1 Genetic diversity, determinants, and dissemination of *Burkholderia pseudomallei*

2 lineages implicated in melioidosis in northeast Thailand

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31 Abstract

32 Melioidosis is an often-fatal neglected tropical disease caused by an environmental bacterium 33 Burkholderia pseudomallei. However, our understanding of the disease-causing bacterial 34 lineages, their dissemination, and adaptive mechanisms remains limited. To address this, we 35 conducted a comprehensive genomic analysis of 1,391 B. pseudomallei isolates collected from 36 nine hospitals in northeast Thailand between 2015 and 2018, and contemporaneous isolates 37 from neighbouring countries, representing the most densely sampled collection to date. Our study identified three dominant lineages with unique gene sets enhancing bacterial fitness. 38 39 indicating lineage-specific adaptation strategies. Crucially, recombination was found to drive 40 lineage-specific gene flow. Transcriptome analyses of representative clinical isolates from each 41 dominant lineage revealed heightened expression of lineage-specific genes in environmental 42 versus infection conditions, notably under nutrient depletion, highlighting environmental 43 persistence as a key factor in the success of dominant lineages. The study also revealed the 44 role of environmental factors - slope of terrain, altitude, direction of rivers, and the northeast 45 monsoons - in shaping B. pseudomallei geographical dispersal. Collectively, our findings highlight persistence in the environment as a pivotal element facilitating *B. pseudomallei* spread, 46 47 and as a prelude to exposure and infection, thereby providing useful insights for informing 48 melioidosis prevention and control strategies.

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50 Introduction

51 Melioidosis, a severe infectious disease, affects an estimated 165,000 cases globally each year, of which 89,000 are fatal¹. The disease is caused by Burkholderia pseudomallei, a Gram-52 53 negative bacillus found in soil and contaminated water across tropical and sub-tropical regions. 54 Historically, limited access to microbiology laboratories for culture-confirmed diagnosis led to underreporting, particularly in lower- and middle-income countries². However, improved 55 infrastructure and awareness have led to increases in reported cases across South Asia, 56 57 Southeast Asia, East Asia³⁻⁸ and Australia^{9,10} In Southeast Asia, the disease incidence is often linked to agriculture practice, particularly during the rainy seasons when rice paddy fields are 58 flooded for planting. The flooded terrain enables the bacterium in the soil to surface, potentially 59 exposing farmers to *B. pseudomallei* and subsequently leading to melioidosis¹¹. Additionally, 60 many cases of melioidosis have been associated with severe weather events^{12–14}. While climate 61 62 likely influences human encounters with *B. pseudomallei*, further investigation is needed to fully 63 understand the mechanisms linking environmental factors to melioidosis epidemiology.

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65 Understanding the population structure, dissemination and adaptation of *B. pseudomallei* in 66 these climatically-challenged endemic regions requires a large-scale, geographically and 67 chronologically densely-sampled, genetic dataset. Previous studies, albeit limited in sample 68 size, have demonstrated that *B. pseudomallei* dissemination is driven by both anthropogenic and environmental factors^{15–18}. Streams^{19,20}, monsoons, typhoons and cyclones^{13,14,21,22} were 69 70 identified as significant contributors to bacterial dissemination, highlighting the importance of 71 bacterial persistence across a range of environmental conditions. B. pseudomallei exhibits 72 remarkable survival capabilities across diverse environments, spanning from wet to dry. nutrient-depleted soil²³⁻²⁸ thereby enabling the bacterium to thrive in various ecological niches. 73 74 Previous studies have noted the temporal and geographical co-existence of multiple B. pseudomallei lineages^{15,16,29}. However, little is known about their distinct genetic content and 75 76 adaptive strategies. Identification of lineage-specific genes associated with bacterial persistence 77 and disease escalation will be essential to develop disease control strategies.

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79 In this study, we conducted a population genomics analysis using combined B. pseudomallei 80 isolates from melioidosis patients across nine provinces in the northeast Thailand including 81 Buriram, Khon Kaen, Mahasarakam, Mukdahan, Nakhon Phanom, Roi-Et, Sisaket, Surin and Udon Thani⁸; totaling 1,265 isolates collected from July 2015 to December 2018. Additionally, 82 83 we incorporated contemporary environmental and clinical collections from Thailand and neighbouring countries^{15,29–34}, consisting of 15 clinical isolates and 111 environmental isolates 84 85 (Figure 1a, Supplementary data 1). Our comprehensive analysis, including a total of 1,391 86 isolates, revealed the population structure, dissemination patterns, and genetic diversity of this 87 bacterium. We identified genetic determinants associated with dominant lineages and investigated their biological functions and expression conditions in three representative isolates. 88 89 each representing a dominant lineage. This provides insights into the strategies employed by B. 90 pseudomallei lineages for successful persistence in the environment, ultimately leading to 91 human exposure and infection.

92

93 Results

Population structure analysis revealed the successful *B. pseudomallei* lineages and mixture of clinical and environmental isolates

96 To define the population structure of clinical and environmental *B. pseudomallei* in a 97 hyperendemic area of northeast Thailand and neighbouring regions (n = 1,391), we performed 98 four independent approaches. PopPUNK³⁵ analysis was performed on genome assemblies

(Supplementary data 2). Additionally, we constructed three maximum-likelihood (ML) 99 100 phylogenies³⁶, each based on different sets of single nucleotide polymorphisms (SNPs): core genomes (n= 77,156 SNPs), core gene multilocus typing³⁷ (cgMLST, n = 46,945 SNPs), and 101 seven-gene multilocus typing genes³⁸ (MLST, n = 31 SNPs). These approaches facilitated the 102 103 grouping of isolates with close genetic similarity into distinct lineages. Notably, both PopPUNK 104 analysis and core genome SNPs phylogeny yielded consistent results (Supplementary Figure 105 1) clustering the population into three dominant lineages (Figure 1b). The average pairwise 106 core SNP distance within each dominant lineage was 549, 351, and 517 SNPs for lineage 1, 2, 107 and 3, respectively, in contrast to the pairwise core SNP distance of 1,087 SNPs within the total 108 population (Figure 1c). This lower pairwise core SNP distance across lineages confirmed the 109 genetic relatedness as defined by PopPUNK and core genome SNP phylogenetic analysis. 110 While coMLST displayed conservation for two out of three dominant lineages. MLST exhibited 111 inconsistencies across dominant lineages with lower phylogenetic resolution and poorer 112 bootstrap support compared to other methodologies (**Supplementary Figure 1**). Consequently, 113 we relied on the population delineated by PopPUNK and core genome SNP phylogeny for 114 subsequent investigations.

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116 The three predominant lineages (denoted as lineage 1 to 3) comprised 312, 297, 125 isolates, 117 respectively. They accounted for 52.8% of the studied population and persisted throughout the 118 sampling period. Interestingly, each lineage peaked during the rainy season, correlating 119 agricultural practices at the onset of rainfall with increased environmental exposure and 120 subsequent melioidosis infections (Figure 1d). Despite the small sample size of environmental 121 isolates, we observed a clustering of these isolates with clinical isolates within each dominant 122 lineage, indicating their core genetic similarities and shared origin. The ratio of environmental to 123 clinical isolates varied across lineages (Chi-square test with Monte Carlo resampling p-value 5.00 x 10⁻⁴. **Supplementary Figure 2**). Due to the substantially lower number of environmental 124 125 isolates used and incomplete geographical distribution matching between clinical and 126 environmental isolates, caution is warranted in interpreting these results. Nevertheless, our 127 findings highlight a mixing of environmental and clinical samples, suggesting that clinical 128 isolates could serve as a surrogate for tracking the dissemination of an environmental 129 bacterium, especially in the absence of equally comprehensive environmental samples.

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131 Genetic evidence identifies patterns of *B. pseudomallei* dissemination in Northeast132 Thailand

133 We examined the dissemination patterns of *B. pseudomallei* in northeast Thailand. Except for 134 Mahasarakam where samples were limited, all three dominant lineages were present across the 135 rest of eight studied provinces (Figure 1a). This prompted us to focus the analysis on these 136 dominant lineages to identify consistent geographical distributions underlying their spread in the 137 region. We generated lineage-specific phylogenies to improve genetic resolution for 138 transmission analysis. Additionally, we reconstructed ancestral histories of provincial origins, 139 quantified the number of inter-provincial transmissions to examine transmission patterns, and 140 estimated the time of the most recent common ancestor of each dominant lineage and its sub-141 lineages (Supplementary Figures 3 – 5, see Methods). This allowed us to link the emergence 142 of these lineages to historical events that might have affected their transmission dynamics. 143 While transmission signals could reflect distinct dissemination patterns in each dominant lineage 144 or its sub-lineages, we also considered the possibility of shared factors leading to a uniform 145 geographical distribution. Notably, we observed consistent dissemination patterns in 14 out of 146 28 provincial pairs across the three dominant lineages (Figure 2a, Supplementary Figure 6). 147 Eight out of these 14 provincial pairs potentially correlated with the slope of terrain altitude 148 between provinces or the natural flow of rivers in the region (Figure 2b). These pairs included 149 "Udon Thani-to-Khon Kaen", "Udon Thani-to-Buriram", "Udon Thani-to-Mukdahan", "Udon 150 Thani-to-Surin", "Khon Kaen-to-Buriram", "Nakhon Phanom-to-Mukdahan", "Roi-Et-to-Surin" and 151 "Surin-to-Sisaket". Northeast Thailand is described as a saucer-shaped plateau, with elevations 152 ranging from over 200 meters above sea level in the northwestern corner (parts of Udon-Thani, and Khon-Kaen) to less than 100 meters in the southeast (parts of Mukdahan and Sisaket), and 153 154 gradually descending toward the Mekong River in the east^{39,40}. The Mun River originates from 155 elevated hills in central Thailand, streaming eastward through Buriram, Surin, Sisaket before 156 merging with the Mekong River. Similarly, the Chi River, a tributary to the Mun, also originates 157 from the central Thailand mountains. The Chi flows eastward through Khon Kaen, 158 Mahasarakam, Roi-Et and converges with the Mun in Sisaket⁴⁰.

Additionally, another set of three out of 14 conserved patterns coincided with the wind direction of the northeastern monsoon during the dry season ("Nakhon Phanom-to-Buriram", "Mukdahanto-Buriram", and "Mukdahan-to-Surin"). Thailand experiences two predominant monsoon seasons: the southwest monsoon from May to October and the northeast monsoon from November to April. The southwest monsoon brings heavy rainfall from the southwest to northeast, often marking the start of the agriculture season (**Figure 1c**)^{41,42}. Conversely, the northeast monsoon brings dry winds from the northeast to southwest. Our observation implies 166 the potential for dry winds to transport aerosolised soil contaminated with B. pseudomallei 167 westward. Despite that previous air sampling during Thailand's rainy season did not detect B. 168 pseudomallei, exploring the impact of dry winds during the northeastern monsoon is essential. 169 This consideration becomes even more pertinent given reports of the long-range transport of 170 particles and small organic matters, such as PM2.5 and PM10, via the northeast monsoon 171 elsewhere in Southeast Asia^{43,44}. Furthermore, we successfully estimated the time of most 172 recent of ancestry for a sub-lineage 1.3, a descendant of lineage 1. Our finding revealed that 173 this sub-lineage emerged around 2011 (95% HPD of 2000-2014, Supplementary Figure 3). 174 The age of this sub-lineage implies that older lineages, such as its parent lineage 1 likely 175 experienced multiple monsoon seasons, which possibly resulted in the observed patterns. Apart from the terrain slope, inland rivers, canal systems⁴⁵ and regular monsoons^{41,42}; various factors 176 177 likely contributed to shaping *B. pseudomallei* dissemination in northeast Thailand. 178 Anthropogenic activities, such as human migration between the provinces, may also contribute 179 to the observed pattern; however without access to comprehensive human movement data, this 180 aspect remains challenging to investigate.

181 Genetic markers potentially contributing to the emergence of successful *B. pseudomallei* 182 lineages

183 The co-existence of multiple *B. pseudomallei* lineages within the same geographical areas and 184 timeframe implies the presence of diverse adaptive strategies which enable them to thrive in a 185 shared ecological niche. While some smaller lineages may be sporadically detected, the 186 persistence of the three dominant lineages throughout the sampling period supports their fitness 187 and successful adaptive strategies in this niche. We next sought to identify genes that were 188 present in isolates that form each dominant lineage, or its sub-lineage; but absent in non-189 dominant lineages (see Methods). Out of total 15,237 genes in the pan-genome outlined from 190 this population (see Methods), 5,577 genes were conserved across the entire population while 191 9,660 genes were variably present (accessory genes). Dominant lineage-specific genes were 192 defined as accessory genes present in ≥95% of isolates within any of the dominant lineages or 193 their sub-lineages, but present in ≤15% of isolates outside these lineages. Among these, 247 194 genes were identified as lineage-specific with their specificity to each dominant lineage and sub-195 lineage tabulated in **Supplementary data 3**. The majority of dominant lineage-specific genes 196 were poorly characterised and annotated as hypothetical proteins (Figure 3). To gain insights 197 into the potential functions, we annotated them using Gene Ontology (GO terms)⁴⁶ which 198 classify them by Biological process, Molecular function, and Cellular component (Figure 3b, see

Methods). Of the 247 dominant lineage genes, GO terms could be assigned to 27 genes for Biological Process, 68 for Molecular Function, and 12 for Cellular component. For genes that could be assigned GO terms, functions involved in "DNA integration", "DNA recombination", and "DNA methylation" might indicate their potential roles in horizontal gene acquisition and protection against incoming foreign DNA through site-specific DNA methylation. Furthermore, GO terms associated with "DNA binding" and "Regulation of DNA-templated transcription" may suggest lineage-specific regulation of the expression of these genes.

206

207 Lineage-specific genes were selectively expressed

208 To delve deeper into the functionality of dominant lineage-specific genes, we explored their 209 expression patterns across both environmental and infection conditions. We selected 210 representative strains - "K96243" (lineage 1 – sub-lineage 1.1), "UKMD286" (lineage 2 – sub-211 lineage 2.1), and "UKMH10" (lineage 3 - sub-lineage 3.2) - all are clinical isolates with pre-212 existing gene expression profiles under infection and environmental conditions^{47–49}. For 213 environmental conditions, K96243 was exposed to water⁴⁷, while UKMD286 and UKMH10 were cultivated in a soil extract medium^{48,49} to mimic *B. pseudomallei* in the environment. For 214 infection conditions, K96243 and UKMD286 were used in murine challenges^{47,48} and isolated 215 216 from mice organs, while UKMH10 was subjected to human plasma⁴⁹ to simulate host infection. 217 This approach facilitated the comparison of differentially expressed lineage-specific genes 218 between environmental and infection conditions. Although each representative strain carried a 219 complete set of lineage-specific genes for their respective sub-lineages (K96243 with 47 genes, 220 UKMD286 with 27 genes, and UKMH10 with 14 genes), their collective representation 221 accounted for 69 out of 247 total lineage-specific genes (27.9%) due to observed genetic 222 diversity within the dominant lineage. Notably, 11 out of 47 lineage-specific genes in K96243 223 and 6 out of 27 lineage-specific genes in UKMD286 were up-regulated in the environmental 224 conditions (Figure 4a to 4c, Supplementary data 4). Notably, none of the lineage-specific 225 genes showed up-regulation during the infection condition. The remaining lineage-specific 226 genes did not exhibit preferential expression in either environmental or infection conditions. The 227 elevated expression level of lineage-specific genes in the environmental condition was 228 unexpected considering that all representative strains were clinical isolates. This observation 229 potentially suggests that dominant lineage-specific genes may play more substantial roles in 230 bacterial environmental survival than in host pathogenicity.

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232 To thrive in its environmental habitat, B. pseudomallei must cope with ranges of physical, 233 chemical and biological stresses such as desiccation, temperature fluctuations, osmotic 234 changes, oxidative stress, UV exposure, nutrient scarcity, changes in pH, exposure to heavy 235 metals, competition from antibiotics released by other microbes, and predation by eukaryotes. 236 We leveraged the extensive condition-wide transcriptome data spanning 62 conditions available 237 for K96243⁴⁷, a representative strain from lineage 1, to compare the expression patterns of 238 lineage-1-specific genes with the rest of genes in the K96243 genome (Figure 4d). Our analysis 239 revealed that lineage-1-specific genes within K96243 exhibited a higher level of gene 240 expression when the bacterial cell experienced nutrient deprivation compared to other genes in 241 K96243 genome (Two-sided Fisher's exact test p-value = 9.23×10^{-5}). This finding implies that lineage 1 might possess an adaptive strategy to persist in nutrient-depleted soil, which is not 242 uncommon in melioidosis endemic areas^{27,28,50}, before being acquired by a human host and 243 244 subsequently causing the disease.

245

246 **Example of lineage-specific genes**

247 The majority of lineage-specific genes were located within genomic islands (GI)^{30,51,52}. These 248 regions are characterised by anomalies in %G+C content or dinucleotide frequency signatures. 249 or the presence of genes associated with mobile genetic elements such as insertion sequence 250 (IS) elements and bacteriophages. Notably, we observed that a cluster of genes specific to 251 lineage 1 (BPSS2060 to BPSS2072) formed a mosaic structure within a putative metabolic island known as GI 16³⁰. Although several variations of GI 16 have been reported (GI16, GI16.1, 252 253 GI16.2, GI16a, GI16b, and GI16b.1)⁵¹, it typically spans 60 kb (*BPSS2051* to *BPSS2090*) and 254 carries several known virulence determinants and genes that enhance metabolic versatility. 255 While certain virulence factors, such as the filamentous haemagglutinin (BPSS2053) required 256 for host cell adhesion and its processing protein were conserved across multiple lineages 257 observed in our study, genes encoding functions that potentially expand the metabolic repertoire 258 were specific to dominant lineages. For example, the mosaic structure of GI 16 (BPSS2060 to 259 BPSS2072), specific to lineage 1 (Supplementary data 3), contains genes involved in 260 alternative nutrient catabolism and anabolism (BPSS2060, BPSS2065, BPSS2067, BPSS2068, 261 and BPSS2072), transcriptional regulation (BPSS2061), and substrate transport (BPSS2064, 262 BPSS2071). Out of 11 lineage-specific genes located in the mosaic structure of GI 16, eight 263 were found to be upregulated during the early phase of nutrient starvation while remaining silent 264 during infections. This finding reflects the functional division of GI16 where its lineage-specific 265 mosaic structure contains genes that contribute to metabolic versatility, while its core structure

encodes virulent determinants associated with disease implications. It is important to note that the structure of GI 16 may vary across different regions due to the plasticity of genomic islands and changes in selection pressures. While this observation is significant for a dominant lineage in northeast Thailand, it may not be generalisable to other geographical locations.

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271 Lineage-specific genes were introduced by homologous recombination

272 Homologous recombination has been shown to play a significant role in facilitating the 273 acquisition and loss of genes, and the generation of mosaic structures within the GI of B. pseudomallei⁵³. To better understand its association with lineage-specific genes, we identified 274 275 recombination events and quantified the rates of recombination in the dominant lineages. The 276 ratio of polymorphisms introduced through recombination compared to those introduced by 277 mutation (r/m) was 3.7, 4.6 and 2.2 for lineages 1, 2, and 3 respectively (**Table 1**). A very high 278 proportion of genes underwent recombination at least once: 99.5% of genes in lineage 1, 99.9% 279 in lineage 2, and 96.6% in lineage 3. Furthermore, every lineage-specific gene within each 280 dominant lineage underwent recombination (Supplementary Figure 8). The bacterial restriction 281 modification (RM) systems prevent the invasion of foreign DNA and restrict gene flow between *B. pseudomallei* lineages³¹. Notably, components of this system including a type I restriction 282 283 system and modification methylase were among dominant lineage-specific genes (BPSL0947-284 BPSL0948 in lineage 1, and their homologues in lineage 2 and 3). They may act as a barrier for 285 homologous recombination and potentially modulate lineage-specific genetic diversity. This 286 highlights the intricate interplay between recombination, lineage-specific genes, and the RM 287 system in shaping the genetic landscape of *B. pseudomallei* in northeast Thailand.

288

289 Discussion

290 Our analysis of B. pseudomallei population genomics enhances our understanding of the 291 evolution and adaptive strategies employed by the dominant lineages in the melioidosis 292 hyperendemic region of northeast Thailand and neighbouring countries. Through an 293 unprecedentedly dense sampling effort between 2015 and 2018, we were able to determine the 294 co-existence of three dominant lineages, characterise their dissemination patterns, and identify 295 lineage-specific genes possibly contributing to their success during the studied period. By 296 analysing transcriptome data from representative strains of each dominant lineage, we gained 297 further insights into their adaptive strategies, particularly emphasising bacterial persistence in 298 the environment as crucial for subsequent host acquisition and infection. Nevertheless, our 299 study has a few limitations.

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301 Due to limited environmental surveillance and the scarcity of environmental isolates in 302 Southeast Asia, our study primarily relied on clinical isolates. Nonetheless, the co-occurrence of 303 clinical and environmental isolates within the same lineage, coupled with the direct acquisition of 304 clinical isolates from the environment, suggests that our findings may hold broader implications 305 for environmental isolates. Furthermore, the lineage classification report in our study is subject 306 to potential alterations over time with the introduction of new data. As B. pseudomallei 307 continuously adapts to environmental pressures, the composition of lineages 1, 2, and 3, along 308 with their respective lineage-specific genes may shift. New lineages with superior fitness or 309 selective advantages, of different adaptive strategies could potentially outcompete the existing 310 dominant lineages. Consequently, this may lead to emergence of new lineage classifications in 311 the future. Continued surveillance efforts including both clinical and environmental samples will 312 be essential. This ongoing monitoring will be pivotal in identifying alterations within the bacterial 313 population, uncovering new adaptive strategies, and evaluating their impact on disease 314 dynamics over time.

315

316 Our understanding of the significance of bacterial persistence as a strategy for successful 317 lineages stems from the observed up-regulation of dominant lineage-specific genes in 318 environmental conditions, along with the increased expression of lineage-1-specific genes 319 during nutrient deprivation. However, it is essential to recognise the genetic diversity within each 320 dominant lineage. The representative strains used in our study carried a subset of the lineage-321 specific genes corresponding to their lineage. As a result, lineage-specific genes absent in 322 these strains, often annotated as hypothetical proteins, remained unexplored in our analysis. 323 Moreover, our detailed characterisation of lineage-1-specific genes using a comprehensive 324 transcriptome dataset of 62 distinct conditions might not capture the full spectrum of conditions 325 encountered by *B. pseudomallei* in its natural habitat. There is a possibility that other adaptive 326 strategies were overlooked in this study, indicating scope for future exploration. Despite these 327 shortcomings, our dataset and analysis currently represent one of the most comprehensive 328 efforts to date. Future research will prioritise the generation of a more extensive condition-wide 329 transcriptome, covering a broader range of conditions and incorporating strains with diverse 330 genetic variations to identify other adaptive strategies employed by B. pseudomallei.

331

332 Our findings underscore the significance of environmental persistence in driving the success of 333 dominant lineages 1 and 2, notably highlighting lineage-1-specific genes in mediating bacterial

334 survival under nutrient depletion. It remains uncertain whether lineage 3 adopts a similar 335 strategy. Nevertheless, our results align with previous soil sampling studies, which consistently 336 observed a higher prevalence of *B. pseudomallei* in nutrient-depleted compared to nutrient-rich 337 soil^{28,50}. Additionally, a molecular evolutionary study also supports the species' long-term 338 adaptation to survive nutrient scarcity²⁷. The northeast region of Thailand, where our samples 339 were primarily collected, has inherently low-fertility soil. This is exacerbated by intensified 340 agriculture, monoculture, excessive synthetic fertilizer use, and poor land management, resulting in depleted soil nutrients and organic matter⁵⁴. This presents a challenging 341 environment for *B. pseudomallei* to thrive, thereby potentially selecting for successful lineages 342 343 with persistent traits as observed in our study.

344

Our analyses also highlight the role of various factors such as differences in terrain altitude^{55,56}. 345 river flow dynamics⁴⁰, and the northeast monsoon⁴¹ contribute to shaping the dissemination 346 347 patterns of *B. pseudomallei*. These drivers of dissemination are influenced by both natural and human activities. For instance, strong winds can carry dried soil particles⁵⁷, potentially 348 349 containing *B. pseudomallei* over distances. Climate change-induced alterations in vegetation cover might expose soil to rainfall and winds⁵⁸, impacting the bacterial spread. Additionally, 350 deforestation can disrupt natural barriers like trees and shrubs, accelerating water runoff^{57,58} and 351 352 potentially facilitating the wide-ranging dissemination of *B. pseudomallei* during flood. 353 Considering these dynamics, the strategy of bacterial persistence likely plays a pivotal role in its 354 widespread dissemination within the region, thereby influencing disease prevalence. Therefore, 355 an effective disease control strategy should integrate both environmental and clinical public 356 health measures to effectively mitigate the impact of melioidosis.

357 Materials and Methods

358 Data collection and bacterial isolates

The *B. pseudomallei* isolates in our study included a cohort⁸ from northeast Thailand gathered between July 2015 to December 2018 consisting of 1,265 clinical isolates. We also incorporated a contemporaneous dataset from Thailand and neighbouring regions^{15,29–34}, comprising 15 clinical and 111 environmental isolates sourced from previous publications. In total, 1,391 *B. pseudomlalei* genomes were used in this study. Their metadata and accession numbers were documented in **Supplementary data 1**.

365

366 The northeast Thailand collection was collected from patients participated in our longitudinal 367 cohort study⁷. The patients were from nine provinces including Udon Thani (n = 230), Mukdahan (n = 198). Roi Et (n = 195), Surin (n = 170), Nakhon Phanom (n = 135), Buriram (n = 123), 368 369 Sisaket (n = 107), Khon Kaen (n = 96) and Maha Sarakham (n = 11), who were admitted to nine 370 hospitals included in our cohort. The ethical approval for the cohort study was obtained from the 371 Ethics Committee of the Faculty of Tropical Medicine, Mahidol University (MUTM 2015-002-001 372 and MUTM 2021-055-01). The isolates were obtained from various clinical samples, including 373 blood (71.8%), pus (12.5%), sputum (9.9%), body fluid (3.5%), urine (1.9%) and tissue (0.4%) (Supplementary data 1). As latent infection accounts for < 5% of the cases, the majority of 374 clinical cases likely directly acquired from the environment⁸. The numbers of enrolled cases 375 376 were lower at the beginning of the study due to the delayed sample collection in some study 377 sites, resulting in inconsistent number of bacterial isolates used across different sites. When 378 applicable, a permutation test was performed to ensure that an unequal number of isolates did 379 not impact the temporal or spatial analysis.

380

381 Culture confirmation of *B. pseudomallei*, DNA extraction and whole genome sequencing

382 All 1,265 B. pseudomallei samples from the northeast Thailand collection were cultured on 383 Ashdown, selective agar plates and confirmed the species using latex agglutination test and 384 matrix-laser absorption ionisation mass spectrometry (MALDI-TOF MS). A single colony from 385 Ashdown agar plate was subjected to culture in Luria-Bertani (LB) broth and subsequently used 386 for DNA extraction. Genomic DNA was extracted using QIAamp DNA Mini Kit (Qiagen, 387 Germany). All genomic DNA were processed for the 150-base-read library preparation and 388 sequencing using Illumina HiSeg2000 system with 100-cycle paired-end runs at Wellcome 389 Sanger Institute, Cambridge UK. An average of 71X read depth was achieved. To control the 390 potential contamination in each sample with other closely related species, we assigned

taxonomic identity using Kraken⁵⁹ v.1.1.1. We then estimated the genome completeness and
 species confirmation using CheckM⁶⁰ v.1.2.2 and FastANI⁶¹ v.1.31, respectively. The quality
 control data of the 1,265 genomes were listed in **Supplementary data 2**.

394

395 Genome assembly and mapping from short read data

Short reads were *de novo* assembled using Velvet v.1.2.10⁶² followed by optimisation as 396 397 previously described²⁹, with their quality scores reported in **Supplementary data 2**. Short reads were also mapped against several reference genomes, including a strain K96243³⁰ (accession 398 399 numbers BX571965 and BX571966) to determine the whole population structure, and lineage-400 specific references to improve the resolution for lineage-specific analyses. We selected K96243³⁰ genome as a population-wide reference due to its origin in northeast Thailand, 401 402 aligning with the geographical focus of our study. Additionally, its well-characterised and 403 complete genome further streamlined subsequent analyses. For all mapping, variants were 404 called using Snippy v.4.6.0 (https://github.com/tseemann/snippy). To avoid mapping errors and 405 false SNPs, we filtered out SNPs covered by less than 10 reads and found in a frequency of 406 less than 0.9.

407

408 **Defining whole population structure**

409 **PopPUNK clustering**

PopPUNK v.2.6.0³⁵ was run on 1,391 assembled genomes. To define the core and accessory distance between each pair of isolates, the assemblies were hashed at different k-mers. The population model was fit using command line "poppunk-runner.py --fit-model --distance -database.dists> --output <database> --full-db --ref-db <database> --min-kmer 15 -max-kmer 31 --max-a-dist 0.53 --K 4 -k-step 2" with the result density of 0.028, transitivity of 0.992, and network score of 0.8961.

416

417 Maximum likelihood phylogenies from core genome SNP, cgMLST and MLST

An alignment of full genome was created by mapping whole genome sequences of each *B. pseudomallei* against a complete genome of K96243³⁰ strain. From this alignment, 4,221 cgMLST loci based on a scheme described in³⁷ were extracted and concatenated to form cgMLST alignment. Additionally, seven MLST loci, as per scheme described in ³⁸ were extracted from the same alignment and concatenated to create the MLST alignment. Core genome SNP alignment was identified from a full genome alignment using snp-sites⁶³ v.2.5.1, with genomic islands⁵¹ masked. Separate maximum likelihood phylogenies were constructed for

425 core genome SNP alignment, cgMLST alignment, and MLST alignment using IQ-TREE³⁶
 426 v.2.0.3. Standard model selection in IQ-TREE determined the best-fit model as
 427 TVM+F+ASC+R6 for all three phylogenies. To access the robustness of the phylogenetic trees,

- 428 a 1,000 bootstrap support was performed for each tree.
- 429

430 Comparison of population structure outlined phylogeny constructed from core genome

431 SNPs, cgMLST, MLST and PopPUNK

432 To test for consistency between phylogenetic trees constructed from core genome SNPs, 433 cgMLST and MLST alignment, we use the R package treespace⁶⁴ v. 1.1.4.3 to explore the tree 434 tip distributions. We compared pairwise tree distances within the first 100 bootstraps within each 435 alignment category (indicative of bootstrap support strength), and across trees generated from 436 different alignment categories (indicating proximity between tree categories). The tree pairwise distances were computed, and principal components (PCs) were derived with eigenvalues 437 438 calculated for different PCs. The similarity among phylogenies from each alignment category 439 was assessed using two PCs dimensions, which jointly accounted for >90% of variability in 440 pairwise distance (Supplementary Figure 1a). The scatter plot of PCs revealed a close 441 clustering of bootstrap trees from core genome SNP and cgMLST, while the bootstrap trees 442 from MLST alignment showed greater dispersion, highlighting less consistency in the trees 443 generated by the MLST approach. We further compared the consistency between the median 444 phylogenetic tree of each alignment category and PopPUNK classification was visually compared using iTOL⁶⁵ (Supplementary Figure 1b). 445

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447 **Specific lineage analysis**

To investigate the dissemination and genetic diversity within each dominant lineage, we conducted individual genome alignments, recombination removal, and maximum likelihood phylogeny. To enhance the sensitivity of variant, we selected closely related genomes as references for each lineage. Specifically, the complete genome *B. pseudomallei* strain K96243 (accession numbers BX571965 and BX571966) served as the mapping reference for lineage 1, while new reference genomes were created for lineages 2 and 3.

454

For lineage 2, we chose a representative isolate 27035_8#57 and subjected it to long-read sequencing on a local MinION sequencer following the manufacturer's standard protocol (Oxford Nanopore Technologies, Oxford, United Kingdom). A complete hybrid assembly of the long-read and short-read sequence data of this strain was performed using Unicycler⁶⁶ v.0.8.4.

The resulting hybrid assembly of the 27035_8#57 genome was employed as the mapping reference for this lineage.

461

In the absence of representative complete genomes for lineage 3, we selected the best quality
de novo assembly of isolate 27035_8#119 and orientated its contigs according to strain K96243
using ABACAS⁶⁷ v.1.3.1. This genome was used as a mapping reference for lineage 3.

465

For lineage-specific mapping, Snippy v.4.6.0 as employed as in the whole population analysis. 466 All genome alignments were subjected to Gubbins⁶⁸ v.3.1.3, a recombination identification tool, 467 468 to detect and remove recombination fragments. This process determined the genetic diversity 469 introduced by horizontally acquired elements and vertically inherited SNPs, thereby producing 470 recombination-free SNP alignments for phylogenetic reconstruction. Maximum-likelihood 471 phylogenies were constructed using recombination-free SNP alignment of each dominant lineage using IQ-TREE³⁶ v.2.0.3 with TVM+F+ASC+R6 and 1,000 replicates of bootstrap 472 473 support. The overall proportion of nodes with ≥80% bootstrap support of lineage-specific 474 phylogenies reached 83.5%.

475

476 Dating the timeline for lineages and sub-lineages

To enable dating analysis, we further divided each lineage into sub-lineages using R package 477 rhierbaps⁶⁹ v.1.1.4. We inferred evolutionary timeline and estimated the age of each lineage and 478 479 sub-lineage based on the isolate's collection date. A Bayesian molecular dating provided in R 480 package BactDating⁷⁰ v.1.1.1. was employed to assess the temporal signals by examining a 481 positive correlation between the isolate's collection date and the root-to-tip distance. 482 Recombination removed phylogenies were used in this analysis. A date-randomisation test, 483 consisting of 100 permutations, was performed to assess the robustness of the temporal signal 484 compared to noise.

485

Notably, the temporal signals were discernable at the sub-lineage level rather than the broader lineage level. Among the 10 sub-lineages, only one sub-lineage (lineage 1.3) exhibited a positive correlation in their clock signals. Given the limited sample size of lineage 1.3, we employed a strict clock model to prevent parameter over-fitting. We ran three independent Markov chain Monte Carlo (MCMC) chains, each spanning at least 100 million iterations, and sampled every 10,000 steps. The prior mutation rate derived from Pearson and colleagues was used. Visual inspection of the trace from each MCMC chain confirmed signal convergence, with 493 effective sampling size values > 200 for key parameters. Visualisation of results were performed
494 using the R package ggtree⁷¹ v.3.10.0 to generate credibility time-calibrated phylogeny for each
495 sub-lineage (Supplementary Figures 3).

496

497 Ancestral state reconstruction analysis

498 Ancestral trait reconstruction was conducted to discern the dissemination patterns of B. 499 pseudomallei among provinces in northeast Thailand, focusing on dominant lineages. Due to 500 varving number of isolates among provinces, the analysis excluded Mahasarakham, which had 501 a limited dataset (n = 11), resulting in the analysis of eight provinces: Buriram, Khon Kaen, 502 Mukdahan, Nakhon Phanom, Roi Et, Sisaket, Surin, and Udon Thani. This approach yielded 28 503 potential province-to-province transmission combinations. To mitigate sampling biases, we sub-504 sampled the phylogeny of each dominant lineage to have an equal number of isolates per province (n = 15 isolates) and permuted 1,000 times. Using the stochastic character mapping 505 function (*make.simmap*) from the R package phytools⁷² v.1.9.16, we conducted 100 simulations 506 507 (nsim = 100) to reconstruct the provincial origins at each node in the sub-sampled phylogeny 508 (1,000 phylogenies per lineage). This allowed us to quantify transition events (Markov jumps) 509 between province pairs and determine the cumulative branch length associated with each 510 province (Markov rewards). A Mann-Whitney U test, with Bonferroni correction for multiple 511 comparisons was applied to compare the transition event counts among provinces 512 (Supplementary Figure 6).

513

514 Pan-genome analysis

All the study genomes were annotated using Prokka⁷³ v.1.14.5, and further used in the pangenome analysis. Each genome has a median of 5,845 coding sequences (CDS) predicted onto each genome with a range of 5,642 to 6,142 CDS per genome. Panaroo⁷⁴ v.1.3.3 was employed to estimate the pan-genome with a sensitive option and a cut-off sequence identity of 92% derived from previous study¹⁵. The number of estimated genes falls within a comparable range to previous studies from a single population^{29,53}.

521

522 Identification of dominant lineage-specific genes

523 We determined lineage-specific genes by assessing their prevalence within dominant lineage or 524 any of their sub-lineages, requiring a high occurrence (95%) within these specific groups while 525 maintaining a low presence in non-dominant lineages. To achieve this, three thresholds were 526 employed: strict (95 % occurrence in dominants vs 5% occurrence in non-dominants),

intermediate (95% in dominants vs 10% in non-dominants), and relaxed (95% in dominants vs
15% in non-dominants). Based on visual examination of gene distribution patterns (Figure 3a,
Supplementary Figure 7), the relaxed threshold was used to maximise the number of genes
included in subsequent analysis.

531

532 Identification of Gene Ontology (GO: terms)

533 Amino acid sequences of lineage-specific genes were submitted to InterPro database⁴⁶ 534 (https://www.ebi.ac.uk/interpro/) which characterised the function of lineage-specific genes 535 based on biological processes, molecular functions, and cellular compartments (**Figure 3**; 536 **Supplementary data 3**).

537

538 Transcriptomic analysis of dominant lineage-specific genes

539 Lineage 1 transcriptome analysis

540 The analysis focused on an expression profile of a strain K96243, which serves as a 541 representative of lineage 1. Data was sourced from microarray experiment generated by Ooi *et* 542 $al.^{47}$ and accessed through the Gene Expression Omnibus (GEO) under accession number 543 GSE43205.

544 To understand the difference between environmental and infection conditions, we compared the 545 expression profile of K96243 being exposed to water and K96243 recovered from infected mice. 546 To simulate environmental conditions, K96243 was cultured to log phase in LB medium, 547 subsequently washed with sterile deionising water, and suspended in water. To emulate 548 infection conditions, BALB mice were infected with 1000 CFU of *B. pseudomallei*, and bacteria 549 were harvested from the lungs three days post-infection. Two replicates were performed for 550 each condition. We retrieved microarray data from GEO using the R package GEOquery⁷⁵ 551 v.2.58.0, and differential gene expression analysis was performed using the R package limma⁷⁶ 552 v.3.58.1.

We used binary expression patterns reported in Ooi *et al.*⁴⁷ to compare the expression profiles of lineage-specific genes against the remaining genes in K96243 across 62 conditions. This enabled the comparison of the count of expressed genes within the lineage-specific category against the remainder of the genes for each condition using Fisher's exact test, with multiple testing adjustments via Benjamini-Hochberg corrections.

558 Lineage 2 transcriptome analysis

This analysis focused on the expression profile of a strain UKMD286, representative of lineage
2. RNAseq data was obtained from an experiment conducted by Ghazali *et al.* ⁴⁸ and accessed
through the European Nucleotide Archive (ENA) (E-MTAB-11200).

To simulate environmental conditions, UKMD286 was cultured in BHIB medium overnight, resuspended, and inoculated into soil extract medium. For infection condition, BALB mice were infected with UKMD286, and the bacteria were harvested from spleens five days post infection. Each experiment was conducted with two replicates. FastQC v.0.11.9 and FastXtool v.0.0.14 were used to pre-processed sequenced reads. Raw reads were aligned to UKDM286 genome using Hisat2⁷⁷ v. 2.2.1 with differential gene expression performed using the R package DESeq2⁷⁸ v.1.40.2.

569 Lineage 3 transcriptome analysis

570 We used a strain UKMH10 to represent the expression profile of lineage 3. Data was originated 571 from an RNAseq experiment conducted by Kong *et al.*⁴⁹ and was accessed through the

572 European Nucleotide Archive (ENA) (PRJEB53338)

573 To replicate environmental conditions, UKMH10 was cultured in LB medium overnight and sub-574 cultured into soil extract medium. To simulate infection conditions, UKMH10 was cultured in LB 575 medium overnight and inoculated into human plasma, then incubated at 37 C to mimic the 576 human body temperature. Bacterial cells were harvested once the absorbance reading of the 577 bacterial cultures at 600 nm (OD600) reached 0.5. Each experiment was performed with two replicates. FastQC v.0.11.9 and FastXtool v.0.0.14 were used to pre-processed sequenced 578 579 reads. Raw reads were aligned to UKMH10 genome using Hisat2⁷⁷ v.2.2.1 with differential gene expression performed using the R package DESeg2⁷⁸ v.1.40.2. 580

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600 Data availability

The genome sequence data presented in this study can be found in online repositories. The ENA under study accession number PRJEB25606 and PRJEB35787. The accession numbers for individual genomes, and their annotated assemblies are listed in **Supplementary data 1**.

604 Code availability

The analyses used public software including Kraken v.1.1.1, CheckM v.1.2.2, FastANI v.1.31,
FastQC v.0.11.9, FastXtool v.0.0.14, Unicycler v.0.8.4, ABACAS v.1.3.1, Velvet v.1.2.10,
Prokka v.1.13.4, Panaroo v.1.2.9, Snippy v.4.6.0, PopPUNK v.2.6.0, IQ-TREE v.2.0.3, Gubbins
v.3.1.3, Hisat2 v.2.2.1 and R packages: treespace v.1.1.4.3, BactDating v.1.1.1, ggtree v.3.10.0,
phytools v.1.9.16, GEOquery v.2.58.0, limma v.3.58.1 and DESeq2 v.1.40.2

610 Ethics and Inclusion statements

The research is led by local researchers and include local contributions throughout all research
process including the study design, study implementation, data ownership, intellectual
properties and authorship for publication.

614 Author contributions

RS, CChe and NC conceived, designed the study and write the original draft. NC and CChe
administrated and supervised the project. NC acquired funding to collect isolates, while CChe
acquired funding for sequencing and downstream analysis. NC, RS, TEW, and NS collected

and identified bacterial isolates. NC, RS and RP collected clinical data. CChe, NRT, NC, RS,
JT, EB and WC performed whole-genome sequencing. NPJD and NC contributed reagents.
NRT and JP contributed software tools. RS, CCho, and CChe performed bioinformatics
analyses. CChe, NC, NRT and JP interpret the analyses. CChe wrote the revised draft. All
authors read and approved the manuscript.

623 **Table 1** Recombination in dominant lineages.

Lineage	Percent of	Percent of	Average r/m			
	CDS	dominant	(numbe	r of SNPs introduced by ation/ SNPs introduced by		
	impacted by	lineage-	recombina			
	recombinatio	specific genes underwent recombination at least once (recombinant gene/lineage- specific genes identified in the reference genome)	substitutions)			
	n at least once (recombinant CDS/total CDSs in the reference genome)		Internal nodes (95% CI)	Terminal nodes (95% CI)	Average (95% CI)	
1	99.5%	100%	2.8	4.6	3.7	
	(5981/6010)	(47/47)	(2.3-3.3)	(4.0-5.3)	(3.3-4.1)	
2	99.9%	100%	5.0	4.3	4.6	
	(5963/5971)	(65/65)	(3.9-6.0)	(3.6-4.9)	(4.0-5.2)	
3	96.6%	100%	1.4	2.9	2.2	
	(5338/5523)	(20/20)	(1.1-1.8)	(2.2-3.7)	(1.8-2.6)	

624

625 Figure and table legend

626 Figure 1 Distribution of *B. pseudomallei* genomes used in this study (a) Geographical 627 representation of the countries and provinces sampled for the 1,391 B. pseudomallei genomes 628 used in this study. Pie-chart summarises the proportion of dominant lineage 1, 2, and 3 629 presented at each location with the chart size proportional to the number of the samples 630 collected (b) An unrooted phylogenetic tree colour-coded by dominant lineages (c) Histogram 631 depicting the distribution of clinical *B. pseudomallei* isolates from the northeast Thailand cohort 632 throughout 2015-2018 sampling period. The shaded blue area represents the period of rainy 633 seasons. (d) Boxplots summarising the pairwised core genome SNP distances among isolates 634 in this study, shown in a logarithmic scale. The distribution is depicted for the entire population 635 and each dominant lineage.

636 Figure 2 Dissemination patterns in northeast Thailand. (a) Province-to-province transmission 637 patterns influenced by northeast Thailand geographical landscape. Nodes present provinces, 638 denoted by abbreviation and ordered by altitude: U-Udon Thani, K-Khon Kaen, B-Buriram, R -639 Roi Et, N-Nakhon Phanom, Su-Surin, M-Mukdahan, and Si-Sisaket. Rivers are depicted in blue 640 with major rivers including the Great Mekong River, the Chi River, and the Mun River and their 641 flow direction annotated. (b) Average altitude of provinces in meters above sea level. Error bars 642 present 95% confidence interval. The northwest provinces exhibit higher altitudes, gradually 643 declining towards the southeast. For (a) and (b) solid arrows illustrating transmission 644 directionality explained by altitude differences. Dotted arrows represent transmission 645 directionality influenced by northeast monsoon winds. Grey arrows signify patterns with unclear 646 explanation.

647 Figure 3 Dominant lineage-specific genes and their Gene Ontology (GO terms). (a) The 648 heatmap represents lineage-specific genes (right) detected in each isolate, aligned with the 649 phylogeny (left). Lineage-specific genes shared across multiple dominant lineages are 650 highlighted in vellow. Lineage-specific genes from lineage 1, 2, 3 are coloured in green, red, and 651 purple, respectively. Additionally, the colour stripes provide information on the lineage and sub-652 lineage membership (b) Bar plots displays the frequency of GO annotations of lineage-specific 653 genes in each dominant lineage categorised by biological process, molecular function, and 654 cellular compartment. The pie-charts summarise the proportion of lineage-specific genes with 655 assigned GO terms (black).

656 Figure 4 Transcriptome analysis of representative strains: K96243 (lineage 1), UKMD286 657 (lineage 2) and UKMH10 (lineage 3) (a to c) Volcano plots demonstrate differential gene 658 expression (DGE) between environmental and infection conditions. Vertical dotted lines 659 represent the statistical cut-off at log two-fold change, while horizontal dotted lines display the 660 statistical cut-off at the adjusted p-value of 0.05 on a negative log scale. Each dot represents a 661 gene, with lineage-1, lineage-2, and lineage-3-specific coloured in green, red, and purple, 662 respectively. (d) Binary expression profile of lineage-1-specific genes across different 663 conditions. A star denotes significant differences in the gene expression profile of lineage-1-664 specific genes compared the remaining genes of strain K96243.

Supplementary Figure 1 Comparison of approaches used in in outlining the population structure (a) A scatter plot displays the first two principal components (PCs) derived from pairwise distances between trees. Each dot represents a bootstrap tree and is colour-coded by the method used to generate the tree: core genome SNP (blue), cgMLST (red) and MLST (green). (b) A scree plot summarises eigenvalues computed for each PCs. (c to d) The median phylogenetic trees constructed from core genome SNP, cgMLST, and MLST and their consistency with PopPUNK clustering method.

Supplementary Figure 2 Distribution of environmental and clinical isolates by each lineage.
Barplots highlight the co-detection of environmental (green) and clinical isolates (red) across
dominant lineages 1, 2, and 3.

Supplementary Figure 3 Lineage 1 specific analysis (a) A recombination removed lineage 1 phylogeny with colour stripes displaying its sub-lineage structure, year of collection, and sampling province (left to right). (b) A map of northeast Thailand showing the distribution of each isolate and the region's river system. (c) Time-calibrated phylogeny of sub-lineage 1.3 with blue error bars indicating 95% highest posterior density interval, with the estimated mutational rate consistent with previous study.

Supplementary Figure 4 Lineage 2 specific analysis (a) A recombination removed lineage 2 phylogeny with colour stripes highlighting sub-lineage structure, year of collection, and sampling province (left to right). (b) A map of northeast Thailand with the region's river system. Dots present the distribution of individual samples.

Supplementary Figure 5 Lineage 3 specific analysis (a) A recombination removed lineage 3
 phylogeny with colour stripes highlighting sub-lineage structure, year of collection, and sampling

687 province (left to right). (b) A map of northeast Thailand showing the distribution of each isolate688 and the region's river system.

Supplementary Figure 6 Transmission patterns and evolutionary time spent at each province. (a to c) Proportion of transition events (Markov jumps) among provincial pairs for lineage 1, 2, and 3 respectively. The pairs were denoted as province1 – province 2, with transitions from province 1 to province 2 shown in red, and transitions from province 2 to province 1 in green. The Man-Whitney U test was conducted for each pair to assess differences in transition frequency by direction, with Bonferroni correction applied for multiple tests. (d to f) Total branch length from provincial trait reconstruction (Markov rewards) for lineage 1, 2, and 3, respectively.

596 **Supplementary Figure 7** Selection criteria for lineage-specific genes. Scatter plots show the 597 frequency distribution of lineage-specific (red) against other genes (black), based on their 598 distribution within the dominant lineages and their sub-lineages (horizontal axis) compared to 599 their distribution in non-dominant lineages.

700 **Supplementary Figure 8** Recombination patterns detected in lineage 1, 2, and 3. From left to 701 right: the recombination-removed phylogeny of each lineage, a stripe representing the sampling 702 year and sub-lineage classification, and heatmaps displaying recombination patterns identified 703 in chromosome 1 and 2. The top orange lines mark the genome coordinates. For each lineage, 704 their respective lineage-specific genes are highlighted in blue at the top of the panel. Each 705 heatmap represents recombination blocks aligned with the phylogeny. Recombination events 706 occurring at the internal nodes are coloured in red, while those occurring at the external 707 branches are coloured in blue. The recombination hotspot is plotted at the bottom of each 708 heatmap.

Supplementary data 1 Epidemiological data, isolate and accession codes for both short reads
 and annotated assembly used in this study deposited in the European Nucleotide Archive (ENA)
 (n = 1,391) (provided as a separate excel file).

712

Supplementary data 2 Quality control information of newly sequenced *B. pseudomallei* from
northeast Thailand collection (n = 1,265) (provided as a separate excel file).

715

Supplementary data 3 List of dominant lineage-specific genes and their Gene Ontology (GO)
 terms (provided as a separate excel file).

718

- 719 Supplementary data 4 List of dominant lineage-specific genes and their expression profile
- under infectious and environmental conditions generated by Ooi *et al.* 2013, Ghazali *et al.* 2023
- and Kong *et al.* 2023 (provided as a separate excel file).

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Figure 2









Figure 4

