes, at the time of necropsy (D21 to D22 after *Salmonella* inoculation). Asterisks (\*) indicate statistically significant difference between HS and LS classes in the cecum (*P* = 0.002), ileum (*P* = 0.005), and MLN, (*P* = 0.009), but not for the tonsils (*P* = 0.273).

**TABLE 1** Growth in Kg according to time (dpi) and pig classes

Shedding class	4 dpi	11 dpi	18 dpi
HS pigs	16.46+/-2.01	21.28+/-2.32	27.80+/-2.72
LS pigs	19.52+/-3.10	24.60+/-3.59	30.87+/-4.15
Control pigs	19.90+/-2.19	25.10+/-2.05	31.76+/-2.60

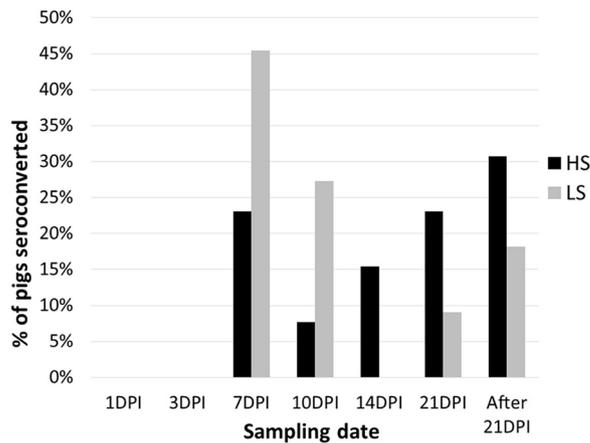
tory response. Altogether, this indicates that, contrary to what was intuitively anticipated (30), a strong proinflammatory response did not lead to high protection. Expression of 13 genes was decreased in HS pigs compared to the control, whereas no downregulated genes were observed in the LS pigs. The downregulated genes in HS are involved in different branches of immunity. TLR1, TLR6, and TLR3 are receptors that recognize pathogen-associated molecular patterns and trigger cytokine production. TLR1 and TLR6 recognize bacterial acylated lipopeptides, while TLR6 recognizes double-stranded RNA molecules. Expression of *GPLY* was slightly lower in HS than in LS; 20 days before inoculation and 1 day after infection, its expression in HS pigs was downregulated compared to controls. At 7 and 14 dpi, the expression of this gene was higher in HS and in LS compared to controls.

**(v) Correlation with the previously described genes.** Kommadath et al. (27) identified four modules of coexpressed genes which are associated with shedding status, and which are also found before inoculation. Within the four modules the “pink” and the “gray” ones are associated with immune response. In our study, 12 genes belonging to the pink module and 6 to the gray module were upregulated in infected pigs (Table 5). It is assumed that coexpressed genes are involved in common biological processes; we used STRING (31) (<https://string-db.org/>) to determine how the genes from our study, belonging to the immune modules associated with shedding status,

**TABLE 2** List of hematological variables with significant differences between control and inoculated pigs, and between high and low shedder pigs<sup>a</sup>

Blood cells	Sampling day	C vs I P value	HS vs LS P value
White cells			
Leucocytes	1 dpi	<b>0.030</b>	0.120
	7 dpi	<b>0.002</b>	0.840
	10 dpi	<b>0.043</b>	0.340
% lymphocytes	1 dpi	<b>0.001</b>	<b>0.040</b>
	7 dpi	<b>0.004</b>	0.170
	10 dpi	<b>0.001</b>	<b>0.030</b>
% basophiles	1 dpi	<b>0.006</b>	<b>0.020</b>
% eosinophiles	1 dpi	<b>0.008</b>	0.320
% monocytes	D0	<b>0.015</b>	0.460
	3 dpi	<b>0.004</b>	<b>0.040</b>
	10 dpi	<b>0.049</b>	0.210
% granulocytes	1 dpi	<b>0.001</b>	<b>0.020</b>
	7 dpi	<b>0.002</b>	0.090
	10 dpi	<b>0.001</b>	<b>0.040</b>
% other cells	10 dpi	<b>0.020</b>	0.660
	21 dpi	<b>0.014</b>	0.330
Red cells			
Red cells	21 dpi	<b>0.006</b>	0.560
Haematocrit	21 dpi	<b>0.010</b>	<b>0.040</b>
Mean cell haemoglobin concn	10 dpi	<b>0.004</b>	0.200
	21 dpi	<b>0.008</b>	0.310
Red blood cell distribution index	3 dpi	<b>0.008</b>	0.884
	7 dpi	<b>0.020</b>	<b>0.030</b>
Platelets	3 dpi	<b>0.020</b>	0.310

<sup>a</sup>Significant differences are shown in bold. C, control; I, inoculated; HS, high shedder; LS, low shedder.



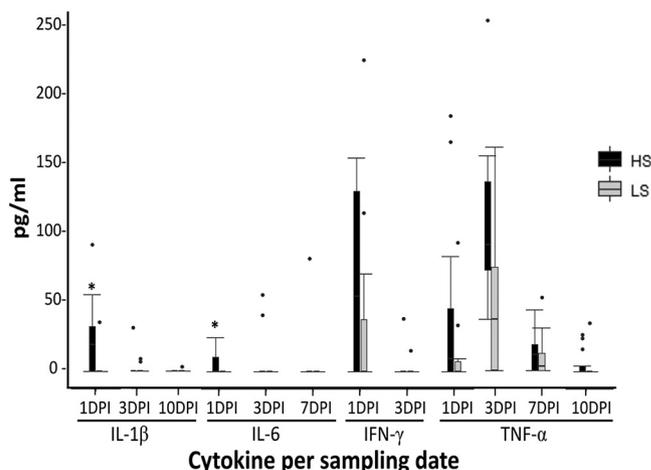
**FIG 4** Percentage of pigs seroconverted by classes (HS versus LS) per sampling date.

could be functionally linked. Interaction networks between our DE genes, present in the Kommadath's pink and gray modules, showed that numerous genes of the pink module belong to the same network (Fig. 6A). However, only few DE genes from our study were found in Kommadath's gray module by using STRING, (Fig. 6B).

**16S microbial population analysis in samples collected during and at the end of experimental infection. (i) Analysis of fecal samples according to shedding class using the QIIME pipeline.** For these analyses, we considered the fecal samples of all three shedding classes and the control group. The 16S microbial population analysis showed a small but statistically significant difference in bacterial abundance as described below.

Alpha diversity analysis showed statistically significant differences in the number of the operational taxonomic units (OTUs) between each class of inoculated pigs (HS, IS, and LS) and the control group ( $P = 1.55E-04$ ,  $8.78E-05$ , and  $1.85E-04$ , respectively) (Fig. 7A). However, no statistically significant differences were detected in the OTU composition between the three classes defined as HS, IS, and LS. Beta diversity analysis results were similar to those obtained with alpha diversity analysis (Fig. 7B). This result shows that *Salmonella* colonization modified gut microbiota composition.

**(ii) Analysis by sample type (feces, cecum, and ileum) using the QIIME pipeline.** Alpha diversity analysis of infected pigs showed statistically significant differences in the number of OTUs present in the ileum compared to the number of OTUs found in feces and the cecum. This was supported by Kruskal-Wallis (pairwise) tests, where the number of OTUs in cecal samples also showed small but statistically significant differ-



**FIG 5** Cytokine production per sampling date. Asterisks (\*) indicate the statistically significant difference between classes HS and LS at  $P < 0.05$ .

**TABLE 3** Comparison of gene expression between HS and LS pigs at 20 (D-20) and 7 (D-7) days before infection and the day of inoculation (D0)<sup>a</sup>

Sampling day	Gene	HS/LS
D-20	<i>IL10</i>	2.3
	<i>CLEC7A</i>	2.3
	<i>GNLY</i>	-1.8
D-7	<i>IL10</i>	3.2
D0	<i>CSF3</i>	-2.3

<sup>a</sup>FC represents the ratio of gene expression in HS pigs compared to LS. Genes with an FC with an absolute value >2, and a *P* value <0.05 are listed.

ences with respect to feces (H value = 6.174; *P* value = 0.013; q value = 0.039) and ileum (H = 4.103; *P* = 0.043; q = 0.086), and the number of observed OTUs in ileum compared with feces (H = 9.081; *P* = 0.003; q = 0.015). These differences are summarized by the principal component analysis (PCA) of unweighted UniFrac distances shown in Fig. 8 and in the alpha diversity boxplot (Fig. 9A). Beta diversity supported the results obtained with the alpha diversity (Fig. 9B). In particular, the composition and/or relative abundances of bacteria in samples differ according to sample type in infected pigs as well as in control group (data not shown).

**(iii) Other metadata observed.** There were no statistically significant differences in OTU composition of fecal samples according to the mother of origin, or sex.

**(iv) Analysis by age and shedding class (FROGS pipeline).** To go further, we investigated the differences in the fecal microbial composition at each date of sampling and in ileum and cecum microbial composition at the time of necropsy among the three shedding classes (i.e., HS, LS, IS). To consider the heterogeneous and relatively high alpha and beta diversities estimated using QIIME, additional analyses were done using FROGS, a pipeline optimized for cases of heterogeneous abundances and high numbers of species (32).

As observed in the first analysis, the  $\alpha$ -diversities did not reveal strong differences except at 2 days after infection (Table 6); the  $\alpha$ -diversity was indeed markedly higher for the LS (the mean Chao1 index was 500.8, 452.3, 371.2 for the LS, IS, and HS, respectively; the mean Shannon index was 4.12, 4.00, and 3.85 for the LS, IS, and HS, respectively). This result underlined that, few days after infection, the shedding classes differed for the richness (Chao1) of their gut microbiota. It should be noted that, after removal of the OTUs assigned to *Salmonella*, the differences of  $\alpha$ -diversity remained significant for Chao1 index 2 days after infection (*P* = 0.005).

The  $\beta$ -diversities (Bray-Curtis index) revealed slight differences among the shedding classes at two and 3 days after infection (*P* value = 0.010 and 0.021, respectively; Fig. 10). In line with this, significant correlations were found by directly fitting bacterial counts on the Bray-Curtis ordination on day 2 and 3 after infection (*P* value = 0.001 and 0.002, respectively). Furthermore, this result was in line with the fact that several OTUs presented differential abundances at these time points (Table 7). The strength of this analysis was supported by the identification of the OTU designated by Cluster\_50, which was one of the OTUs assigned to *Salmonella*, and representing the main part of this genus' abundance (>95%). According to the bacteriological measurements, this cluster likely corresponded to the inoculated strain. It was found enriched in the HS on day 2 and 3 after infection. Among the other OTUs, whose abundance was significantly different, several facultative anaerobes or aero-tolerant genera were found enriched in the HS (i.e., OTU assigned to the *Bacteroidetes* phylum like *Bacteroides* and *Coprobacter* and to the *Proteobacteria* phylum like *Actinobacillus*, *Escherichia* and *Desulfuvibrio*). Surprisingly, despite the overall high percentage of *Firmicutes* in the fecal microbiota, only few *Firmicutes*-associated OTU (*Ruminococcaceae*, *Lactobacillaceae*) were significantly different between the HS, IS and LS phenotypes.

**(v) Functional enrichment analysis.** The functional enrichment analysis revealed that the main differences among the three shedding classes were observed on day 2,

**TABLE 4** Comparison of gene expression between HS and control pigs or between LS and control pigs at 1, 3, 7, and 14 dpi<sup>a</sup>

LS/ control				Gene	HS/ control			
1 dpi	3 dpi	7 dpi	14 dpi		1 dpi	3 dpi	7 dpi	14 dpi
6.20	6.7	2.1		<i>S100A12</i>	49.6	20.5	2.1	2.0
16.80				<i>IDO</i>	33.0	8.3		
4.90	4.2			<i>ALOX5AP</i>	14.4	7.4	2.4	
4.80	6.3			<i>S100A9</i>	12.0	12.6	2.4	
4.60	4.2			<i>CASP4</i>	10.6	7.9		
3.80	3.4			<i>ARG2</i>	9.0	5.7		
4.40	6.6	2.0		<i>IL15</i>	8.0	7.6		
4.30	2.9			<i>TLR4</i>	6.1	3.3		
3.20	4.0			<i>IGSF6</i>	5.8	5.5		
3.10	4.0			<i>TREM1</i>	4.9	4.1	2.5	
3.90	2.2			<i>SOCS3</i>	4.9	2.3		
	7.6			<i>MX1</i>	4.1	11.8	3.6	
2.50	3.4			<i>CASP1</i>	3.2	6.0		
				<i>MYD88</i>	3.1	2.1		
				<i>SOCS1</i>	2.8			
				<i>CD14</i>	4.2			
2.00				<i>SDCBP</i>	2.1			
2.10				<i>TNFRSF1A</i>	2.0			
	3.9			<i>IL10</i>		6.6		
	4.2			<i>MX2</i>		5.5		
	3.6			<i>CEBPB</i>		5.0	2.3	
3.20	3.7			<i>IL18</i>		4.4		
				<i>CLEC7A</i>		3.9	2.1	
2.50	2.5			<i>MDA5</i>		3.5		
	2.5			<i>INFGR1</i>		3.2		
	2.2			<i>EIF2AK2</i>		3.1	2.2	
				<i>TLR8</i>		2.7		
				<i>IL23A</i>		2.4		
		2.0		<i>CASP8</i>		2.1	2.9	
				<i>CD86</i>		2.0		
				<i>INFAR2</i>		2.0		
		2.0	3.3	<i>CCL4</i>			2.6	3.0
		3.7	5.1	<i>GNLY</i>	-4.2		3.7	
				<i>HSP70</i>				-2.3
				<i>HSP60</i>	-2.0			
				<i>TGFBR3</i>	-2.1			
				<i>TLR6</i>	-2.1			
				<i>CSF3</i>	-2.7			
				<i>TLR1</i>	-2.9			
				<i>GATA3</i>	-3.2			
				<i>TNFSF5</i>	-3.2			
				<i>SLADQ</i>	-3.2			
				<i>SLADR</i>	-4.2			
				<i>TBX21</i>	-4.6			
				<i>TLR3</i>	-3.2	-2.3		

<sup>a</sup>FC represents the ratio of gene expression between LS infected pigs compared to controls or between HS infected pigs compared to controls. Genes with an FC with an absolute value >2 and a *P* value <0.05 are listed. Gray cells represent a FC that is higher in HS pigs than in LS.

3, and 17 after infection. However, for clarity, only comparisons between LS and HS are described Table 8. These differences included (i) an enrichment of short-chain fatty acid production in the HS (i.e., pathway noted P162-PWY) and (ii) enrichments of several menaquinol biosynthesis pathways. Menaquinone biosynthesis pathways are essential in electron transport and ATP generation in all Gram-positive, and anaerobically respiring Gram-negative bacteria. This result suggested that anaerobic respiration is induced in HS by using electron acceptors other than molecular oxygen (O<sub>2</sub>) such as nitrate (NO<sub>3</sub><sup>-</sup>), fumarate, sulfate (SO<sub>4</sub><sup>2-</sup>), or sulfur (S). These pathways were highly induced but only at 3 dpi (Table 8). On day 17 after infection, similar pathways were

**TABLE 5** List of DE genes in this study that belong to the pink and gray modules of coexpressed genes associated with the shedding phenotype, as defined by Kommadath et al. (27)

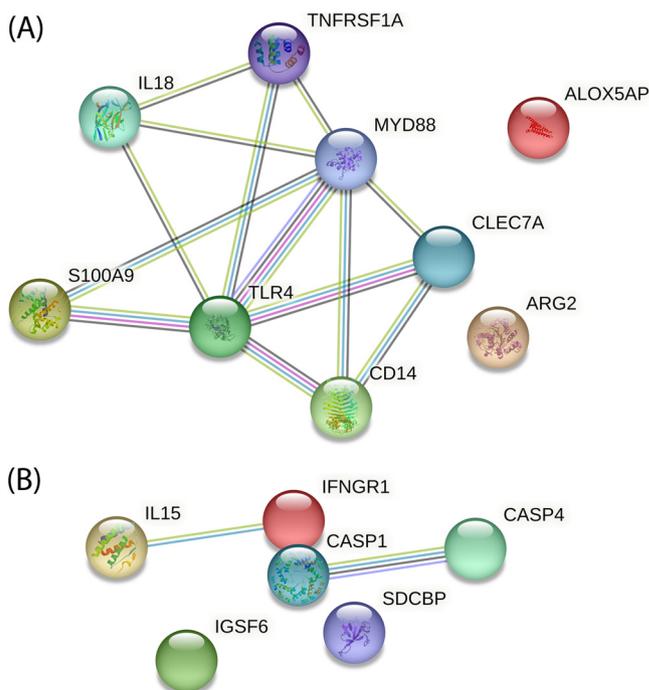
Pink module	Grey module
<i>ALOX5AP</i>	<i>CASP1</i>
<i>ARG2</i>	<i>CASP4</i>
<i>CD14</i>	<i>IFNGR1</i>
<i>CLEC7A</i>	<i>IGSF6</i>
<i>IL18</i>	<i>IL15</i>
<i>MYD88</i>	<i>SDCBP</i>
<i>S100A9</i>	
<i>S100A12</i>	
<i>TLR4</i>	
<i>TNFRSF1A</i>	

detected but mainly from the comparison of the IS and LS with high levels of menaquinol biosynthesis in the IS (data not shown).

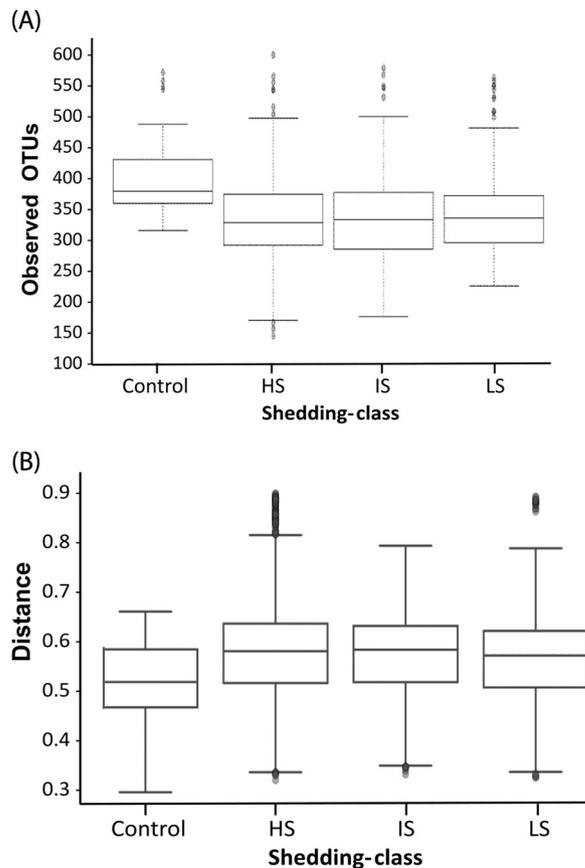
**DISCUSSION**

In our study, we observed that pigs experimentally infected with the same dose of a monophasic variant of *Salmonella* Typhimurium exhibited very different shedding levels (differences of more than 7 log) and we could distinguish three groups of shedding, namely, high, intermediate, and low shedders, based on a kinetic analysis. Such different shedding patterns were described in a previous experimental study of *S.* Typhimurium infection in pigs (26), introducing the notions of persistent and low shedding.

In addition, in our study, when pigs were high shedders, they also exhibited a significantly higher level of *Salmonella* in mesenteric lymph nodes, ileum, and cecum



**FIG 6** Protein-protein interaction networks corresponding to genes in our experiment that belong to the modules of coexpressed genes associated with shedding phenotype as defined by Kommadath et al. (27). The nodes of the networks correspond to genes and include, when it is known, a 3D representation of the protein. The edges, corresponding to known interaction, are colored according to the source of information: coexpression evidence in black, neighborhood evidence in green, protein homology in pastel blue, cooccurrence evidence in blue, experimental evidence in purple, and database evidence in cyan blue. (A) Pink module. (B) Gray module.

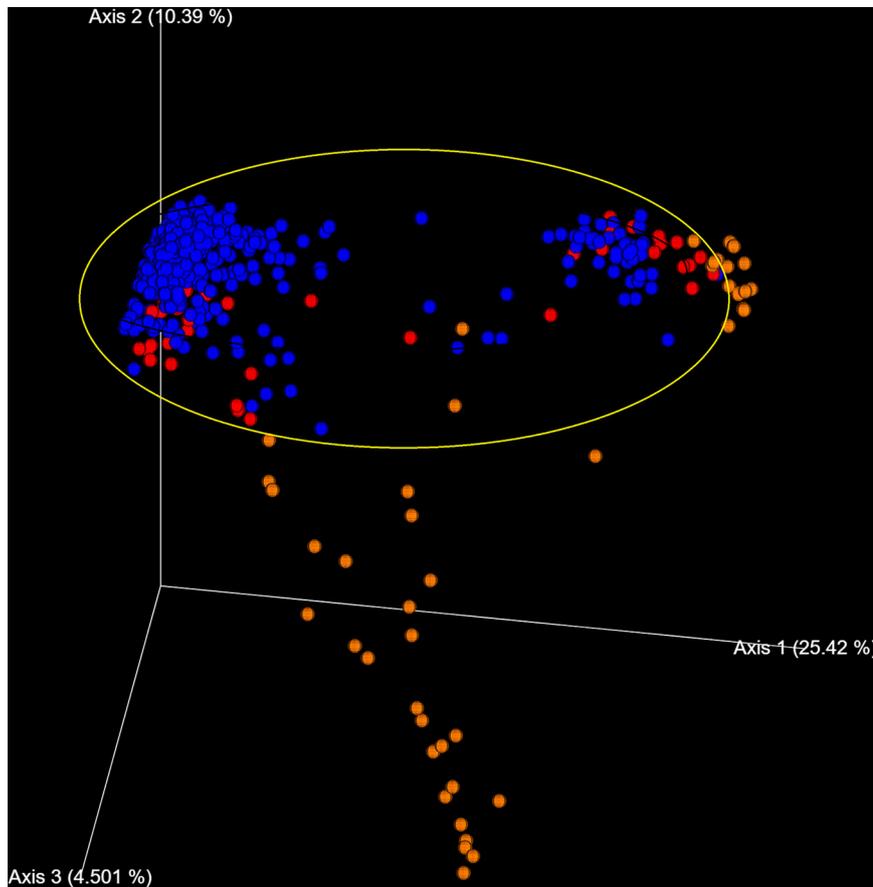


**FIG 7** Alpha and beta diversity according to shedding class. (A) Alpha diversity. Fecal samples belonging to the three classes described as HS, IS, and LS show a consistent reduction of the observed OTUs in respect to the control group. No statistically significant differences were detected between HS, IS, and LS with this 16S analysis. (B) Beta diversity (PERMANOVA) class.  $P = 0.001$  when controls are compared to HS, IS, or LS;  $P = 0.003$  when HS are compared to LS;  $P = 0.018$  when HS are compared to IS;  $P = 0.173$  for when IS are compared to LS.

compared to low and intermediate shedders. In line with result, the high shedder phenotype was found to be related to high cecal colonization (4). We also demonstrate that, these different shedding patterns were not linked to the mother or sex because high and low shedder pigs can originate from a same mother and were equally detected in males and females.

As HS have a much higher transmission rate, they constitute a key target for epidemiological investigation and disease management. Consequently, defining the conditions that favor the HS phenotype is a prerequisite to control the reservoir of contamination within a population. Moreover, to lead these interventions, we need to improve our knowledge on markers in gut microbiota (33) and/or in immune response (19, 25, 26) that could promote the high excretion in pigs.

In the first part of this study, we analyzed immunological parameters that could represent a signature of the shedding status and could identify high shedders before infection. It may be intuitively assumed that a less robust immune response could result in a high shedding level after infection (30). Contrary to that, before infection, we found that there was no difference in the level of production of the cytokines IL-1 $\beta$ , IL-6, TNF- $\alpha$ , and IFN- $\gamma$  between HS and LS pigs. In addition, we analyzed the expression of 70 immune-related genes before infection and we found only very few differences in gene expression between the shedding classes. Surprisingly, among these few genes, the comparison of expression levels in LS and HS indicated that the latter could be less effective in mounting an immune response. Indeed, at 20 and 7 days before

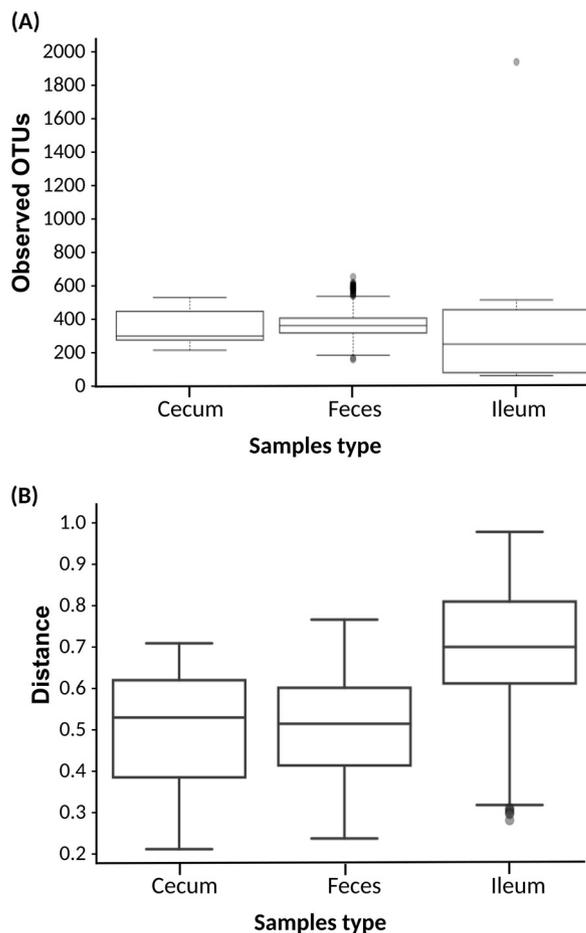


**FIG 8** Unweighted UniFrac PCA. PCA was generated from unweighted UniFrac distances. Samples collected from ileum (orange) are clearly separated (an exception was made for few samples) from fecal (blue), and cecal (red) clusters (yellow oval).

inoculation, the *IL-10* gene, encoding a protein with potent antiinflammatory activities, was more expressed in HS pigs. In the same way, at 7 days before infection, expression of *GNLY*, encoding a protein with robust antimicrobial activity against various pathogens, including *Salmonella*, was slightly less expressed in HS pigs (34). Similarly, *CSF3*, that may reduce *Salmonella* shedding upon infection in pigs (35), was less expressed in HS the day of infection. On the other hand, early after infection, compared to noninfected pigs, expression of *GNLY* and *CSF3* but also that of *TLR1*, 3, and 6, key players of innate response development, was downregulated only in HS pigs (Table 4). Downregulation of expression of these genes could be related to a partial immunosuppressive activity exerted on HS by the invading *Salmonella*.

In agreement with what was observed in previous studies (26, 36), we found that infected pigs exhibited an increased production of the key proinflammatory cytokines IL-1 $\beta$ , IL-6, and TNF- $\alpha$  and that HS pigs were higher producers of these cytokines compared to LS pigs (26). It is important to note that at 1 dpi, the strong proinflammatory response could not be explained by higher *Salmonella* colonization, as the HS and LS phenotypes were mainly related to the level of *Salmonella* after the second day postinfection (Fig. 2). Consistent with other reports, we observed a peak in the production of the Th1-produced cytokine IFN- $\gamma$  at 1 dpi and HS pigs were again higher producers compared to LS (25, 26). After 3 dpi the differences between infected and control pigs decreased, showing that the burst of immune response is observed in the first days after infection and correlated with the high level of *Salmonella* shedding observed from 2 dpi.

In agreement with other studies (19, 26, 27), we found that the peak in the modulation of gene expression was as early as 1 dpi and that HS pigs developed a stronger



**FIG 9** Alpha and beta diversity analysis boxplot according to samples type (feces, cecum, and ileum). (A) Alpha diversity. Small, but statistically significant differences can be noted with respect to feces ( $H$  value = 6.174,  $P$  value = 0.013,  $q$  value = 0.039) and ileum ( $H$  = 4.103,  $P$  = 0.043,  $q$ -value = 0.086), and the number of observed OTUs in ileum compared with feces ( $H$  = 9.081,  $P$  = 0.003,  $q$  = 0.015). (B) Beta diversity. Bacterial composition in ileum shows a statistically significant difference from the bacterial composition in fecal and cecal samples (Table 7).

immune response. Kommadath et al. (27) have described four modules of coexpressed genes associated with the shedding status; two of them, the pink and the gray, are related to immune response. A significant number of our DE genes belonged to these immune-related modules. In the pink module, most of the genes are involved in the activation of the transcriptional activator NF- $\kappa$ B. Recognition of the lipopolysaccharide by TLR4 and CD14 recruits the downstream adaptor component MYD88, leading to the final activation of NF- $\kappa$ B. By fostering the expression of MYD88 and CD14, TREM1 acts as an amplifier of the TLR4 signaling (37). CLEC7A is an innate receptor whose dimerization after recognition of  $\beta$ -glucans leads to recruitment of SYK then to the activation of NF- $\kappa$ B (38). S100A12 and S100A9 may be recognized by the innate receptors RAGE and TLR4; RAGE may activate different signaling pathways also resulting in NF- $\kappa$ B activation. Similarly, recognition of the master proinflammatory cytokine TNF- $\alpha$  by TNFRSF1A, one of its receptors, also induces NF- $\kappa$ B activation.

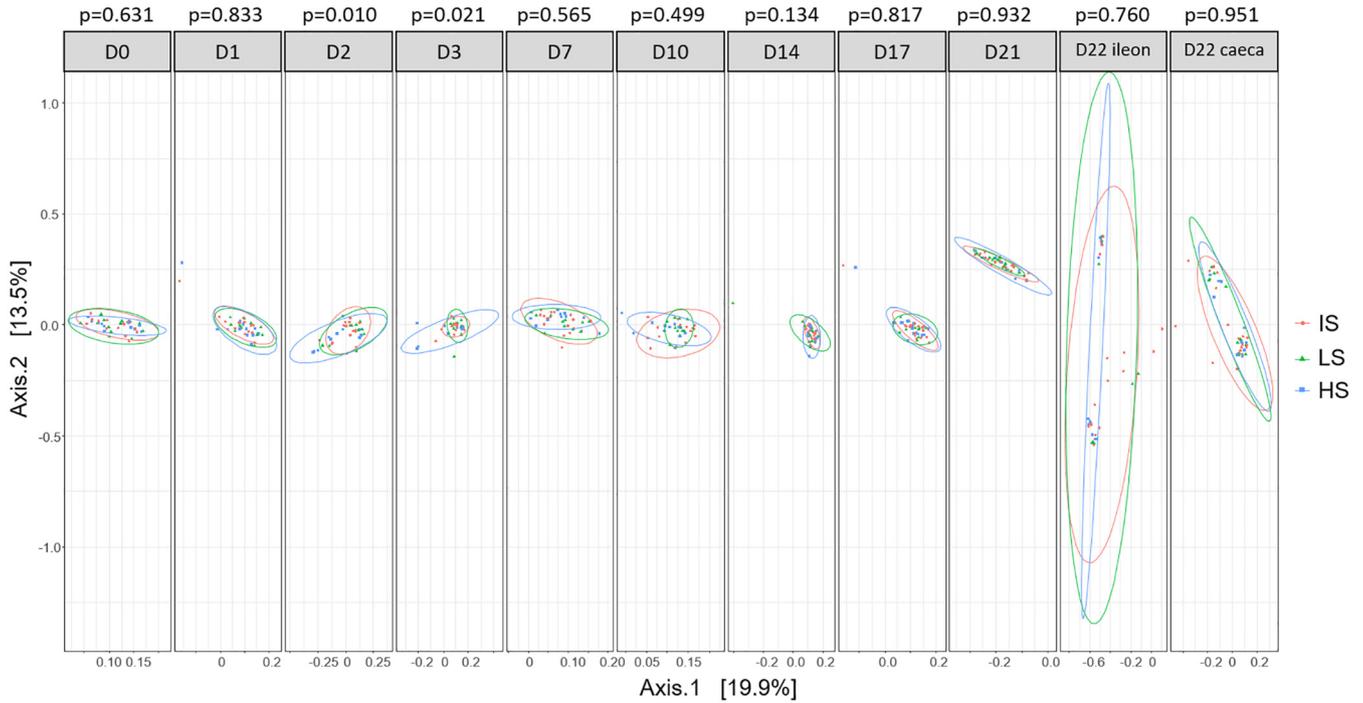
It is noteworthy that, within the top 10 upregulated genes as early as 1 dpi, we found genes exhibiting an immunosuppressive activity like *IDO*, which mediates regulatory T-cell proliferation and exerts antiproliferative activity on effector T cells (39), and *SOCS3*, an inhibitor of cytokine signaling (40). The increased expression of these four genes, with alternative activities, probably illustrated that in order to prevent the pathological consequences following the triggering of the immune response, infected pigs also trigger an early antiinflammatory response concomitant with the

**TABLE 6** One-way ANOVA tests comparing the Chao1 and Shannon  $\alpha$ -diversity<sup>a</sup>

Sampling day and sample type	F value	P value
D0 feces		
Chao1	0.107	0.899
Shannon	0.521	0.598
D1 feces		
Chao1	1.949	0.157
Shannon	2.931	0.066
D2 feces		
Chao1	6.467	<b>0.004</b>
Shannon	4.810	<b>0.014</b>
D3 feces		
Chao1	0.740	0.485
Shannon	1.079	0.352
D7 feces		
Chao1	0.383	0.684
Shannon	1.360	0.270
D10 feces		
Chao1	2.308	0.114
Shannon	2.624	0.086
D14 feces		
Chao1	0.379	0.687
Shannon	1.308	0.283
D17 feces		
Chao1	0.527	0.595
Shannon	0.200	0.819
D21 feces		
Chao1	1.436	0.251
Shannon	0.505	0.608
D22 ileum		
Chao1	0.125	0.883
Shannon	0.133	0.876
D22 cecum		
Chao1	1.133	0.332
Shannon	1.337	0.274

<sup>a</sup>The comparison among the three shedding categories HS, IS, and LS the day of infection (D0) and at different days postinfection is shown. Significant *P* values (*P* < 0.05) are in bold.

establishment of the proinflammatory one. Nevertheless, the stronger immune response in HS is puzzling because one could have anticipated that the high shedding level was the consequence of a less robust immune response (30). However, different hypotheses could solve this apparent paradox. First, the LS could develop a more efficient or a faster immune response and by the time gene expression was analyzed the immune response had returned to the steady state. However, this seems unlikely because gene expression analyses were done very early after infection, at 1 dpi. Another hypothesis relies on the strong interaction between gut microbiota and immune response. Several studies agree indeed that the local inflammation, resulting from the immune response elicited by *Salmonella*, triggers changes in the gut environment, which promote the growth of *Salmonella* and prompt changes in the gut microbiota composition (28, 41). Our data strongly argued for this hypothesis. The fact that the proinflammatory response is observed just before the high level of *Salmonella* colonization suggested that intestinal infection with *S. Typhimurium* results in inflammation, which induced dysbiosis favoring the HS phenotypes.



**FIG 10** Principal coordinates analyses of the Bray-Curtis  $\beta$ -diversity indexes at every time point. The ellipses represent the distribution of the samples collected in each shedding category, HS (high shedders), IS (intermediate shedders), LS (low shedders). The  $P$  values of the PERMANOVA comparison of the shedding classes are reported above each plot. From D0 to D21, samples correspond to feces.

This inflammatory response is thought to enable *Salmonella* to compete with the resident gut microbiota using tetrathionate produced by the oxidation of thiosulfate by reactive oxygen species (ROS) or due to its resistance to defensive molecules such as ROS or lipocalin-2, which is secreted by neutrophils and limits the acquisition of iron (42). This hypothesis could explain why the strong inflammatory response observed in HS is correlated with the highest *Salmonella* shedding levels (43–45). The analysis of the fecal microbiota composition and of the putative functions of this microbiota supported this last hypothesis.

The 16s rRNA microbial community profiling showed that  $\alpha$ -diversity indexes as

**TABLE 7** OTUs presenting differential abundances in feces among the HS and LS shedding classes<sup>a</sup>

Sampling day	OTU	Log <sub>2</sub> fold change	Phylum	Family	Genus	P.adj
D1	Cluster_61	3.29	Firmicutes	Ruminococcaceae	[ <i>Eubacterium</i> ] coprostanoligenes group	8.49E-03
D2	Cluster_50	-4.37	Proteobacteria	Enterobacteriaceae	<i>Salmonella</i>	1.64E-04
	Cluster_44	-5.51	Bacteroidetes	Bacteroidaceae	<i>Bacteroides</i>	3.26E-04
	Cluster_145	-5.89	Fusobacteria	Fusobacteriaceae	<i>Fusobacterium</i>	1.92E-06
	Cluster_172	-2.98	Bacteroidetes	Bacteroidaceae	<i>Bacteroides</i>	4.13E-03
	Cluster_423	-6.99	Proteobacteria	Pasteurellaceae	<i>Actinobacillus</i>	2.30E-08
	Cluster_316	-6.34	Proteobacteria	Pasteurellaceae	<i>Actinobacillus</i>	1.55E-07
	Cluster_250	-5.52	Firmicutes	Lachnospiraceae	[ <i>Acetivibrio</i> ] ethanol gignens group	9.17E-06
	Cluster_18	3.92	Cyanobacteria	Unknown_family	Unknown genus	7.03E-04
	Cluster_148	-2.77	Bacteroidetes	Marinifilaceae	<i>Odoribacter</i>	3.26E-03
	Cluster_182	-3.28	Bacteroidetes	Barnesiellaceae	<i>Coprobacter</i>	3.78E-03
D3	Cluster_44	-7.80	Bacteroidetes	Bacteroidaceae	<i>Bacteroides</i>	2.21E-03
	Cluster_21	-3.49	Proteobacteria	Enterobacteriaceae	<i>Escherichia-Shigella</i>	2.21E-03
	Cluster_50	-4.72	Proteobacteria	Enterobacteriaceae	<i>Salmonella</i>	1.26E-03
	Cluster_62	-3.79	Bacteroidetes	Marinifilaceae	<i>Butyricimonas</i>	1.09E-03
	Cluster_145	-12.43	Fusobacteria	Fusobacteriaceae	<i>Fusobacterium</i>	6.81E-04
	Cluster_134	-2.92	Proteobacteria	Desulfovibrionaceae	<i>Desulfovibrio</i>	3.17E-03
	Cluster_172	-3.67	Bacteroidetes	Bacteroidaceae	<i>Bacteroides</i>	2.21E-03
	Cluster_148	-3.60	Bacteroidetes	Bacteroidaceae	<i>Odoribacter</i>	1.07E-03
D10	Cluster_26	-4.09	Firmicutes	Lactobacillaceae	<i>Lactobacillus</i>	1.27E-03

<sup>a</sup>OTUs enriched in HS and LS are shown in dark gray and light gray cells, respectively. The log<sub>2</sub> fold change and corresponding adjusted  $P$  value (P.adj) are reported. Only the most abundant features ( $N_{seq} > 1,000$ ) are reported.

**TABLE 8** MetaCyc pathways presenting differential abundances in feces among the HS and LS shedding classes<sup>a</sup>

Sampling day	MetaCyc pathway	Log <sub>2</sub> fold change	Description	P.adj
D2	TYRFUMCAT-PWY	-3.10	L-tyrosine degradation I	3.469E-03
	P163-PWY	1.89	L-lysine fermentation to acetate and butanoate	6.136E-03
	P281-PWY	3.27	3-phenylpropanoate degradation	3.060E-03
	PWY-6562	5.12	Norspermidine biosynthesis	5.793E-05
D3	GLYCOCAT-PWY	4.60	Glycogen degradation I (bacterial)	9.416E-03
	P162-PWY	1.84	L-lysine fermentation to acetate and butanoate	9.416E-03
	P163-PWY	1.87	Purine nucleobases degradation I (anaerobic)	6.375E-03
	P281-PWY	2.51	Glycolysis V (Pyrococcus)	6.406E-03
	PPGPPMET-PWY	3.04	ppGpp biosynthesis	1.585E-03
	PWY-5837	2.40	Superpathway of menaquinol-8 biosynthesis I	8.554E-03
	PWY-5838	1.86	Superpathway of menaquinol-7 biosynthesis	9.416E-03
	PWY-5840	1.83	Superpathway of menaquinol-9 biosynthesis	9.42E-03
	PWY-5845	1.99	Superpathway of menaquinol-6 biosynthesis I	9.42E-03
	PWY-5850	1.99	Ubiquinol-7 biosynthesis (prokaryotic)	9.42E-03
	PWY-5860	2.13	Superpathway of demethylmenaquinol-8 biosynthesis	9.42E-03
	PWY-5861	2.05	Superpathway of demethylmenaquinol-9 biosynthesis	9.42E-03
	PWY-5862	2.13	Superpathway of phylloquinol biosynthesis	9.42E-03
	PWY-5863	2.36	Superpathway of menaquinol-10 biosynthesis	8.55E-03
	PWY-5896	1.99	Superpathway of menaquinol-11 biosynthesis	9.42E-03
	PWY-5897	1.93	Superpathway of menaquinol-12 biosynthesis	9.42E-03
	PWY-5898	1.93	Superpathway of menaquinol-13 biosynthesis	9.42E-03
	PWY-5899	1.93	Superpathway of geranylgeranyldiphosphate biosynthesis I (via mevalonate)	9.42E-03
	PWY-7013	1.99	Protein N-glycosylation (bacterial)	1.58E-03
	PWY-7328	1.40	Superpathway of UDP-N-acetylglucosamine-derived O-antigen building blocks biosynthesis	9.43E-03
	THREOCAT-PWY	3.78	Superpathway of L-threonine metabolism	7.81E-04
D7	PWY-6892	5.19	Thiazole biosynthesis I ( <i>E. coli</i> )	8.46E-03
D10	P562-PWY	-1.82	Myo-inositol degradation I	3.62E-03
D17	THREOCAT-PWY	2.96	Superpathway of L-threonine metabolism	9.57E-03

<sup>a</sup>Pathways enriched in HS and LS are written in dark gray and light gray, respectively. The log<sub>2</sub> fold change and corresponding adjusted *P* value (*P*.adj) are reported.

well as  $\beta$ -diversity differed according to the origin of the sample (ileum content, cecal content, feces) in accordance with previous results (46). When samples were split by sample type, the influence of the infection on the ordination of the microbiota was observed. *Salmonella* infection significantly modified fecal microbiota composition, as both alpha and beta diversities were modified in the different groups of infected pigs (HS, IS, LS) compared to the control pigs. This result is in line with the data obtained in chickens and pigs (5, 47).

Differences in OTUs suggested that it is possible to distinguish HS, IS, and LS at several time points (i.e., 2 dpi and 3 dpi). This was observed after an analysis performed with FROGS. The overall  $\alpha$ -diversity was thus higher in the LS at 2 dpi, which is consistent with the results showing that low diversity is often correlated with dysbiosis and susceptibility to infection. Moreover, these differences arrived just after the proinflammatory burst. Among the OTUs involved in these differences, we observed an enrichment of *Ruminococcaceae* and of *Cyanobacteria* in LS compared to HS. This result is consistent with those of Argüello et al. (17) who noticed that some members of the *Clostridia* were more abundant and there was an overall relative increase in cellulolytic microbiota (*Ruminococcus* and *Prevotella*) in nonshedders and noninfected pigs during the weaning and growing stages. In contrast, numerous other bacterial genera were significantly increased in HS compared to LS, showing a dramatic modification of fecal microbiota composition in HS. Besides one OTU assigned to *Salmonella* that likely corresponded to the inoculated strain, other *Enterobacteriaceae* such as *Escherichia*, but also several microaerophilic and aero-tolerant taxa (*Lactobacillaceae*, *Pasteurellaceae*) were more abundant in HS compared to LS. This may reflect higher levels of epithelial oxygenation in the HS, even though several anaerobic bacteria are also more abundant in HS (*Fusobacteria* and *Bacteroidetes* families).

In line with these observations, we found that several functional pathways that may be involved in anaerobic respiration were enriched in the HS. Anaerobic respiration is a

metabolic route favoring *Salmonella* over commensal bacteria, which are unable to use host-derived electron acceptors like ROS (48). For example, at 3 dpi, numerous superpathways related to menaquinol biosynthesis are induced in HS compared to LS. Menaquinone biosynthesis pathways are essential in electron transport and ATP generation in anaerobically respiring Gram-negative bacteria, which have been described as a major way of *Enterobacteriaceae* like *Salmonella* to overgrow commensal bacteria, which used fermentation (49). Anaerobic respiration may be induced in HS by using oxygen as an electron acceptor. Moreover, electron acceptors other than molecular oxygen (O<sub>2</sub>), such as nitrate (NO<sub>3</sub><sup>-</sup>), fumarate, sulfate (SO<sub>4</sub><sup>2-</sup>), or sulfur (S) can be used. We also observed that glycogen degradation pathway is induced in HS compared to LS. Glycogen metabolism in numerous bacteria can be derived from a common pathway and glycogen reserves enable survival when energy consumption is low. The endogenous glucose polymer, glycogen, appears to play an important role in colonization, since mutants that are unable to synthesize or degrade glycogen have significant colonization defects. In support of the hypothesis that *E. coli* relies on internal carbon stores to maintain colonization during periods of famine, Jones et al. (50) found that by providing a constant supply of a readily metabolized sugar in the animal's drinking water, the competitive disadvantage of *E. coli* glycogen metabolism mutants is rescued. In line with this, McMeechan et al. (51) found that the glycogen production plays a role in the survival of *Salmonella* during intestine colonization. An important implication of these observations is that in HS, the relatively high abundance of *Enterobacteriaceae*, including *Salmonella* and *E. coli*, was related to a high level of glycogen use as the primary carbon and energy storage molecule. Two pathways related to amino acid degradation and fermentation, namely, "L-lysine fermentation to acetate and butanoate" and the pathway "superpathway of L-threonine metabolism," were also enriched in the HS (the second being chimeric and including catabolic pathways). Not all amino acids are equally suitable for fermentation, and differences in bacterial growth are observed when single amino acids are compared as sole energy sources in culture media. The highest growth is observed from catabolism of glutamate, arginine, glycine, serine, phenylalanine, and tyrosine (52). Recently, evidence has emerged that aromatic amino acids (phenylalanine, tyrosine, and tryptophan) can be fermented to phenylpropionic acid metabolites, phenylacetic acid, and 4-hydroxyphenyl-acetic acid, which are abundant in feces. The organisms involved include several species of *Bacteroides*, *Eubacterium hallii*, and *Clostridium barlettii* (53). These observations may explain the enrichment of the two pathways related to amino acid degradation and fermentation, insofar as several OTUs assigned to *Bacteroides* were found enriched in the HS.

In contrast, only few pathways were found enriched in the LS compared to the HS. This included the "tyrosine degradation pathway." This observation was consistent with lower levels of *Salmonella* and *E. coli* in LS, given that this pathway is absent in these two taxa.

In our study, the role of short-chain fatty acid (SCFA) production remained unclear, insofar as we expected to observe low levels of SCFA production in the HS. SCFAs, such as butyrate, serve as a major source of energy for the intestinal enterocytes and signals to the host. Moreover, there is considerable interest in these butyrogens since butyrate reduces proinflammatory signals and has a protective role in maintaining a healthy colon. Butyrogenic gut bacteria belong to *Lachnospiraceae* and *Ruminococcaceae* (54). Consistent with the protective activity of SCFA, *Ruminococcaceae* were more abundant in LS than in HS (Table 7). However, the *Lachnospiraceae* were more abundant in HS than in LS. Moreover, most butyrate produced in the human intestine is assumed to derive from carbohydrates, and in our MetaCyc pathways, only the "L-lysine fermentation to acetate and butanoate" pathway was differentially detected. This point requires further investigation, including direct quantification of SCFA production at every time point in a similar pig model.

**Conclusion.** Overall, this study demonstrated that when pigs are orally inoculated with *Salmonella*, heterogeneity in *Salmonella* shedding is observed. Hierarchical clustering on AULC permitted to clustering of pigs into three groups: high (HS), intermediate (IS) and low shedder (LS) pigs. The HS pigs had a higher *Salmonella* shedding level and a higher level of organ colonization, and they grew significantly slower than the control and LS pigs.

Our analysis of immune cell counts and gut microbiota composition did not allow us to identify clear predictive biomarkers for these shedding phenotypes. Interestingly, we observed a higher Interleukin-10 gene expression in the future HS pigs at two time points before infection. Interleukin 10 is a cytokine with potent antiinflammatory properties that plays a central role in limiting host immune response to pathogens. *Salmonella* inoculation induces a profound and rapid modification of the immune response, which comes just before the modification of the fecal microbiota composition. The immune response is higher in HS than in LS and therefore does not appear to be a protective response. Furthermore, the strong proinflammatory response observed in gene expression and cytokine production at 1 dpi could be responsible of the high *Salmonella* levels. This is in line with the results obtained in mouse models where the proinflammatory response tends to provide a competitive advantage to *Salmonella* compared to other bacteria of the gut microbiota. This hypothesis is supported by the analysis of the gut microbiota. *Salmonella* infection alters both alpha and beta diversity and many bacterial taxa are differentially abundant between the HS and LS. Furthermore, functional enrichment analysis reveals that strong differences between the three shedding classes are observed on day 2, 3, and 17 postinfection. Modification in bacterial taxa and functions strongly supports the idea that many bacteria in HS use anaerobic respiration rather than fermentation, which is favored in LS. This could of course be related to the utilization and adaption to the inflammatory response detected in the HS. Again, this would be clearly in line with the hypotheses put forward in mouse model studies.

## MATERIALS AND METHODS

**Experimental infection: study design and animals.** The experiment was performed at Anses BSL3 facilities, which have an agreement for animal experimentation delivered by the Direction Départementale de la Protection des Populations des Côtes d'Armor (Departmental Directorate for Protection of the Population; Anses registration no. C-22-745-1). The *in vivo* trial protocol was approved by the Ethics Committee on Animal Research no. 16, of the French Ministry of Research (APAFiS license no. 16570-2018083015249240). Anses, Inrae and University of Surrey approved the design of the animal experiment.

The trial was conducted using 45 specific-pathogen-free (SPF) Large White piglets born at Anses Ploufragan's protected animal facilities. The piglets were weaned when they were 4 weeks old and were distributed into 5 units: one unit with 5 piglets in the same pen (control group), and four units divided in two pens, each with 10 piglets (5 per pen). Those 40 pigs represent the inoculated group. The 45 piglets were born from 5 sows and in each pen, piglets were distributed to avoid a maternal effect between the pens. Moreover, piglets were distributed following a random-cluster criteria balancing by sex and weight.

Seven days before inoculation with *Salmonella*, the 45 piglets were tested and confirmed to be free of *Salmonella*. Fecal samples from each piglet were analyzed in accordance with the NF U47-102 method (55). On day 0, at 7 weeks age, the 40 piglets were inoculated with a monophasic variant of *Salmonella* Typhimurium strain, as previously described (56). Briefly, piglets were orally inoculated with 10 mL of solution containing a bacterial solution of  $10^8$  CFU per mL in tryptone salt broth (TSB) (bioMérieux) using a cannula connected to a screw syringe. The five control piglets received 10 mL of sterile TSB broth.

**Monitoring and sample collection during experimental infection.** During the experimental trial, rectal temperatures and clinical signs were monitored daily, whereas weight and food intake were recorded weekly. Rectal temperatures over 40.0°C indicated a pyrexia.

Three tubes of blood were collected at the jugular vein before infection at day 20 and 7 before inoculation (dbi) and then at 0, 1, 3, 7, 10, 14, 17, and 21 days postinoculation (dpi). One EDTA-stabilized tube was intended to measure total blood cells directly after sampling; another tube to determine anti-*Salmonella* antibody levels and cytokines in serum. These tubes were centrifuged at  $2300 \times g$  for 5 min and sera were stored at  $-20^\circ\text{C}$  until analyses.

Blood was also sampled in a third EDTA-stabilized tube for gene expression analysis. Blood was immediately suspended in RNA later (Qiagen) and stored at  $-80^\circ\text{C}$  until tested.

Feces (at least 30 g) were collected *per rectum* 7 days before inoculation and at day 1, 2, 3, 10, 14, 17, and 21 dpi for the inoculated groups and 7, 14, and 21 dpi for the control group for *Salmonella* detection and enumeration.

At 21 dpi, piglets were euthanized by intravenous injection with an overdose of tiletamine and zolazepam (Zoletil 100). The subsequent necropsies primarily examined organs and tissues in the abdominal and thoracic cavities. The tonsils, mesenteric lymph nodes (MLNs), and intestinal contents of the ileum and cecum were collected for *Salmonella* detection and enumeration.

The fecal samples and the intestinal contents of the ileum and cecum taken during the postmortem examination were immediately stored and transported on ice to the laboratory where 1 g of each sample was rapidly recovered in a cryotube and stored at  $-80^\circ\text{C}$  until used for fecal gut microbiota analysis. Enumeration and detection of *Salmonella* was performed on 25 g of each sample.

**Enumeration and detection of *Salmonella* during experimental infection.** Enumeration of *Salmonella* in samples were performed as previously described by Cevallos et al. (56). If enumerations for

*Salmonella* spp. were negative (below the enumeration limit; 10 CFU/g), the NF U47-102 method was used to confirm the presence or absence of *Salmonella* in the sample. Absence of *Salmonella* excretion for control piglets was checked throughout the assay.

To evaluate total *Salmonella* shedding for each pig over the course of the study, the CFU/g in feces from each sampling point was log-normalized and plotted to calculate the area under the log curve (AULC) (19, 26). Then, a hierarchical clustering (function hclust) was performed on AULC data to cluster pigs in three classes according to their level of *Salmonella* shedding (high, intermediate, and low shedders, respectively, named HS, IS, and LS). A variance analysis (function lm) and a Newman and Keuls test (function SNK.test, library agricolae), were performed to confirm that these three classes were significantly different.

For the data on level of *Salmonella* obtained from necropsy, we compared the contamination levels of each group at the two dates of necropsy with a Kruskal-Wallis test ( $P < 0.05$ ). The *Salmonella* shedding levels at each sampling day after infection were also compared using the Kruskal-Wallis test ( $P < 0.05$ ). All the statistical analyses were performed using R software (R version 4.0.4).

**Evaluation of the immune response. (i) Total blood cell counts.** Blood cell counts were performed directly after sampling from EDTA blood samples using an MS9.5 hematology analyzer (Melet Schloesing Laboratoires). Thirty parameters per sample were obtained (see Table S1 in the supplemental material).

**(ii) Level of *Salmonella* antibodies.** For antibody screening, an IDEXX Swine *Salmonella* Ab test (IDEXX) was used following the manufacturer protocol. The latter test has a sensitivity of 99.1% and a specificity of 99.4%. The presence or absence of antibodies to *Salmonella* in the sample was determined by calculating the S/P ratio corresponding to the absorbance value at 650 nm of the sample (S) over the mean absorbance value of the positive control (P). The results were expressed as a percentage of optical density (OD%). Seroconversion was defined as previously described by Cevalos et al. in 2019 (57) and according to the manufacturer's recommendations.

**(iii) Level of cytokines in serum.** Porcine cytokines were quantified using the Elisa kit from Bio-technique for TNF- $\alpha$ , IL-1 $\beta$ , and IL-6, (Bio-technique R&D system) and using the Elisa kit from Invitrogen for and IFN- $\gamma$  (FisherScientific). For these 3 sets of data, statistical analyses of results were performed with R software (R version 4.0.4).

**(iv) RNA extraction for transcriptomic immune response studies.** After thawing, 600  $\mu$ L of each blood sample was centrifuged, the supernatant was discarded and the NucleoSpin RNA kit (Macherey-Nagel) was used to purify the RNA as described by the manufacturer. The RNA concentration was measured using the ND-1000 Nanodrop spectrophotometer. Two hundred ng of total RNA, 0.25  $\mu$ g of Oligo-d(T<sub>20</sub>) (Eurogentec), and 0.25  $\mu$ g of random primer (Promega) were denatured at 75°C for 5 min then incubated on ice for 5 min. The reverse transcription reaction was carried out in a final volume of 25  $\mu$ L containing 1 mM dNTP, 30 U/ $\mu$ g of RNA of AMV reverse transcriptase (Promega), and 1 U/ $\mu$ L of RNasin (Promega), at 42°C for 60 min, then at 94°C for 5 min. After reverse transcription, cDNAs were purified using the QIAquick PCR purification kit (Qiagen). cDNA yield was measured using the ND-1000 Nanodrop spectrophotometer, and concentrations were adjusted to 50 ng/ $\mu$ L.

**(v) Quantitative RT PCR and transcriptomic immune response analysis.** Seventy primer pairs were used in the study to compare immune gene expression in HS and LS pigs; their sequences are indicated in Table S2. BioMark, a high-throughput PCR device from Fluidigm, was used to perform the qPCR according to manufacturer's recommendations. A total of 63 ng of purified cDNA was amplified using the pooled primers to a final concentration of 50 nM and a thermal cycling consisting of 5 min at 95°C, followed by 18 cycles of 15 s at 95°C, and 4 min at 60°C, and a final holding step at 4°C. Thereafter, the protocol recommended by the supplier was followed (Fluidigm quick references PN 100 to 5875 B1, and PN 100 to 9791 B1). The software Fluidigm real-time PCR analysis was used to determine the Cq values of each sample/primer pair couple. The Cq were determined by the auto detector method with a quality control of 0.65 and a linear baseline correction. The fold change (FC) in gene expression between HS and LS or between control and infected pigs was calculated by the 2<sup>- $\Delta\Delta$ Cq</sup> method (58). An unpaired *t* test was used to determine the statistically significantly differentially expressed (DE) genes. The fold changes (FC) greater than 2 and less than -2, with a *P* value < 0.05 were considered for further analyses.

**16S rRNA microbial profiling using next-generation sequencing. (i) Sample size and metadata collection.** In total, 458 samples were collected, including fecal ( $n = 369$ ) and intestinal tract samples (cecum,  $n = 45$  and ileum,  $n = 44$ ).

Metadata consisted of animal ID, animal dam ID, sex, sampling dates before and after inoculation, sample type (feces, ileum, cecum contents), *Salmonella* level (log<sub>10</sub>CFU/g), and shedding classes (LS, IS, HS) of each pig.

All samples were collected in sterile centrifuge tubes containing 50% glycerol and stored on cool packs until moved to -80°C. Samples remained frozen until the time of DNA extraction.

**(ii) DNA extraction and 16S sequencing.** Total DNA was extracted using the DNeasy PowerSoil kit (Qiagen) following the manufacturer's instructions, and 16S (V3-V4 region) amplification was performed using the following primers Illumina\_16S\_341F (5'-TCGTCGGCAGCGTCAGATGTGTATAAGAGACAGCC TACGGGNGGCWGCAG) and Illumina\_16S\_805R (5'-GTCTCTGGGCTCGGAGATGTGTATAAGAGACAGGAC TACHVGGGTATCTAATC) as described by Zheng et al. in 2015 (59). 16S rRNA amplicons were sent to the Earlham Institute (UK) where sequencing was performed using an Illumina MiSeq sequencer obtaining 300-bp paired-end raw reads.

**(iii) Pig gut microbiota analysis using Qiime pipeline.** Raw sequences were analyzed using Qiime2 (version 2019.10.0) software (60, 61). Briefly, raw sequences were demultiplexed (Demux Qiime tool), trimmed and low-quality sequences were discarded (Dada2 Qiime tool option -p-trim-left 15 -p-trunc-len 220 (62)). These sequences were used to generate a MAFFT tree (63) and used to construct a phylogeny with Fasttree2 (64) that was referenced against the Greengene database (65) to obtain

operational taxonomic units (OTUs) that can be clustered according to the available metadata. OTU composition was analyzed using a variety of different statistical analyses and data visualization available within Qiime2 software: Alpha rarefaction (shows the OTUs number/sequencing depth); Alpha distribution (OTUs diversity within each group); Beta distribution (OTUs diversity between each group); and OTUs boxplot and principal component analysis (PCA), which can be calculated considering the presence/absence of OTUs Jaccard) or considering the OTUs relative abundance (UniFrac based). OTU tables were also imported in Orange3 statistical software (66) to explore different statistical analyses to test for associations between *Salmonella* shedding status and changes in the gut microbiota such as K-means and MDS. Metadata were validated to be suitable for the use in Qiime2 by using Keemei Google Docs plug-in (67).

**(iv) Pig gut microbiota analysis using FROGS pipeline.** Data were uploaded on FROGS analysis pipeline (32) which was used for all further steps of gut microbiota characterization. First, paired-end reads from each sample were clustered by allowing a mismatch rate of 0.1. They were next selected using an expected read size of 300 bp, and a total amplicon size ranging from 400 bp to 600 bp with a mean of 460 bp. The resulting sequences were clustered using Swarm (68). For this, we used 1 and 1 as the values of aggregation distance parameters (for the denoising and final clustering steps, respectively). OTUs, including chimeric sequences, were then removed using VSearch (69). Other quality control steps included removal of very rare OTUs (relative abundance < 0.0005% of the total read numbers) and those including sequences matching phiX sequences recorded in a specific data bank (32). Finally, the resulting OTUs were classified using an NCBI BLAST+ search within the Silva SSU 123 database (70, 71).

Diversity assessment was based on the Chao1, Shannon  $\alpha$ -diversity indexes and the Bray-Curtis  $\beta$ -diversity index. Alpha diversity within each shedding category were compared using one-way ANOVA, computed using the dedicated native function of R environment (72). Comparison of  $\beta$ -diversity indexes were conducted using PERMANOVA tests performed using the `adonis()` function of the R-package Vegan (73); bacterial counts were fitted onto the  $\beta$ -diversity ordination using the `envfit()` function of the R-package Vegan. Differential abundances were assessed following the hypothesis that abundances in each sample followed negative binomial distributions. Under this scheme, the abundance may be modeled by fitting a generalized linear model. Significant logarithmic fold change ratios were detected using Wald tests and Benjamini-Hochberg adjustment for multiple testing ( $P < 0.01$ ). The computations were performed using the R-package DESeq2 (74).

**(v) Functional metagenomic predictions based on pig gut microbiota analysis.** Functional gene families and MetaCyc pathways were predicted using the PICRUSt2 package (75). MetaCyc pathways were aggregated at the superpathway level using the MetaCyc database (76).

## SUPPLEMENTAL MATERIAL

Supplemental material is available online only.

**TABLE S1**, DOCX file, 0.02 MB.

**TABLE S2**, DOCX file, 0.03 MB.

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## REFERENCES

1. EFSA, ECDC. 2019. The European Union One Health 2018 Zoonoses Report. EFSA J 17:276.
2. Gopinath S, Lichtman JS, Bouley DM, Elias JE, Monack DM. 2014. Role of disease-associated tolerance in infectious superspreaders. Proc Natl Acad Sci U S A 111:15780–15785. <https://doi.org/10.1073/pnas.1409968111>.
3. Lawley TD, Bouley DM, Hoy YE, Gerke C, Relman DA, Monack DM. 2008. Host transmission of *Salmonella enterica* serovar Typhimurium is controlled by virulence factors and indigenous intestinal microbiota. Infect Immun 76:403–416. <https://doi.org/10.1128/IAI.01189-07>.
4. Menanteau P, Kempf F, Trotureau J, Virlogeux-Payant I, Gitton E, Dalifard J, Gabriel I, Rychlik I, Velge P. 2018. Role of systemic infection, cross contaminations and super-shedders in Salmonella carrier state in chicken. Environ Microbiol 20:3246–3260. <https://doi.org/10.1111/1462-2920.14294>.
5. Bearson SMD, Allen HK, Bearson BL, Looft T, Brunelle BW, Kich JD, Tuggle CK, Bayles DO, Alt D, Levine UY, Stanton TB. 2013. Profiling the gastrointestinal microbiota in response to Salmonella: low versus high Salmonella shedding in the natural porcine host. Infect Genet Evol 16:330–340. <https://doi.org/10.1016/j.meegid.2013.03.022>.
6. Bonardi S. 2017. Salmonella in the pork production chain and its impact on human health in the European Union. Epidemiol Infect 145:1513–1526. <https://doi.org/10.1017/S095026881700036X>.

7. Rostagno MH, Callaway TR. 2012. Pre-harvest risk factors for *Salmonella enterica* in pork production. *Food Res International* 45:634–640. <https://doi.org/10.1016/j.foodres.2011.04.041>.
8. Ivanek R, Osterberg J, Gautam R, Sternberg Lewerin S. 2012. Salmonella fecal shedding and immune responses are dose- and serotype- dependent in pigs. *PLoS One* 7:e34660. <https://doi.org/10.1371/journal.pone.0034660>.
9. Nollet N, Houf K, Dewulf J, De Kruif A, De Zutter L, Maes D. 2005. Salmonella in sows: a longitudinal study in farrow-to-finish pig herds. *Vet Res* 36:645–656. <https://doi.org/10.1051/vetres:2005022>.
10. Massacci FR, Morelli A, Cucco L, Castinel A, Ortenzi R, Tofani S, Pezzotti G, Estellé J, Panicià M, Magistrali CF. 2020. Transport to the slaughterhouse affects the salmonella shedding and modifies the fecal microbiota of finishing pigs. *Animals (Basel)* 10:676. <https://doi.org/10.3390/ani10040676>.
11. D'Incau M, Salogni C, Giovannini S, Ruggeri J, Scali F, Tonni M, Formenti N, Guarneri F, Pasquali P, Alborali GL. 2021. Occurrence of Salmonella Typhimurium and its monophasic variant (4, [5], 12:i:-) in healthy and clinically ill pigs in northern Italy. *Porcine Health Manag* 7:34. <https://doi.org/10.1186/s40813-021-00214-1>.
12. Enekave E, Hong S, Mather AE, Boxrud D, Taylor AJ, Lappi V, Johnson TJ, Vannucci F, Davies P, Hedberg C, Perez A, Alvarez J. 2018. Salmonella enterica serotype 4,[5],12:i:- in Swine in the United States Midwest: an emerging multidrug-resistant clade. *Clin Infect Dis* 66:877–885. <https://doi.org/10.1093/cid/cix909>.
13. Garrido V, Migura-García L, Gaitán I, Arrieta-Gisasola A, Martínez-Ballesteros I, Fraile L, Grilló MJ. 2021. Prevalence of Salmonella in free-range pigs: risk factors and intestinal microbiota composition. *Foods (Basel, Switzerland)* 10:1410. <https://doi.org/10.3390/foods10061410>.
14. Sun H, Wan Y, Du P, Bai L. 2020. The epidemiology of monophasic Salmonella Typhimurium. *Foodborne Pathog Dis* 17:87–97. <https://doi.org/10.1089/fpd.2019.2676>.
15. Hurd HS, Gailey JK, McKean JD, Rostagno MH. 2001. Rapid infection in market-weight swine following exposure to a Salmonella Typhimurium-contaminated environment. *Am J Vet Res* 62:1194–1197. <https://doi.org/10.2460/ajvr.2001.62.1194>.
16. Martins RP, Collado-Romero M, Arce C, Lucena C, Carvajal A, Garrido JJ. 2013. Exploring the immune response of porcine mesenteric lymph nodes to Salmonella enterica serovar Typhimurium: an analysis of transcriptional changes, morphological alterations and pathogen burden. *Comp Immunol Microbiol Infect Dis* 36:149–160. <https://doi.org/10.1016/j.cimid.2012.11.003>.
17. Argüello H, Estellé J, Leonard FC, Crispie F, Cotter PD, O'Sullivan O, Lynch H, Walia K, Duffy G, Lawlor PG, Gardiner GE. 2019. Influence of the Intestinal Microbiota on Colonization Resistance to Salmonella and the Shedding Pattern of Naturally Exposed Pigs. *mSystems* 4. <https://doi.org/10.1128/mSystems.00021-19>.
18. Argüello H, Estellé J, Zaldívar-López S, Jiménez-Marín Á, Carvajal A, López-Bascón MA, Crispie F, O'Sullivan O, Cotter PD, Priego-Capote F, Morera L, Garrido JJ. 2018. Early Salmonella Typhimurium infection in pigs disrupts Microbiome composition and functionality principally at the ileum mucosa. *Sci Rep* 8:7788. <https://doi.org/10.1038/s41598-018-26083-3>.
19. Huang T-H, Uthe JJ, Bearson SMD, Demirkale CY, Nettleton D, Knetter S, Christian C, Ramer-Tait AE, Wannemuehler MJ, Tuggle CK. 2011. Distinct peripheral blood RNA responses to Salmonella in pigs differing in Salmonella shedding levels: intersection of IFNG, TLR and miRNA pathways. *PLoS One* 6:e28768. <https://doi.org/10.1371/journal.pone.0028768>.
20. Uthe JJ, Bearson SMD, Qu L, Dekkers JC, Nettleton D, Rodriguez Torres Y, O'Connor AM, McKean JD, Tuggle CK. 2011. Integrating comparative expression profiling data and association of SNPs with Salmonella shedding for improved food safety and porcine disease resistance. *Anim Genet* 42:521–534. <https://doi.org/10.1111/j.1365-2052.2010.02171.x>.
21. Huang T, Huang X, Shi B, Wang F, Feng W, Yao M. 2018. Regulators of Salmonella-host interaction identified by peripheral blood transcriptome profiling: roles of TGFβ1 and TRP53 in intracellular Salmonella replication in pigs. *Vet Res* 49:121. <https://doi.org/10.1186/s13567-018-0616-9>.
22. Meurens F, Berri M, Auray G, Melo S, Levast B, Virlogeux-Payant I, Chevalyre C, Gerdtts V, Salmon H. 2009. Early immune response following Salmonella enterica subspecies enterica serovar Typhimurium infection in porcine jejunal gut loops. *Vet Res* 40:5. <https://doi.org/10.1051/vetres:2008043>.
23. Collado-Romero M, Martins RP, Arce C, Moreno Á, Lucena C, Carvajal A, Garrido JJ. 2012. An in vivo proteomic study of the interaction between Salmonella Typhimurium and porcine ileum mucosa. *J Proteomics* 75: 2015–2026. <https://doi.org/10.1016/j.jprot.2012.01.001>.
24. Martins RP, Lorenzi V, Arce C, Lucena C, Carvajal A, Garrido JJ. 2013. Innate and adaptive immune mechanisms are effectively induced in ileal Peyer's patches of Salmonella Typhimurium infected pigs. *Dev Comp Immunol* 41:100–104. <https://doi.org/10.1016/j.dci.2013.04.020>.
25. Uthe JJ, Wang Y, Qu L, Nettleton D, Tuggle CK, Bearson SM. 2009. Correlating blood immune parameters and a CCT7 genetic variant with the shedding of Salmonella enterica serovar Typhimurium in swine. *Vet Microbiol* 135:384–388. <https://doi.org/10.1016/j.vetmic.2008.09.074>.
26. Knetter SM, Bearson SMD, Huang T-H, Kurkiewicz D, Schroyen M, Nettleton D, Berman D, Cohen V, Lunney JK, Ramer-Tait AE, Wannemuehler MJ, Tuggle CK. 2015. Salmonella enterica serovar Typhimurium-infected pigs with different shedding levels exhibit distinct clinical, peripheral cytokine and transcriptomic immune response phenotypes. *Innate Immun* 21: 227–241. <https://doi.org/10.1177/1753425914525812>.
27. Kommadath A, Bao H, Arantes AS, Plastow GS, Tuggle CK, Bearson SMD, Guan LL, Stothard P. 2014. Gene co-expression network analysis identifies porcine genes associated with variation in Salmonella shedding. *BMC Genomics* 15:452. <https://doi.org/10.1186/1471-2164-15-452>.
28. Drumo R, Pesciaroli M, Ruggeri J, Tarantino M, Chirullo B, Pistoia C, Petrucci P, Martinelli N, Moscatti L, Manuali E, Pavone S, Picciolini M, Ammendola S, Gabai G, Battistoni A, Pezzotti G, Alborali GL, Napolioni V, Pasquali P, Magistrali CF. 2015. Salmonella enterica Serovar Typhimurium exploits inflammation to modify swine intestinal microbiota. *Front Cell Infect Microbiol* 5:106.
29. Barba-Vidal E, Castillejos L, Roll VFB, Cifuentes-Orjuela G, Moreno Muñoz JA, Martín-Orúe SM. 2017. The Probiotic Combination of *Bifidobacterium longum* subsp. *infantis* CECT 7210 and *Bifidobacterium animalis* subsp. *lactis* BPL6 Reduces Pathogen Loads and Improves Gut Health of Weaned Piglets Orally Challenged with Salmonella Typhimurium. *Front Microbiol* 8:1570. <https://doi.org/10.3389/fmicb.2017.01570>.
30. Nathan C, Ding A. 2010. Nonresolving inflammation. *Cell* 140:871–882. <https://doi.org/10.1016/j.cell.2010.02.029>.
31. Snel B, Lehmann G, Bork P, Huynen MA. 2000. STRING: a web-server to retrieve and display the repeatedly occurring neighbourhood of a gene. *Nucleic Acids Res* 28:3442–3444. <https://doi.org/10.1093/nar/28.18.3442>.
32. Escudé F, Auer L, Bernard M, Mariadassou M, Cauquil L, Vidal K, Maman S, Hernandez-Raquet G, Combes S, Pascal G. 2018. FROGS: find, Rapidly, OTUs with Galaxy Solution. *Bioinformatics (Oxford, England)* 34:1287–1294. <https://doi.org/10.1093/bioinformatics/btx791>.
33. Kim HB, Isaacson RE. 2017. Salmonella in swine: microbiota interactions. *Annu Rev Anim Biosci* 5:43–63. <https://doi.org/10.1146/annurev-animal-022516-022834>.
34. Wang Z, Choice E, Kaspar A, Hanson D, Okada S, Lyu SC, Krensky AM, Clayberger C. 2000. Bactericidal and tumoricidal activities of synthetic peptides derived from granulysin. *J Immunol* 165:1486–1490. <https://doi.org/10.4049/jimmunol.165.3.1486>.
35. Bearson BL. 2016. Molecular profiling: catecholamine modulation of gene expression in *Escherichia coli* O157:H7 and Salmonella enterica Serovar Typhimurium. *Adv Exp Med Biol* 874:167–182. [https://doi.org/10.1007/978-3-319-20215-0\\_7](https://doi.org/10.1007/978-3-319-20215-0_7).
36. Uthe JJ, Royae A, Lunney JK, Stabel TJ, Zhao S-H, Tuggle CK, Bearson SMD. 2007. Porcine differential gene expression in response to Salmonella enterica serovars Choleraesuis and Typhimurium. *Mol Immunol* 44: 2900–2914. <https://doi.org/10.1016/j.molimm.2007.01.016>.
37. Ornatowska M, Azim AC, Wang X, Christman JW, Xiao L, Joo M, Sadikot RT. 2007. Functional genomics of silencing TREM-1 on TLR4 signaling in macrophages. *Am J Physiol Lung Cell Mol Physiol* 293:L1377–84. <https://doi.org/10.1152/ajplung.00140.2007>.
38. Gross O, Gewies A, Finger K, Schäfer M, Sparwasser T, Peschel C, Förster I, Ruland J. 2006. Card9 controls a non-TLR signalling pathway for innate antifungal immunity. *Nature* 442:651–656. <https://doi.org/10.1038/nature04926>.
39. Wu H, Gong J, Liu Y. 2018. Indoleamine 2, 3-dioxygenase regulation of immune response (Review). *Mol Med Rep* 17:4867–4873.
40. Rottenberg ME, Carow B. 2014. SOCS3 and STAT3, major controllers of the outcome of infection with Mycobacterium tuberculosis. *Semin Immunol* 26:518–532. <https://doi.org/10.1016/j.smim.2014.10.004>.
41. Litvak Y, Mon KKZ, Nguyen H, Chanthavixay G, Liou M, Velazquez EM, Kutter L, Alcantara MA, Byndloss MX, Tiffany CR, Walker GT, Faber F, Zhu Y, Bronner DN, Byndloss AJ, Tsois RM, Zhou H, Bäumlér AJ. 2019. Commensal Enterobacteriaceae protect against salmonella colonization through oxygen competition. *Cell Host Microbe* 25:128–139.e5. <https://doi.org/10.1016/j.chom.2018.12.003>.
42. Faber F, Thiennimitr P, Spiga L, Byndloss MX, Litvak Y, Lawhon S, Andrews-Polymenis HL, Winter SE, Bäumlér AJ. 2017. Respiration of

- Microbiota-Derived 1,2-propanediol drives *Salmonella* expansion during Colitis. *PLoS Pathog* 13:e1006129. <https://doi.org/10.1371/journal.ppat.1006129>.
43. Lopez CA, Rivera-Chávez F, Byndloss MX, Bäumlér AJ. 2015. the periplasmic nitrate reductase NapABC supports luminal growth of *Salmonella enterica* Serovar Typhimurium during colitis. *Infect Immun* 83:3470–3478. <https://doi.org/10.1128/IAI.00351-15>.
  44. Lopez CA, Winter SE, Rivera-Chávez F, Xavier MN, Poon V, Nuccio S-P, Tsois RM, Bäumlér AJ. 2012. Phage-mediated acquisition of a type III secreted effector protein boosts growth of salmonella by nitrate respiration. *mBio* 3. <https://doi.org/10.1128/mBio.00143-12>.
  45. Winter SE, Thiennimitr P, Winter MG, Butler BP, Huseby DL, Crawford RW, Russell JM, Bevins CL, Adams LG, Tsois RM, Roth JR, Bäumlér AJ. 2010. Gut inflammation provides a respiratory electron acceptor for *Salmonella*. *Nature* 467:426–429. <https://doi.org/10.1038/nature09415>.
  46. Zhao W, Wang Y, Liu S, Huang J, Zhai Z, He C, Ding J, Wang J, Wang H, Fan W, Zhao J, Meng H. 2015. The dynamic distribution of porcine microbiota across different ages and gastrointestinal tract segments. *PLoS One* 10:e0117441. <https://doi.org/10.1371/journal.pone.0117441>.
  47. Kempf F, Menanteau P, Rychlik I, Kubasová T, Trotureau J, Virlogeux-Payant I, Schaeffer S, Schouler C, Drumo R, Guitten E, Velge P. 2020. Gut microbiota composition before infection determines the *Salmonella* super- and low-shedder phenotypes in chicken. *Microb Biotechnol* 13:1611–1630. <https://doi.org/10.1111/1751-7915.13621>.
  48. Winter SE, Bäumlér AJ. 2011. A breathtaking feat: to compete with the gut microbiota, *Salmonella* drives its host to provide a respiratory electron acceptor. *Gut Microbes* 2:58–60. <https://doi.org/10.4161/gmic.2.1.14911>.
  49. Rivera-Chávez F, Bäumlér AJ. 2015. The pyromaniac inside you: salmonella metabolism in the host gut. *Annu Rev Microbiol* 69:31–48. <https://doi.org/10.1146/annurev-micro-091014-104108>.
  50. Jones SA, Jorgensen M, Chowdhury FZ, Rodgers R, Hartline J, Leatham MP, Struve C, Krogfelt KA, Cohen PS, Conway T. 2008. Glycogen and maltose utilization by *Escherichia coli* O157:H7 in the mouse intestine. *Infect Immun* 76:2531–2540. <https://doi.org/10.1128/IAI.00096-08>.
  51. McMeechan A, Lovell MA, Cogan TA, Marston KL, Humphrey TJ, Barrow PA. 2005. Glycogen production by different *Salmonella enterica* serotypes: contribution of functional glgC to virulence, intestinal colonization and environmental survival. *Microbiology (Reading, Engl)* 151:3969–3977. <https://doi.org/10.1099/mic.0.28292-0>.
  52. Smith EA, Macfarlane GT. 1998. Enumeration of amino acid fermenting bacteria in the human large intestine: effects of pH and starch on peptide metabolism and dissimilation of amino acids. *FEMS Microbiol Ecol* 25: 355–368. <https://doi.org/10.1111/j.1574-6941.1998.tb00487.x>.
  53. Rowland I, Gibson G, Heinken A, Scott K, Swann J, Thiele I, Tuohy K. 2018. Gut microbiota functions: metabolism of nutrients and other food components. *Eur J Nutr* 57:1–24. <https://doi.org/10.1007/s00394-017-1445-8>.
  54. Bui TP, Ritari J, Boeren S, de Waard P, Plugge CM, de Vos WM. 2015. Production of butyrate from lysine and the Amadori product fructoselysine by a human gut commensal. *Nat Commun* 6:10062. <https://doi.org/10.1038/ncomms10062>.
  55. AFNOR. NF U47-102, Animal health analysis methods — isolation and identification of any *Salmonella* serotype or of specified *Salmonella* serotypes among mammals. France 2008.
  56. Cevallos-Almeida M, Houdayer C, Rose V, Bailly Y, Paboeuf F, Fablet C, Denis M, Kerouanton A. 2018. Colonization of pigs experimentally infected with a monophasic variant of *Salmonella* Typhimurium. *Foodborne Pathog Dis* 15:576–582. <https://doi.org/10.1089/fpd.2018.2427>.
  57. Cevallos-Almeida M, Fablet C, Houdayer C, Dorenlor V, Eono F, Denis M, Kerouanton A. 2019. Longitudinal study describing time to *Salmonella* seroconversion in piglets on three farrow-to-finish farms. *Vet Rec Open* 6:e000287.
  58. Livak KJ, Schmittgen TD. 2001. Analysis of relative gene expression data using real-time quantitative PCR and the 2(-Delta Delta C(T)) Method. *Methods* 25:402–408. <https://doi.org/10.1006/meth.2001.1262>.
  59. Zheng W, Tsompana M, Ruscitto A, Sharma A, Genco R, Sun Y, Buck MJ. 2015. An accurate and efficient experimental approach for characterization of the complex oral microbiota. *Microbiome* 3:48. <https://doi.org/10.1186/s40168-015-0110-9>.
  60. Bolyen E, Rideout JR, Dillon MR, Bokulich NA, Abnet CC, Al-Ghalith GA, Alexander H, Alm EJ, Arumugam M, Asnicar F, Bai Y, Bisanz JE, Bittinger K, Brejnrod A, Brislawn CJ, Brown CT, Callahan BJ, Caraballo-Rodríguez AM, Chase J, Cope EK, Da Silva R, Diener C, Dorrestein PC, Douglas GM, Durall DM, Duvallet C, Edwardson CF, Ernst M, Estaki M, Fouquier J, Gauglitz JM, Gibbons SM, Gibson DL, Gonzalez A, Gorlick K, Guo J, Hillmann B, Holmes S, Holste H, Huttenhower C, Huttley GA, Janssen S, Jarmusch AK, Jiang L, Kaehler BD, Kang KB, Keefe CR, Keim P, Kelley ST, Knights D, et al. 2019. Author Correction: reproducible, interactive, scalable and extensible microbiome data science using QIIME 2. *Nat Biotechnol* 37:1091. <https://doi.org/10.1038/s41587-019-0252-6>.
  61. Bolyen E, Rideout JR, Dillon MR, Bokulich NA, Abnet CC, Al-Ghalith GA, Alexander H, Alm EJ, Arumugam M, Asnicar F, Bai Y, Bisanz JE, Bittinger K, Brejnrod A, Brislawn CJ, Brown CT, Callahan BJ, Caraballo-Rodríguez AM, Chase J, Cope EK, Da Silva R, Diener C, Dorrestein PC, Douglas GM, Durall DM, Duvallet C, Edwardson CF, Ernst M, Estaki M, Fouquier J, Gauglitz JM, Gibbons SM, Gibson DL, Gonzalez A, Gorlick K, Guo J, Hillmann B, Holmes S, Holste H, Huttenhower C, Huttley GA, Janssen S, Jarmusch AK, Jiang L, Kaehler BD, Kang KB, Keefe CR, Keim P, Kelley ST, Knights D, et al. 2019. Reproducible, interactive, scalable and extensible microbiome data science using QIIME 2. *Nat Biotechnol* 37:852–857. <https://doi.org/10.1038/s41587-019-0209-9>.
  62. Callahan BJ, McMurdie PJ, Rosen MJ, Han AW, Johnson AJ, Holmes SP. 2016. DADA2: high-resolution sample inference from Illumina amplicon data. *Nat Methods* 13:581–583. <https://doi.org/10.1038/nmeth.3869>.
  63. Katoh K, Misawa K, Kuma K, Miyata T. 2002. MAFFT: a novel method for rapid multiple sequence alignment based on fast Fourier transform. *Nucleic Acids Res* 30:3059–3066. <https://doi.org/10.1093/nar/gkf436>.
  64. Price MN, Dehal PS, Arkin AP. 2010. FastTree 2—approximately maximum-likelihood trees for large alignments. *PLoS One* 5:e9490. <https://doi.org/10.1371/journal.pone.0009490>.
  65. DeSantis TZ, Hugenholtz P, Larsen N, Rojas M, Brodie EL, Keller K, Huber T, Dalevi D, Hu P, Andersen GL. 2006. Greengenes, a chimera-checked 16S rRNA gene database and workbench compatible with ARB. *Appl Environ Microbiol* 72:5069–5072. <https://doi.org/10.1128/AEM.03006-05>.
  66. Demars J, Curk T, Erjavec A, Gorup C, Hocevar T, Milutinovic M, et al. 2013. Orange: data mining toolbox in Python. *J Machine Learning Res* 14: 2349–2353.
  67. Rideout JR, Chase JH, Bolyen E, Ackermann G, Gonzalez A, Knight R, et al. 2016. Keemei: cloud-based validation of tabular bioinformatics file formats in Google Sheets. *PeerJ PrePrints* 4:e1670v1.
  68. Mahé F, Rognes T, Quince C, de Vargas C, Dunthorn M. 2014. Swarm: robust and fast clustering method for amplicon-based studies. *PeerJ* 2: e593. <https://doi.org/10.7717/peerj.593>.
  69. Rognes T, Flouri T, Nichols B, Quince C, Mahé F. 2016. VSEARCH: a versatile open source tool for metagenomics. *PeerJ* 4:e2584. <https://doi.org/10.7717/peerj.2584>.
  70. Camacho C, Coulouris G, Avagyan V, Ma N, Papadopoulos J, Bealer K, Madden TL. 2009. BLAST+: architecture and applications. *BMC Bioinformatics* 10:421. <https://doi.org/10.1186/1471-2105-10-421>.
  71. Quast C, Pruesse E, Yilmaz P, Gerken J, Schweer T, Yarza P, Peplies J, Glöckner FO. 2013. The SILVA ribosomal RNA gene database project: improved data processing and web-based tools. *Nucleic Acids Res* 41:D590–6. <https://doi.org/10.1093/nar/gks1219>.
  72. R development Core Team. R: a language and environment for statistical computing. Version 4.0. 2 (Taking Off Again). R Foundation for Statistical Computing, Vienna, Austria. 2020.
  73. Oksanen J, Blanchet FG, Kindt R, Legendre P, Minchin PR, O'hara RB, et al. 2013. Package 'vegan'. *Community Ecology Package*, Version 2:1–295.
  74. Love MI, Huber W, Anders S. 2014. Moderated estimation of fold change and dispersion for RNA-seq data with DESeq2. *Genome Biol* 15:550. <https://doi.org/10.1186/s13059-014-0550-8>.
  75. Douglas GM, Maffei VJ, Zaneveld JR, Yurgel SN, Brown JR, Taylor CM, Huttenhower C, Langille MGI. 2020. PICRUSt2 for prediction of metagenome functions. *Nat Biotechnol* 38:685–688. <https://doi.org/10.1038/s41587-020-0548-6>.
  76. Caspi R, Billington R, Fulcher CA, Keseler IM, Kothari A, Krummenacker M, Latendresse M, Midford PE, Ong Q, Ong WK, Paley S, Subhraveti P, Karp PD. 2018. The MetaCyc database of metabolic pathways and enzymes. *Nucleic Acids Res* 46:D633–D639. <https://doi.org/10.1093/nar/gkx935>.