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Functional genetic variation in *pe/ppe* genes contributes to diversity in *Mycobacterium tuberculosis* lineages and potential interactions with the human host

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Introduction: Around 10% of the coding potential of *Mycobacterium tuberculosis* is constituted by two poorly understood gene families, the *pe* and *ppe* loci, thought to be involved in host-pathogen interactions. Their repetitive nature and high GC content have hindered sequence analysis, leading to exclusion from whole-genome studies. Understanding the genetic diversity of *pe/ppe* families is essential to facilitate their potential translation into tools for tuberculosis prevention and treatment.

Methods: To investigate the genetic diversity of the 169 *pe/ppe* genes, we performed a sequence analysis across 73 long-read assemblies representing seven different lineages of *M. tuberculosis* and *M. bovis* BCG. Individual pe/ppe gene alignments were extracted and diversity and conservation across the different lineages studied.

Results: The *pe/ppe* genes were classified into three groups based on the level of protein sequence conservation relative to H37Rv, finding that >50% were conserved, with indels in *pe_pgrs* and *ppe_mptr* sub-families being major drivers of structural variation. Gene rearrangements, such as duplications and gene fusions, were observed between *pe* and *pe_pgrs* genes. Inter-lineage diversity revealed lineage-specific SNPs and indels.

Discussion: The high level of *pe/ppe* genes conservation, together with the lineage-specific findings, suggest their phylogenetic informativeness. However, structural variants and gene rearrangements differing from the reference were also identified, with potential implications for pathogenicity. Overall, improving our knowledge of these complex gene families may have insights into pathogenicity and inform the development of much-needed tools for tuberculosis control.

KEYWORDS

Mycobacerium tuberculosis, genomics, MTBC, diversity, pe/ppe family of genes

1. Introduction

Tuberculosis (TB) disease, caused by bacteria of the *Mycobacterium tuberculosis* complex (MTBC), is a major global public health problem with drug resistance making its control

difficult (World Health Organization, 2021). The available vaccine, Bacillus Calmette-Guérin (BCG), has limited efficacy and recent attempts to develop more productive vaccines have been unsuccessful, in part due to the insufficient understanding of host-pathogen interactions (Sable et al., 2019). The MTBC genome has a low overall genetic diversity and a striking clonal population structure, with nine lineages (L1-L9), which are postulated to have different impacts on pathogenesis, disease diagnosis, treatment outcome and vaccine efficacy (Coscolla and Gagneux, 2014; Tientcheu et al., 2016, 2017). Of the nine phylogeographic lineages identified, three are referred to as evolutionarily "ancient" (L1, L5, L6), and three "modern" (L2, L3, and L4). While some genetic differences between lineages have been identified (Napier et al., 2020), the molecular mechanisms responsible for differences in pathogenesis and virulence remain largely unknown.

The H37Rv *M. tuberculosis* (*Mtb*) genome has unique pe(n=100)and ppe (n=69) genes, which are found in larger numbers in pathogenic mycobacteria compared to saprophytic or avirulent species (Gey van Pittius et al., 2006; Akhter et al., 2012; McGuire et al., 2012), and therefore suggested to play a role in pathogenicity and virulence. These two families constitute ~10% of the Mtb coding potential and have a conserved N-terminal domain, within which signature prolineglutamate (PE) and proline-proline-glutamate (PPE) motifs can be identified in most of the protein products (Cole et al., 1998). In contrast, the C-terminal sequences are more variable and of various sizes. Their evolution and expansion have been proposed to be linked to a series of duplication events of the early secreted antigenic target 6kDa (ESAT-6) gene clusters (Gey van Pittius et al., 2006; Abdallah et al., 2007), together with insertions/deletions (indels) and homologous recombination (Medha and Sharma, 2021). Often, pe/ppe genes are hotspots of polymorphisms and recombination, showing higher diversity than the rest of the genome, while others are conserved across lineages, implying different functional roles (Talarico et al., 2005, 2008; Karboul et al., 2008; McEvoy et al., 2012; Copin et al., 2014; Phelan et al., 2016).

Despite the function of PE and PPE proteins being poorly understood, some have been demonstrated to have various roles in host-pathogen interactions and immune evasion. Their subcellular localization requires them to be secreted by the ESX system (Ates, 2020), with PPE38 playing an essential role in the secretion of PE_PGRS and PPE_MPTR proteins (Ates et al., 2018). The disruption of *ppe38* observed in Beijing strains (L2) has been associated with a hypervirulent phenotype (Ates et al., 2018), thereby demonstrating how strain-specific structural variants may affect pathogenesis and virulence of different MTBC lineages. The *pe/ppe* proteins are highly immunogenic and, therefore, promising targets for vaccine and diagnostic development (Qian et al., 2020). The apparent polymorphic and repetitive nature of these genes was proposed as a source of antigenic variation (Talarico et al., 2008; Tundup et al., 2008; Akhter et al., 2012); however, highly conserved T-cell epitopes have been found among *pe_pgrs* genes (Copin et al., 2014), which contradict this theory.

To provide a better understanding of the role of *pe/ppe* genes in pathogenesis, immune evasion and complement immunogenic assays and evaluations of vaccine candidates, there is a need to fully characterize the genetic diversity across the different MTBC lineages. However, *pe/ppe* genes have been systematically excluded from analyzes due to the difficulties in reliably aligning sequences to the high GC repetitive regions (Phelan et al., 2016; Meehan et al., 2019; Ates, 2020). Although the availability of high throughput short sequencing technologies has revolutionized the study of MTBC genetic diversity, an increased number

of coverage blind spots in short-read sequencing occurs in *pe/ppe* genes (Modlin et al., 2021). This limitation can be overcome by long-read sequencing technologies, such as the PacBio and Oxford Nanopore Technology platforms (Elghraoui et al., 2017). To characterize these elusive genes and genetic variants, we have performed an *in-silico* analysis of the 169 *pe/ppe* gene sequences across 73 MTBC isolates with (near-)complete assembled genomes, representing seven different lineages. We have classified the *pe/ppe* genes based on their conservation profiles across the MTBC, identifying lineage-specific markers among the conserved genes and lineage patterns responsible for disrupted protein sequences, likely to have functional consequences. Overall, using long read sequence data, we provide the first comprehensive analysis of the genetic diversity among the *pe/ppe* families to assist the development of TB control tools.

2. Materials and methods

2.1. Selection of samples, culture and sequencing

A total of 73 PacBio assemblies were used for the analysis. Ten samples were cultured at LSHTM CL3 laboratories and sequenced for this study, sourced from TB patients in the Karonga district (Malawi) between 2001 and 2009. Briefly, Mtb clinical isolates derived from patient's sputum were cultured to mid-log phase (optical density=0.6-0.8) in Middlebrook 7H9 supplemented with 0.05% Tween 80 and 10% albumin-dextrosecatalase (ADC) at 37°C in roller bottles. DNA was extracted from passage 2 by heat-inactivation followed by the CTAB-chloroform-isoamyl alcohol method (Somerville et al., 2005). DNA samples were sequenced with single-molecule real-time (SMRT) sequencing technology from Pacific Biosciences (PacBio) RSII through The Applied Genomics Center at LSHTM. To generate genome assemblies, de novo methods were performed on the raw sequencing data from the ten isolates together with other 27 samples previously sequenced (Phelan et al., 2018; Gomez-Gonzalez et al., 2019), using Flye software (Kolmogorov et al., 2019). These assembled genomes were base corrected using Illumina short-reads using Pilon software (Walker et al., 2014). The Illumina short-reads from the different samples used for the assembly improvement were publicly available from previous studies (for accession numbers, see Supplementary Table S1). The remaining 35 assembled genomes studied were publicly available and sourced from the ENA (for accession numbers, see Supplementary Table S1). To ensure robust inference, only high-quality assemblies with a maximum of 8 contigs were included in the analysis. Lineage and sub-lineage profiling were performed using TB-Profiler software (Phelan et al., 2019).

2.2. Whole-genome population genetics analysis

The H37Rv reference genome (ASM19595v2) was used for the population genetics analysis. Snippy software (Seemann, 2015) was used to simulate reads from assemblies and to call variants (SNPs and indels with a minimum coverage of 10 and a minimum fraction differing from reference of 0.9) at a whole-genome level against the H37Rv reference genome. No regions were excluded from this analysis. The R packages PopGenome (Pfeifer et al., 2014) and SeqinR (Charif and Lobry, 2007) were used for the population genetics analysis. In brief, Nei's π nucleotide

diversity per site (SNP π), indel diversity per site (indel π) and absolute divergence (dxy) were calculated in sliding windows throughout the genome for the different populations (e.g., ancient and modern lineages). The average of the three parameters was calculated for the comparison between populations. The dN/dS pairwise ratios were calculated by concatenating the coding regions relative to the reference H37Rv. Statistical differences in diversity and divergence parameters between gene functional groups were calculated using analysis of variance (ANOVA), where *p*-values were corrected by multiple comparisons using Tukey's Honest Significant Differences (HSD) test. Functional groups were considered as defined previously (Cole et al., 1998). IQ-TREE software (Nguyen et al., 2015) was used for the phylogenetic reconstruction of maximum-likelihood trees using a GTR+I+G substitution model using SNPs and/or indels alignments of the samples analyzed. The NCBI prokaryotic genome annotation pipeline PGAP (Tatusova et al., 2016) was used to annotate the genomes and validate gene rearrangements. Differences in annotation calls between H37Rv and other isolates were investigated using BLAST searching and matching.

2.3. *Pe/ppe* gene extraction, alignment and classification

The pe and ppe gene alignments were generated using a customized pipeline. In brief, non-pe/ppe flanking genes were found in the assemblies using blast software (Zhang et al., 2000) and used as anchors to extract the sequence, which were subsequently aligned with MAFFT software (Katoh and Standley, 2013). Where flanking genes were in different contigs or could not be mapped to the assemblies, genes were considered missing in samples. Single pe/ppe gene alignments were obtained relative to the H37Rv sequences and curated manually if necessary. SNPs and indels for each gene were obtained using the H37Rv reference. Levels of disruption that these variants caused on the protein sequence were assigned (0 = no variants or synonymous SNPs; 1 = non-synonymous SNPs; 2 = in-frame indels; 3=SNPs or frameshifts leading to changes in start/stop codons, deletions of >50% of the coding region, or completely missing or insertions >1,000 bp). To investigate whether individual genes were conserved across the different lineages, each pe/ppe gene was classified into one of the three classes or categories: conserved (C), structurally non-conserved (S), and unique k-mer profile (K) (see Supplementary Figure S1). Briefly, for each gene alignment, if two or more isolates were assigned a value of 3 as described above, the gene was considered structurally non-conserved (class S). In some genes, some samples had a high density of SNPs in some regions while still maintaining the same sequence length as the reference. Other genes had samples that contained completely novel sequences insertions. In an attempt to characterize the presence of these, DSK software (Rizk et al., 2013) was used to count k-mers. For each gene alignment, the k-mer profile was obtained. Those that did not show structural variants but had enrichments of unique k-mers because of SNPs or indels, were considered as class K.

2.4. Illumina short-read data analysis

A database of ~30k isolates ("30k dataset") with short-read Illumina data and representing every lineage (L1-L6 and *M/bovis*) was used (Napier et al., 2020). Short-reads were aligned to the reference with BWA-MEM (Li and Durbin, 2009), and the coverage per gene per sample was calculated with BEDTools software (Quinlan and Hall, 2010). The coverage was normalized by four housekeeping genes (*gyrA*, *gyrB*, *rpoB* and *rpoC*) and compared between *pe/ppe* genes and the rest of the genome. For the comparison between groups, *pe/ppe* genes were divided into the previously explained categories (C, S, K). In some cases, categories were combined if samples sizes were small. Statistical differences in the means between categories were assessed using T-tests.

2.5. The *pe* and *ppe* genes sequence analysis

For the population genetics analysis of the individual *pe/ppe* genes, the alignments obtained by the previous pipeline were used. Population genetics parameters (nucleotide and indel diversity and divergence) for individual genes were calculated using PopGenome R package (Pfeifer et al., 2014). The BUSTED method was used to calculate *dN/dS* ratios (Murrell et al., 2015). Identification of known domains was performed with Pfam software (Mistry et al., 2021). T-tests were applied to calculate the statistical differences for nucleotide and indel diversity between the different domains or gene groups. AlphaFold software (Jumper et al., 2021) was used for the prediction of protein structure models. For all variants identified in *pe/ppe* genes, fixation index (F_{ST}) values were calculated to assess allele differences across lineages. As a validation of variants with F_{ST} values of 1 (perfect differentiation), allele frequencies in the 30 k dataset were obtained (Napier et al., 2020). For the consideration of lineage-specific variants, an allele frequency of 0 in other lineages and >0.95 in the corresponding lineage was required.

2.6. Data availability

The sequence data supporting the conclusions of this article have been deposited in the ENA (Supplementary Table S1 for accession numbers).

3. Results

3.1. Genome-wide SNP and indel nucleotide diversity

A total of 73 clinical MTBC isolates with PacBio long-read sequencing data and complete genomes (Phelan et al., 2018; Gomez-Gonzalez et al., 2019) were included in the analysis (Supplementary Table S1). These isolates represented eight different lineages of the MTBC, including ancient (n: 11 L1, 2 L5, 7 L6), modern (n: 20 L2, 5 L3, 27 L4 including H37Rv and H37Ra) and one from each of *M. bovis* BCG and L8 (see Methods and Supplementary Table S1 for detailed information). The maximum SNP distance differences by lineage were > 350 SNPs, ensuring there was genetic diversity among isolates. All genomes were aligned to the reference H37Rv, and a total of 20,144 polyallelic sites and 6,632 indels were identified genome-wide

across the 73 isolates. Differences in per SNP and indel nucleotide diversity (π) and absolute divergence (dxy) between the ancient and the modern lineages were observed in genomic regions containing *pe/ppe* genes (Figure 1A). The regions where high SNP or indel diversity was observed ($\pi > 9 \times 10^{-4}$) coincided with highly homologous co-localized genes and recombination hotspots (*ppe3*, *pe_pgrs3/4*), or previously described highly diverse loci (*ppe1*, *pe_pgrs9/10*, *pe_pgrs50*, *pe_pgrs53-57*, *ppe55*, *ppe57-59*) (Phelan et al., 2016), where diversity was found suggestive of lineage-specific structural patterns.

Overall, a higher mean diversity across the whole-genome was obtained among ancient lineages (SNP π = 3.96×10⁻⁴; indel π = 9×10⁻⁵) than within modern lineages (SNP π = 2.31×10⁻⁴; indel π = 6×10⁻⁵), despite L4 having a high value of SNP π (Supplementary Figure S2). There was significantly higher SNP and indel diversity in *pe/ppe* genes compared to other gene functional groups (*p* < 0.01; Supplementary Table S2). Similarly, *dxy* for both SNPs and indels between the ancient and modern lineages was significantly higher in the *pe/ppe* gene families compared to other functional groups (*p* < 0.01; Supplementary Table S2), suggesting its genetic diversity contributes



(A) Whole genome SNP nucleotide and indel diversity. From top to bottom, the first track shows nucleotide diversity along the chromosome, with the peaks over 0.001 highlighted in a box. The *pe/ppe* genes in the peaks of nucleotide diversity are annotated. Green bars show where *pe/ppe* genes are located along the genome. The second track shows nucleotide diversity in ancient lineages. The third track shows absolute divergence between ancient and modern lineages. The fourth track shows nucleotide diversity in modern lineages. Line in read represents SNPs diversity and in blue indel diversity. (B) Maximum likelihood phylogenetic tree reconstructed with whole genome SNPs (n = 20,144). (C) Maximum likelihood phylogenetic tree reconstructed in blue, modern lineages in pink.

to lineage differentiation and can potentially classify MTBC lineages. Maximum-likelihood phylogenetic trees constructed using the genome-wide SNPs and indels resulted in the expected clustering by lineage (Figures 1B,C).

3.2. Conservation and disruption of the *pe* and *ppe* families across the MTBC lineages

Individual alignments for the *pe/ppe* genes were obtained (see Methods) to overcome the mapping problems in their repetitive and GC rich regions. The level of disruption caused by variants, relative to the H37Rv reference, was assigned for each isolate and each gene (see Methods and Supplementary Figure S1). The number of truncated or absent *pe/ppe* genes per lineage varied from 4 (L4.9) to \geq 30 in the most distant lineages (L5/6/8 or M. bovis BCG), while the number of genes with complete conserved protein sequences per lineage was on average 109 for L4, decreasing to 60 for the most distant lineages on the phylogenetic tree (Figure 2). Overall, isolates had >55% of their *pe/ppe* genes relatively conserved, only harboring non-synonymous SNPs at most (median 118, range 93-163). Additionally, the 169 pe/ppe genes were classified into three different classes based on the presence of structural variants, namely those are: (i) conserved (C) (79/169; 27 pe, 20 pe_pgrs and 32 ppe), (ii) structurally non-conserved (S) (85/169; 9 pe, 40 pe_pgrs and 36 ppe), and (iii) with a unique k-mer profile (K) (5/169; 4 pe_pgrs and 1 ppe) (see Methods, Supplementary Tables S3, S4; Supplementary Figure S1). The genes in class K were those with large numbers of polymorphisms that could not be classified otherwise. Based on this classification, 46% of the *pe/ppe* genes were found conserved across the MTBC lineages analyzed (for a list of conserved genes see Supplementary Table S5).

To support the classification of the genes into the three classes, we analyzed short-read sequencing data from ~30 k dataset (Napier et al., 2020). Mean normalized coverage of the *pe/ppe* genes (0.74) was found to be lower than the rest of the genome (0.93; p < 0.01). There was the expected depletion in coverage in repetitive regions; however, not all *pe/ppe* genes fell in coverage blind spots. Because of their repetitive regions, both class K and S *pe/ppe* genes had lower mean coverage (combined: 0.67) compared to class C (0.82) or the rest of the genome (0.93; p < 0.01; Supplementary Figure S3A). Identifying the genes with lowest coverage values revealed 70 *pe/ppe* genes in troughs of low coverage had been classified into the two non-conserved classes (S, K), highlighting difficulties in robustly characterizing their variants using a short-read alignment approach.

Subfamilies V (as defined by Gey van Pittius et. al. (Gey van Pittius et al., 2006)) of *pe/ppe* genes (mainly formed by *pe_pgrs* and *ppe_mptr*) are known to carry the most repetitive and polymorphic regions. Within the *pe* family, although 41% of the genes in subfamily V were structurally conserved, including 20 *pe_pgrs* (Supplementary Figure S4A), most of the structural diversity found across our isolates was observed in this group. Frequent differences in predicted protein lengths were found driven by deletions (Supplementary Figure S4B). Interestingly, the *dN/dS* ratio in *pe_pgrs* was 0.57 compared to 1.20 in the rest of the *pe* genes, suggesting negative selection effects. In contrast with the *pe* family, *ppe* subfamilies II and III harbored disruptive variants, including frameshifts, IS6110 insertions, or other changes in the open reading frame (ORF) leading to



FIGURE 2

Heatmap showing the structural classification of each gene for each sample. Each row represents a separate sample, following the order based on the phylogenetic tree shown on the left. Genes on columns, pe family on the left, ppe family on the right. In green, genes without variants or synonymous SNPs; in yellow, genes with non-synonymous SNPs; in orange, genes with in-frame indels; in red, genes with frameshifts, changes in start/stop codons or large deletions. Top track shows the sub-family of each gene based on a previous classification (Gey van Pittus et al., 2006). Bottom track summarizes the structural classification of each gene across all samples in one of the following categories: structurally conserved (class C) in green, structural variants (class S) in red and unique k-mer profile (class K) in yellow. Barplot on the right shows the distribution of genes with each type of variant by sample.

gene fusions (Supplementary Figure S4C). Subfamily V *ppe_mptr* genes accounted for the highest level of disruption in protein sequence, with 16 genes in class S (16/24). In general, the largest variation in gene length was given by different numbers of the pentapeptide repeat (MPTR domain) and the integration of the IS6110 insertion (Supplementary Figure 4D).

3.3. Nucleotide diversity in *pe/ppe* genes

SNP and indel diversity were calculated for the 169 *pe/ppe* gene sequence alignments across the classes (C, S, K) (Supplementary Tables S3, S4). As expected, indel diversity in genes

from class S was significantly higher than in class C (S mean indel π =5.85×10⁻⁴, C mean indel π =8.3×10⁻⁵; *p*<0.001; Figure 3A). However, there were no significant differences of SNP π between classes. SNP π was heterogenous among class C and S genes (range 0 to >0.002). As expected, due to its polymorphic nature, the class K genes (*n*=5) had a higher SNP diversity (mean SNP π <7×10⁻⁴). A very weak correlation between SNP and indel diversity at a gene level was found (Spearman's rho=0.042; Supplementary Figure S5).

Overall, the *pe_pgrs* subfamily accounted for the majority of the indel diversity compared to other *pe* and *ppe* genes (Figure 3B), but diversity in the individual genes varies significantly (range indel π from <2×10⁻⁵ to >2×10⁻³). Interestingly, among *ppe* gene subfamilies,



Boxplots of SNP and indel diversity in the 169 *pe/ppe* genes compared by (A) gene classification; (B) gene family and (C) domain within gene family. Outliers with π > 0.005 in (A) and (B) and π > 0.01 in (C) have been removed from figure. Adjusted *p*-value at (*) 5%, (**) 1%, (***) 0.1% or (****) 0.01% significance levels. *ppe_svp* (IV) genes showed higher values of SNP and indel π than *ppe_mptr* (V) (Supplementary Figure S5). In accordance with the rest of the genome, *pe/ppe* genes in ancient lineages had a higher SNP π than modern counterparts (ancient mean SNP π =6.7×10⁻⁴; modern mean SNP π =4.2×10⁻⁴). Intra-lineage diversity was studied for those genes with representatives in greater than 5 lineages. There was a total of 34 and 32 genes where SNP or indel diversity, respectively, was zero for at least 4 of the 5 lineages studied, representing a situation where diversity is being driven by a single lineage or inter-lineage differences. Diversifying selection was found in 19 *pe*, 16 *pe_pgrs* and 19 *ppe* genes (*dN/dS*>1.5; genome-wide average 0.71). Despite showing selection pressure, thirty of these genes belonged to class *C*, without structural variants (Supplementary Tables S3, S4). Genome-wide, only the "insertion sequences" functional (gene ontology) group showed a *dN/dS* ratio>1 suggesting positive selection.

Diversity in the different domains showed that PE and PPE domains have low indel diversity (Figure 3C), suggesting specific structural conservation. In addition, these PE and PPE domains had a higher SNP nucleotide diversity than indel diversity (p<0.01) except in the *pe_pgrs* subfamily. Indel diversity was greater after the PE domain in *pe_pgrs* genes (p<0.01), while diversity in the *ppe* genes and the rest of the *pe* family was predominantly a result of SNPs.

3.4. Large insertions

Class S genes had a high abundance of large insertions that could be distinguished into two types: (i) >25 bp insertions with >70% identity to same or other *pe/ppe* genes, and (ii) the integration of the IS6110 insertion sequence. The former corresponds mostly to sequences of repetitive regions in *pe_pgrs* and *ppe_mptr* genes, which could result from homologous recombination, and follow a lineagespecific pattern in several cases. Duplications were also identified, including an extra copy of ppe53 found in all isolates except L4.3 to L4.9 and L8 with the same N-terminal but different C-terminal domain (Supplementary Figure S6). The integration of IS6110 was observed in regions around *pe/ppe* genes, which were similar across the different lineages (Supplementary Figure S7). Thirteen genes (1 pe and 12 ppe, including 9 ppe_mptr) were found to harbor IS6110 in at least one isolate (Supplementary Table S6), which was responsible for the disruption of the protein sequence. Genes known to have IS6110 inserted, such as ppe34, had a high frequency of integration (34 isolates, including all L2-3), leading to two shorter ORFs compared to H37Rv-PPE34 (Supplementary Figure S8), confirmed with PGAP annotation. The ppe38-40 loci are a hotspot for the insertion of IS6110. This genomic region as annotated in the H37Rv reference is rarely found in clinical isolates, but one often encounters the ppe71 duplication (Ates et al., 2018). We observed the two esx flanking genes and *ppe71* in many isolates (n = 38/72; 52.8%), including the laboratory strains H37Rv and H37Ra (Supplementary Figure S9). Nevertheless, all Beijing (L2.1.1) isolates had only a single copy, which was truncated losing the PPE domain by the insertion of IS6110, and has been found to suppress the secretion of PE_PGRS and PPE_MPTR proteins (Ates et al., 2018). The contiguous gene, ppe39, has been described in an extended version in Beijing isolates (Han et al., 2015). Most isolates (except H37Rv/Ra and L4.6/L4.9) had an extra ~268 residues at the N-terminal, which included a PPE domain that appears truncated by the integration of IS6110 in the reference genome.

3.5. Other gene fusions and duplications

We found 10 pairs of *pe/ppe* genes that showed potential gene fusions compared to the H37Rv reference, including the fusion of the PE and PGRS domains of adjacent genes. The pe_pgrs4/3 (L2) and pe_pgrs20/19 (L1) loci are two examples of the fusion of domains in single lineages, where a large deletion covering the end of the upstream gene leads to the merging of these two adjacent genes forming a pe_ pgrs gene (Figure 4A). Using AlphaFold software, the predicted protein structure of the PE_PGRS4/3 fusion in L2 revealed a PE_ PGRS protein highly similar to PE_PGRS3 and PE_PGRS4 (Figure 4B). For ppe6/5, ppe8/7, pe_pgrs12/13, pe_pgrs50/49, pe_ pgrs55/56 and ppe67/66 gene pairs, the ORF continued until the end of the second gene due to a frameshift caused by a small indel (Figure 4A). Interestingly, *pe_pgrs12* and *pe_pgrs55* have a PE domain, while in the downstream genes *pe_pgrs13* and *pe_pgrs56* this domain is absent, only showing PGRS motifs, and therefore, their combination leads to a PE_PGRS-like structure inferred by AlphaFold software (Figure 4C). Likewise, for ppe6/5 and ppe8/7, the ppe5 and ppe7 loci do not have any PPE domain, thereby the gene fusion leads to a PPE_ MPTR-like structure. Finally, there are four *pe/ppe* genes in *Mtb* annotated as pseudogenes (pe21/pe_pgrs36 and ppe48/ppe47), where small indels causing frameshifts led to a change in ORF and the consequent formation of single PE_PGRS-or PPE_MPTR-like genes (Figure 4A). All these gene fusions were confirmed by PGAP annotation.

The *pe_pgrs3* locus is a recombination hotspot (Phelan et al., 2016), and several large indels were identified when aligned to the H37Rv reference, including insertions linked to duplication of repetitive regions. These observations confirm the non-conserved nature of the *pe_pgrs3* gene, and make it difficult to characterize with the usual variant calling pipelines. Surprisingly, the protein sequences obtained from the aligned region showed a duplication of *pe_pgrs3* in almost every sample analyzed (Figure 4A), confirmed by the annotation of the assemblies obtained by PGAP. The two pe_pgrs3 genes identified were highly similar differing in the presence/absence of the C-terminal domain from H37Rv-pe_pgrs3 gene, which also shows lineage-specific patterns. Despite the lack of concordance with the reference, we observed a significant degree of conservation within lineages. The *pe_pgrs3* gene is duplicated in *M. bovis* and *M. canetti*, and until now, it was believed not to be duplicated in Mtb (De Maio et al., 2020). Laboratory strains (H37Rv and H37Ra) retained a unique gene, which combines N-terminal and C-terminal from the ancestral two copies, while the other lineages carried two copies of the gene differing from M. bovis (L1, L3 or L4) in the C-terminal domain of one of the copies (Supplementary Figure S10).

3.6. Lineage-specific SNPs and indels in *pe/ppe* genes

A total of 3,649 SNPs and 1,319 indels were identified among the *pe/ppe* genes, from which 459 SNPs and 122 indels were found in the class C genes. A completely conserved protein sequence across all lineages was only found for seven genes (*ppe7*, *pe9*, *pe13*, *pe19*, *pe22*, *pe25* and *pe_pgrs40*), including *ppe7* that differed from the H37Rv reference by a 1 bp insertion. The existence of inter-without intra-lineage diversity in some genes suggested a potential lineage-specific



PGRS13 (C) and PE_PGRS36 (D); in pink PE_PGRS3 (B), PE_PGRS12 (C) and PE21 (D).

pattern. We performed a principal component analysis with SNPs and indel matrices for each sample (excluding L8). Clustering by lineage was clear for indels, and with sub-groups for some lineages being observed using SNPs (Supplementary Figure S11). Following the hypothesis of having lineage-specific markers within these genes, we built three maximum likelihood phylogenetic trees with only these SNPs, indels, and both (Supplementary Figure S12). Sixteen genes with >1.5% of its coding region being polymorphic sites or with a unique *k-mer* profiler were removed to reconstruct the SNP tree. The topologies of the trees obtained with SNPs and indels were different; however, both showed a clear clustering by lineage, suggesting lineagespecific patterns.

The fixation index (F_{ST}) was calculated to identify these lineagespecific polymorphisms, comparing one lineage against the others for each of the variants found in *pe/ppe* genes across the 72 available genomes. Overall, 83 SNPs and 8 indels (including SNPs and frameshifts leading to disrupted proteins) were identified with an F_{ST} of 1 in one lineage within our dataset and with an allele frequency > 0.95 in the corresponding lineage within the ~30 k dataset (Table 1; Supplementary Table S7).

Differences between H37Rv annotation and the rest of genomes were also found in the PGAP annotation calls. The differences were mostly due to the lineage diversity, duplications or gene fusions already discussed; however, some small sequences (\leq 500bp) were

identified in some samples as *pe/ppe* family proteins. In brief, three small sequences were found upstream of *ppe59* in various samples of different lineages, and all L3 and almost all L2 had a sequence (492–647 bp) predicted as PE family protein upstream of *pe_pgrs35*. In L8, a 1,665 bp sequence not present in any other lineage was predicted as PE family protein, which corresponds with one of the "new *pe/ppe* genes" discovered in L8 (Ngabonziza et al., 2020).

4. Discussion

The *pe* and *ppe* genes are important MTBC loci, but are routinely excluded form whole-genome sequencing studies, especially those using short sequence data, due to difficulties in accurately mapping their repetitive and polymorphic regions (Meehan et al., 2019). In a recent study, Marin et al. (2022) found several *pe/ppe* genes with good mappability and variant detection with short-read sequencing platforms that could be included in WGS analysis with confidence. The majority of these genes belonged to class C in our analysis, congruent with a better coverage of the conserved *pe/ppe* genes found also in previous studies (Gómez-González et al., 2022), while class K genes were found with lower scores for Illumina mappability. Long read sequencing technologies can be of use to overcome this problem (Gómez-González et al., 2022). We analyzed PacBio assemblies to

Lineage	Class C Gene [mutation]	Class S Gene [mutation]	Class K Gene [mutation]
Ancient	ppe4 [A185A], ppe28 [C144W], pe_pgrs44 [G478G]	ppe5 [S1765F], ppe8 [I3250F; 9889_9890insATA***], ppe12 [R545K], pe_pgrs14 [A246A], pe_pgrs47 [G383G]	
L1	pe4 [K164N], ppe2 [T412T], pe_pgrs5 [G225D], pe_pgrs7 [G951R], pe_pgrs11 [G280R], ppe13 [G336G], ppe44 [G59V], ppe61 [T257M], ppe63 [Y365N]	ppe5 [11273V], ppe8 [139_139del, V118A], pe_pgrs14 [G668D], pe_pgrs47 [S20S], ppe64 [G306S]	
L2	ppe17 [P167L], pe14 [A106A], pe16 [A96A], pe24 [G216V], pe_pgrs58 [A314V]	pe_pgrs22 [G730G]	
L3	<i>pe3</i> [S175P], <i>pe4</i> [F197S], <i>ppe4</i> [L52M], <i>ppe10</i> [W147S], <i>pe_pgrs7</i> [G405G], <i>pe17</i> [T285I], <i>pe_pgrs30</i> [T600N], <i>pe26</i> [S330L], <i>ppe48</i> [I64L]	pe1 [G369R], ppe5 [G960A], ppe8 [D741N, S1920F], pe_pgrs6 [A124V], pe_pgrs10 [G799G], ppe19 [F4V], ppe33 [G22S], ppe54 [G2189S], ppe64 [63_64del*]	pe_pgrs45 [G437G]
L2/3		pe10 [337_337del****]	
L4		lipY [A58G]	
L4.1		<i>pe_pgrs16</i> [1968_1969insG*]	
L5	ppe2 [E140G, D431N], ppe3 [E448D], ppe14 [T293M], pe12 [L217F], pe_pgrs24 [L101R], pe_pgrs30 [R115L], ppe29 [A366P], ppe31 [H188Y], ppe36 [I25I], pe_pgrs39 [A109T], pe_pgrs40 [D29D], pe26 [G160S], pe_pgrs44 [A439A], pe_pgrs59 [G22D]	ppe8 [F414V], ppe12 [G378S], pe_pgrs22 [G118G], ppe16 [G349R], pe_pgrs23 [G280G], ppe24 [S716R], pe_pgrs32 [E76D; A483T], ppe37 [V124M], pe_pgrs42 [G125G], ppe43 [449_454del*], lipY [F129S], pe_pgrs55 [1411_1411del*]	<i>ppe18</i> [H234R]
L6	ppe1 [P298P], ppe3 [M450T], ppe10 [G288A], ppe13 [N244N], ppe23 [S37P], pe_pgrs43 [W1503R]	pe1 [P494L], ppe16 [1279_1283del*], ppe45 [W75* ^(*)], ppe56 [6586_6586del*]	
M. bovis	pe3 [P255T], ppe10 [W8 ^{*(*)}], ppe20 [V94A], pe_pgrs30 [A172V]	pe1 [G26R], ppe8 [G2403G], pe_pgrs15 [L113L], ppe25 [925_927del**], pe_pgrs41 [S26N]	

TABLE 1 Lineage-or clade-specific variants.

*Truncated protein; **In-frame; ***Leads to gene fusion; ****Delayed stop codon.

provide the most comprehensive picture to date of genetic diversity in all 169 pe/ppe genes. The sequence analysis revealed a large amount of both conservation and diversity across the pe/ppe families. As expected, we observed greater nucleotide diversity in *pe/ppe* genes compared to the rest of the genome, especially in some clustered loci (e.g., pe_pgrs53-57, ppe57-59), with some of them predicted to be pathogenicity islands (Xie et al., 2014). The diversity is driven not only by SNPs but also by indels, including the integration of IS6110, for which several transposition sites have been identified among pe/ ppe genes, especially within members of the ppe subfamily V (ppe_ mptr) (Yesilkaya et al., 2005; Namouchi and Mardassi, 2006; McEvoy et al., 2009, 2012; Reyes et al., 2012). Consistent with previous findings (Reves et al., 2012), we observed a tendency of occurrence of IS6110 insertions in genomic regions with *pe/ppe* genes, including some genes exhibiting lineage-specific patterns. The ppe38-40 represents a known hotspot for IS6110 integration with consequences for strain virulence (McEvoy et al., 2009; Ates et al., 2018). SNP and indel diversity were heterogeneous across the genes. The class S genes displayed greater indel diversity but a similar SNP diversity to class C. The main source of diversity in *pe_pgrs* genes was identified after the PE domain, mainly driven by indels. In contrast, diversity was more often the result of SNPs in other *pe/ppe* genes. In line with previous work, the dN/dS ratios obtained broadly varied across individual genes across *pe/ppe* families (Copin et al., 2014).

Evidence of homologous recombination, especially in repetitive regions of *pe/ppe* (Karboul et al., 2008; Phelan et al., 2016), and events of gene conversion (Karboul et al., 2006) have been described. For example, homologous recombination due to repetitive nature of the PGRS domain has been previously suggested to occur in the

pe_pgrs4/3 locus (Phelan et al., 2016), but no duplications of these genes have been previously described. We found a second copy of *pe_pgrs3* in most of the isolates, with a similar configuration of the one found in M. bovis and M. canetti (De Maio et al., 2020). Due to the similarity to the ancestral configuration, it is possible that recombination events have resulted in the loss of one copy in H37Rv and related strains, which could be suggestive of reductive evolution in Mtb. Several gene fusions compared to the annotated H37Rv were also identified in this study. Some genes were found in single lineages (e.g., *pe_pgrs20/19*), while others were in all isolates (e.g., *ppe48/47*). Interestingly, the four *pe/ppe* genes annotated as pseudogenes, organized in two operons in H37Rv, were found to form a single ORF in most isolates leading to a potentially functional protein. This lack of consistency between the H37Rv annotated sequences and the predicted protein sequences in the clinical isolates could potentially mislead and hinder the capture of variants when using sequence alignment methods. Given the lineage-specific structural variants observed, the use of lineage reference genomes could be a possibility to overcome the difficulties in capturing variants in genes where variation must be being missed due to substantial differences with the H37Rv reference, potentially improving the mappability with shortread sequencing methods. Other studies (Chitale et al., 2022) have also suggested the use of revised reference genomes of H37Rv including changes in some *pe/ppe* genes found, which would overall improve the accuracy of variant detection.

The inter-lineage diversity found in some *pe/ppe* genes, together with its substantial impact on the phylogenetic differences between the ancient and the modern lineages, suggested the presence of lineage-specific variants in these regions that could be phylogenetically

informative. We identified numerous lineage-specific SNPs and indels validated in $\sim 30 \text{ k}$ *Mtb* isolates with whole-genome sequencing data. Protein disruption was a frequent outcome of the lineage-specific indels, which considering the putative role in host-pathogen interactions of these proteins, could provide insights into different behavior between strains. One limitation of short-read sequencing data for the validation work was the lack of accuracy in detecting big indels, especially among repetitive regions.

All *pe/ppe* genes were classified based on the conservation observed across the 72 isolates. Structural variants, such as frameshifts, changes in start and stop codons, and large deletions, were responsible for the classification of numerous genes as non-conserved, which often were identified across one or multiple sub-lineages, showing an otherwise conserved profile within lineage. Member of subfamilies V were found in higher numbers among the non-conserved genes. Importantly, this classification was based on the alignment to the H37Rv sequence, which, as shown, does not always represent the functional locus. However, on average, more than half of the *pe/ppe* gene members per sample were found to be conserved, suggesting an important role. The various levels of diversity that different genes display have been proposed to imply non-redundant functions (Copin et al., 2014). Nevertheless, the complex gene layout found in the different lineages requires more investigation to understand the functional consequences of the variation observed. One difficulty is the lack of structural data for *pe/ppe* proteins that restricts the prediction of functional consequences. However, novel in silico tools, such as AlphaFold software (Jumper et al., 2021), can be of assistance.

Some pe/ppe proteins have demonstrated to act as modulators of the immune response and consequently, multiple epitopes have been characterized on these proteins being investigated as targets for vaccine development (Medha and Sharma, 2021). T-cell epitopes are found in the conserved PE domains of *pe_pgrs* genes rather than the variable sequences, supporting the hypothesis that this conservation favors infection (Copin et al., 2014; Fishbein et al., 2015). It is plausible that gene fusions where loci with missing pe/ppe domains are transcribed with an upstream locus lead to a functional protein. On this premise, understanding the structural diversity of these genes and its consequent effect on these proteins is crucial for its potential use in vaccine development, which ideally would target conserved sequences across the different lineages. Homolka et al. (2016) showed how diversity in *ppe18* could significantly impact the effectivity of a vaccine candidate. Additionally, the role of these proteins in cell wall localisation and small molecule transportation means they could be explored as drug targets (Ates, 2020).

In conclusion, although there is significant variation in *pe/ppe* genes, with this study we have found that some are relatively conserved. We have provided a list of conserved genes that could be included in whole-genome sequencing analysis rather than being excluded, especially as they can be phylogenetically informative. These proteins play an essential role in host-pathogen interactions, and therefore it is important to elucidate their function and the potential impact of diversity on pathogenicity and virulence. Future studies in a larger number of isolates, combined with functional characterization, will lead to insights that can assist with the control of the tuberculosis disease.

Data availability statement

The datasets presented in this study can be found in online repositories. The names of the repository/repositories and accession number(s) can be found in the article/Supplementary material.

Author contributions

SC, JP, and TGC conceived and directed the project. PG-G and AG undertook sample processing and DNA extraction. PG-G performed bioinformatic and statistical analyzes under the supervision of SC, JP, and TGC. LT provided data. SC led the generation of sequence data, with assistance from MH and LT. PG-G, SC, JP, and TGC interpreted results. PG-G wrote the first draft of the manuscript with inputs from JP and TGC. PG-G, JP, and TGC compiled the final manuscript. All authors contributed to the article and approved the submitted version.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fmicb.2023.1244319/ full#supplementary-material

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