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 38 study identified three dominant lineages with unique gene sets enhancing bacterial fitness, 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 46 highlight persistence in the environment as a pivotal element facilitating *B. pseudomallei* spread, 47 and as a prelude to exposure and infection, thereby providing useful insights for informing 48 melioidosis prevention and control strategies.

49

#### 50 **Introduction**

51 Melioidosis, a severe infectious disease, affects an estimated 165,000 cases globally each year, 52 of which 89,000 are fatal<sup>1</sup>. The disease is caused by *Burkholderia pseudomallei*, a Gram-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 55 underreporting, particularly in lower- and middle-income countries<sup>2</sup>. However, improved 56 infrastructure and awareness have led to increases in reported cases across South Asia, 57 Southeast Asia, East Asia<sup>3-8</sup> and Australia<sup>9,10</sup> In Southeast Asia, the disease incidence is often 58 linked to agriculture practice, particularly during the rainy seasons when rice paddy fields are 59 flooded for planting. The flooded terrain enables the bacterium in the soil to surface, potentially 60 exposing farmers to *B. pseudomallei* and subsequently leading to melioidosis<sup>11</sup>. Additionally, 61 many cases of melioidosis have been associated with severe weather events<sup>12–14</sup>. While climate 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 69 and environmental factors<sup>15–18</sup>. Streams<sup>19,20</sup>, monsoons, typhoons and cyclones<sup>13,14,21,22</sup> were 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, 73 nutrient-depleted soil<sup>23–28</sup> thereby enabling the bacterium to thrive in various ecological niches. 74 Previous studies have noted the temporal and geographical co-existence of multiple *B.*  75 *pseudomallei* lineages<sup>15,16,29</sup>. However, little is known about their distinct genetic content and 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 82 Udon Thani<sup>8</sup>; totaling 1,265 isolates collected from July 2015 to December 2018. Additionally, 83 we incorporated contemporary environmental and clinical collections from Thailand and 84 neighbouring countries<sup>15,29–34</sup>, consisting of 15 clinical isolates and 111 environmental isolates 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 88 investigated their biological functions and expression conditions in three representative isolates, 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**

# 94 **Population structure analysis revealed the successful** *B. pseudomallei* **lineages and**  95 **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. PopPUN $K^{35}$  analysis was performed on genome assemblies

99 (**Supplementary data 2**). Additionally, we constructed three maximum-likelihood (ML) 100 phylogenies<sup>36</sup>, each based on different sets of single nucleotide polymorphisms (SNPs): core 101 genomes (n= 77,156 SNPs), core gene multilocus typing<sup>37</sup> (cgMLST, n = 46,945 SNPs), and 102 seven-gene multilocus typing genes<sup>38</sup> (MLST,  $n = 31$  SNPs). These approaches facilitated the 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 cgMLST 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 10-4 124 , **Supplementary Figure 2**). Due to the substantially lower number of environmental 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 Northeast**  132 **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, 153 and Khon-Kaen) to less than 100 meters in the southeast (parts of Mukdahan and Sisaket), and 154 gradually descending toward the Mekong River in the east<sup>39,40</sup>. 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 Sisaket<sup>40</sup>.

159 Additionally, another set of three out of 14 conserved patterns coincided with the wind direction 160 of the northeastern monsoon during the dry season ("Nakhon Phanom-to-Buriram", "Mukdahan-161 to-Buriram", and "Mukdahan-to-Surin"). Thailand experiences two predominant monsoon 162 seasons: the southwest monsoon from May to October and the northeast monsoon from 163 November to April. The southwest monsoon brings heavy rainfall from the southwest to 164 northeast, often marking the start of the agriculture season (Figure 1c)<sup>41,42</sup>. Conversely, the 165 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 176 from the terrain slope, inland rivers, canal systems<sup>45</sup> and regular monsoons<sup>41,42</sup>; various factors 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<br>193 their sub-lineages, but present in ≤15% of isolates outside these lineages. Among these, 247 193 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)<sup>46</sup> which 198 classify them by Biological process, Molecular function, and Cellular component (Figure 3b, see

199 Methods). Of the 247 dominant lineage genes, GO terms could be assigned to 27 genes for 200 Biological Process, 68 for Molecular Function, and 12 for Cellular component. For genes that 201 could be assigned GO terms, functions involved in "DNA integration", "DNA recombination", and 202 "DNA methylation" might indicate their potential roles in horizontal gene acquisition and 203 protection against incoming foreign DNA through site-specific DNA methylation. Furthermore, 204 GO terms associated with "DNA binding" and "Regulation of DNA-templated transcription" may 205 suggest lineage-specific regulation of the expression of these genes.

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### 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 water $47$ , while UKMD286 and UKMH10 were 214 cultivated in a soil extract medium<sup>48,49</sup> to mimic *B. pseudomallei* in the environment. For 215 infection conditions, K96243 and UKMD286 were used in murine challenges<sup>47,48</sup> and isolated 216 from mice organs, while UKMH10 was subjected to human plasma<sup>49</sup> 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.

231

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 K96243 $47$ , 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 =  $9.23 \times 10^{-5}$ ). This finding implies that 242 lineage 1 might possess an adaptive strategy to persist in nutrient-depleted soil, which is not 243 uncommon in melioidosis endemic areas<sup>27,28,50</sup>, before being acquired by a human host and 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 252 island known as GI 16<sup>30</sup>. Although several variations of GI 16 have been reported (GI16, GI16.1, 253 GI16.2, GI16a, GI16b, and GI16b.1)<sup>51</sup>, 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 (B*PSS2061*), 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

266 encodes virulent determinants associated with disease implications. It is important to note that 267 the structure of GI 16 may vary across different regions due to the plasticity of genomic islands 268 and changes in selection pressures. While this observation is significant for a dominant lineage 269 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.*  274 *pseudomallei*<sup>53</sup>. To better understand its association with lineage-specific genes, we identified 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 282 B. pseudomallei lineages<sup>31</sup>. Notably, components of this system including a type I restriction 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.

#### 300

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<sup>28,50</sup>. Additionally, a molecular evolutionary study also supports the species' long-term 338 adaptation to survive nutrient scarcity<sup>27</sup>. 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, 341 resulting in depleted soil nutrients and organic matter<sup>54</sup>. This presents a challenging 342 environment for *B. pseudomallei* to thrive, thereby potentially selecting for successful lineages 343 with persistent traits as observed in our study.

344

345 Our analyses also highlight the role of various factors such as differences in terrain altitude<sup>55,56</sup>. 346 river flow dynamics<sup>40</sup>, and the northeast monsoon<sup>41</sup> contribute to shaping the dissemination 347 patterns of *B. pseudomallei*. These drivers of dissemination are influenced by both natural and 348 human activities. For instance, strong winds can carry dried soil particles<sup>57</sup>, potentially 349 containing *B. pseudomallei* over distances. Climate change-induced alterations in vegetation 350 cover might expose soil to rainfall and winds<sup>58</sup>, impacting the bacterial spread. Additionally, 351 deforestation can disrupt natural barriers like trees and shrubs, accelerating water runoff<sup>57,58</sup> and 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**

359 The *B. pseudomallei* isolates in our study included a cohort<sup>8</sup> from northeast Thailand gathered 360 between July 2015 to December 2018 consisting of 1,265 clinical isolates. We also incorporated 361 a contemporaneous dataset from Thailand and neighbouring regions<sup>15,29–34</sup>, comprising 15 362 clinical and 111 environmental isolates sourced from previous publications. In total, 1,391 *B.*  363 *pseudomlalei* genomes were used in this study. Their metadata and accession numbers were 364 documented in **Supplementary data 1**.

365

366 The northeast Thailand collection was collected from patients participated in our longitudinal 367 cohort study<sup>7</sup>. The patients were from nine provinces including Udon Thani (n = 230), Mukdahan 368 (n = 198), Roi Et (n = 195), Surin (n = 170), Nakhon Phanom (n = 135), Buriram (n = 123), 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%) 374 (**Supplementary data 1**). As latent infection accounts for < 5% of the cases, the majority of 375 clinical cases likely directly acquired from the environment<sup>8</sup>. The numbers of enrolled cases 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 HiSeq2000 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

391 taxonomic identity using Kraken $^{59}$  v.1.1.1. We then estimated the genome completeness and 392 species confirmation using CheckM $^{60}$  v.1.2.2 and FastANI $^{61}$  v.1.31, respectively. The quality 393 control data of the 1,265 genomes were listed in **Supplementary data 2**.

394

# 395 **Genome assembly and mapping from short read data**

396 Short reads were *de novo* assembled using Velvet v.1.2.10<sup>62</sup> followed by optimisation as 397 previously described<sup>29</sup>, with their quality scores reported in **Supplementary data 2**. Short reads 398 burre also mapped against several reference genomes, including a strain K96243 $^{30}$  (accession 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 401 K96243 $30$  genome as a population-wide reference due to its origin in northeast Thailand, 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**

410 PopPUNK v.2.6.0<sup>35</sup> was run on 1,391 assembled genomes. To define the core and accessory 411 distance between each pair of isolates, the assemblies were hashed at different k-mers. The 412 population model was fit using command line "poppunk-runner.py --fit-model --distance 413 <database.dists> --output <database> --full-db --ref-db <database> --min-kmer 15 –max-kmer 414 31 --max-a-dist 0.53 --K 4 –k-step 2" with the result density of 0.028, transitivity of 0.992, and 415 network score of 0.8961.

416

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

418 An alignment of full genome was created by mapping whole genome sequences of each *B.*  419 *pseudomallei* against a complete genome of K96243<sup>30</sup> strain. From this alignment, 4,221 420 cgMLST loci based on a scheme described in $37$  were extracted and concatenated to form 421 cgMLST alignment. Additionally, seven MLST loci, as per scheme described in  $38$  were 422 extracted from the same alignment and concatenated to create the MLST alignment. Core 423 genome SNP alignment was identified from a full genome alignment using snp-sites<sup>63</sup> v.2.5.1, 424 with genomic islands<sup>51</sup> masked. Separate maximum likelihood phylogenies were constructed for

425 core genome SNP alignment, cgMLST alignment, and MLST alignment using IQ-TREE<sup>36</sup> 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 treespace<sup>64</sup> 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 437 distances were computed, and principal components (PCs) were derived with eigenvalues 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 445 compared using iTOL<sup>65</sup> (**Supplementary Figure 1b**).

446

### 447 **Specific lineage analysis**

448 To investigate the dissemination and genetic diversity within each dominant lineage, we 449 conducted individual genome alignments, recombination removal, and maximum likelihood 450 phylogeny. To enhance the sensitivity of variant, we selected closely related genomes as 451 references for each lineage. Specifically, the complete genome *B. pseudomallei* strain K96243 452 (accession numbers BX571965 and BX571966) served as the mapping reference for lineage 1, 453 while new reference genomes were created for lineages 2 and 3.

454

455 For lineage 2, we chose a representative isolate 27035\_8#57 and subjected it to long-read 456 sequencing on a local MinION sequencer following the manufacturer's standard protocol 457 (Oxford Nanopore Technologies, Oxford, United Kingdom). A complete hybrid assembly of the 158 Iong-read and short-read sequence data of this strain was performed using Unicycler<sup>66</sup> v.0.8.4.

459 The resulting hybrid assembly of the 27035\_8#57 genome was employed as the mapping 460 reference for this lineage.

461

462 In the absence of representative complete genomes for lineage 3, we selected the best quality 463 de novo assembly of isolate 27035\_8#119 and orientated its contigs according to strain K96243 464 using ABACAS $^{67}$  v.1.3.1. This genome was used as a mapping reference for lineage 3.

465

466 For lineage-specific mapping, Snippy v.4.6.0 as employed as in the whole population analysis. 467 All genome alignments were subjected to Gubbins<sup>68</sup> v.3.1.3, a recombination identification tool, 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 472 lineage using IQ-TREE<sup>36</sup> v.2.0.3 with TVM+F+ASC+R6 and 1,000 replicates of bootstrap 473 support. The overall proportion of nodes with ≥80% bootstrap support of lineage-specific<br>474 phylogenies reached 83.5%. phylogenies reached 83.5%.

475

## 476 **Dating the timeline for lineages and sub-lineages**

477 To enable dating analysis, we further divided each lineage into sub-lineages using R package 478 rhierbaps<sup>69</sup> v.1.1.4. We inferred evolutionary timeline and estimated the age of each lineage and 479 sub-lineage based on the isolate's collection date. A Bayesian molecular dating provided in R 480 package BactDating<sup>70</sup> 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

486 Notably, the temporal signals were discernable at the sub-lineage level rather than the broader 487 lineage level. Among the 10 sub-lineages, only one sub-lineage (lineage 1.3) exhibited a 488 positive correlation in their clock signals. Given the limited sample size of lineage 1.3, we 489 employed a strict clock model to prevent parameter over-fitting. We ran three independent 490 Markov chain Monte Carlo (MCMC) chains, each spanning at least 100 million iterations, and 491 sampled every 10,000 steps. The prior mutation rate derived from Pearson and colleagues was 492 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 ggtree<sup>71</sup> 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 varying 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 505 province (n = 15 isolates) and permuted 1,000 times. Using the stochastic character mapping 506 function (*make.simmap*) from the R package phytools<sup>72</sup> v.1.9.16, we conducted 100 simulations 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**

515 All the study genomes were annotated using Prokka<sup>73</sup> v.1.14.5, and further used in the pan-516 genome analysis. Each genome has a median of 5,845 coding sequences (CDS) predicted onto 517 each genome with a range of 5,642 to 6,142 CDS per genome. Panaroo<sup>74</sup> v.1.3.3 was 518 employed to estimate the pan-genome with a sensitive option and a cut-off sequence identity of 519 92% derived from previous study<sup>15</sup>. The number of estimated genes falls within a comparable 520 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),

527 intermediate (95% in dominants vs 10% in non-dominants), and relaxed (95% in dominants vs 528 15% in non-dominants). Based on visual examination of gene distribution patterns (**Figure 3a,**  529 **Supplementary Figure 7**), the relaxed threshold was used to maximise the number of genes 530 included in subsequent analysis.

531

## 532 **Identification of Gene Ontology (GO: terms)**

533 Amino acid sequences of lineage-specific genes were submitted to InterPro database<sup>46</sup> 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.<sup>47</sup> 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 GEOquery<sup>75</sup> 551 v.2.58.0, and differential gene expression analysis was performed using the R package limma<sup>76</sup> 552 v.3.58.1.

553 We used binary expression patterns reported in Ooi *et al.<sup>47</sup>* to compare the expression profiles 554 of lineage-specific genes against the remaining genes in K96243 across 62 conditions. This 555 enabled the comparison of the count of expressed genes within the lineage-specific category 556 against the remainder of the genes for each condition using Fisher's exact test, with multiple 557 testing adjustments via Benjamini-Hochberg corrections.

## 558 **Lineage 2 transcriptome analysis**

559 This analysis focused on the expression profile of a strain UKMD286, representative of lineage 560 2. RNAseq data was obtained from an experiment conducted by Ghazali *et al.* <sup>48</sup> and accessed 561 through the European Nucleotide Archive (ENA) (E-MTAB-11200).

562 To simulate environmental conditions, UKMD286 was cultured in BHIB medium overnight, 563 resuspended, and inoculated into soil extract medium. For infection condition, BALB mice were 564 infected with UKMD286, and the bacteria were harvested from spleens five days post infection. 565 Each experiment was conducted with two replicates. FastQC v.0.11.9 and FastXtool v.0.0.14 566 were used to pre-processed sequenced reads. Raw reads were aligned to UKDM286 genome 567 using Hisat2 $^{77}$  v. 2.2.1 with differential gene expression performed using the R package 568 DESeq $2^{78}$  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.<sup>49</sup>* 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 578 replicates. FastQC v.0.11.9 and FastXtool v.0.0.14 were used to pre-processed sequenced 579 reads. Raw reads were aligned to UKMH10 genome using Hisat2 $^{77}$  v.2.2.1 with differential gene 580 expression performed using the R package DESeq2<sup>78</sup> v.1.40.2.

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# 600 **Data availability**

601 The genome sequence data presented in this study can be found in online repositories. The 602 ENA under study accession number PRJEB25606 and PRJEB35787. The accession numbers 603 for individual genomes, and their annotated assemblies are listed in **Supplementary data 1**.

#### 604 **Code availability**

605 The analyses used public software including Kraken v.1.1.1, CheckM v.1.2.2, FastANI v.1.31, 606 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, 607 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 608 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, 609 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**

611 The research is led by local researchers and include local contributions throughout all research 612 process including the study design, study implementation, data ownership, intellectual 613 properties and authorship for publication.

#### 614 **Author contributions**

615 RS, CChe and NC conceived, designed the study and write the original draft. NC and CChe 616 administrated and supervised the project. NC acquired funding to collect isolates, while CChe 617 acquired funding for sequencing and downstream analysis. NC, RS, TEW, and NS collected 618 and identified bacterial isolates. NC, RS and RP collected clinical data. CChe, NRT, NC, RS, 619 JT, EB and WC performed whole-genome sequencing. NPJD and NC contributed reagents. 620 NRT and JP contributed software tools. RS, CCho, and CChe performed bioinformatics 621 analyses. CChe, NC, NRT and JP interpret the analyses. CChe wrote the revised draft. All 622 authors read and approved the manuscript.

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



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 yellow. 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.

665 **Supplementary Figure 1** Comparison of approaches used in in outlining the population 666 structure (a) A scatter plot displays the first two principal components (PCs) derived from 667 pairwise distances between trees. Each dot represents a bootstrap tree and is colour-coded by 668 the method used to generate the tree: core genome SNP (blue), cgMLST (red) and MLST 669 (green). (b) A scree plot summarises eigenvalues computed for each PCs. (c to d) The median 670 phylogenetic trees constructed from core genome SNP, cgMLST, and MLST and their 671 consistency with PopPUNK clustering method.

672 **Supplementary Figure 2** Distribution of environmental and clinical isolates by each lineage. 673 Barplots highlight the co-detection of environmental (green) and clinical isolates (red) across 674 dominant lineages 1, 2, and 3.

675 **Supplementary Figure 3** Lineage 1 specific analysis (a) A recombination removed lineage 1 676 phylogeny with colour stripes displaying its sub-lineage structure, year of collection, and 677 sampling province (left to right). (b) A map of northeast Thailand showing the distribution of 678 each isolate and the region's river system. (c) Time-calibrated phylogeny of sub-lineage 1.3 with 679 blue error bars indicating 95% highest posterior density interval, with the estimated mutational 680 rate consistent with previous study.

681 **Supplementary Figure 4** Lineage 2 specific analysis (a) A recombination removed lineage 2 682 phylogeny with colour stripes highlighting sub-lineage structure, year of collection, and sampling 683 province (left to right). (b) A map of northeast Thailand with the region's river system. Dots 684 present the distribution of individual samples.

685 **Supplementary Figure 5** Lineage 3 specific analysis (a) A recombination removed lineage 3 686 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 isolate 688 and the region's river system.

689 **Supplementary Figure 6** Transmission patterns and evolutionary time spent at each province. 690 (a to c) Proportion of transition events (Markov jumps) among provincial pairs for lineage 1, 2, 691 and 3 respectively. The pairs were denoted as province1 – province 2, with transitions from 692 province 1 to province 2 shown in red, and transitions from province 2 to province 1 in green. 693 The Man-Whitney U test was conducted for each pair to assess differences in transition 694 frequency by direction, with Bonferroni correction applied for multiple tests. (d to f) Total branch 695 length from provincial trait reconstruction (Markov rewards) for lineage 1, 2, and 3, respectively.

696 **Supplementary Figure 7** Selection criteria for lineage-specific genes. Scatter plots show the 697 frequency distribution of lineage-specific (red) against other genes (black), based on their 698 distribution within the dominant lineages and their sub-lineages (horizontal axis) compared to 699 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.

709 **Supplementary data 1** Epidemiological data, isolate and accession codes for both short reads 710 and annotated assembly used in this study deposited in the European Nucleotide Archive (ENA) 711  $(n = 1,391)$  (provided as a separate excel file).

712

713 **Supplementary data 2** Quality control information of newly sequenced *B. pseudomallei* from 714 northeast Thailand collection ( $n = 1,265$ ) (provided as a separate excel file).

715

716 **Supplementary data 3** List of dominant lineage-specific genes and their Gene Ontology (GO) 717 terms (provided as a separate excel file).

#### 718

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

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Figure 2







![](_page_32_Figure_3.jpeg)

# Figure 4

![](_page_33_Figure_1.jpeg)