

1 **Large-scale investigation for antimicrobial activity reveals novel defensive species across**  
2 **the healthy skin microbiome**

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31 **Abstract**

32

33 The human skin microbiome constitutes a dynamic barrier that can impede pathogen invasion by  
34 producing antimicrobial natural products. Gene clusters encoding for production of secondary  
35 metabolites, biosynthetic gene clusters (BGCs), that are enriched in the human skin microbiome  
36 relative to other ecological settings, position this niche as a promising source for new natural  
37 product mining. Here, we introduce a new human microbiome isolate collection, the EPithelial  
38 Isolate Collection (EPIC). It includes a large phylogenetically diverse set of human skin-derived  
39 bacterial strains from eight body sites. This skin collection, consisting of 980 strains is larger and  
40 more diverse than existing resources, includes hundreds of rare and low-abundance strains, and  
41 hundreds of unique BGCs. Using a large-scale co-culture screen to assess 8,756 pairwise  
42 interactions between skin-associated bacteria and potential pathogens, we reveal broad antifungal  
43 activity by skin microbiome members. Integrating 287 whole isolate genomes and 268  
44 metagenomes from sampling sites demonstrates that while the distribution of BGC types is stable  
45 across body sites, specific gene cluster families (GCFs), each predicted to encode for a distinct  
46 secondary metabolite, can substantially vary. Sites that are dry or rarely moist harbor the greatest  
47 potential for discovery of novel bioactive metabolites. Among our discoveries are four novel  
48 bacterial species, three of which exert significant and broad-spectrum antifungal activity. This  
49 comprehensive isolate collection advances our understanding of the skin microbiomes  
50 biosynthetic capabilities and pathogen-fighting mechanisms, opening new avenues towards  
51 antimicrobial drug discovery and microbiome engineering.

52

## 53 Introduction

54  
55 Human skin represents a first line of defense against mechanical and chemical insults while  
56 maintaining homeostasis<sup>1-7</sup>. One of the most significant roles of the skin is to act as a barrier to  
57 invading pathogens. This physical barrier is fortified by a diverse microbiome composed of  
58 bacteria, fungi, viruses, and microeukaryotes<sup>8</sup>. The surface area of the skin is estimated to be  
59 around 30 m<sup>2</sup>, including appendages such as hair follicles and sweat ducts<sup>5</sup>, making it one of the  
60 most expansive direct host-microbe interfaces in the body. Across the body, unique  
61 microenvironments form on the skin with characteristic moisture levels, pH, and lipid content,  
62 driving the composition of the associated microbial community<sup>7,9-12</sup>. Far from mere bystanders,  
63 members of the skin microbiome help maintain the integrity of the skin barrier through both  
64 direct and indirect defense mechanisms<sup>9-11</sup>.

65  
66 The skin microbiome fulfills diverse additional functional roles including establishment of  
67 immune tolerance and sensing pathogens<sup>13-15</sup>. Skin microbiota indirectly compete against  
68 potential pathogenic invaders by consuming limited nutrients and acidifying the skin surface<sup>1,16</sup>.  
69 Constituents of the microbiome also engage in direct defense of the skin through mechanisms  
70 such as production of antimicrobial molecules. Recent efforts to mine the human microbiome for  
71 genes encoding bioactive metabolites have revealed a rich biosynthetic potential<sup>17</sup>. Given that  
72 bacteria inhabit different niches within the human body and are exposed to different competitors  
73 in the environment, we expect that skin microbiota produce metabolites that are relevant to the  
74 niche that they colonize. An example is lugdunin, a non-ribosomal thiazolidine cyclic peptide  
75 produced by *Staphylococcus lugdunensis* which is commonly found in nasal cavities. Lugdunin  
76 is bactericidal against methicillin-resistant *S. aureus* and strains of vancomycin resistant  
77 *Enterococcus* but shows no toxicity toward primary human erythrocytes or neutrophils<sup>18</sup>. An  
78 additional example is *Cutibacterium acnes*, commonly found at sebaceous body sites, which  
79 produces cutimycin, a thiopeptide with anti-staphylococcal activity<sup>19</sup>. Further, numerous human  
80 skin coagulase-negative *Staphylococcus* commensal species produce lantibiotics that inhibit *S.*  
81 *aureus*<sup>18,20-23</sup>.

82  
83 Characterization of antimicrobial products from the skin microbiota has historically focused on  
84 easily cultivated, high-abundance organisms, particularly *Staphylococcal* species, targeting the  
85 Gram-positive skin pathogen *S. aureus*<sup>18-20,24-27</sup>. Recent studies have shown these abundant skin  
86 microbes harbor a wide diversity of biosynthetic gene clusters (BGCs) in their genomes<sup>28-30</sup>.  
87 However, the skin microbiome also contains numerous species of low abundance<sup>31</sup> with  
88 unexplored biosynthetic potential. These species likely harbor many more uncharacterized  
89 BGCs. Given that currently characterized BGCs come from a limited set of microbial genera,  
90 and most BGCs remain uncharacterized<sup>32,33</sup>, the skin microbiome represents a potentially rich,  
91 untapped source of antimicrobial compounds. This potential is further amplified by the presence  
92 of distinct and unique BGCs, likely to encode for the synthesis of novel bioactive compounds.  
93 These facts suggest that most of the chemical diversity encoded with skin-associated microbial  
94 genomes remains unknown. The ongoing discovery of new bacterial species on the skin through  
95 combined metagenomic assembly and cultivation approaches, exemplified by the recent Skin  
96 Microbial Genome Collection (SMGC) identifying 174 new bacterial species, further  
97 underscores the richness of this ecosystem<sup>31</sup>. Identifying the yet-unknown metabolites produced

98 in this niche will not only deepen our understanding of skin biology but likely unveil novel  
99 therapeutic molecules.

100  
101 We have developed the The EPithelial Isolate Collection (EPIC), a microbial biorepository of  
102 6,540 bacterial strains isolated from 1,060 mammalian samples. This new collection is derived  
103 from epithelial sites such as the skin, oral, and nasal barriers in humans, swine, non-human  
104 primates, horses, chickens, goats, donkeys, and cows. Here, we detail the extensive  
105 characterization of healthy human skin isolates included in EPIC, along with matched  
106 metagenomes from eight distinct human skin body sites. This subset of EPIC includes 980  
107 bacterial isolates, 287 whole genomes, and 268 metagenomes (**Figure 1**). Our collection is much  
108 larger and significantly more diverse than previous collections<sup>31</sup>, comprising hundreds of rare  
109 and low-abundance strains, many of which are not included in previously reported  
110 collections<sup>31,34</sup>. We use a large-scale solid-phase co-culture screen to assess pairwise interactions  
111 between skin-associated bacteria and potential pathogens spanning Gram-negative, Gram-  
112 positive and fungal pathogens. This data reveals broad antifungal activity by members of the skin  
113 microbiome. Comparing gene cluster families (GCFs) allowed us to visualize hundreds of GCFs  
114 and demonstrate most of them are unique to our collection. Finally, we describe four novel skin-  
115 associated bacterial species, unique to this collection, that exert significant antifungal activity.  
116 The BGCs of these novel species are distinct from those in related species of the same genera.  
117 This rich and diverse isolate collection enables systematic exploration of skin microbiome  
118 chemistry, illuminates the breadth of antagonistic interactions within the skin microbiome, and  
119 will provide new molecular insights into how commensal bacteria defend against pathogenic  
120 invasion.

## 121 **Results**

122  
123 To create our skin microbiome collection, we recruited 34 healthy volunteers to provide samples  
124 from 8 distinct body sites (**Figure 2A**) with biogeographic ranges from moist (nares, umbilicus,  
125 toe web space), rarely moist or dry (antecubital fossa and volar forearm), and sebaceous (alar  
126 crease, back, and occiput) microenvironments (**Table S1**). Samples from 17 participants were  
127 processed for culture on several media types to capture phylogenetic and strain diversity across  
128 body sites. Shotgun metagenomic sequencing from all 34 participants was used to assess  
129 microbial community composition, diversity, and agreement between metagenomes and cultured  
130 isolates from each sample. The strain repository is named the EPithelial Isolate Collection  
131 (EPIC) and contains 980 skin-associated bacterial strains from 136 skin samples. This makes  
132 EPIC much larger and more diverse than published studies<sup>31</sup>. For example, 74.3% of the isolates  
133 are classified outside of the skin-dominant genera *Staphylococcus* and *Corynebacterium*,  
134 compared to 23.1% of isolates in the Skin Bacterial Culture Collection (SBCC)<sup>31</sup>. Furthermore,  
135 EPIC contains skin isolates belonging to an additional 24 genera not represented in the SBCC  
136 (**Table S2**).

### 137 138 **The EPIC library includes rare, low-abundance, and phylogenetically diverse bacteria**

139  
140 Using culture-based techniques, representative bacterial strains from each sampled body site and  
141 across individuals were isolated. To maximize the phylogenetic diversity of EPIC, each skin  
142 swab was cultured using multiple types of media, including media components to suppress

143 *Staphylococcus* growth and to foster *Corynebacterium* and lower abundance *Actinomycetota*  
144 growth. Taxonomic classification of isolates was accomplished with full length 16S rRNA  
145 sequencing and, for a subset, whole-genome sequencing (**Table S2, S3**). In total, the 980 skin  
146 isolates represent at least 70 species (95% ANI threshold for species delineation<sup>35</sup>) spanning 40  
147 genera. Most isolates fall within the phylum Actinomycetota (n=529), followed by Bacillota  
148 (n=177), Pseudomonadota (n=27), and Bacteroidota (n=4). *Micrococcus*, *Staphylococcus*, and  
149 *Corynebacterium* species were most frequently isolated from most body sites. We also isolated  
150 rare and phylogenetic diverse species across different genera, such as *Kocuria*,  
151 *Aestuariimicrobium*, *Kyotococcus*, *Nesterenkonia*, *Microbacterium*, *Brachybacterium*, *Rothia*,  
152 *Dietzia*, and *Dermabacter*. Distinct compositions of bacterial isolates were derived from each  
153 site, with the nares showing the greatest taxonomic diversity (**Figure 2A**); where 52 isolates  
154 belonging to 12 different genera were collected.

155  
156 Metagenomic sequencing assessed concordance between culture-dependent versus culture-  
157 independent profiling (**Figure 2BC**). Similar to previous findings<sup>16,36,37</sup>, *Cutibacterium*,  
158 *Corynebacterium*, *Micrococcus*, and *Staphylococcus* are the most abundant skin colonizers  
159 across all sampled body sites. Focusing on low-abundance genera, defined as a relative  
160 abundance of < 1% in metagenomes, EPIC isolates exhibit considerable overlap with this group  
161 (**Figure 2C**). This indicates that EPIC consists of common (high abundance) skin colonizers and  
162 rare (low abundance) skin colonizers, spanning the phylogenetic diversity of the skin  
163 metagenome. Because purely aerobic culturing conditions were used, EPIC is deficient in  
164 *Cutibacterium* species due to their anaerobic nature (**Figure 2C**).

165  
166 In summary, we have created a comprehensive library of skin bacterial isolates encompassing  
167 species typically present at very low levels for further exploration. We begin this exploration by:  
168 1) mining EPIC for antimicrobial activity, using a large pairwise solid-phase screen; and 2)  
169 annotating and analyzing newly discovered BGCs.

## 170 171 **A pairwise interaction screen reveals widespread inhibition of pathogens by skin-associated** 172 **bacterial strains**

173  
174 Previous studies that investigated the bioactive and antimicrobial capacity of skin microbes have  
175 focused on highly abundant species, such as *Staphylococcus* spp. Here, we developed a pairwise  
176 interaction screen to score the ability of phylogenetically diverse taxa, including low-abundance  
177 species, to provide colonization resistance to diverse human pathogens in a contact-independent  
178 manner (**Figure 3A; Table S4**). Through co-culture on solid media, we scored pairwise  
179 interactions between all 398 isolates with a pathogen panel comprising 22 Gram-positive and  
180 Gram-negative bacteria, and fungi (8,756 total interactions). Briefly, EPIC isolates are grown for  
181 7 days to allow accumulation of secreted metabolites, followed by inoculation with the pathogen  
182 in the same well but in a spatially distant location. Interactions are scored as neutral or no  
183 inhibition (0), partial inhibition (1), or complete or full inhibition (2) (**Figure 3A**). Widespread  
184 inhibition of Gram-positive and fungal pathogens by EPIC isolates occurred, while antagonism  
185 towards Gram-negative pathogens was less common (**Figure 3BC**). Hierarchical clustering  
186 shows that isolates not exhibiting any inhibition include diverse genera. This highlights the  
187 species- and even strain-level variation in interaction patterns, where some strains of the same  
188 species may strongly inhibit the growth of specific pathogens while others do not (**Figure 3B**).

189  
190 Given the prominence of broad-spectrum fungal inhibition, we quantified the spectrum of  
191 activity for isolates inhibiting any one of the fungal pathogens. *Cryptococcus neoformans* is the  
192 most susceptible, with 129 isolates capable of completely inhibiting its growth. We identified 75  
193 and 82 isolates able to completely inhibit the growth *Candida albicans* and *Candida* sp.,  
194 respectively. Over 25 isolates have full inhibition against *Aspergillus flavus* and *Trichosporon*  
195 *asahii* (**Figure 3B**). Together, more than 30 isolates displayed broad spectrum activity against *C.*  
196 *neoformans*, *Candida* sp., and *C. albicans* (**Figure S1**). A subset of 84 isolates with strong *C.*  
197 *albicans* inhibitory activity was randomly selected to evaluate their ability to inhibit the  
198 multidrug-resistant pathogen *Candida auris*. *C. auris* primarily spreads through skin colonization  
199 and invasive infections are associated with mortality rates exceeding 60%<sup>38-41</sup>. Of this subset,  
200 40% of the isolates spanning diverse genera displayed partial to complete growth inhibition of *C.*  
201 *auris* (**Table S5**). Global bioactivity patterns were then assessed to determine that inhibitory  
202 profiles of EPIC isolates cluster by genus rather than body site (**Figure 3BD**). This supports a  
203 framework where closely related species display similar bioactivity profiles, independent of the  
204 local microenvironment. We developed a summarized inhibition score to further quantify  
205 antagonistic activity of each genus against each pathogen group. Isolates in the genera  
206 *Citricoccus*, *Staphylococcus*, *Kocuria*, *Micrococcus*, *Microbacterium*, *Brevibacterium*, and  
207 *Sphingobacterium* displayed the greatest fungal inhibition compared to Gram-positive and Gram-  
208 negative pathogens (**Figure 3D**). Six isolates from the genera *Brevibacterium*, *Microbacterium*,  
209 *Sphingobacterium*, and *Staphylococcus* show inhibition against all five fungal pathogens tested  
210 (**Figure S1**).

211  
212

### 213 **The EPIC expands the known biosynthetic potential of the skin microbiome**

214

215 Our discovery of widespread antimicrobial action among the species in our collection suggests  
216 significant novel biosynthetic capacity within the skin microbiome. To explore this biosynthetic  
217 potential, we sequenced the genomes of 287 strains displaying potent inhibitory activity.  
218 Genomes were dereplicated at 99% average nucleotide identity, resulting in a set of 182 distinct  
219 genomes for annotation of BGCs for<sup>42</sup>. The distribution of broad-level BGC types was similar  
220 across body sites and within taxonomic groups (**Figure 4AB**). To comprehensively understand  
221 the diversity of BGCs in skin-associated microbes, we similarly predicted BGCs across 621  
222 genomes or MAGs from the SMGC, a recently established reference collection of microbial  
223 genomes from human skin<sup>31</sup>. Using unified clustering of BGCs from the SMGC and whole  
224 genomes generated in this study, we find 1,960 distinct gene-cluster families (GCFs)<sup>43</sup> (**Table**  
225 **S6, S7**). Of the 305 GCFs found within EPIC skin isolate genomes, only 12 GCFs (3.9%)  
226 correspond to characterized BGCs from the MIBiG database, a repository of BGCs for which  
227 metabolic information is available (**Figure 4C; Table S8**). The 12 GCFs found in MIBiG,  
228 include BGCs encoded by well-studied genera associated with skin, such as BGCs for the  
229 synthesis of aureusimines<sup>44</sup> and dehydroxynocardamine<sup>45</sup> from *Staphylococcus* and  
230 *Corynebacterium*, respectively. The remaining known GCFs are associated with environmental  
231 taxa, such as *Bacillus* and *Streptomyces*. While some of the 305 GCFs found in EPIC genomes  
232 are also present in the SMGC, the majority, 54.4%, spanning at least 30 types of BGCs from 28  
233 genera, are unique to EPIC (**Figure 4C**).

234

235 To infer potential novelty of antimicrobial small molecules that may be encoded by these GCFs,  
236 we investigated the resistome of the skin microbiome, focusing on genes encoding specific  
237 antibiotic resistance enzymes that may confer self-protection<sup>46</sup>. Resistance to macrolides and  
238 beta-lactams were the most common across body sites; however, overall, a low prevalence of  
239 antimicrobial resistance (AMR) genes within the skin metagenome is observed (**Figure S2A**).  
240 Similarly, through assessing whole genomes of cultured isolates, only 41 (22.5%) of 182  
241 genomes encoded antibiotic resistance protein homologs. When considering resistance protein  
242 homologs and protein variants predicted to confer resistance, such as mutations in ribosomal  
243 protein rpsL conferring resistance to aminoglycosides, 122 of 182 genomes contained at least  
244 one resistance determinant (**Figure S2B, Table S9**).

245  
246 To understand whether particular body sites exhibit greater biosynthetic potential relative to  
247 others, we examined the presence of predicted GCFs in our skin metagenomes (**Figure S3,**  
248 **S4**)<sup>47,48</sup>. The number of distinct GCFs discovered as a function of sequencing depth at each body  
249 site across multiple individuals was assessed through rarefaction<sup>49,50</sup> (**Figure 4D**). Sebaceous  
250 body sites exhibit the clearest saturation for discovery of GCFs, consistent with prior  
251 observations that these sites are dominated by *Cutibacterium acnes*, which carry a limited set of  
252 GCFs<sup>8,19,37,51</sup>.

#### 253 254 **Discovery of novel skin bacterial species with strong antifungal activity**

255  
256 Using whole genome sequencing from EPIC isolates, three *Corynebacterium* species and one  
257 *Brachybacterium* species lacking cultured representatives but are predicted as novel species  
258 based on metagenome assembly in the SMGC were identified. In addition, four predicted novel  
259 species that are not represented in either the Genome Taxonomy Database or the SMGC were  
260 discovered. Each of these species belongs to a distinct but known genus: *Aestuariimicrobium*,  
261 *Corynebacterium*, *Kocuria*, and *Brevibacterium* (**Table S3**). To further assess the uniqueness of  
262 these species, we searched public metagenomes in the NCBI Sequence Read Archive to  
263 determine the environmental distribution<sup>52,53</sup>. *Corynebacterium* isolate LK952 is most commonly  
264 found in skin metagenomes, followed by gut and other human-associated metagenomes.  
265 Similarly, *Aestuariimicrobium* isolate LK1188 and *Brevibacterium* isolate LK1337 are detected  
266 in human skin metagenomes. The *Kocuria* isolate LK960 was identified in human skin  
267 metagenomes and other diverse environments, suggesting a larger host range (**Figure 5A**).

268  
269 Each of these novel species displayed complete inhibition of fungal pathogens with limited  
270 antibacterial activity (**Figure 5B**). The *Kocuria* isolate completely inhibits *C. albicans*, *C. auris*,  
271 and *A. flavus* and although to a lesser extent, also inhibits *Trichosporon asahii* (**Figure 5C**). The  
272 novel *Corynebacterium* species exerts moderate antifungal activity while the *Brevibacterium*  
273 isolate displays less potent antifungal activity, and *Aestuariimicrobium* has no antifungal  
274 activity. Prediction and annotation of BGCs reveal the new species encode one to four BGCs  
275 each (**Figure S4**).

276  
277 Finally, we assessed the novelty of each species' BGC-ome, the collection of BGCs for a single  
278 genome, by querying the presence of predicted BGCs in genomes of known species from each  
279 genus<sup>54</sup>. Intriguingly, the BGC-ome of the novel *Corynebacterium* isolate is substantially  
280 divergent from other *Corynebacterium* genomes, with several genes missing that are present

281 within homologous BGC of other species (**Figure 5D**). *Aestuariimicrobium* contains a single  
282 BGC that is predicted to encode for a unique terpene cluster (**Table S4**). This comparative  
283 genetic analysis confirmed that BGCs encoded in each predicted novel species genomes largely  
284 have <95% amino acid identity to homologous BGCs from related species in their respective  
285 genera (**Figure 5D-F**).

286  
287

## 288 Discussion

289

290 The human skin microbiome plays an essential role in barrier maintenance and host immune  
291 regulation. In addition, it prevents pathogen invasion, indicating vast and largely unknown  
292 antimicrobial biosynthetic potential. Here, we present a human skin microbiome collection,  
293 including hundreds of strains, including cultured isolates for at least 7 new species, and  
294 metagenomes from 34 participants across 8 body sites. The isolate collection is several times  
295 larger and more diverse than existing resources and was designed to include far more taxa that  
296 are rare or low in abundance. Using this collection, we explore the expansive degree of  
297 antagonism skin microbes can exhibit towards human pathogens. Using large-scale solid-phase  
298 bioassays, we tested hundreds of strains for activity against 22 pathogens, and after assessing  
299 8,756 pairwise interaction, we discover broad antimicrobial activity, most notably antifungal  
300 activity, in scores of strains that include new species. This confirms extensive antimicrobial  
301 biosynthetic potential, which we then explore in individual genomes, describing numerous new  
302 BGCs across multiple genera. Our collection thus has significant potential for the expansion of  
303 our knowledge of the biosynthetic resources within the skin microbiome.

304

305 Recently, broad intraspecies antagonism on the skin was reported within *S. epidermidis* strains  
306 isolated from facial skin<sup>55</sup>. Our findings reveal broader antagonistic networks across taxonomic  
307 lineages. Skin isolates spanning four bacterial phyla illustrate extensive cross-kingdom  
308 antimicrobial capacity with enrichment against fungal pathogens. This further suggest that inter-  
309 kingdom antagonism may be as significant as intraspecies competition in shaping microbial  
310 community assemblages on human skin.

311

312 Our resource is one of the most diverse human barrier site isolate collections. Nevertheless,  
313 *Micrococcus* is overrepresented in our strain collection. This finding is consistent with previous  
314 studies in isolating *Micrococcus* from human skin<sup>56,57</sup>. They are prevalent members of the skin  
315 microbiome, strict aerobes, and non-fastidious. *Micrococcus* are frequently found in indoor air<sup>58</sup>,  
316 which could be the result of skin shedding. Despite the overrepresentation, our culture isolation  
317 methods also captured fastidious bacteria, including novel, rare, and low-abundance skin species  
318 that appear at less than 1% abundance in metagenomes.

319

320 Based on whole-genome based taxonomic classification, we determined that eleven isolates are  
321 representative of eight novel species belonging to distinct but known genera including  
322 *Aestuariimicrobium*, *Corynebacterium*, *Kocuria*, *Brevibacterium*, and *Brachybacterium*. Six of  
323 these isolates, representing 4 novel species genomes, lacked culture representatives. Five  
324 additional isolates representing another 4 novel species, lacked both culture representatives and  
325 genomes. Three of the species in the genera *Corynebacterium*, *Kocuria*, and *Brevibacterium*,  
326 exert partial to complete inhibition of human fungal pathogens, while exerting no inhibition of



327 Gram-positive and Gram-negative pathogens. We find that BGCs from all four novel species  
328 exhibit both substantial amino acid sequence divergence and gene content variation relative to  
329 orthologous BGCs in other species from their genera. This is likely due to genetic drift or other  
330 evolutionary processes that occur during the process of speciation<sup>59</sup>. Thus, compared to the  
331 extensively studied filamentous actinomycetes, which have been heavily mined<sup>60</sup>, the BGCs  
332 from rare and low-abundant genera on the skin have received less attention but are likely to be  
333 valuable sources of novel antimicrobials<sup>61</sup>. Our finding demonstrates that the skin microbiome  
334 can serve as a rich reservoir for exploring such rare actinomycetes and our collection makes this  
335 possible.

336  
337 By surveying BGC distributions of isolates across 8 body sites, we found that the skin  
338 microbiome encodes broad classes of BGCs. This finding expands our knowledge of the  
339 distribution of BGCs in the human microbiome, including gut, vagina, airways, skin, and oral<sup>17</sup>.  
340 At least 32 BGC types and 305 distinct GCFs were identified within genomes of sequenced skin  
341 isolates, spanning body sites and microenvironments, suggesting the presence of a rich repertoire  
342 of specialized metabolites. Further, this indicates that there is a high degree of specialization  
343 among the microbial inhabitants, shaping skin microbiome composition through establishing  
344 colonization resistance. In addition, our analysis identified GCFs that belong to environmental  
345 taxa such as *Bacillus* and *Streptomyces*. This observation might be attributed to the skin's  
346 continuous exposure to the environment, which potentially allows for the, usually transient,  
347 residence of environment-associated taxa and their respective BGCs on the skin<sup>62</sup>.

348  
349 Most of the GCFs identified in EPIC skin genomes are distinct from previously-characterized  
350 BGCs, with 166 GCFs absent in skin-associated microbial genomes from the SMGC<sup>31</sup>. Through  
351 assessment of metagenomes, we find that rarely moist or dry sites had the highest numbers of  
352 GCFs discovered as a function of sequencing depth, suggesting that these sites have the greatest  
353 potential for isolating microbes with novel chemistries. Notably, our approach for profiling the  
354 presence of GCFs, using a catalog gathered from the EPIC and the SMGC, is likely to miss some  
355 BGCs from taxa not represented across the two genome collections.

356  
357 Our understanding of the antibiotic resistome offers a strategic approach to natural product  
358 discovery and target identification<sup>63</sup>. Self-resistance machinery in the producer organism  
359 typically co-localize within BGCs encoding enzymatic modules to synthesize the active  
360 compound<sup>63</sup>. Using self-resistance as a guide has led to the discovery of thiotentronic acid  
361 antibiotics through identification of a putative fatty acid synthase resistance gene in  
362 *Salinospora*<sup>64</sup> and pyxidicycline through pentapeptide repeat proteins in *Myxobacteria*<sup>65</sup>. We  
363 evaluated the presence of AMR genes within our skin bacterial isolates and metagenomic  
364 samples and find that the prevalence of AMR genes within the skin metagenome is low across  
365 body sites. Overall, these findings are consistent with reports that healthy host-associated  
366 microbiomes have a lower prevalence of AMR genes compared to diseased states<sup>66,67</sup>. This  
367 suggests that antimicrobial molecules produced likely target alternative mechanisms from  
368 currently available antibiotics, which emphasizes the need for characterizing antimicrobial  
369 molecules and their mode of actions on the human skin.

370  
371

372 Human skin is a critical defense barrier, hosting a unique microbiome that can produce  
373 specialized metabolites to protect their niche and in turn the host. This study revealed a large and  
374 phylogenetically diverse set bacterial species with significant antifungal activity. These findings  
375 expand our understanding of antimicrobial production within the human microbiome across  
376 skin<sup>18,19,25,68</sup>, oral<sup>69</sup>, and nasal ecosystems<sup>45</sup>. While the broad ecology and colonization resistance  
377 function of the skin microbiome is well documented, its specific role in defense against fungal  
378 pathogens is less understood. This is a critical gap given that fungi are implicated in numerous  
379 dermatological conditions, such as pityriasis versicolor, seborrheic dermatitis, atopic  
380 dermatitis<sup>70,71</sup>, and chronic wounds<sup>72,73</sup>, but they also exist as commensals within the skin  
381 microbiome<sup>3</sup>. Our findings demonstrate the remarkable capacity of skin-associated bacteria to  
382 inhibit fungal growth and expansion through production of secreted antifungal molecules. This  
383 opens new lines of investigation for identifying novel and safe antifungal compounds serving a  
384 dual role of regulating the commensal fungal communities and preventing the proliferation of  
385 pathogenic fungi on the skin<sup>3</sup>.

386  
387

## 388 **Methods**

### 389 **Participant recruitment**

390 We recruited participants at the University of Wisconsin-Madison under an Institutional Review  
391 Board approved protocol. Inclusion criteria included age >18 yr. Participation in the study was  
392 completely voluntary and participants were able to stop at any time. Participants received no  
393 payment for being a part of the study.

394

### 395 **Sample collection**

396 We collected samples for metagenomic sequencing by wetting a sterile foam swab in nuclease-  
397 free water and swabbing approximately a 1 in. x 1 in. area of the selected site on the right-hand  
398 side of the participant's body. We swabbed the area approximately 15 times in a downward  
399 motion with constant pressure while rotating the swab. We collected the swab into a 2.0 ml  
400 BioPure Eppendorf tube containing 300 µl Lucigen MasterPure™ Yeast Cell Lysis solution.  
401 Tubes were labeled with subject ID, body site, visit number, and date of collection and stored at -  
402 80°C until DNA extraction.

403

404 For culturing samples, we used the Copan Diagnostics ESwab. The swab was wet in nuclease-  
405 free water, and the sample was taken from approximately a 1 in. x 1 in. area of the selected site  
406 on the left-hand side of the participant's body. We swabbed the area approximately 15 times in a  
407 downward motion with constant pressure while rotating the swab. We then placed the swab into  
408 the Copan Diagnostics ESwab collection tube, which contained 1 ml of liquid Amies media. We  
409 labeled the tubes with subject ID, body site, and date of collection and stored at 4°C for up to 24  
410 hours until processing.

411

### 412 **Strain isolation and storage**

413 Within 24 hours of collection, we added 100 µl of the Amies media from the culture sample  
414 collection tube to 900 µl of sterile water to create a 1:10 dilution. We then transferred 100 µl of  
415 the diluted sample to each of 3 agar plates: Brain Heart Infusion (BHI) + 50 mg/L mupirocin,  
416 BHI + 0.1% Tween 80 + 50 mg/L mupirocin, and Trypticase Soy Agar with 5% Sheep's blood

417 (blood agar) + 50 mg/L mupirocin. Blood agar plates were purchased premade, and we spread  
418 750  $\mu$ l of 1 mg/ml mupirocin to the top of the agar and allowed to soak into the media. We  
419 distributed the diluted sample evenly across the plate using sterile glass beads. We incubated the  
420 inoculated plates at 28°C for 48 to 72 hours, depending on colony formation. After incubation,  
421 we chose distinct colonies based on color, size, morphology, and opacity and struck onto a new  
422 BHI + 0.1% Tween 80 plate. We grew these strains for 24 hours or until fully grown. If the  
423 isolate was not a pure culture, we replated the isolate until it was a pure culture. Once a pure  
424 culture was obtained, we inoculated an overnight liquid culture in 3 ml BHI + 0.1% Tween 80.  
425 We stored our strains long-term at -80°C in a 2.0 mL cryotube by combining 900  $\mu$ L overnight  
426 liquid culture with 900  $\mu$ L 30% glycerol.

427

### 428 **Strain library**

429 Of the strains isolated in our study, 451 underwent 27F 16S rRNA gene Sanger sequencing  
430 (Functional Biosciences, Madison, WI; see below). We classified sequences to genus-level using  
431 either RDP<sup>74</sup> or NCBI Blast. We identified 279 isolates with whole genomes to the species level  
432 by running their genomes through autoMLST<sup>75</sup>.

433

### 434 **Colony PCR**

435 We performed Colony PCR on the overnight liquid culture of 451 isolates in a 25 $\mu$ L reaction  
436 containing 12.5 $\mu$ L EconoTaq® PLUS 2X PCR Master Mix by Lucigen, 1 $\mu$ L of 10 $\mu$ M 27F 16S  
437 rRNA primer, 1 $\mu$ L of 10 $\mu$ L 1492R 16s rRNA primer, 10 $\mu$ L nuclease-free water, and 0.5 $\mu$ L of the  
438 overnight liquid culture. We amplified the 16S rRNA gene using the following settings: initial  
439 denaturation at 95°C for 10 minutes, followed by 30-40 cycles of 95°C for 30 seconds, annealing  
440 at 54°C for 30 seconds, and extension at 72°C for 60 seconds, with a final extension at 72°C for 5  
441 minutes and a hold at 4°C indefinitely. We confirmed amplification of the 16s rRNA gene by gel  
442 electrophoresis.

443

### 444 **16S Sanger sequencing**

445 We cleaned the PCR product using the Sigma-Aldrich GenElute PCR Clean-Up kit, following kit  
446 directions. We submitted clean PCR product to Functional Biosciences in Madison, WI for  
447 Sanger sequencing of the 27F end<sup>76</sup>. Quality trimmed FASTA files were inputted into RDP  
448 Classifier for genus-level identification<sup>74</sup>.

449

### 450 **Microbiome DNA extractions**

451 We performed microbiome DNA extractions on a set of swabs collected into 300 $\mu$ L Lucigen  
452 Master-Pure Yeast Cell Lysis buffer and frozen at -80°C. Samples were thawed on ice prior to  
453 extraction. We incorporated the extraction methods and data from Swaney and Kalan (2022)<sup>77</sup>.  
454 We captured the community composition across body sites and microenvironments through  
455 metagenomics profiling.

456

### 457 **Bioassay**

458 Fresh bacterial isolates were struck from isolation plates or from freezer stock. We tested the  
459 bioassays in 12-well plates containing 3mL BHI solid agar in each well. A single colony from  
460 each skin isolate was streaked in a half-moon shape onto the left half each well of two 12-well  
461 plates. We incubated the plate for 7 days at 28°C. After 7 days, we spotted 3  $\mu$ L of a 1:10  
462 dilution of overnight liquid cultures from 22 different pathogens onto the right side of the well.

463 Two 12-well control plates with media alone were included as a positive control for pathogen  
464 growth. Plates were again incubated for 7 days before scoring inhibition of growth on a scale  
465 from 0 to 3. The scoring scale was as follows; 0 – No inhibition, 1 – Slight inhibition, or more  
466 transparent than control, 2 – Medium inhibition, or zone of inhibition, 3 – Full inhibition, no  
467 pathogen growth. Inhibition scores were further simplified, grouping slight and medium  
468 inhibition together. The simplified scores are as follows: 0 – No inhibition, 1 – Slight inhibition,  
469 2 – Full inhibition. Photos were taken of each well and uploaded along with scores to a database.  
470

471 To increase throughput, we adapted the bioassay method to using 24-well plates in the spring of  
472 2020. Liquid overnight cultures of skin isolates are grown in 3 mL BHI + 0.1% Tween 80. The  
473 bioassays were done in 24-well plates containing 1.5 mL 0.5X BHI + 0.1% Tween solid agar in  
474 each well. We spotted 1.5µL of the liquid overnight skin isolate culture onto the left half of the  
475 well. Plates were incubated at 28°C for 5 to 7 days. Slow-growing isolates were inoculated on  
476 day 0, while fast-growing isolates were introduced on day 2. On day 7, we inoculated 1µL of  
477 1:10 diluted overnight pathogen cultures onto the right side of each well. Pathogens include  
478 Gram-positive bacteria (*Bacillus cereus*, *Bacillus subtilis*, *Enterococcus faecalis*, *Micrococcus*  
479 *luteus*, *Mycobacterium smegmatis*, *Staphylococcus aureus*, *Staphylococcus epidermidis*), Gram-  
480 negative bacteria (*Acinetobacter baumannii*, *Citrobacter freundii*, *Enterobacter cloacae*,  
481 *Escherichia coli*, *Klebsiella oxytoca*, *Proteus vulgaris*, *Pseudomonas aeruginosa* PAO1,  
482 *Pseudomonas aeruginosa* 27873, *Serratia marcescens* 8055), and fungi (*Aspergillus flavus*,  
483 *Candida albicans* K1, *Candida* sp., *Cryptococcus neoformans*, *Trichosporon asahii*, and  
484 *Candida auris* B11211). Plates were incubated at 28°C for 3 days. After 3 days of incubation,  
485 skin isolates were scored as before.  
486

#### 487 **gDNA extractions for whole genome sequencing**

488 We extracted bacterial gDNA from plated isolates using the Sigma-Aldrich GenElute Bacterial  
489 Genomic DNA Kit. Library preparation and whole-genome sequencing on Illumina NextSeq 550  
490 were performed at the SeqCenter sequencing facility (Pittsburgh, PA, USA).  
491

#### 492 **Genome assembly, representative selection through dereplication analysis, and phylogeny** 493 **construction**

494 We processed sequencing data for quality and adapters using fastp (v0.20.0)<sup>78</sup> with parameters “-  
495 -detect\_adapter\_for\_pe -f 20”. Subsequently, short-read assemblies were constructed using  
496 Unicycler (v0.4.7)<sup>79</sup> with default settings. Dereplication of the full set of 287 genomic assemblies  
497 at 99% identity using dRep (v3.2.2)<sup>80</sup> with parameters “--S\_algorithm fastANI -sa 0.99” led to  
498 the selection of 182 distinct representative genomes (**Table S2**). Genomes were taxonomically  
499 classified using GTDB-Tk (v1.7.0)<sup>81</sup> with GTDB release 202<sup>82</sup>. GTDB-Tk was unable to assign  
500 species designations for 11 genomes. To further assess whether these genomes corresponded to  
501 novel species, we computed their average nucleotide identity comparing to genomes in the  
502 recently established SMGC database<sup>31</sup> using FastANI<sup>35</sup> (**Table S3**). Six of the genomes featured  
503 >95% average nucleotide identity (ANI) and >20% coverage to genomes from the SMGC and  
504 were thus regarded as known species. Genomic dereplication further indicated that two of the  
505 five remaining genomes represented the same novel species belonging to the genus of *Kocuria*.  
506 A phylogeny of the EPIC was created based on universal ribosomal proteins with GToTree  
507 (v1.6.36)<sup>83</sup> and pared down to representative isolates from dereplication using PareTree<sup>84</sup>.  
508 Visualization of the phylogeny was performed using iTol<sup>85</sup>.

509

## 510 **Annotation of biosynthetic gene clusters, determination of gene cluster families and** 511 **subsequent profiling in metagenomes**

512

513 We used antiSMASH (v6.0.0) to predict the presence of BGCs for our dereplicated set of isolate  
514 genomes as well as metagenomic assembled genomes (MAGs) from skin microbiomes by  
515 Kashaf *et al.* 2022 using the parameters: “--taxon bacteria --genefinding-tool prodigal --  
516 fullhmmer --asf --cb-general --cb-subclusers --cb-knownclusers --cc-mibig --rre --pfam2go”.  
517 Following BGC annotation, we grouped analogous BGCs across genomes into GCFs using BiG-  
518 SCAPE(v1.1.4)<sup>43</sup> with requests for inclusion of singletons, hybrid classifications to be turned off,  
519 and a mixed analysis to be performed. A network graph with GCFs as nodes was constructed  
520 through parsing BiG-SCAPE mix clustering results and linking nodes based on shared  
521 annotations for GCFs stemming from antiSMASH type classifications of member BGCs.  
522 Visualization of the network was performed using igraph with the layout layout\_nicely.

523

524 BiG-MAP (version committed on March 22, 2023)<sup>48</sup> was used to profile the presence of BGCs  
525 from the analysis in 268 metagenomes from Swaney *et al.* 2022. We ran BiG-MAP with default  
526 parameters except for BiG-MAP.family.py, where we requested a distance cutoff of 0.3 for BiG-  
527 SCAPE based grouping of GCFs instead of 0.2. GCFs were regarded as present in a single  
528 metagenome if their full coverage was  $\geq 50\%$ , full normalized RPKM was  $\geq 0.1$ , core gene  
529 coverage was  $\geq 75\%$  and core gene normalized RPKM was  $\geq 3.0$ . These parameters were  
530 selected based on manual examination of observed distributions (**Figure S4**). Using these  
531 parameters, 158 different GCFs were identified by BiG-MAP in at least one metagenome. These  
532 corresponded to 148 distinct GCFs in the original BiG-SCAPE analysis, with ten BiG-SCAPE  
533 GCFs having two representatives regarded as separate GCFs by BiG-MAP.family.py.  
534 Discordance is likely because BiG-MAP performs a preliminary collapsing of similar BGCs  
535 using MASH and runs BiG-SCAPE with only representative BGCs from this initial clustering.  
536 For rarefaction analysis of GCF discovery as a function of metagenomic sequencing depth, we  
537 referenced coarser BiG-SCAPE GCF designations of representative gene clusters measured by  
538 BiG-MAP to avoid inflated metrics of GCF discovery (**Figure 4, S5**).

539

## 540 **Antibiotic resistance within the skin microbiome**

541

542 To predict the collection of antibiotic resistance (AMR) genes, the antibiotic resistome, of the  
543 human skin microbiome, we used the Comprehensive Antibiotic Resistance Database's (CARD)  
544 Resistance Gene Identifier (RGI) software (v6.0.1)<sup>46</sup>. Briefly, we surveyed both isolate genomes  
545 and metagenomes using RGI to assess the presence of known AMR resistance genes in the  
546 CARD database. Filtered metagenomic short read sequences were compared to the reference  
547 AMR sequences in CARD via the k-mer alignment (KMA) algorithm. Metagenomes were  
548 considered to have the AMR gene present if metagenomic reads covered greater than or equal to  
549 70% of the reference CARD AMR sequence and MAPq score greater than or equal to 50. AMR  
550 genes were grouped by the class of drugs they conferred resistance to and if a gene confers  
551 resistance to multiple drugs it counted toward both. Prevalence for antimicrobial resistance  
552 across each body site was calculated (number of subject samples from the body site with a gene  
553 confirming resistance to an antibiotic divided by the total number of subject samples from the  
554 site [n = 34]).

555  
556 To assess whole bacterial isolate genomes, we searched for “Perfect” and “Strict” matches of  
557 AMR genes in the CARD. We further required the percentage length of the reference sequence  
558 and identity of matches to be  $\geq 90\%$ . We grouped AMR genes by the class of drugs they  
559 conferred resistance to and if a gene confers resistance to multiple drugs it counted toward both.  
560 The prevalence of antimicrobial resistance within the bacterial isolates from each species was  
561 calculated by the number of isolates, with at least one gene confirming resistance to an antibiotic,  
562 divided by the total number of species isolates assessed.

### 563 564 **Assessment of novel species distributions across public metagenomes and comparison of** 565 **their biosynthetic content to BGCs from known species in their respective genera**

566  
567 For each of the four novel species identified, the Branchwater webserver  
568 (<https://branchwater.sourmash.bio/>; accessed October)<sup>52</sup> was used to identify public  
569 metagenomes from NCBI’s SRA database which feature them. To regard a species as present in  
570 a metagenome, we required a cANI value of at least 95% to our query genome, a threshold  
571 commonly used to delineate species<sup>86</sup>, and a containment value of at least 0.5. The abon  
572 program, within the zol suite (v1.3.11)<sup>54</sup>, was used with default settings to assess the  
573 conservation of BGCs from each novel species across genomes for other species belonging to the  
574 same genus. Comprehensive databases were independently set up for each genus by gathering  
575 genomes belonging to them from GTDB R214<sup>87</sup>.

### 576 577 **Data availability**

578 Supplemental Material S1 to S9 can be found on GitHub ([https://github.com/Kalan-](https://github.com/Kalan-Lab/SkinBioassayStudy)  
579 [Lab/SkinBioassayStudy](https://github.com/Kalan-Lab/SkinBioassayStudy)) as SuppTable.xlsx. Whole genome assemblies are publicly available  
580 from NCBI under BioProject PRJNA803478. Metagenomic sequencing data are publicly  
581 available in the Sequence Read Archive (SRA) under BioProject PRJNA763232.

### 582 583 **Code availability**

584 Code used for analyses and figures is available on GitHub ([https://github.com/Kalan-](https://github.com/Kalan-Lab/SkinBioassayStudy)  
585 [Lab/SkinBioassayStudy](https://github.com/Kalan-Lab/SkinBioassayStudy)).

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595  
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### 597 598 599 **Figure Legends**

600

601 **Figure 1: Comprehensive workflow for skin microbiome sampling and genomic analysis.**  
602 Skin samples were collected from each participant from eight different body sites for  
603 metagenomic sequencing. A second set of samples was collected for strain isolation.  
604 Metagenomic sequencing was performed with taxonomic assignment using Kraken2 and  
605 Bracken. For strain isolation, samples were plated on selective media to isolate pure bacterial  
606 colonies and identified by 16S rRNA gene sequencing or whole genome sequencing. Skin  
607 isolates were assessed for antimicrobial activity using a large-scale solid-phase biological assay.  
608 Biosynthetic gene clusters and gene cluster families were annotated with antiSMASH/BiG-  
609 SCAPE and BiG-MAP for genomes and metagenomes, respectively. Antibiotic resistance genes  
610 were called using the Resistance Gene Identifier (RGI) and the Comprehensive Antibiotic  
611 Resistance Gene (CARD) database.

612  
613 **Figure 2: Strain library identification at each body site.** (A) Individual pie charts represent  
614 the community structure at each site. Pie labels indicate body site, font colors indicate body site  
615 type (sebaceous, moist, rarely moist), and the n-value of each pie chart is shown in parentheses.  
616 Bacterial composition is shown within the pie charts. (B) Relative abundance of taxa across eight  
617 body sites comparing cultured versus metagenomic datasets. (C) Scatter plot depicts relative  
618 abundance of individual genera as derived from metagenomic (y-axis) and cultured datasets (x-  
619 axis). Each dot represents a specific genus, with its position reflecting the relative abundance  
620 across the two dataset types. Genera are color-coded according to their respective phyla.

621  
622 **Figure 3: Bacteria from the human skin exhibit specific and broad antifungal activity.** (A)  
623 Examples for each inhibition profiles: complete, partial, and none. (B) Quantification of  
624 inhibition (none [light pink], partial, and complete [dark pink]) profiles. Y-axis shows the  
625 number of isolates and x-axis lists members of the pathogen panel. (C) Hierarchical clustering of  
626 pairwise interactions where skin bacteria are tested against a panel of pathogens. Growth patterns  
627 of pathogens are scored from 0 (no inhibition) to 2 (complete or full inhibition). Rows represent  
628 individual isolates annotated by body site and genus. Columns represent a single pathogen  
629 grouped by Gram-positive, Gram-negative, and fungal species. (D) Inhibition score summary for  
630 each represented genus against fungal (top), Gram-negative (middle), and Gram-positive  
631 (bottom) pathogens. Dot size indicates number of isolates. Color indicates phylum of isolates.

632  
633 **Figure 4: Charting biosynthetic potential of the EPIC library.** (A) Relative abundance of  
634 different biosynthetic gene cluster (BGC) types found across eight body sites. Y-axis depicts  
635 relative abundance, while x-axis indicates distinct body sites. Each color within the bars  
636 corresponds to a unique BGC type. (B) A phylogeny of 182 de-replicated isolate genomes with  
637 the inner track corresponding to the genus and the outer barplot depicting the number of BGCs  
638 identified in the genome by antiSMASH. (C) A network of 305 GCF (nodes) identified in the  
639 genomes using BiG-SCAPE. Edges indicate common antiSMASH-based BGC types shared  
640 between GCFs. The size of the nodes corresponds to the number of genomes a GCF was found in  
641 and the color composition indicates proportion of BGCs belonging to a GCF originating from  
642 EPIC, SMGC, or MIBiG. (D) Cumulative GCF discovery by BiG-MAP is plotted as a function  
643 of aggregate sequencing depth at each body site across multiple individuals using skin  
644 metagenomes from Swaney et al. 2022. Metagenomes are ordered from lowest to highest  
645 sequencing depth.

646

647 **Figure 5: Identification of novel skin species.** (A) Branchwater analysis of NCBI SRA  
648 metagenomes with query genomes of novel species show associations with human skin  
649 metagenomes. Colors indicate metagenome type each genome was found in. (B) Biological  
650 activity of *Aestuariimicrobium* LK1188 (first), *Brevibacterium* LK1337 (second),  
651 *Corynebacterium* LK952 (third), and *Kocuria* LK960 (fourth) using co-culture screening. Solid  
652 line indicates inhibition scores of against Gram-positive (red), Gram-negative (blue) and fungal  
653 pathogens (green). Dotted line indicates the average inhibition score of all isolates tested in each  
654 corresponding genus against each pathogen type. (C) Pairwise interaction between (in columns)  
655 LK952 (first), LK960 (second), LK13337 (third), and pathogen control (fourth) against fungal  
656 pathogens (in rows) *C. albicans* (first), *C. auris* (second), *T. asahii* (third), and *A. flavus* (fourth).  
657 (D-F) Investigating novelty of BGC-ome of (D) *Corynebacterium* LK952, (E) *Kocuria* LK960,  
658 and (F) *Brevibacterium* LK1337 to representative genomes from each genus. Colors indicate  
659 BGC type predicted in each species. X-axis indicates the average amino acid identity of a single  
660 BGC identified compared to the target genome. Y-axis indicates proportion of genes in identified  
661 BGC co-located in the target genome.

662  
663 **Figure S1: Fungal inhibition by skin isolates.** The top panel illustrates a stacked bar plot,  
664 where each color corresponds to a specific bacterial genus, showing the distribution of strain  
665 counts (y-axis) within each genus. The bottom matrix focuses on the intersections among  
666 different fungal pathogen types. Rows are labelled to represent distinct fungal pathogens, and  
667 each column signifies the overlapping occurrence of these pathogens across sampled sets. Cells  
668 within the matrix are filled to illustrate the presence of a fungal pathogen type in the intersecting  
669 sets, with filled cells in the same column connected by a horizontal line. To the left of this  
670 matrix, bar charts corresponding to the row labels indicate the total number of instances for each  
671 fungal pathogen type.

672  
673 **Figure S2: Low prevalence of antibiotic resistance genes in the EPIC library.** A) Prediction  
674 of antibiotic resistance genes in the skin metagenomes. Rows indicate antibiotic classes.  
675 Columns indicate body sites. Colors indicate prevalence of antibiotic resistance across each body  
676 site. B) Profiling of antibiotic resistance genes in 287 whole genomes from cultured skin  
677 isolates.

678  
679 **Figure S3: Representative BGC coverage and normalized RPKM metrics reported by BiG-**  
680 **MAP.** Distribution of the normalized RPKM metric is shown across all metagenomes from  
681 Swaney et al. 2022 for full (A) and core (B) regions in representative BGCs. Similarly,  
682 distributions of the proportion of sites across full (C) and core (D) regions of GCFs is also  
683 shown. Only non-zero datapoints are shown.

684  
685 **Figure S4: The distribution of GCFs across metagenomes.** The normalized RPKM of core  
686 regions for GCFs (rows) partitioned by their BGC annotation class (row groups) is shown across  
687 metagenomes (columns) divided according to body site of sampling (column groups). Grey  
688 indicates the GCF was not detected for a particular metagenome at the required cutoffs. Only  
689 GCFs found in five or more metagenomes are shown.

690



691 **Figure S5: Visualization of BGCs in novel, skin-associated species, including (A)**  
692 **Corynebacterium LK952, (B) Kocuria LK960, (C) Brevibacterium LK1337, and (D)**  
693 **Aestuariimicrobium LK1188. Colors correspond to the BGC-likeness of each gene.**

694  
695

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697

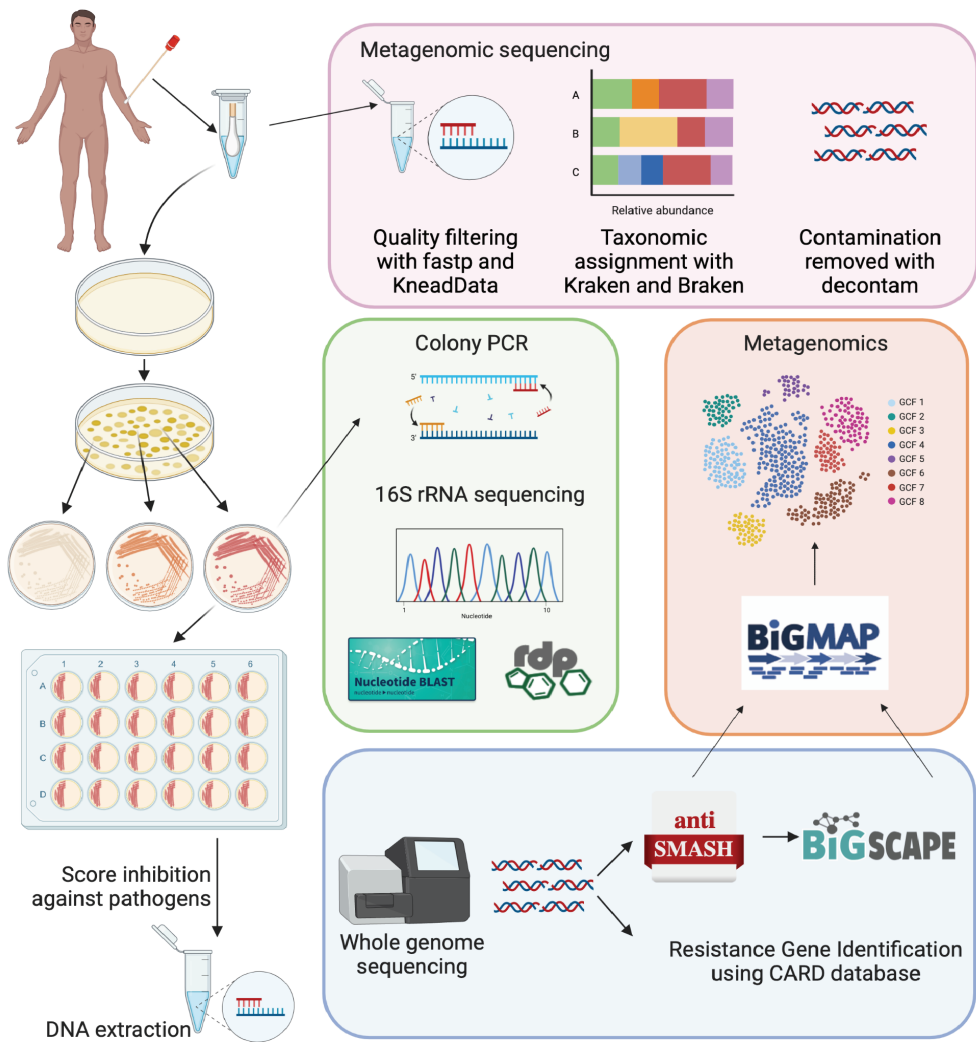
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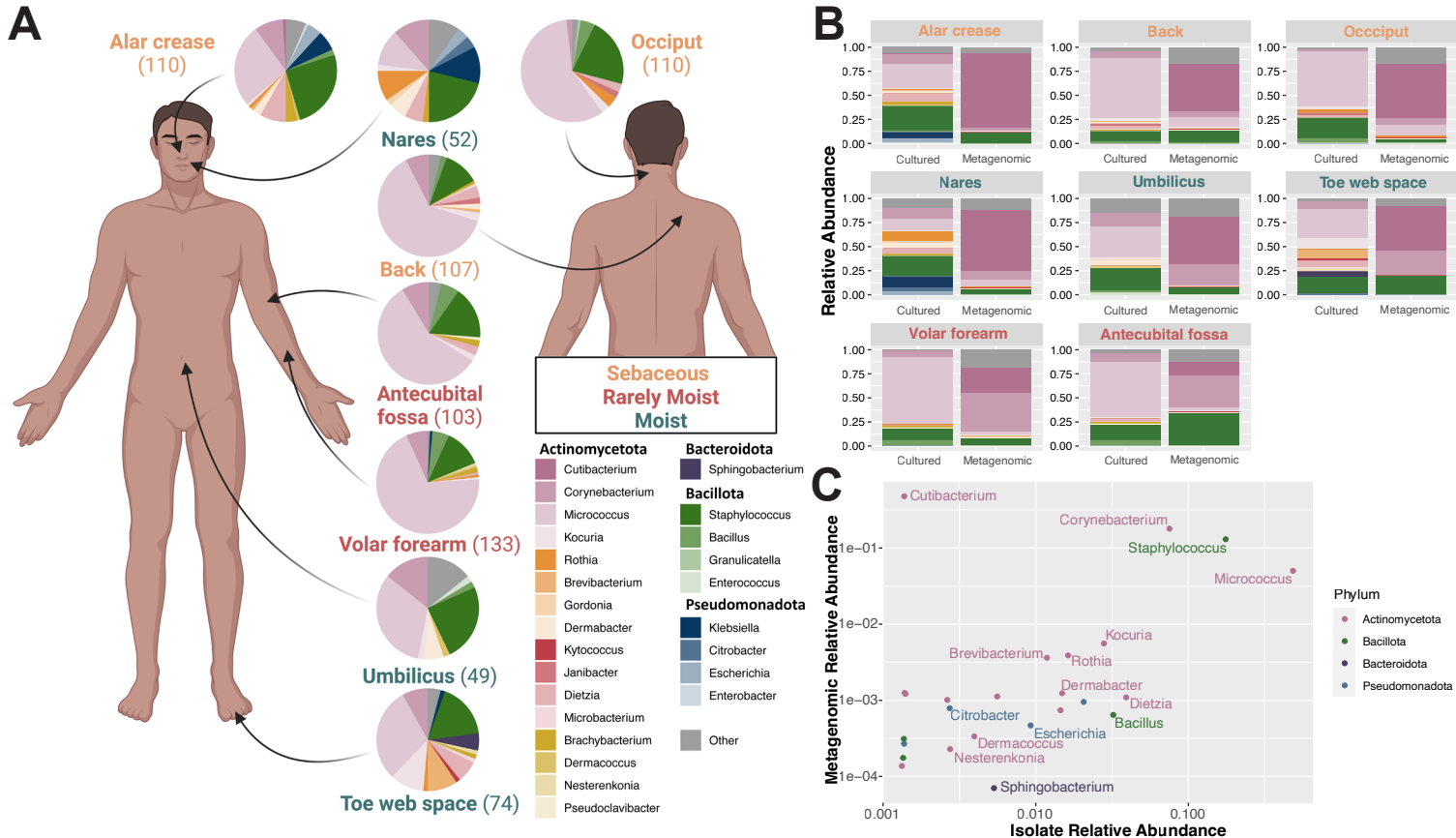
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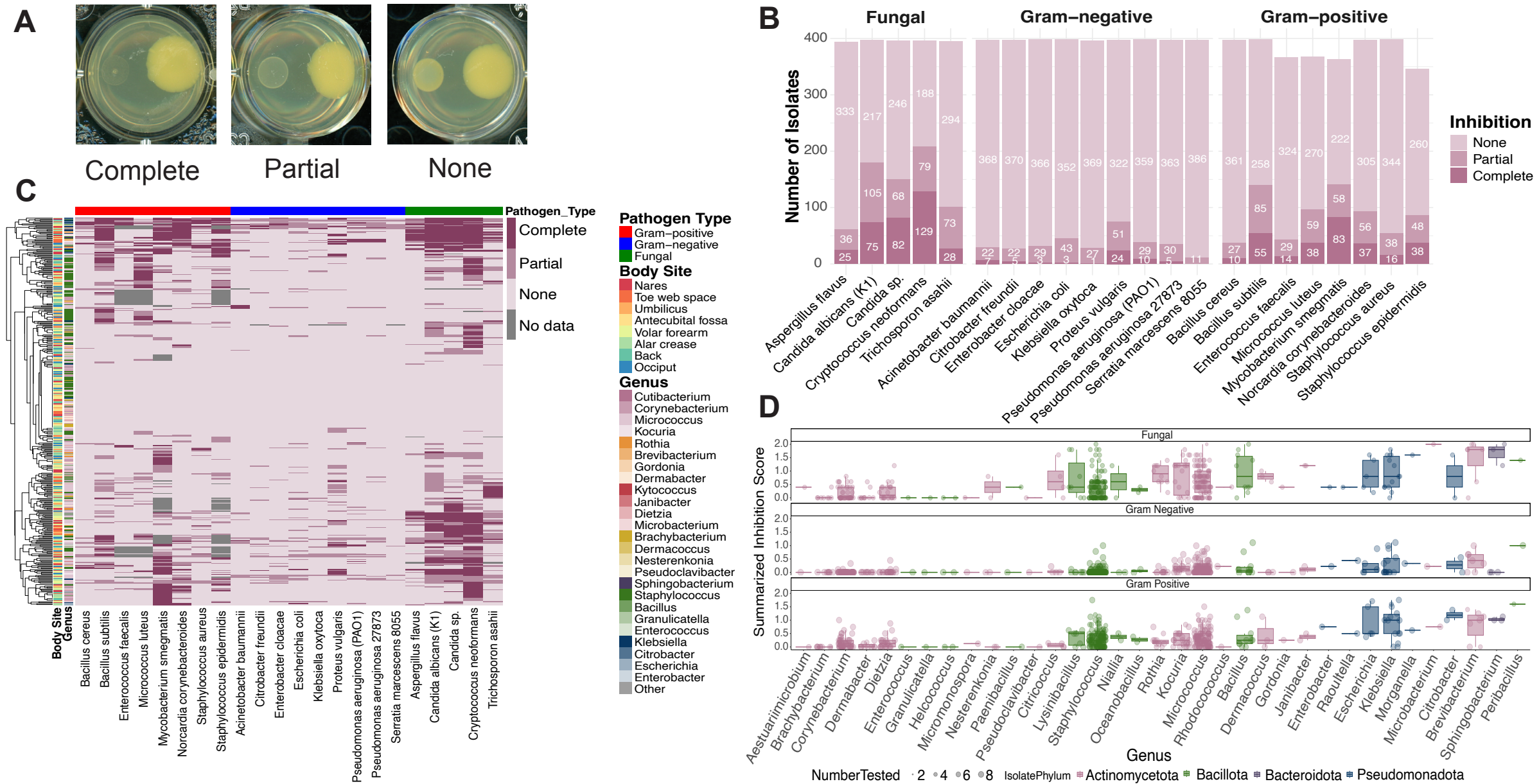
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**Figure 1: Comprehensive workflow for skin microbiome sampling and genomic analysis.** Skin samples were collected from each participant from eight different body sites for metagenomic sequencing. A second set of samples was collected for strain isolation. Metagenomic sequencing was performed with taxonomic assignment using Kraken2 and Bracken. For strain isolation, samples were plated on selective media to isolate pure bacterial colonies and identified by 16S rRNA gene sequencing or whole genome sequencing. Skin isolates were assessed for antimicrobial activity using a large-scale solid-phase biological assay. Biosynthetic gene clusters and gene cluster families were annotated with antiSMASH/BiG-SCAPE and BiG-MAP for genomes and metagenomes, respectively. Antibiotic resistance genes were called using the Resistance Gene Identifier (RGI) and the Comprehensive Antibiotic Resistance Gene (CARD) database.

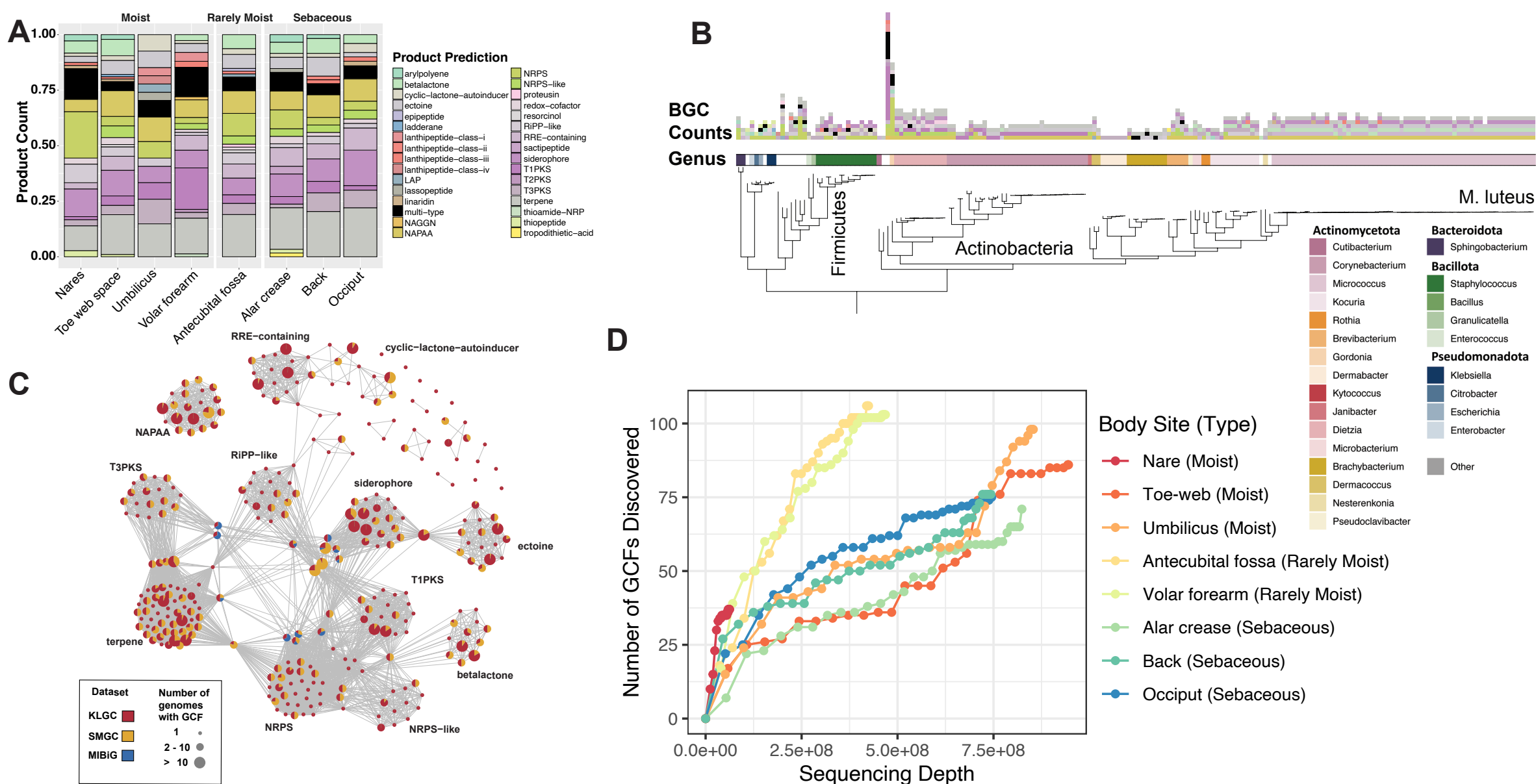


**Figure 2: Strain library identification at each body site. (A)** Individual pie charts represent the community structure at each site. Pie labels indicate body site, font colors indicate body site type (sebaceous, moist, rarely moist), and the n-value of each pie chart is shown in parentheses. Bacterial composition is shown within the pie charts. **(B)** Relative abundance of taxa across eight body sites comparing cultured versus metagenomic datasets. **(C)** Scatter plot depicts relative abundance of individual genera as derived from metagenomic (y-axis) and cultured datasets (x-axis). Each dot represents a specific genus, with its position reflecting the relative abundance across the two dataset types. Genera are color-coded according to their respective phyla.

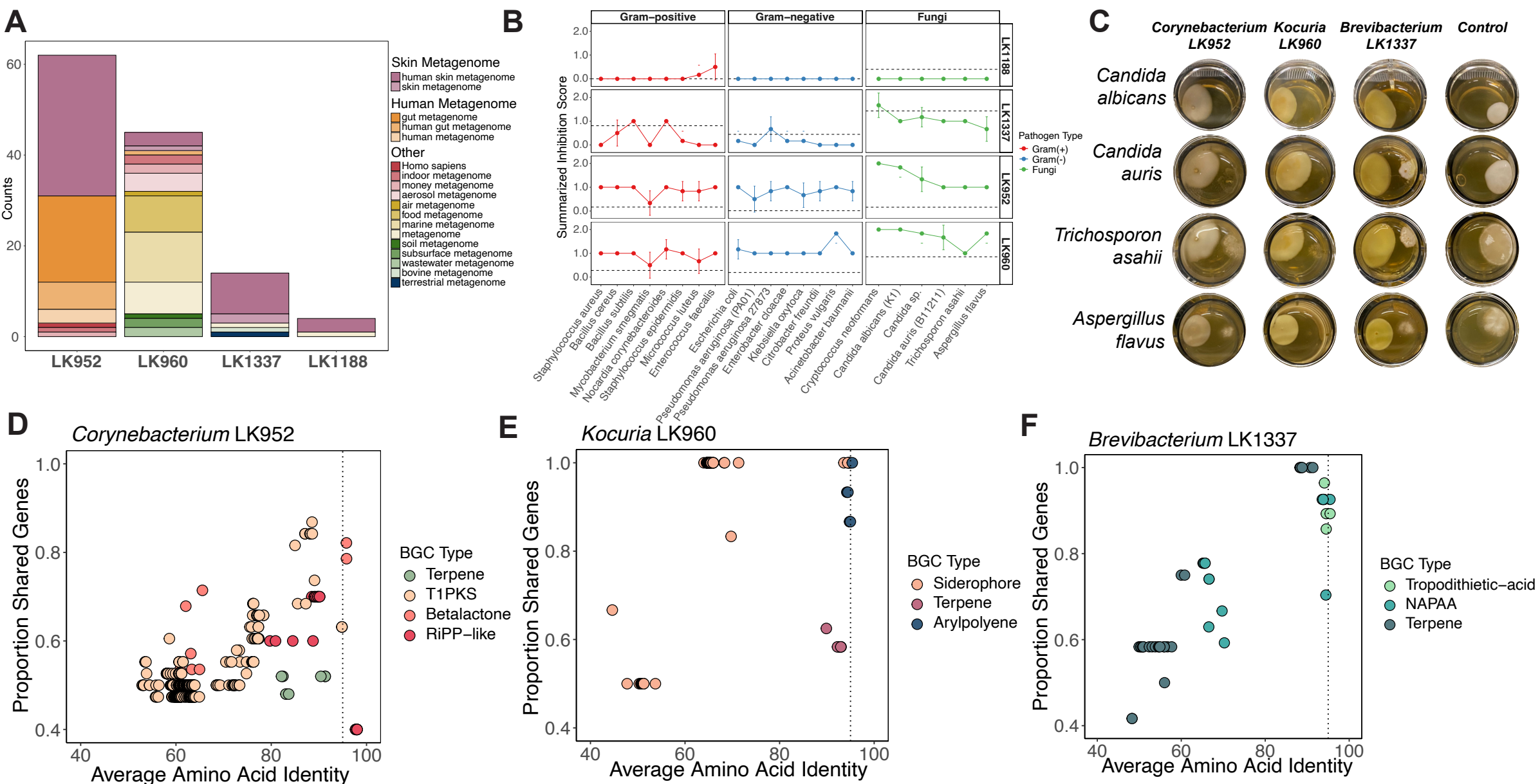


**Figure 3: Bacteria from the human skin exhibit specific and broad antifungal activity.** (A) Examples for each inhibition profiles: complete, partial, and none. (B) Quantification of inhibition (none [light pink], partial, and complete [dark pink]) profiles. Y-axis shows the number of isolates and x-axis lists members of the pathogen panel. (C) Hierarchical clustering of pairwise interactions where skin bacteria are tested against a panel of pathogens. Growth patterns of pathogens are scored from 0 (no inhibition) to 2 (complete or full inhibition). Rows represent individual isolates annotated by body site and genus. Columns represent a single pathogen grouped by Gram-positive, Gram-negative, and fungal species. (D) Inhibition score summary for each represented genus against fungal (top), Gram-negative (middle), and Gram-positive (bottom) pathogens. Dot size indicates number of isolates. Color indicates phylum of isolates.





**Figure 4: Charting biosynthetic potential of the EPIC library.** (A) Relative abundance of different biosynthetic gene cluster (BGC) types found across eight body sites. Y-axis depicts relative abundance, while x-axis indicates distinct body sites. Each color within the bars corresponds to a unique BGC type. (B) A phylogeny of 182 deprecated isolate genomes with the inner track corresponding to the genus and the outer barplot depicting the number of BGCs identified in the genome by antiSMASH. (C) A network of 305 GCF (nodes) identified in the genomes using BiG-SCAPE. Edges indicate common antiSMASH-based BGC types shared between GCFs. The size of the nodes corresponds to the number of genomes a GCF was found in and the color composition indicates proportion of BGCs belonging to a GCF originating from EPIC, SMGC, or MIBiG. (D) Cumulative GCF discovery by BiG-MAP is plotted as a function of aggregate sequencing depth at each body site across multiple individuals using skin metagenomes from Swaney et al. 2022. Metagenomes are ordered from lowest to highest sequencing depth.



**Figure 5: Identification of novel skin species.** (A) Branchwater analysis of NCBI SRA metagenomes with query genomes of novel species show associations with human skin metagenomes. Colors indicate metagenome type each genome was found in. (B) Biological activity of *Aestuariimicrobium* LK1188 (first), *Brevibacterium* LK1337 (second), *Corynebacterium* LK952 (third), and *Kocuria* LK960 (fourth) using co-culture screening. Solid line indicates inhibition scores of against Gram-positive (red), Gram-negative (blue) and fungal pathogens (green). Dotted line indicates the average inhibition score of all isolates tested in each corresponding genus against each pathogen type. (C) Pairwise interaction between (in columns) LK952 (first), LK960 (second), LK13337 (third), and pathogen control (fourth) against fungal pathogens (in rows) *C. albicans* (first), *C. auris* (second), *T. asahii* (third), and *A. flavus* (fourth). (D-F) Investigating novelty of BGC-ome of (D) *Corynebacterium* LK952, (E) *Kocuria* LK960, and (F) *Brevibacterium* LK1337 to representative genomes from each genus. Colors indicate BGC type predicted in each species. X-axis indicates the average amino acid identity of a single BGC identified compared to the target genome. Y-axis indicates proportion of genes in identified BGC co-located in the target genome.