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- 1 Pathotyping the Zoonotic Pathogen Streptococcus suis: Novel Genetic Markers to
- 2 Differentiate Invasive Disease-Associated Isolates from Non-Disease Associated
- 3 Isolates from England and Wales.
- 4
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Abstract [limit: 250 words; word count: 248]

Streptococcus suis is one of the most important zoonotic bacterial pathogens of pigs causing significant economic losses to the global swine industry. *S. suis* is also a very successful coloniser of mucosal surfaces and commensal strains can be found in almost all pig populations worldwide, making detection of the S. suis species in asymptomatic carrier herds of little practical value in predicting the likelihood of future clinical relevance. The value of future molecular tools for surveillance and preventative health management lies in the detection of strains that genetically have increased potential to cause disease in presently healthy animals. Here we describe the use of genome-wide association studies to identify genetic markers associated with the observed clinical phenotypes i) invasive disease or ii) asymptomatic carriage on the palatine tonsils of pigs on UK farms. Subsequently we designed a multiplex-PCR to target three genetic markers that differentiated 115 S. suis isolates into disease-associated and non-disease associated groups; performing with a sensitivity of 0.91, specificity of 0.79, negative predictive value of 0.91, and positive predictive value of 0.79 in comparison to observed clinical phenotypes. We describe evaluation of our pathotyping tool, using an out-of-sample collection of 50 previously uncharacterised *S. suis* isolates, in comparison to existing methods used to characterise and subtype *S. suis* isolates. In doing so, we show our pathotyping approach to be a competitive method to characterise *S. suis* isolates recovered from pigs on UK farms, and one that can easily be updated to incorporate global strain collections.

## Introduction

Streptococcus suis (S. suis) is one of the most important bacterial pathogens of
pigs causing significant economic losses to the swine industry worldwide (1).
The infectious agent is responsible for a wide range of clinical manifestations,
including septicaemia with sudden death, meningitis, endocarditis, arthritis, and
pneumonia amongst other diseases (2). <i>S. suis</i> is also a zoonotic pathogen
associated with exposure to pigs or pork-derived products (3). Although cases in
Europe are infrequently reported, in recent years the surveillance and number of
reported human infections has increased substantially in Southeast Asia (4-9).
Importantly, S. suis is not only an invasive pathogen but also a very
successful coloniser of mucosal surfaces (10). In fact, the upper respiratory tract
of pigs, in particular the palatine tonsils, is considered to be both the natural
habitat of <i>S. suis</i> and a principal route of invasion; although the bacterium can
also be recovered from the gastrointestinal and genital tracts (2). Colonisation of
adult pigs is common in almost all pig populations sampled, meaning that
transfer of <i>S. suis</i> from sow to piglet during parturition and suckling is an
important route of transmission (10).
Several methods exist to investigate strain diversity and identify
phylogenetic groups of <i>S. suis</i> . Simple biochemical tests cannot always
differentiate <i>S. suis</i> from <i>S. suis</i> -like strains when performed on cultured isolates
recovered from diseased animals, and to date remain of little practical use in
differentiating invasive disease-associated strains from asymptomatic
commensal-like strains both of which may contribute sub-clinically to the
respiratory microflora of colonised pigs (1). Other existing methods used to

characterise and subtype S. suis as part of epidemiological studies have recently

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been the subject of a comprehensive review by Xia et al. (11). Each approach has its limitations often requiring either large amounts of sample DNA, which is labour intensive and cumbersome, or high levels of technical competence making the comparison of results between laboratories difficult. To date, serotyping remains the most widely used method to subtype S. suis isolates and is an important part of the routine diagnostic procedure (2, 12). A total of 35 serotypes have been described for *S. suis* based on differences

in the capsular polysaccharide antigens, but since their original descriptions evidence now exists for the reclassification of a number of serotypes as other Streptococcus species meaning current opinion considers there to be just 29 "true" *S. suis* serotypes (namely 1-19, 21, 23-25, 27-31, and 1/2) (13). Serotype 2 predominates among clinical cases of disease in most countries, although serovars 1-9, 14 & 1/2 have all been documented as being of clinical importance in certain geographical locales (14-18). As a result, serotyping has been used as a proxy for predicting the virulence potential of *S. suis* isolates. However, the use of serotyping alone as a predictor of virulence has the limitation that strains of the same serotype can vary substantially in virulence (19, 20).

Given the limitations of serotyping to reliably predict virulence potential of S. suis strains other markers have been investigated. A wide range of homologs of bacterial virulence factors and virulence-associated factors found in other Gram-positive organisms has been shown to affect the virulence of *S. suis* strains through targeted mutagenesis studies (21-23). However, clear association with specific roles in the onset and development of disease has not been found for many proposed factors (24, 25). Despite this, the 'virulence-associated markers' (rather than virulence factors *per se*) extracellular protein factor (EF, encoded by

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the epf gene) (26) and muramidase-released protein (MRP, encoded by the mrp gene) (27), as well as, the thiol-activated toxin hemolysin, suilysin (SLY, encoded by the sly gene) (28, 29) have been extensively used to predict the virulence potential of *S. suis* strains in certain mainly European countries, particularly for strains of serotype 2 (17, 24, 30). Unfortunately, genotyping of *epf, mrp*, and/or sly also fails to provide clear classification of a *S. suis* isolate as virulent (or not) because isogenic mutants devoid of such factors have been found to be as virulent as their respective parental strains, emphasising the importance of their consideration as virulence associated markers rather than true virulence factors per se (31). Advances in sequencing technologies now allow whole-genome

sequencing (WGS) of multiple strains of the same species, including S. suis (32-36). This explosion in the amount of detailed genetic information has allowed Bayesian analysis of population structure and the investigation of *S. suis* recombination rates, revealing enormous species diversity and significant genomic differences between *S. suis* isolates responsible for systemic disease in pigs when compared to non-clinical isolates recovered from the upper respiratory tract (35). Indeed, in 2015 Weinert et al. proposed loss of proteinencoding sequences had led to a smaller systemic disease-associated genome with increased virulence potential and an overrepresentation of genes encoding previously reported virulence-factors associated with S. suis (35).

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Minimum core genome (MCG) sequence typing is a recently described typing scheme that also takes advantage of the increase in available S. suis WGS data, using population genetics-based sub-divisions for strain identification and typing (33, 37). MCG sequence typing exploits advances in next-generation

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sequencing to identify novel regions of the core-genome that can be used to identify and type *S. suis* isolates into "MCG groups" that can later be associated with clinical phenotypes. In fact, during its design, MCG group 1 was reported as being assigned to all highly virulent isolates tested and associated with the greatest occurrence of previously reported virulence genes (33). However, MCG sequence typing like multilocus sequence typing (MLST), also described for S. suis (38), is difficult to apply to routine diagnostic testing and can sometimes lack the discriminatory power to differentiate bacterial strains into virulent and avirulent sub-populations, limiting its usefulness in epidemiological studies. The aim of this study was to design and then evaluate a pathotyping tool

to predict the virulence potential of *S. suis* isolates using genome-wide association studies, a so-far unexploited method for the identification of S. suis virulence-associated markers. The statistical power to allow the identification of robust associations between genotype and phenotypes including virulence in many different bacterial species is now possible due to the rapid increases in the availability of detailed WGS data (39, 40). Here we have combined WGS data with high-quality clinical metadata in order to identify genetic markers in the S. suis accessory genome (i.e. genes absent from one or more isolates or unique to a given isolate) associated with i) invasive disease or ii) asymptomatic carriage on the palatine tonsils of pigs on UK farms. Subsequently, we designed a multiplex-PCR (mPCR) to target three genetic markers that differentiated 115 S. suis isolates into i) invasive disease-associated and ii) non-disease associated groups. We also describe evaluation of our pathotyping tool (generalised linear model and mPCR), using an out-of-sample collection of 50 previously uncharacterised *S. suis* isolates, in comparison to existing methods used to

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characterise and subtype *S. suis* isolates. In doing so, we show our approach to be a competitive method to subtype S. suis isolates recovered from pigs on UK farms, and one that can easily be updated to incorporate global strain collections.

## **Materials and Methods**

<b>Bacterial isolates.</b> Two groups of <i>S. suis</i> isolates were used in this study a) a
training collection of 115 isolates and b) an out-of-sample test collection of 50
previously uncharacterised isolates. <b>The original training collection</b> was used
to identify genetic markers which could differentiate <i>S. suis</i> isolates into i)
invasive disease-associated and ii) non-disease associated phenotypic groups.
The 'training' collection consisted of laboratory reference strain P1/7
(NC_012925) originally recovered from an ante- <i>mortem</i> blood culture from a pig
that died with meningitis in the United Kingdom (32, 41). The other 114 isolates
of the training collection were a subset recovered from pigs on farms in England
and Wales during routine diagnostic investigations at the Animal Health and
Veterinary Laboratories Agency (AHVLA; now the Animal and Plant Health
Agency, APHA) in 2010, and contribute to a larger collection previously
described in 2015 by Weinert et al. (35). Well-defined phenotypic metadata were
available based on which each isolate was categorised as being associated with
invasive <i>S. suis</i> disease (n=53; recovered from systemic sites in the presence of
clinical signs (arthritis, meningitis, septicaemia) and/or gross pathology
consistent with <i>S. suis</i> infection) or as being non-disease associated (n=62;
recovered from the tonsil or trachea-bronchus of pigs without any typical signs
of streptococcal disease but diagnosed with disease unrelated to <i>S. suis</i> , such as
enteric disease). The out-of-sample test collection was used to evaluate our
pathotyping tool. Out-of-sample forecasting is a common approach used to
evaluate the performance of binary diagnostic tests. To avoid reducing statistical
power, rather than split the training collection, an additional out-of-sample 'test'
collection was put together consisting of 23 invasive disease-associated

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(recovered from systemic, non-respiratory locations of pigs diagnosed with S. suis disease at the APHA during 2013) and 27 non-disease associated isolates (recovered from material scraped from the palatine tonsils of pigs exhibiting no signs of S. suis disease on farms in England and Wales between June 2013 and May 2014). Site of recovery, ante-, and post-mortem findings of all isolates described in this study are summarised in Table S1. Identification of genetic markers associated with observed clinical

**phenotype.** Genetic markers to pathotype *S. suis* were identified using positive detection data of putative protein-encoding sequences making up the *S. suis* accessory-genome (i.e. genes absent from one or more isolates or unique to a given isolate). The accessory genome was taken from Weinert et al. (35). Briefly, de novo assemblies of Illumina fastq reads were produced, protein-encoding genes were then identified and used in MCL clustering to find orthologue groups, which were manually checked. Two complementary genome-wide association studies i) the univariate Chi-squared test for independence and ii) the multivariate Discriminant Analysis of Principal Components (DAPC) were combined to define a preliminary list of genetic markers associated with the observed clinical phenotypes i) invasive disease or ii) asymptomatic carriage on the palatine tonsils of pigs. The Chi-squared test for independence, implemented in the R package: *stats* (42), was used to compare the observed positive detection of protein-encoding sequences with expected frequencies, in doing so calculating a test statistic that if greater than the critical value was reason enough to reject the null hypothesis of independence (p-value < 0.05). Bonferroni adjustment  $(\alpha/n)$  was used to control for family-wise error associated with multiple sampling.

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DAPC (43, 44), implemented in the R package: adegenet (45, 46), was used to identify genetic differences between pre-defined phenotypic groups. The total amount of original variation retained in the DAPC model affected which genetic markers contributed most to the separation of genetic structures. As a result, four independent DAPC analyses were performed retaining 60, 70, 80 or 90% of the original genetic variation, and the 1% of ranked genetic markers contributing most to the discrimination of pre-defined phenotypic groups was then analysed and genetic markers consistently output by two or more DAPC analyses taken forward as candidates for pathotyping *S. suis*.

Analysis of the distribution of previously reported virulence factors **associated with** *S. suis disease.* Protein-encoding sequences present in P1/7, taken from the list of previously published virulence and virulence-associated factors complied as part of a comprehensive review by Fittipaldi et al. (24), were extracted from GenBank (Table S2). P1/7 protein-encoding sequences were used as tBLASTn queries against a bespoke BLAST database consisting of the draft genome assemblies of all isolates described in this study. Amino acid level matches to >80% of >80% of the total length of each translated protein-encoding sequence were considered hits.

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**Selection of genetic markers to pathotype** *S. suis.* Logistic regression analysis in the form of a generalised linear model (GLM) with backwardsstepwise selection using penalised likelihood ratio tests, implemented in the R package: logistf (47), was used to identify the fewest statistically significant (pvalue <0.05) markers to differentiate *S. suis* isolates into pre-defined i) invasive disease-associated and ii) non-disease associated groups. A receiver operating characteristic (ROC) curve, implemented in the R package: ROCR (48), was used

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to visualise the GLM performance metrics true positive rate (sensitivity) and false positive rate (1-specificity) in comparison to the observed clinical phenotype (considered to be the 'gold-standard' in this study), and a cutoff threshold selected to convert the real-valued output (fitted values) of the logistic regression (probability of causing invasive disease) into a binary class decision: invasive disease-associated (1)/non-disease associated (0). As no cutoff was optimal according to all possible performance criteria, cutoff choice involved a trade-off between different performance metrics where low false negative rate (1-sensitivity, analogous to Type II error) was chosen as the most valuable performance metric for pathotyping *S. suis*, with a view to establish and then maintain a pig population free of invasive disease-associated strains. All statistical analyses were performed using the standard R environment for statistical computing and graphs (version 3.1.1) (49). **Identification of** *S. suis-***species specific genetic markers.** We designed a mPCR to target genetic markers associated with observed clinical phenotype, along with a *S. suis* species-specific marker as a positive control. The most conserved protein-encoding sequences of the *S. suis* core-genome (i.e. present in all isolates) were used to select a species-specific marker to complement the pathotyping markers. To do this, all annotated protein-encoding sequences of S. suis strain P1/7 were used as BLASTn queries against a bespoke BLAST database of all *de novo* assemblies and known *S. suis* complete genome sequences. Protein-encoding sequences with identities >95% across >80% of the

total length of each query sequence were then used to query the NCBI non-

redundant (nr) database to identify matches only to S. suis.

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Multiplex-PCR and detection of PCR amplicons. The online software, primer3 version 4.0.0 (http://primer3.ut.ee) was used to design mPCR primers. All mPCR primers were designed to target conserved regions within the proteinencoding sequence of genetic markers (as opposed to flanking regions) and are summarised in Table 1. Primers were designed to have similar physical characteristics, enabling simultaneous amplification under the same thermal cycling conditions and in multiplex reactions. Primer length (21-30 bp), GC content (40-60%), melting temperature (>68 °C if possible, but at least 60 °C), and expected amplicon size (100-1000 bp) were based on the manufacturer's recommendations for primer design using the Multiplex PCR *Plus* kit (Qiagen). Consistency between the positive detection of genetic markers and primer matches was investigated using BLASTn. Prior to ordering, all primers were queried against the NCBI nr nucleotide database to check for non-S. suis DNA matches. Primers were synthesised by Sigma-Aldrich (Haverhill, United Kingdom) and delivered in solution (TE buffer; 10 mM Tris-Cl, 1 mM EDTA [pH 8.0]) at a stock concentration of 100 µM; primers were used at a working stock concentration of 20  $\mu$ M. All mPCRs were performed using the Multiplex PCR Plus Kit (Qiagen), and unless otherwise stated contained the same reagents except for template DNA. The reaction mixture (50 µl) for each mPCR consisted of 25 µl 2x Multiplex PCR Master Mix, 5 μl 10x CoralLoad Dye, 10 μl RNase-free water, 0.2 μM (final concentration) of each primer, and 10 ng template DNA. The three-step thermal cycling program for all reactions was as follows: 95 °C for 5 min, followed by 35

cycles of (denaturation) 95 °C for 30 s, (annealing) 66 °C for 90 s, and (extension)

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72 °C for 90 s; with a final extension of 68 °C for 10 min using a T100 Thermal Cycler (Bio-Rad). PCR products were analysed by gel electrophoresis using 2% (wt/vol) UltraPure Agarose (Invitrogen) gels made with 1x TBE buffer, and contained 1x SYBR Safe DNA gel stain (Invitrogen). Running time was 60 min at a constant 100 V. Results were visualised using a GelDoc imager (BioRad). Where appropriate, mPCR products were purified using the QIAquick PCR Purification Kit (Qiagen) as per the manufacturer's instructions and Sanger sequenced using the Source Bioscience Lifesciences sequencing service. Returned sequencing data was aligned with reference sequences of the target protein-encoding sequence using CodonCode Aligner software (CodonCode Corporation). The approximate limit of detection of the mPCR was estimated from 10fold serial dilutions of *S. suis* genomic DNA of known concentration. DNA extracted from four isolates of the training collection representing invasive disease-associated (SS002 and SS004) and non-disease associated (LSS011 and LSS027) phenotypes/genotypes was mixed in equal quantities so that templates for each mPCR amplicon would be present in all reactions. A series of 10-fold dilutions were then performed to create mPCR templates of decreasing concentration. The limit of detection was considered to be the lowest concentration of template DNA from which all predicted mPCR amplicons, after 35 thermal cycles, were easily visible under UV transillumination. To evaluate the specificity of the mPCR assay for *S. suis*, field isolates of Streptococcaceae commonly recovered from the upper respiratory tract of pigs on farms in England and Wales were used as a panel of negative controls. The

collection included isolates of Streptococcus gallolyticus, Streptococcus orisratti,

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Streptococcus pneumoniae, and Streptococcus uberis, sourced from BBSRC research project: BB/L003902/1. In addition, commensal Pasteurellaceae including Actinobacillus indolicus, Actinobacillus minor, Actinobacillus porcinus, and Haemophilus parasuis (Nagasaki and SW140) were also included, as well as DNA from an Alcaligenaceae isolate of *Bordetella bronchiseptica* RB50 (NC\_002927) (50).

Comparison of our pathotyping tool to existing methods used to **subtype disease-associated isolates of** *S. suis.* To compare our pathotyping tool (GLM and mPCR) to published methods used to subtype disease-associated isolates of *S. suis*, the molecular serotype, virulence-associated gene (*epf, mrp*, and sly) profile, MLST, and MCG sequence type were all determined in silico. For comparison of our pathotyping tool against each existing method the original training collection was used to 'train' a model that was then applied to the outof-sample 'test' collection.

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Traditional serotyping (by capillary precipitation) data was unavailable for all *S. suis* isolates described in this study, therefore, molecular 'serotyping' was performed using an adaptation (for in silico use) of the mPCR assays described by Liu et al. (51). Primer sequences were used as BLASTn queries and nucleotide level matches to >95% of the total length of each primer sequence were considered hits. The distance between hits was compared to reported PCR amplicon sizes. Isolates that could not be assigned to one of the 35 (1-34 & 1/2) originally described *S. suis* serotypes were deemed non-serotypable (NT). Differentiation of molecular 'serotypes' 1 from 14 and 2 from 1/2 was performed using the published method described by Athey et al. (52). All isolates, in particular those deemed to be NT, were confirmed to be S. suis using a

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combination of i) biochemical profile (API 20 Strep), ii) MLST data, and iii) recN sequence homology (53).

Virulence-associated gene profiling was performed using an adaptation (for in silico use) of the method described by Silva et al. (54). Again mPCR and singleplex-PCR primer sequences were used as BLASTn queries and nucleotide level matches to >95% to the total length of each primer sequence were considered hits. The distance between hits compared to reported PCR amplicon sizes. Logistic regression (as described above) using the prevalence of *epf, mrp*, and/or sly as the GLM explanatory variables was used to classify all isolates as i) invasive disease-associated or ii) non-disease associated.

MLST was performed using the online software MLST version 2.0 (http://cge.cbs.dtu.dk) (55).

MCG sequence typing was performed using an adaptation (for *in silico* use) of the method described by Zheng et al. (37). Multiplex-PCR primer sequences were used as BLASTn queries and nucleotide level matches to >95% of the total length of each primer sequence were considered hits. The distance between hits compared to reported mPCR amplicon sizes. Nucleotide sequences between primer sequence matches were then extracted, aligned against the MCG typing reference strain GZ1 (GenBank: CP000837), and the 10 SNPs of interest called allowing isolates to be assigned to one of the seven reported MCG groups for S. suis.

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McNemar's Chi-squared Test for Count Data, implemented in the R package: stats (42), was used to test for statistically significant differences in the sensitivities and specificities of two binary diagnostic tests in a paired study. The Weighted Generalised Score Statistic for Comparison of Predictive Values as

proposed by Kosinski (56), implemented in the R package: DTComPair (57), was

- 348 used to test for significant differences in (negative and positive) predictive
- 349 values of two binary diagnostic tests.

Results

<b>Design of a pathotyping tool for </b> <i>S. suis.</i> Genetic markers to pathotype <i>S. suis</i>
were identified using positive detection data of 7261 putative protein-encoding
sequences making up the <i>S. suis</i> accessory-genome (35). To do this, the output of
two complementary genome-wide association studies were combined to define a
preliminary list of 497 genetic markers associated with the observed clinical
phenotypes i) invasive disease or ii) asymptomatic carriage on the palatine
tonsils of pigs. A multistep process was used to reduce the preliminary list to a
number suitable for logistic regression analysis, retaining genetic markers only if
i) positively detected in >50% of invasive disease-associated and <50% of non-
disease associated isolates (and vice versa <50% of invasive disease-associated
and <50% of non-disease associated isolates; n=88 remaining), ii) protein-
encoding sequence length was >500 bp (based on the manufacturer's
recommendations for primer design using the Qiagen Multiplex PCR <i>Plus</i> kit;
n=44 remaining), and iii) not predicted to be a mobile genetic element, such as a
phage gene, integrase or transposon (based on Prokka annotations; n=14
remaining). A GLM with backwards-stepwise selection using penalised likelihood
ratio tests was then used for the final selection of genetic markers, two
associated with invasive disease and one associated with asymptomatic carriage
(Table 1). A receiver operating characteristic (ROC) curve was used to visualise
the GLM performance metrics true positive rate (sensitivity) and false positive
rate (1-specificity), and select the cutoff threshold of 0.43 to be used to convert
the real-valued output (fitted values) of the GLM into a binary class decision:
invasive disease-associated/non-disease associated (Table S1). In comparison to
the observed clinical metadata, considered the 'gold-standard' in this study, our

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three genetic markers subtyped the 115 S. suis isolates of the training collection with a sensitivity of 0.91, specificity of 0.79, negative predictive value of 0.91, and positive predictive value of 0.79 (Table S3(a)).

At present, WGS is not readily available for routine surveillance studies in veterinary diagnostics laboratories, therefore, we designed a mPCR to target the three genetic markers selected to pathotype *S. suis.* In addition to genetic markers selected to differentiate *S. suis* isolates into i) invasive diseaseassociated and ii) non-disease associated groups, we also incorporated a S. suis species-specific marker into our mPCR assay. To do this, we first identified the most conserved protein-encoding sequences contributing to the *S. suis* core genome (i.e. present in all isolates) and selected SSU0577 as a novel S. suis species-specific marker, that had a minimum nucleotide sequence identity of 98.15% across the total length of the 918 bp protein-encoding sequence.

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Evaluation of our pathotyping mPCR with the original training **collection.** Figure 1 shows an example of the mPCR amplicon patterns after gel electrophoresis on a 2% (wt/vol) agarose gel and photographed under UV transillumination. Amplicons of size 722 bp correspond to the S. suis speciesspecific marker (SSU0577), and were produced by all isolates of the training collection irrespective of invasive disease-associated/non-disease associated phenotype or genotype. Other amplicons, of size 211 bp and 347 bp correspond to the invasive disease-associated markers SSU0207 and SSU1589 respectively, and amplicons of size 892 bp correspond to the non-disease associated marker SSUST30534.

To determine the analytical sensitivity of the mPCR the approximate limit of detection was estimated from 10-fold serial dilutions of S. suis genomic DNA of

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known concentration. The limit of detection was estimated to be  $\sim 0.0001$  ng of S. suis genomic DNA (equivalent to  $\sim$ 45 genome copies), the lowest concentration of template DNA from which all predicted mPCR amplicons, after 35 thermal cycles, were easily visible under UV transillumination (data not shown). To evaluate the specificity of our mPCR for *S. suis*, field isolates of Streptococcaceae, Pasteurellaceae, and Alcaligenaceae commonly recovered from the upper respiratory tract of pigs on farms in England and Wales were used as a panel of negative controls. No mPCR amplicons, after 35 thermal cycles and gel electrophoresis, were visible under UV transillumination for any of the panel of ten negative controls (data not shown). Evaluation of our pathotyping tool with an out-of-sample collection. Further evaluation of our pathotyping tool (GLM and mPCR) was done using an out-of-sample test collection of 50 previously uncharacterised (genetically) S. suis isolates (23 invasive disease-associated and 27 non-disease associated). Template DNA extracted from each of the 50 isolates produced the 722 bp mPCR amplicon corresponding to the *S. suis* species-specific marker SSU0577. For each isolate, the presence/absence of mPCR amplicons was then input into the GLM and the cutoff threshold of 0.43 applied to the fitted-values to generate the binary classification decision. Table 2(a) summarises the classification of the outof-sample test collection isolates in comparison to the observed clinical metadata, resulting in a sensitivity of 0.83, specificity of 1.00, negative predictive value of 0.87, and positive predictive value of 1.00.

Comparison of our pathotyping tool to existing methods used to

**subtype disease-associated isolates of** *S. suis.* To compare our pathotyping

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tool to the use of serotype as a proxy to predict the virulence potential of *S. suis* isolates, the serotypes most frequently recovered from diseased pigs (1-9, 14 & 1/2) were considered a marker of disease association and all other serotypes considered markers of non-disease association. Table 2(b) summarises the classification of the out-of-sample test collection isolates in comparison to the observed clinical metadata, and shows the use of molecular serotypes 1-9, 14 & 1/2 to predict disease-association performed with a sensitivity of 0.87 (n=3 type II errors), not statistically different from our new mPCR pathotyping tool (McNemar's Chi-squared test for count data p-value = 0.31731). Other performance metrics for the molecular serotype-based approach were a significantly worse positive predictive value of 0.77 (weighted generalised score statistic for comparison of predictive values p-value = 0.01149) and a significantly worse specificity of 0.78 (n=8 type I errors, McNemar's Chi-squared test for count data p-value = 0.01431); no statistically significant difference in negative predictive value was observed (weighted generalised score statistic for comparison of predictive values p-value = 0.90553). To compare our pathotyping tool to the use of *epf, mrp* and/or *sly* for the identification of virulent *S. suis* strains, first a GLM was fitted to the prevalence data of these virulence-associated genes in the original 'training' collection of S. suis isolates and, using the same selection criteria as previously described for the pathotyping markers, a ROC curve used to select the cutoff of 0.12 to convert the GLM fitted values into a binary class decision. The predict function, implemented in the R package: *logistf* (47), was then used to generate fitted values for the isolates in the out-of-sample test collection (Table S1). Table 2(c)

summarises the classification of the out-of-sample test collection isolates as

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invasive disease-associated/non-disease associated based on the positive detection of *epf, mrp* and/or *sly* in comparison to the observed clinical phenotype. The combined virulence-associated markers performed with a sensitivity of 0.96 (n=1 type II errors), again not statistically different from our new mPCR pathotyping tool (p-value = 0.08326). Other performance metrics for the virulence-associated genotyping approach were a significantly worse positive predictive value of 0.46 (p-value = 2.97708e<sup>-7</sup>; incidentally performing no better than chance (Exact binomial test p-value = 1)), and a significantly worse specificity of 0.04 (p-value = 3.41417e<sup>-7</sup>). The negative predictive value was calculated to be 0.50, worse but not a statistically significant difference (pvalue = 0.07853). We compared our pathotyping tool to the use of the King et al. MLST scheme (38) as a proxy to predict the virulence potential of *S. suis* isolates. Sequence type (ST) 1 was assigned to 70% of disease-associated isolates and 3% of non-disease associated isolates of the training collection (Table S1). As ST1 is mostly associated with disease in both pigs and humans in Europe (12) we used assignment to ST1 as a binary classifier to indicate disease-association in comparison to the observed clinical metadata. Table 2(d) summarises the classification of the out-of-sample test collection isolates as invasive diseaseassociated/non-disease associated based on the assignment to ST1 in comparison to the observed clinical phenotype. Assignment to ST1 performed with a sensitivity of 0.70 (n=7 type II errors), worse in comparison to our pathotyping tool but not a statistically significant difference (p-value = 0.08326). The negative predictive value was calculated to be 0.79, again worse but not a

statistically significant difference (p-value = 0.08294). Other performance

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metrics (specificity and positive predictive value) were found to be identical in comparison to our pathotyping tool.

Finally, we compared our pathotyping tool to the use of the Zheng *et al.* MCG typing scheme (33, 37), one of the most recent typing schemes that exploits advances in next-generation sequencing to identify virulent S. suis strains. MCG group 1 was assigned to 77% of disease-associated isolates and 3% of nondisease associated isolates of the training collection (Table S1). Together with the report of MCG group 1 being assigned to all highly virulent isolates tested during design of the typing scheme (33), we used assignment to MCG group 1 as a binary classifier to indicate disease-association; performance in comparison to the observed clinical metadata is summarised in Table 2(e). Assignment to MCG group 1 performed with a sensitivity of 0.78 (n=5 type II errors), again worse in comparison to our pathotyping tool but not a statistically significant difference (p-value = 0.31731). The negative predictive value was calculated to be 0.84, also worse but not a statistically significant difference (p-value = 0.31725). Other performance metrics (specificity and positive predictive value) were found to be identical in comparison to our pathotyping tool.

## **Discussion**

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We have described the design of a pathotyping tool (GLM and mPCR) exploiting the identification of genetic markers in the *S. suis* accessory-genome (i.e. genes absent from one or more isolates or unique to a given isolate) associated with the observed clinical phenotypes i) invasive disease or ii) asymptomatic carriage on the palatine tonsils of pigs on UK farms. Initial analyses of the original training collection were unable to identify any single genetic marker of invasive disease prevalent in >95% of invasive disease-associated isolates and not positively identifiable in <5% of non-disease associated isolates. Furthermore, we found over half (n=40) of published putative "virulence-factors", extracted from the previous comprehensive review by Fittipaldi et al. (24) and present in P1.7, did not show a strong relationship with observed clinical phenotype as they were either i) positively detected in the *S. suis* core-genome (i.e. prevalent in all isolates; n=38) or ii) not detected by our methods in any of the 115 isolates of the training collection (n=2; data not shown). The reason for this is unclear, although could be an effect of previous studies being limited to small numbers of isolates often restricted to serotype 2 (58), and of varied and inconsistent animal models between research groups (25).

To avoid restricting our analyses to previously published reports and not taking full advantage of the statistical power of our WGS data set, we used two complementary genome-wide association studies and then logistic regression analysis for the final selection of genetic markers to pathotype *S. suis*. Using logistic regression analysis also allowed for the possibility that multiple genetic markers might best describe the *S. suis* pathotype. Our pathotyping markers assigned the 115 *S. suis* isolates of the original training collection to phenotypic

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groups (disease-associated/non-disease associated) with a sensitivity of 0.91 i.e. the proportion of isolates recovered from systemic sites and predicted to be disease causing isolates. A specificity of 0.79 i.e. the proportion of isolates recovered from the upper respiratory tract of pigs without any typical signs of S. suis infection and predicted to be non-disease associated isolates. A negative predictive value of 0.91 i.e. the proportion of isolates predicted to be non-disease associated that were actually recovered from the upper respiratory tract of pigs without any typical signs of *S. suis* infection. As well as, a positive predictive value of 0.79 i.e. the proportion of isolates predicted to be associated with invasive disease that were actually recovered from a systemic site.

An important caveat of our pathotyping tool design is consideration of the observed clinical phenotype associated with each isolate as the 'gold standard' to characterise S. suis isolates as disease-associated or non-disease associated. In the absence of an agreed superior approach, clinical metadata was used to assign S. suis isolates to one of two phenotypic groups and it is acknowledged that such an approach is not perfect as not all additional factors can be accounted for, such as host-immune status, concurrent infections, or environmental conditions that could influence the susceptibility of a host to *S. suis*-associated disease. Indeed, reports of *in vivo* challenge studies can be readily found in the *S. suis* literature, although most describe data limited to a small number of isolates, often restricted to serotype 2 (58), and under very different conditions making the extrapolation of findings difficult to interpret. An ideal standard would require an agreed panel of isolates for which a series of consistently controlled experimental infection challenge studies had been undertaken using pigs of identical immune status and genetics. However, in order for this to happen

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experts in the field must first agree on a suitable model and set of well-defined criteria to score virulence (25, 59, 60).

Another important caveat of our pathotyping tool design is the source of S. suis isolates of the original training collection that were deemed to be nondisease associated. While all efforts were made to accurately define invasive disease-associated and non-disease associated phenotypic groups it should be acknowledged that non-disease associated isolates of the original training collection were recovered from routine submissions to the APHA in 2010 and that these pigs were not healthy, even though they did not show signs of typical streptococcal disease; instead clinical features were consistent with different non-infectious diseases or disease caused by other non-S. suis infectious agents. Indeed, 13 isolates of the original training collection deemed to be non-disease associated by phenotype were predicted by our pathotyping tool to have the potential to cause invasive disease. These 13 type I errors (or 'false' positives) in comparison to the observed clinical metadata could in fact be true predictions and examples of *S. suis* strains with the potential to cause invasive disease being carried in the upper respiratory tract of pigs on UK farms. Therefore, it is possible that the mortality of these 13 pigs was due to clones of isolates recovered from the palatine tonsils or trachea-bronchus yet was not identified as so due to a concurrent or opportunistic infection presenting a more obvious phenotype, such as diarrhoea. Such an observation is supported by evidence in the literature reporting that virulent strains of *S. suis* can be isolated from the tonsils of pigs without obvious streptococcal disease (61, 62), which is likely to represent carriage of invasive disease-causing stains by pigs that have mounted an effective immune response.

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We deemed the false negative rate (1-sensitivity) to be the most valuable performance metric for a S. suis pathotyping tool in order to establish and maintain a pig population free of invasive disease-associated *S. suis* strains. During out-of-sample testing the false negative rate of 0.17 corresponded to four false negatives (or type II errors), where non-disease associated pathotyping tool predictions were made for isolates linked with invasive disease clinical metadata. It is interesting to speculate at the reasons for such observations. Often S. suis strains are described as opportunistic or secondary pathogens that without a weakened host immune status (due to stress or concurrent infection) would normally be carried asymptomatically, contributing to the normal oral microflora of pigs. This may be the explanation for the differences observed between our pathotyping tool prediction and the observed clinical phenotype, again emphasising the fallibility of the phenotype assigned when it is based on field sampling without carefully controlled infection challenge data. Comparison to published methods revealed our molecular pathotyping tool to be a competitive method to subtype *S. suis* isolates, even though the necessarily small number of clinically phenotyped isolates in the out-of-sample collection limited the statistical power of the comparison. Comparing the commonly used performance metrics sensitivity, specificity, negative predictive value, and positive predictive value we found the use of i) serotypes 1-9, 14 & 1/2, ii) a GLM based on the positive detection of virulence-associated markers epf, mrp and/or sly, iii) assignment to MLST 1, and iv) assignment to MCG group

1 performed with statistically similar sensitivities in comparison to our

pathotyping tool. However, the trade off for high sensitivities was significantly

worse specificities and negative predictive values when using serotypes 1-9, 14

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& 1/2 or the virulence-associated markers: *epf, mrp* and/or *sly*, in certain cases performing no better than chance (p-value =1) in comparison to our pathotyping tool. Over all, the performance of our pathotyping tool was at least statistically similar and competitive with, and in some cases, better than previously described methods for assessing the clinical significance of *S. suis* isolates. Similarly, performance of our newly proposed *S. suis* species-specific marker (SSU0577) was encouraging. An important part of our pathotyping tool, due to the presence of *S. suis*-like organisms such as *Streptococcus orisratti* in the pig upper respiratory tract, we acknowledge that the specificity of SSU0577 for S. suis and not S. suis-like organisms needs to be extended and studied further against markers such as recN.

At present the role in pathogenesis of our newly defined pathotyping markers is unknown. Based on predicted biological functions (Table 1) we speculate that marker SSU0207, predicted to be a copper exporting ATPase, might allow *S. suis* to avoid copper toxicity inside phagocytes as copper homeostasis has been shown to be important in many bacterial species (63-65). The marker SSU1589 is annotated as a Type I restriction-modification (RM) system S protein in S. suis strain P1/7. Ubiquitous among prokaryotes, Type I RM systems are large multifunctional protein complexes thought to defend host bacterium from foreign DNA borne by bacteriophages, and have recently been described in P1/7 and *S. suis* strains isolated in the Netherlands (66, 67). Considered primitive immune systems in bacteria, it has been proposed that the range of functions RM systems may have should be expanded to include stabilising mobile genetic elements or gene regulation, potentially providing evolutionary fitness advantages and virulence under certain conditions (68).

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usefulness evaluated.

Indeed, the proposed role in protection against foreign DNA may merely be a coincidental benefit of these functions (69). In fact, a Type I RM system in Streptococcus pneumoniae which can undergo genetic recombination with truncated variants of the same gene to generate alternative variants with different methylation specificities could control global changes in gene expression (70). In Streptococcus pneumoniae there is a selection for variants of this genetic switching *in vivo*, indicating a role in systemic disease. Our third genetic marker (SSUST30534), a putative sugar ABC transporter, was positively associated with the non-disease associated phenotype (asymptomatic commensal-like carriage). The practical application of the genetic marker positively associated with asymptomatic carriage might not be immediately obvious but its statistical significance in the GLM is noteworthy. Indeed, gene loss (of so-called 'antivirulence genes') in the evolution of bacterial pathogens from non-pathogenic commensal strains could be a mechanism of fine tuning pathogen genomes for maximal fitness in new host environments; in short when regulation of invasion, replication and transmission processes is

altered, virulence can emerge (71). Indeed, genome reduction via gene loss and

pseudogenisation associated with enhanced pathogenicity has been described in

other bacteria, such as *Rickettsia* spp., *Shigella* spp. and *Yersinia* spp. (71).

Genome reduction through the loss of genes, potentially interfering with host

infection, has also been proposed in *S. suis* (35). Therefore, as the elimination of

the genetic marker associated with asymptomatic carriage from the GLM could

not be done without a statistically significant loss of fit it was retained and its

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In conclusion, we foresee a useful clinical application of our pathotyping tool in preventative programs aimed at monitoring the health status of pigs and identification of subclinical carriers of invasive disease-associated S. suis strains in the upper respiratory tract. Our approach can easily be updated to incorporate global strain collections (such as, from North America and Southeast Asia) to identify geographically-dependent phenotypes. This could contribute to a lower prevalence of disease attributed to S. suis among pig populations and consequently a reduction in the usage of antibiotics in the swine industry, as well as a reduction in zoonotic transmission of this pathogen through improved surveillance of pig populations.

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Figure 1. Agarose gel showing the expected amplicon sizes of our three genetic markers with the Streptococcus suis-specific marker from 14 isolates of the training collection. Agarose gel containing multiplex-PCR amplicons produced from genomic DNA of eight invasive disease-associated, and six non-disease associated isolates of *S. suis* recovered from pigs on farms in England and Wales. PCR amplicons were electrophoresed on a 2% (wt/vol) agarose gel containing 1x SYBR Safe DNA gel stain for 60 minutes at a constant 100 V and photographed under UV transillumination. Multiplex-PCR amplicon patterns matched anticipated amplicon patterns based on in silico analyses for all isolates described in this study. Isolate names are indicated above lanes. Lane M contains 1x Bioline HyperLadder 100 bp Plus DNA ladder with sizes indicated on the left (bp). Multiplex-PCR amplicon sizes are indicated on the right (bp).

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Table 1. Multiplex-PCR primer details. Multiplex-PCR primers were designed using the online software primer3 (version 4.0.0, http://primer3.ut.ee) and designed to target conserved regions within the protein-encoding sequence of genetic markers (as opposed to flanking regions). Primers were designed to have similar physical characteristics, enabling simultaneous amplification under the same thermal cycling conditions and in multiplex reactions. GenBank identifier prefixes "SSU" and "SSUST3" correspond to Streptococcus suis P1/7 (NC\_012925) (32) and Streptococcus suis ST3 (NC\_015433) (71) respectively.

Primer name	Primer sequence (5' - 3')	Marker of	Multiplex-PCR	Predicted biological function
			amplicon size (bp)	(Interpro)
SSU0207_0735F	TTACAAGAACAGGCAAGACAGTCGCC	Disease-association	211	Copper exporting ATPase 1
SSU0207_0945R	GCTGCTTTATAATCTGGGTGTTCGTTG			
SSU1589_0460F	CCTTTAATGCAGGGACAAAAGTGAGCTC	Disease-association	347	Type I restriction-modification (RM)
SSU1589_0806R	CCCATAATCTTACAGTTAACTTCCTTGC			system S protein
SSUST30534_0368F	ATCCCTCCCAATAAAAGATTTGGATGC	Non-disease association	892	Putative sugar ABC transporter
SSUST30534_1259R	TTTTCGAGCTCTCCATACACTGCTTCTG			
SSU0577_0086F	CAGGTAGTTTGGGCTTAGCTTCATCAGG	Streptococcus suis sp.	722	Sporulation regulator (WhiA)
SSU0577_0807R	TGGATGCTGAATTCGCAACTGGGCAATC			

Table 2. Contingency tables used to calculate the performance metrics summarising the classification of Streptococcus suis isolates in the out-of-sample test collection (n=50). Contingency tables used to calculate and summarise the performance metrics of two binary diagnostic tests. Each table compares the observed clinical phenotype (considered the 'gold-standard' in this study) to the use of the a) newly described pathotyping markers, b) serotypes: 1-9 and 1/2, c) Virulence-associated markers: epf, mrp, and/or sly, d) Multilocus sequence type (MLST): 1, and **e**) assignment to Minimum Core Genome (MCG) sequence type: 1 as markers of invasive disease.

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		Phen	otype		
	50	23	27		
	Total population	Phenotype positive	Phenotype negative		
	19	19	0	1.00	0.00
CR	mPCR positive	True positive	False positive	Positive predictive rate	False discovery rate
mPCR	31	4	27	0.13	0.87
	mPCR negative	False negative	True negative	False omission rate	Negative predictive rate
	•	0.83	0.00	0.90	
		True positive rate	False positive rate	F <sub>1</sub> score	
		0.17	1.00		•
		False negative rate	True negative rate		

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966 b)

		Phen	otype		
	50	23	27		
	Total population	Phenotype positive	Phenotype negative		
.5	26	20	6	0.77	0.23
9, 14 & 1/	Positive	True positive	False positive	Positive predictive rate	False discovery rate
Serotype: 1-9, 14 & 1/2	24	3	21	0.13	0.88
	Negative	False negative	True negative	False omission rate	Negative predictive rate
		0.87	0.22	0.82	
		True positive rate	False positive rate	F <sub>1</sub> score	
		0.13	0.78		•
		False negative rate	True negative rate		

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968 c)

			Phen	otype		
		50	23	27		
		Total population	Phenotype positive	Phenotype negative		
associated f, mrp & sly		48	22	26	0.46	0.54
		Positive	True positive	False positive	Positive predictive rate	False discovery rate
Virulence-associated markers: epf, mrp & sly	2	1	1	0.50	0.50	
	Negative	False negative	True negative	False omission rate	Negative predictive rate	
		I	0.96	0.96	0.62	
			True positive rate	False positive rate	F <sub>1</sub> score	
			0.04	0.04		•
			False negative rate	True negative rate		

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970 d)

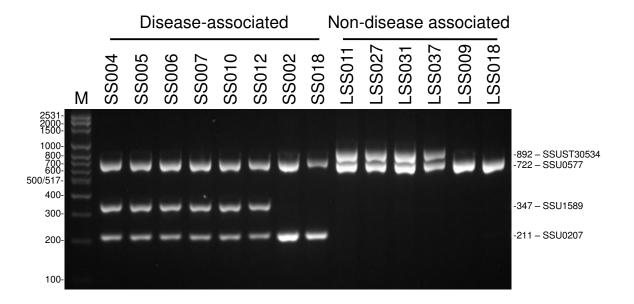
		Phen	otype		
	50	23	27		
	Total population	Phenotype positive	Phenotype negative		
	16	16	0	1.00	0.00
Multilocus Sequence Type: 1	Positive	True positive	False positive	Positive predictive rate	False discovery rate
ocus Seq Type: 1	34	7	27	0.21	0.79
Multile	Negative	False negative	True negative	False omission rate	Negative predictive rate
		0.70	0.00	0.82	
		True positive rate	False positive rate	F <sub>1</sub> score	
		0.30	1.00		•
		False negative rate	True negative rate		

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972 e)

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			Phen	otype		
		50	23	27		
		Total population	Phenotype positive	Phenotype negative		
е		18	18	0	1.00	0.00
ım Core Genorr group: 1		Positive	True positive	False positive	Positive predictive rate	False discovery rate
Minimum Core Genome group: 1	32	5	27	0.16	0.84	
	Negative	False negative	True negative	False omission rate	Negative predictive rate	
			0.78	0.00	0.	
			True positive rate	False positive rate	F <sub>1</sub> score	
			0.22	1.00		1
			False negative rate	True negative rate		

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Primer name	Primer sequence (5' - 3')	Marker of	Multiplex-PCR	Predicted biological function
			amplicon size (bp)	(Interpro)
SSU0207_0735F	TTACAAGAACAGGCAAGACAGTCGCC	Disease-association	211	Copper exporting ATPase 1
SSU0207_0945R	GCTGCTTTATAATCTGGGTGTTCGTTG			
SSU1589_0460F	CCTTTAATGCAGGGACAAAAGTGAGCTC	Disease-association	347	Type I restriction-modification (RM)
SSU1589_0806R	CCCATAATCTTACAGTTAACTTCCTTGC			system S protein
SSUST30534_0368F	ATCCCCTCCCAATAAAAGATTTGGATGC	Non-disease association	892	Putative sugar ABC transporter
SSUST30534_1259R	TTTTCGAGCTCTCCATACACTGCTTCTG			
SSU0577_0086F	CAGGTAGTTTGGGCTTAGCTTCATCAGG	Streptococcus suis sp.	722	Sporulation regulator (WhiA)
SSU0577_0807R	TGGATGCTGAATTCGCAACTGGGCAATC			

Table 2. Contingency tables used to calculate the performance metrics summarising the classification of Streptococcus suis isolates in the out-of-sample test collection (n=50). Contingency tables used to calculate and summarise the performance metrics of two binary diagnostic tests. Each table compares the observed clinical phenotype (considered the 'gold-standard' in this study) to the use of the a) newly described pathotyping markers, b) serotypes: 1-9 and 1/2, c) Virulence-associated markers: epf, mrp, and/or sly, d) Multilocus sequence type (MLST): 1, and e) assignment to Minimum Core Genome (MCG) sequence type: 1 as markers of invasive disease.

a)

		Phen	otype		
	50	23	27		
	Total population	Phenotype positive	Phenotype negative		
	19	19	0	1.00	0.00
mPCR	mPCR positive	True positive	False positive	Positive predictive rate	False discovery rate
	31	4	27	0.13	0.87
	mPCR negative	False negative	True negative	False omission rate	Negative predictive rate
		0.83	0.00	0.90	
		True positive rate	False positive rate	F <sub>1</sub> score	
		0.17	1.00		•
		False negative rate	True negative rate		

MOZ

b)

		Phenotype			
	50	23	27		
	Total population	Phenotype positive	Phenotype negative		
2	26	20	6	0.77	0.23
Serotype: 1-9, 14 & 1/2	Positive	True positive	False positive	Positive predictive rate	False discovery rate
e: 1-	24	3	21	0.13	0.88
Serotyp	Negative	False negative	True negative	False omission rate	Negative predictive rate
		0.87	0.22	0.82	
		True positive rate	False positive rate	F <sub>1</sub> score	
		0.13	0.78		•
		False negative rate	True negative rate		

c)

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		Phen	otype		
	50	23	27		
	Total population	Phenotype positive	Phenotype negative		
	48	22	26	0.46	0.54
Virulence-associated markers: epf, mrp & sly	Positive	True positive	False positive	Positive predictive rate	False discovery rate
	2	1	1	0.50	0.50
	Negative	False negative	True negative	False omission rate	Negative predictive rate
		0.96	0.96	0.62	
		True positive rate	False positive rate	F <sub>1</sub> score	
		0.04	0.04		•
		False negative rate	True negative rate		

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d)

		Phenotype			
	50	23	27		
	Total population	Phenotype positive	Phenotype negative		
	16	16	0	1.00	0.00
Multilocus Sequence Type: 1	Positive	True positive	False positive	Positive predictive rate	False discovery rate
ocus Seq Type: 1	34	7	27	0.21	0.79
Multile	Negative	False negative	True negative	False omission rate	Negative predictive rate
		0.70	0.00	0.82	
		True positive rate	False positive rate	F <sub>1</sub> score	
		0.30	1.00		•
		False negative rate	True negative rate		

e)

			Phenotype			
		50	23	27		
		Total population	Phenotype positive	Phenotype negative		
Minimum Core Genome group: 1		18	18	0	1.00	0.00
	ıp: 1	Positive	True positive	False positive	Positive predictive rate	False discovery rate
	grou	32	5	27	0.16	0.84
		Negative	False negative	True negative	False omission rate	Negative predictive rate
			0.78	0.00	0.	
			True positive rate	False positive rate	F <sub>1</sub> score	
			0.22	1.00		
			False negative rate	True negative rate		