Main manuscript for:

Bacterial DNA on the skin surface overrepresents the viable skin microbiome

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Abstract

The skin microbiome provides vital contributions to human health. However, the spatial

organization and viability of its bacterial components remain unclear. Here we apply culturing,

imaging, and molecular approaches to human and mouse skin samples, and find that the skin

surface is colonized by fewer viable bacteria than predicted by bacterial DNA levels. Instead,

viable skin-associated bacteria are predominantly located in hair follicles and other cutaneous

invaginations. Furthermore, we show that the skin microbiome has a uniquely low fraction of

viable bacteria as compared to other human microbiome sites, indicating that most bacterial

DNA on the skin surface is not associated with viable cells Additionally, a small number of

bacterial families dominate each skin site and traditional sequencing methods overestimate both

the richness and diversity of the skin microbiome. Finally, we performed an in-vivo skin

microbiome perturbation-recovery study using human volunteers. Bacterial 16S rRNA gene

sequencing revealed that, while the skin microbiome is remarkably stable even in the wake of

aggressive perturbation, repopulation of the skin surface is driven by the underlying viable

population. Our findings help explain the dynamics of skin microbiome perturbation, as bacterial

DNA on the skin surface can be transiently perturbed but is replenished by a stable underlying

viable population. These results address multiple outstanding questions in skin microbiome

biology with significant implications for future efforts to study and manipulate it.

Significance statement

This study provides a crucial update to the skin microbiome paradigm by showing that viable bacteria of the skin microbiome are primarily localized to hair follicles and other sub-cutaneous structures rather than the skin surface. The native distribution of skin-associated bacteria has not been previously evaluated, and here we show that the skin surface has few intact bacteria while deeper structures are replete with bacterial contents. We used orthogonal approaches to evaluate the source of bacterial DNA and the dynamics of bacterial repopulation on the skin surface to develop an updated model of the skin microbiome. By adjusting the current understanding of the skin microbiome to match this model, we will be able to address outstanding questions in the field.

Main text

Introduction

The skin is the largest organ in the human body, providing roughly 25 square meters for

potential host-microbe interactions (*1*). It facilitates our tactile interactions with the world,

separates us from the dangers of our daily lives, and has the incredible ability to regenerate itself

every 20-30 days (*2*). Human skin is also home to the organisms that comprise the skin

microbiome, which has been shown to have important roles in human health. For example, the

human skin microbiome affects immune system education (*3*, *4*), wound healing (*5–7*),

colonization resistance (*8*), modulation of gene expression in the skin (*9*), and may have a role in

development (*10*). Despite the many contributions of the skin microbiome to human health, there

are important questions that have not been addressed by traditional methods: Why is the skin

microbiome stable across months in longitudinal studies yet easily perturbed upon transient

environmental changes like swimming (*11*)? Why are there so many anaerobic bacteria on an

organ exposed to the air? And why is it so difficult to stably colonize the skin with new microbes

without strong perturbations like abrasion? Traditional methods of sampling, culturing, and

sequencing the bacterial members of the skin microbiome have been indispensable for

establishing the skin microbiome field. Here, we extend these foundational studies with

additional approaches to differentiate between viable and nonviable bacteria in order to begin to

address these important questions.

Results

Fluorescence in-situ hybridization reveals few bacteria on the skin surface

The predominant method by which skin microbiomes have been studied is through

sequencing of DNA from swabbed skin areas (*12*). However, despite the fact that the skin

microbiome is often depicted as a well-mixed coating of microbes on the skin surface (*8, 13*), it

remains unclear whether the DNA from the skin surface reflects the underlying biology of the

skin microbiome. Additionally, while sequencing and culture-based studies have demonstrated

that bacteria extend into deeper portions of the skin, the spatial distribution of the skin

microbiome as a whole has not been well characterized (*4, 14, 15*). To address this gap, we

determined the spatial distribution of bacterial cells in the skin using the universal bacterial FISH

probe EUB338, which hybridizes to bacterial 16S rRNA (*16*).

We first used EUB338 FISH on biopsied healthy adult human facial tissues and found

that the skin surface contains very few bacteria (Figure 1A). In contrast, clusters of bacteria were

found within hair follicles and other cutaneous skin structures like comedos (Figure 1B). There

are several possible explanations for the lack of FISH staining on the skin surface. For example,

viable surface bacteria may have been eliminated by sterilization of the biopsied area

(we address this possibility below). Alternatively, the surface bacteria could exist in a state like

stationary phase that has fewer EUB388-hybridizable ribosomes, or by skin microbiome

bacterial species being less prone to FISH probe hybridization. However, we confirmed that

EUB338 staining works well in stationary-phase cells of multiple skin microbiome bacterial

species (*Cutibacterium acnes, Staphylococcus epidermidis, Micrococcus luteus,* and

Corynebacterium striatum) (Figure 1 – figure supplement 1).

To quantify our FISH staining of human skin, we calculated the ratio of the mean

fluorescence within an area of interest (a hair follicle, the skin surface, or other cutaneous

structures) to the mean fluorescence outside of the area of interest. We refer to this ratio as an

enrichment score. The median enrichment score for human follicles was 11.24 (Figure 1D). In

contrast, the skin surface (stratum corneum) had an enrichment score of just 0.188. This pattern

of observing many bacteria in follicles but few on the skin surface held true when using FISH

probes specific for *C. acnes*, one of the most abundant bacterial species in the human skin

microbiome (Figure 1C). For *C. acnes*, the enrichment score was 3.14 for follicles as compared

to 0.67 for the stratum corneum (Figure 1D). We note that while the majority of intact skin-

associated bacteria were not associated with the skin surface, there are some visible bacteria on

the skin surface, especially in areas near sub-surface structures (Figure 1A-B). This indicates that

FISH staining is capable of detecting bacteria on the skin surface. Thus, our results are consistent

with previous reports that bacteria can be cultured from skin surface swabs but also extend these

findings to demonstrate that the skin surface has fewer intact bacteria than deeper skin structures.

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Figure 1: Bacterial FISH staining of human tissue. (A-C) Scale bar = 20 μm. The bottom left
corner of each diagram shows a schematic of the hair follicle in white and the anatomic location
of each image frame in yellow. DAPI staining is shown in blue in all parts. EUB338
hybridization is shown in red for all images. (A) Human tissues stained with the pan-bacterial
FISH probe EUB338 show little bacterial presence at the skin surface. (B) Human tissues stained
with EUB338 show abundant bacterial signal that is concentrated in hair follicles, pilosebaceous
units, and other cutaneous structures. (C) Human tissues stained with a *C. acnes* specific FISH
probe (in green) demonstrate the same overall spatial organization as those stained with
EUB338. (D) Quantification enrichment scores showing the median and interquartile range.
Significance was calculated using the Mann-Whitney test. *$P \leq 0.05$, **$P \leq 0.01$. N = 8 for
human follicle, N = 6 for human follicle (*C. acnes*), N = 6 for human stratum corneum, N = 5 for
human stratum corneum (*C. acnes*) where "N" represents different follicles or stratum corneum
sections. The human tissue samples shown in Figure 1 were obtained from adult facial tissues
(cheek and forehead).

**16S rRNA FISH with EUB338 on
stationary-phase skin microbiome species**

Figure 1 – figure supplement 1: **Fluorescence in situ hybridization on stationary-phase skin microbiome bacterial species.** All bacterial species were grown in appropriate conditions to stationary phase. The pan-bacterial FISH probe EUB338 was used. Hybridization is indicated in red.

Bacterial DNA on the skin surface overrepresents the number of viable bacteria

DNA sequencing of samples collected by swabbing the skin surface is the most common method used to evaluate the skin microbiome because it is simple, noninvasive, and has been shown to result in higher consistency than other sampling methods ([17]). However, traditional metagenomic and 16S rRNA gene sequencing do not differentiate DNA from viable and inviable bacteria. We thus implemented a method that allowed us to quantitatively determine the fraction of DNA in a sample that is specifically from intact bacteria. The ability to molecularly differentiate between viable and inviable cells would enable us to both validate our FISH results using an orthogonal method on untreated, living skin, and address the possibility that the low numbers of intact surface bacteria observed by FISH was due to biopsy sterilization. To this end, we utilized the cell-impermeable small molecule propidium monoazide (PMA), which binds irreversibly to double-stranded DNA upon photoactivation to inhibit PCR amplification (Figure 2A) ([18]). When PMA photoactivation is performed before the cell lysis step of DNA isolation, the genomic DNA inside viable bacteria is protected from PMA binding because PMA is cell-impermeable, while cell-free DNA or DNA within permeabilized bacteria becomes PMA-bound. To quantify bacterial DNA, we combined the use of PMA with droplet digital PCR (PMA-ddPCR). Calculating the ratio of ddPCR counts between samples without PMA and samples with PMA allowed us to generate a viability score for any given bacterial population. A similar approach has been used to assess the viability of bacteria from environmental waste water samples ([19]).

To assure that PMA-ddPCR would allow us to reliably gauge the fraction of viable cells in a population, we first validated that it generated the expected results using known ratios of either exponentially-growing or stationary-phase *E. coli* cultures combined with heat-killed *E.*

coli cells (Figure 2 – figure supplement 1A). To determine if PMA-ddPCR also works for skin-

resident bacterial species, we also applied PMA-ddPCR to four of the most common skin

microbiome bacteria, *S. epidermidis, C. acnes, M. luteus,* and *C. striatum.* We performed serial

dilutions of each bacterial culture and in parallel performed both CFU plating and ddPCR with or

without PMA. In every case, the amount of DNA in a sample (assessed by PMA-ddPCR)

correlated well with the number of culturable bacteria in high and low bacterial abundance

scenarios (Figure 2 – figure supplement 1B). Together, these controls confirm that PMA-ddPCR

is a reliable method for assessing the amount of DNA in a sample present within intact bacteria.

We next applied PMA-ddPCR to human skin microbiome samples by swabbing the skin

of four healthy human volunteers at eight sites (glabella, retroauricular crease, lower back, hair

shaft, antecubital fossa, popliteal fossa, nares, and dorsal forearm) (Figure 2B, Figure 2 – figure

supplement 2A, B). PMA-ddPCR revealed that the viability scores for these sites ranged between

0.02 and 0.12 (0 represents a fully-nonviable population, 1.0 represents a fully-viable

population), indicating that the majority of bacterial DNA found on the skin surface is not

associated with viable cells (Figure 2C). To investigate whether this was a skin-specific

phenomenon, we tested several non-skin microbiome sites (tongue, saliva, plaque, and feces).

We found that in all non-skin microbiome sites, the viability score was significantly higher than

for the skin, ranging from 0.4 (saliva) to 0.87 (feces) (Figure 2C).

We next sought to address whether our PMA-ddPCR viability scores accurately represent

of the number of viable bacteria on the skin surface. The low number of detectable bacteria on

the skin surface (either by FISH or PMA-ddPCR) does not imply that the skin surface is sterile

but rather that the majority of the bacterial DNA on the skin surface is not from these viable

cells. Indeed, culturing bacteria directly from the skin is common (*8*). To quantify the number of

viable bacteria directly, we plated a small amount of each sample using the standard conditions

for culturing skin microbes (5% sheep blood in tryptic soy agarose plates incubated both

aerobically and anaerobically). To determine whether PMA-ddPCR or traditional ddPCR better

represented the number of viable skin microbiome bacteria, we compared our results to a

standard curve generated with known numbers of *S. epidermidis*. For each sample, the PMA-

ddPCR quantification closely matched this standard curve while the samples lacking PMA

showed no overlap (Figure 2 – figure supplement 1C, D). The highest-abundance skin

microbiome species are readily culturable, such that quantifying the DNA from viable skin

surface bacteria should be able to accurately predict colony forming units (CFU) upon plating.

Quantifying the bacterial DNA in skin microbiome samples without the use of PMA resulted in

DNA quantities that were, on average, 82 times higher than predicted by the standard curve,

while the use of PMA brought this value down to just 1.3 (Figure 2 – figure supplement 1E).

Using ddPCR counts to predict CFU showed similar results, as ddPCR in the absence of PMA

yielded values that predicted CFU counts 58.5 times greater than those measured, while PMA-

ddPCR yielded values that predicted CFU counts that were on average only 1.28 times greater

than the actual cultured CFU (Figure 2 – figure supplement 1F). Since we expect a ratio near 1,

these findings lend further support for our conclusion that PMA-ddPCR reflects the viable

microbiome better than approaches lacking PMA.

Our FISH results from biopsied skin suggested that viable bacteria may be protected

below the skin surface. To test this hypothesis in untreated skin, we collected facial follicle

contents and used PMA-ddPCR to compare the viability of facial follicle contents and skin

surface samples (Figure 2D). The average viability for follicle contents was greater than 10-fold

higher for follicle contents than for the skin surface, confirming that cells in hair follicles are

more viable than those on the skin surface. Together, our results from FISH on biopsies and

PMA-ddPCR on skin swabs independently support the conclusion that the skin surface is

populated by few viable bacterial cells, indicating that the surface of healthy, non-sterilized

human skin is sparsely colonized. While these data indicate that bacterial DNA on the skin

surface is predominantly not associated with viable bacterial cells, we note that our results do not

suggest that there are no viable cells on the skin surface. Rather, our data indicate that the

majority of the bacterial DNA on the skin surface is not within bacteria such that using PMA

provides a much more accurate estimation of the viable skin microbiome.

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Figure 2: PMA-ddPCR and viability scores for human skin and non-skin microbiomes . (A)
Schematic of the PMA-ddPCR workflow. (B) Sampling scheme showing each skin site that was
sampled. Colors indicate site-type (sebaceous in blue, moist in green, dry in red). (C) PMA-
ddPCR on skin and non-skin microbiome sites shows that the viability score of the skin
microbiome is significantly lower than other microbiome sites. ****$P \leq 0.0001$ for Student's T
Test on pooled skin and non-skin samples. Four volunteers contributed skin and non-skin
microbiome samples. Additional samples were collected from some individuals and represent
biological replicates. N= 8 for glabella, N = 6 for retroarticular crease, N = 5 for lower back, hair
shaft, nares, and dorsal forearm, N = 3 for antecubital fossa, tongue, saliva, and plaque, N = 2 for
popliteal fossa, and N = 1 for human feces. Each human skin sample site consists of samples

from four different individuals. Some volunteers were sampled multiple times on different days
(at least two weeks apart). For glabella, one volunteer was sampled 4 times, one volunteer was
sampled 2 times, and two volunteers were sampled 1 time. For retroauricular crease, two
volunteers were sampled 2 times, and two volunteers were sampled 1 time. For lower back, one
volunteer was sampled 2 times and three volunteers were sampled 1 time. For hair shaft, all
samples came from one volunteer. For antecubital fossa, three volunteers were sampled 1 time.
For popliteal fossa two volunteers were sampled 1 time. For nares, one volunteer was sampled 2
times and three volunteers were sampled 1 time. For dorsal forearm, one volunteer was sampled
2 times and three volunteers were sampled 1 time. Tongue, saliva, and plaque all represent 1
sample from three different individuals. For raw ddPCR counts, see Figure 2 – figure supplement
2A, B. (D) PMA-ddPCR on follicle contents and forehead swabs from 5 individuals. Mean
viability score for follicle contents is 0.15 and for forehead is 0.013.
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Figure 2 – figure supplement 1: PMA-ddPCR and sampling controls. (A) PMA-ddPCR validation controls on known ratios of heat-killed and exponentially-growing *E. coli* cultures. PMA-ddPCR performed on a population of exponentially-growing cells resulted in a viability score of 0.903,. PMA-ddPCR performed on a population of 100% heat-killed cells in exponential growth resulted in a viability score of 0. Populations consisting of 50% (by volume) heat-killed and 50% exponentially-growing cells exhibited an average viability score of 0.506. PMA-ddPCR performed on a population of stationary-phase cells resulted in a viability score of 0.965. PMA-ddPCR performed on a population of 100% heat-killed cells in stationary phase resulted in a viability score of 0. Populations consisting of 50% (by volume) heat-killed and 50% stationary-phase cells exhibited an average viability score of 0.621. Mean and standard deviation are shown. (B) Standard curves showing correlation between PMA-ddPCR counts and CFU for four skin microbiome bacterial species when grown to stationary phase. Open data points indicate samples without the use of PMA and closed data points indicate that PMA was used. Horizontal lines connect paired samples. (C) Standard curve showing correlation between PMA-ddPCR counts and CFU for exponentially-growing *Staphylococcus epidermidis*. 95% confidence interval is shown in the green shaded region. (D) Standard curve generated using *Staphylococcus epidermidis* cultures shown in green. Shading represents 95% confidence interval. Open and closed circles represent skin microbiome samples that did (closed circles) or did not (open circles) receive PMA treatment. Dark or light gray shading represents 95% confidence interval for skin microbiome samples. Paired samples are connected to show the downward shift in DNA abundance with the inclusion of PMA. (E) Comparison of the predicted ddPCR counts to measured ddPCR counts based on CFU for samples that were treated with PMA (closed circles) and samples that were not treated with PMA (open circles) (mean for PMA-treated samples is 1.31, mean for untreated samples is 82.2) . (F) Comparison of the predicted CFU to measured CFU based on ddPCR for samples that were treated with PMA (closed circles) and samples that were not treated with PMA (open circles) (mean for PMA treated samples is 1.28, mean for untreated samples is 58.5). (G) The effect of including 0.1% Triton X100 in swabbing buffer used for skin microbiome sampling. (H) Comparison of viability scores obtained with the two methods of DNA isolation used.

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290 **Figure 2 – figure supplement 2: Copies per 20 µL ddPCR reaction without (A) and with (B)**
291 **the use of PMA.** Data in A and B were used to calculate the viability score shown in Figure 2.
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Traditional sequencing methods overestimate skin microbiome richness and diversity

Like many microbiomes, the existing knowledge of the skin microbiome is heavily based upon bacterial 16S rRNA gene amplicon sequencing, which was developed to assess bacterial populations while avoiding biases introduced by culturing methods. However, our findings suggest that using 16S rRNA gene amplicon sequencing to study the skin microbiome is not entirely unbiased, as most of the DNA in these samples is not from viable bacteria and traditional 16S rRNA gene amplicon sequencing does not differentiate between DNA originating from live or dead cells. The inability of 16S rRNA gene amplicon sequencing to differentiate between these two types of bacterial populations has been mentioned as a potential downfall of the method (*8*). To evaluate how accurately traditional 16S rRNA gene amplicon sequencing captures the living skin microbiome composition, we utilized PMA followed by 16S rRNA gene amplicon sequencing (an approach we refer to as PMA-seq) (*20*). By sequencing pairs of matched samples with PMA treatment (PMA-seq) and without PMA treatment (traditional sequencing), we were able to explore how closely the microbiome compositions obtained from traditional sequencing methods resembled the viable microbiome composition obtained by PMA-seq (Figure 3A). These experiments established that at each skin site sampled, as compared to traditional sequencing, the PMA-treated samples were less rich (richness, R, is a measure of the number of identifiable bacterial taxa) and less diverse (diversity, H, is measured by the Shannon diversity index) (Figure 3B, C). Furthermore, samples that had greater richness in traditional sequencing (R_{trad}) showed proportionally larger decreases in richness and Shannon diversity with PMA-seq (R_{PMA} and H_{PMA}) (Figure 3B, C). These results suggest that, although it appears by traditional sequencing that there is a wide range of richness values at different skin sites (1-30 different taxa), in reality the richness across the skin microbiome at different body sites is

relatively similar and low (1-10 different taxa). Thus, though there appears to be a wide range of

diversity in the skin microbiome by traditional sequencing, PMA-seq indicates that this is

generally an overestimation at any given skin site. In fact, our results suggest that the viable skin

microbiome tends to be dominated by a relatively small number of taxa at most sites. An

independent analysis of glabella samples from 5 healthy volunteers including contamination

controls revealed the same trends that traditional sequencing overrepresented both diversity and

richness with or without contaminant removal (Figure 6 – figure supplement 3).

Figure 3: Relative abundance and change in richness and diversity of traditional sequencing compared to PMA-seq. (A) Relative abundance of all sequenced bacterial taxa at the family level. Paired bars represent data from traditional sequencing (left) and PMA-seq

(right). Samples are ordered by increasing richness in traditional sequencing. Labels below each

pair of bars indicate each sample's donor, replicate, and site (for example, HV1.1 RAC indicates

Healthy Volunteer 1, replicate sample 1, retroauricular crease). HS: hair shaft, RAC:

retroauricular crease, VF: volar forearm, PF: popliteal fossa, TW: toe web, AF: antecubital fossa.

*Relative abundance data for Staphylococcaceae was determined using forward-read sequencing

information only. Samples with fewer sequencing reads than PBS controls are not displayed. All

identified bacterial taxa with corresponding colors can be found in Figure 3 – figure supplement

1. (B) The richness changes between traditional sequencing (R_{trad}) and PMA-seq (R_{PMA}) are

demonstrated by plotting the change in richness (ΔR) against R_{trad}. Colors represent different site

types and shapes represent different sample sites. The shaded gray region represents the 95%

confidence interval for the linear regression. (C) The Shannon diversity changes between

traditional sequencing (H_{trad}) and PMA-seq (H_{PMA}) are demonstrated by plotting the change in

diversity (ΔH) against H_{trad}.

A

Figure 3 – figure supplement 1: Full list of identified taxa with corresponding colors. (A)
Full list of identified bacterial groups shown in Figure 3.

343 **Most bacterial groups are overrepresented by traditional sequencing methods**

344   The metrics of richness and diversity offer important information regarding how the

345   composition of the skin microbiome changes at different body sites between traditional

346   sequencing and PMA-seq. We further wanted to understand the changes in the relative

347   abundance of specific bacterial taxa. To quantify taxon-level PMA-dependent changes, we

348   developed a PMA-index (I_{PMA}) for each bacterial taxon, which is calculated as follows:

349 $I_{PMA} = \frac{A_{PMA}}{(A_{PMA}) + (A_{trad})}$, where A_{PMA} is the relative abundance by PMA-seq and A_{trad} is the relative

350   abundance by traditional sequencing. A low PMA-index (below 0.5) indicates that the taxon in

351   question is overrepresented by traditional sequencing, while a high PMA index (above 0.5)

352   indicates that the taxon in question is underrepresented by traditional sequencing (Figure 4A).

353   We note that these values represent enrichment relative to the rest of the sequences, a

354   measurement that is distinct from the viability score. Calculating PMA-indices revealed that the

355   abundances of most bacterial taxa at any given body site are overestimated by traditional

356   sequencing, as most taxa had PMA-index values close to 0 (Figure 4A). 16S rRNA copy

357   numbers could affect these values but the 16S copy numbers of the most abundant species were

358   all within a roughly two-fold range, such that this effect is minor compared to the trends

359   observed (*21*).

360   The four most abundant bacterial families (*Propionibacteriaceae*, *Staphylococcaceae,*

361   *Corynebacteriaceae*, and *Micrococcaceae*) made up 93% of total sequencing reads (96% of

362   PMA-seq reads and 91% of traditional sequencing reads) and demonstrated interesting family-

363   level PMA-index patterns. The family *Propionibacteriaceae* includes a major component of the

364   skin microbiome*, C. acnes*, which has been shown by traditional sequencing to comprise

365   upwards of 50% of the skin microbiome irrespective of site type (*8*). PMA-seq revealed that

traditional sequencing accurately represents *Propionibacteriaceae* abundance in sebaceous sites

(demonstrated by a PMA-index close to 0.5), but overrepresents *Propionibacteriaceae* in moist

and dry sites (PMA-indices of 0.2-0.3). Furthermore, *Propionibacteriaceae* dominated sebaceous

sites (accounting for >75% of all viable bacteria in most sebaceous samples), but did not

dominate moist or dry sites (their viable abundance did not exceed 50% of all viable bacteria in

any of those samples) (Figure 4 A, B). Similar to the *Propionibacteriaceae,* bacteria in the

family *Staphylococcaceae* appeared to be well-represented across all body sites (Figure 4A, B

and Figure 4 – figure supplement 2).

Bacteria in the family *Corynebacteriaceae* are also considered main constituents of the

skin microbiome, but our results showed that traditional sequencing overestimates the abundance

of *Corynebacteriaceae* at every skin site except for the nares. For example, traditional

sequencing identified a high abundance of *Corynebacteriaceae* in the popliteal fossa, but PMA-

seq showed that these reads were largely of inviable origin (Figure 4B). Previous studies have

demonstrated that *Corynebacteria* are readily cultured from nasal isolates, which supports our

PMA-seq finding that viable members of this taxon are abundant in the nares but not at most

other skin sites (*22*).

Interestingly, *Micrococcaceae* were overrepresented by traditional sequencing at every

site except for the hair shaft. In the hair shaft, *Micrococcaceae* were abundant by PMA-seq but

almost undetectable by traditional sequencing. As shown in Figure 4B, the increase in the

relative proportion of viable *Micrococcaceae* detected by PMA-seq corresponds to a decrease in

viable *Propionibacteriaceae*, suggesting that *Micrococcaceae* may not be detected by traditional

sequencing because of the high abundance of DNA from inviable *Propionibacteriaceae.* These

results suggest that most skin sites are colonized by a relatively small number of bacterial

families, that different families distinctly colonize different skin sites, and that the majority of the

additional bacterial DNA on the skin surface comes from inviable bacteria.

Figure 4: PMA-index and change in relative abundance between traditional and PMA-seq.
(A) The PMA index for each bacterial taxon that was present in at least 4 samples is shown here
as an average between samples of the same sample site shown in Figure 3. Color indicates PMA
index value. Saturation indicates confidence (sigma) in the PMA index value and was calculated
using the standard deviation of PMA index across the samples that went into that pixel. Bacterial
taxa are ordered by decreasing overall relative abundance. Each square represents the average of
at least four samples taken from different individuals. PMA-index is calculated by comparing the
relative abundance of a given taxon as measured by PMA-seq (A_{PMA}) to the sum of the relative
abundance for that taxon in both traditional sequencing (A_{trad}) and PMA-seq. (B) Relative
abundance at each body site for the top three most abundant (overall) bacterial taxa as assessed
by traditional sequencing and PMA-seq. Colors of bars correspond to colors in Figure 2F.
*Relative abundance data for Staphylococcaceae was determined using forward-read sequencing
information only.

Figure 4 – figure supplement 1: Viability score for three skin sites using lysostaphin and

***Staphylococcus*-specific PCR primers.** (A)The three skin sites with the most abundant bacterial

DNA are shown. Half of each sample was treated with lysostaphin prior to DNA isolation to

assess how the viability score would change. ddPCR was performed on each sample using both

16S primers (white bars) and *Staphylococcus*-specific (black bars) primers. The dashed line

indicates the average viability score of non-skin microbiome sites (0.66). N=3 for all.

414   **The spatial organization and viability of mouse skin microbiome is similar to that of**

415   **humans**

416   While our human skin microbiome FISH and PMA studies strongly agree, we are unable

417   to directly compare the two approaches on the same samples. For a system in which we could

418   perform both FISH and PMA-ddPCR on the same samples we turned to mouse skin, which also

419   enabled us to ask if our findings are human-specific. We assessed the spatial distribution of

420   bacterial cells in mouse skin tissue using the universal bacterial EUB338 FISH probe with tissue

421   from K14-H2B-GFP mice. We observed the same bacterial distributions as seen in the human

422   tissues: a high abundance of bacteria in hair follicles (enrichment score of 15.26) with relatively

423   few bacteria on the skin surface (enrichment score of 0.21) (Figure 5A, D). To test how fur

424   impacts the presence of bacteria on the skin surface, we performed FISH staining on skin from

425   nude mice (SKH1-Hrhr Elite) and found similar bacterial distributions (follicle-associated

426   enrichment score of 10.79 compared to 1.13 for the skin surface) (Figure 5B, D). The significant

427   numbers of bacteria observed in hair follicles suggests that the absence of surface bacteria is not

428   merely due to the relatively clean conditions in which laboratory mice are housed. Because the

429   mouse tissue was untreated with any sterilizing agents prior to dissection, these results also

430   suggest that the low numbers of skin-surface-associated bacteria found in the human tissue

431   samples are not simply a result of pre-biopsy sterilization.

432   Since we did not detect many bacteria on the surface of any of the skin samples tested,

433   we sought a positive control to confirm that our FISH staining can visualize bacteria on the skin

434   surface if they are present. For example, it is possible that the dry, acidic nature of the skin

435   affects FISH efficacy, leading to little staining at the skin surface. To this end, we applied *E. coli*

436   cells to dorsal mouse skin tissue after removing it from the animal. This tissue was then

processed in the same way as the human and other mouse tissue. FISH staining revealed many

bacteria on the surface of these samples, confirming that this technique can be used to reliably

visualize bacteria on the skin surface (Figure 5C). As a negative control, we also confirmed that

a probe encoding the reverse complement of the EUB338 FISH probe (NONEUB338) did not

significantly hybridize to the skin surface or follicles (Figure 5C). These results support the

conclusions from our FISH experiments on human tissues by demonstrating EUB338 staining

works well in the biological context of the skin surface.

We next assessed the viability of bacteria in the mouse skin microbiome using PMA-

ddPCR. As further evidence that our previous findings are not human-specific, the PMA-ddPCR-

based viability score for mouse skin microbiome sites was similar to the average viability score

for human skin sites (0.066 and 0.045 respectively) and was much lower than the viability score

for the mouse or human fecal microbiome (0.98 and 0.66 respectively). (Figure 5E). These

results indicate that, despite having distinct skin biology, both humans and mice have an

abundance of bacterial DNA on the skin surface that is not associated with viable cells.

451
Figure 5: Bacterial FISH staining of mouse tissue (A-D) and comparison of mouse viability
scores and human viability scores. (A). Tissues from a K14-H2B_GFP mouse stained with
EUB338 show abundant bacterial signal in hair follicles but not on the skin surface. (B) Tissues
from SKH1-Hrhr Elite nude mice also show bacterial presence concentrated to cutaneous
structures and not at the skin surface. (C) E. coli applied to C57BL/6 mouse tissue was stained
with either EUB338 (in red) or its complementary strand control probe NONEUB338 (in
yellow). (D) Quantification enrichment scores showing the median and interquartile range.
Significance was calculated using the Mann-Whitney test. *P ≤ 0.05, **P ≤ 0.01, N = 6 for hairy
mouse follicle, nude mouse follicle, and hairy mouse stratum corneum, N = 5 nude mouse
stratum corneum. (E) The PMA-ddPCR-based viability scores for mouse skin microbiomes are
much lower than for mouse fecal microbiomes (0.66 and 0.98 respectively). These viability
scores for mouse sites are very similar to those for humans (0.066 and 0.045 for skin
microbiomes, 0.98 and 0.66 for fecal microbiomes).

Skin surface repopulation is driven by stable communities of bacteria below the skin

surface

Finally, we sought to determine the functional significance of our findings in the context

of skin microbiome surface perturbation and repopulation. Specifically, we performed a

perturbation-recovery experiment in which the skin microbiome of the forehead from healthy

volunteers was sampled at time T=0 to establish a baseline microbial community, sterilized with

benzalkonium chloride, and then sampled 3, 24, and 48 hours later to monitor recovery (Figure

6A). At each sampling timepoint, half of the sample was treated with PMA prior to DNA

isolation and subsequent 16S rRNA gene sequencing. Focusing on the PMA-untreated samples,

we found that 4 of the 5 volunteers demonstrated significant reduction in the absolute amount of

DNA upon sterilization (Figure 6B), but that the overall bacterial composition of the skin

microbiome remained relatively stable (Figure 6A, all bacteria listed in Figure 6 – figure

supplement 2.). It should be noted that, although the skin microbiome composition found in

volunteer 4 is markedly different than the communities identified in the other volunteers, the

species identified are still common constituents of a healthy skin microbiome. Additionally,

volunteer 4 had a lower abundance of bacterial DNA overall, which might lead to the lower

relative abundance of *Propionibacteriaceae.*

Our findings recapitulate the previously-reported paradox that the skin microbiome is

both unstable at short timescales and stable at long timescales (*11*, *23*). Importantly, analysis of

the PMA-treated samples enabled us to understand the recovery dynamics. Specifically, we used

Bray-Curtis dissimilarity to monitor the extent of population recovery (Figure 6C). We found

that comparing samples with and without the use of PMA to the PMA-treated baseline

community resulted in a consistent pattern across all individuals. Over the course of the

experiment, the communities converged back to the PMA-treated baseline samples but showed

no consistent patterns of similarity relative to the PMA-untreated baseline samples (Figure 6 –

figure supplement 2A). This result supports the hypothesis that bacterial DNA on the skin

surface can be readily removed and is continuously replenished by viable bacterial populations in

protected reservoirs below the skin surface.

Examining repopulation dynamics more finely at the ASV level revealed that ASVs that

were lost upon surface sterilization often re-appeared over time, consistent with the model that

surface repopulation can be driven by subsurface populations. Specifically, we found that after

surface sterilization, the skin is repopulated by the same ASVs as were present before

sterilization (Figure 6 – figure supplement 2B). In Figure 6 – figure supplement 2B we highlight

specific examples in which a given ASV appears in the live-cell population (+PMA) prior to

surface sterilization (T0), disappears from the live-cell population following sterilization (T3),

and later re-appears in the live-cell population either 24 or 48 hours later. Together, these results

suggest that bacteria thrive in protected areas below the skin surface and that the DNA of their

dead remains accumulates on the skin surface. In this way, the bacterial DNA on the skin surface

acts as a fingerprint of the communities below.

Figure 6: Skin microbiome perturbation and recovery. (A) Bacterial relative abundance in

each individual over the 48 hours following perturbation. 0 hour represents baseline, pre-

perturbation community. Whether a sample was treated with PMA is indicated by (-) and (+). (B)

Quantification of bacterial DNA recovery over the 48 hours following perturbation. DNA was

quantified using ddPCR. (C) Bray-Curtis dissimilarity of each individual over the 48 hours

following perturbation. Red data points are comparing PMA-treated samples to the PMA-treated

baseline sample. Blue data points are comparing PMA-untreated samples to the PMA-treated

baseline sample. Dashed vertical line indicates the point of perturbation.

k__Bacteria.p__Actinobacteria.c__Actinobacteria.o__Actinomycetales.f__Propionibacteriaceae
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515
516 **Figure 6 – figure supplement 1: List of bacteria identified in perturbation recovery.** The
517 bacterial groups listed here correspond to the entire sequencing dataset shown in Figure 6.
518 Bacteria are listed in order of relative abundance.

Figure 6 – figure supplement 2: Skin microbiome perturbation and recovery. (A) Bray-Curtis dissimilarity of each individual over the 48 hours following perturbation. Red data points are comparing PMA-treated samples to the PMA-untreated baseline sample. Blue data points are comparing PMA-untreated samples to the PMA-untreated baseline sample. Dashed vertical line indicates the point of perturbation. (B) ASV-level relative abundance converted into binary dataset in which each ASV present at greater than 1% appears in yellow and each absent ASV appears in black. Columns are ASVs and rows are PMA-treated or PMA-untreated perturbation recovery samples. ASVs indicated in red follow a specific pattern of repopulation in which the ASV is present in the live cells (+PMA) at T=0, disappears from the live cells after perturbation (T=3), and then recovers in the live population at either T=24 or T=48.

530
Figure 6 – figure supplement 3. Contamination removal performed on 600nt sequencing
data. The Shannon diversity changes between traditional sequencing (H_{trad}) and PMA-seq
(H_{PMA}) are demonstrated by plotting the change in diversity (ΔH) against H_{trad} (A-C). The
richness changes between traditional sequencing (R_{trad}) and PMA-seq (R_{PMA}) are demonstrated
by plotting the change in richness (ΔR) against R_{trad} (D-F). Analysis with no decontamination
removal (A, D), decontamination removal using the Decontam program in R with a 0.1 threshold
(B, E), and decontamination removal using the Decontam program in R with a 0.2 threshold (C,
F) are all shown. Symbols in red represent samples associated with the perturbation recovery
data (Figure 6). Symbols in blue represent PBS DNA isolation controls. Symbols in black
represent follicle contents obtained from single follicles. The shaded red region represents the
95% confidence interval for the linear regression performed using only the perturbation recovery
samples (red symbols).
543

Discussion

Here we used both imaging and PMA-based methods to demonstrate that the skin surface is
sparsely colonized by bacteria. This central finding holds true across skin from human biopsies,
healthy swabbed volunteers, hairy mice, and nude mice. The skin microbiome has garnered a
great deal of attention as a means for educating the immune system, combatting pathogens, and
promoting wound healing, and multiple groups are pursuing skin probiotics (*24*). Our findings
have significant implications for the mechanisms underlying these skin microbiome functions, as
well as for the ability to manipulate skin microbiome composition. For example, our findings
support previous work suggesting that the key function of immune education by the skin
microbiome occurs within hair follicles (*4*), though cell-free DNA on the skin surface could also
impact the immune system. Our findings also support previous work showing that individual
pores are colonized by clonal bacterial populations (*25*). Meanwhile, our results may help to
explain why stably colonizing the skin surface with exogenous bacteria has proved to be difficult
and often requires abrasion (*3*). Disruption of the stratum corneum (skin surface layers), which
occurs during skin abrasion, may allow access to the deeper layers of tissue where the stable
bacterial populations reside. Thus, targeting the bacteria within hair follicles may represent a
better strategy for stably manipulating the skin microbiome or educating the immune system.
While it has been known for some time that bacteria inhabit hair follicles (*4*, *15*), our findings
extend this knowledge and suggest that the viable bacteria of the skin microbiome are primarily
restricted to these sites. This is further supported by our finding that bacteria can be readily
cultured from the skin surface, but at far less abundance than suggested by the amount of
bacterial DNA present on the skin.

Our results also offer insight into the apparent paradox of skin microbiome stability, in

which skin microbiome composition has been found to be both stable over long time scales and

susceptible to transient perturbations (*11*, *23*). The skin is primarily colonized by viable bacteria

in pilosebaceous units and other skin invaginations and the bacterial DNA on the skin surface is

mostly from dead bacteria. We show that surface bacterial DNA could be easily washed away,

while the viable subsurface bacteria remain unperturbed by washing, persisting as a small but

stable source of new bacterial DNA that replenishes the skin surface over time. This model also

explains the counter-intuitive abundance of obligately anaerobic bacterial species associated with

the skin microbiome. Human skin cells are also born below the skin surface and are pushed to

the surface as they die. In this way, the life cycle of the skin microbiome may resemble the

process of epithelial cell turnover in the skin.

Our findings also raise fundamental questions that will need to be addressed by future

studies focused on why the skin surface is poorly colonized. The accessible DNA in non-skin

microbiomes is generally representative of viable bacterial cells, as all non-skin microbiomes

tested had viability scores above 0.4. This is in stark contrast to the skin microbiome, which had

viability scores between 0.02 and 0.12. Even saliva, which contains many antimicrobial

compounds (*26*), had a viability score nearly four times greater than the highest viability score

for a skin microbiome. The uniquely low viability score associated with the skin could be

explained by passive mechanisms, like bacterial DNA adhering to the skin long after a bacterial

cell dies. Alternatively, there could be active mechanisms like bacterial killing on the skin

surface by factors like antimicrobial peptides produced by epithelial cells, competition between

bacterial species, or exposure to harsh environmental factors such as starvation, UV radiation, or

desiccation. Another open question is whether similar trends hold true for non-bacterial

components of the skin microbiome like fungi and viruses.

Materials and Methods

Human and mouse tissue processing

Human tissue was obtained through the Skin Translational Research Core within the Skin

Biology and Diseases Resource-based Center (SBDRC) at the University of Pennsylvania.

Normal adult human skin was obtained during skin cancer surgery procedures. After the

cancerous tissue is removed, normal (non-cancerous) tissue must also be removed to ensure a

cosmetic repair. These tissue portions are normally discarded, but were collected for imaging.

Surgical scrub solution was used prior to the procedure (alcohol/iodine or chlorhexidine).

Tissues were embedded in room temperature OCT immediately upon excision and stored at -

80°C. Tissues were obtained from the forehead of a 32-year-old female and the cheek of an 81-

year-old male. Tissues were obtained through the Core B- Skin Translational Research Core at

the SBDRC as part of the "Fresh normal and diseased human skin" service.

All mice were housed and maintained in a certified animal facility and all experiments

were conducted according to USA Public Health Service Policy of Humane Care and Use of

Laboratory Animals. All protocols were approved by the Institutional Animal Care and Use

Committee, protocol #1867-17 (Princeton University). Dorsal skin from K14-H2B-GFP,

C57BL/6, and SKH1-Elite nude mice was used for fluorescence experiments. The fur from K14-

H2B-GFP and C57BL/6 mice was shaved using clippers prior to dissection. Dorsal tissue was

removed, cut into thin (~5mm) strips, and embedded in room temperature OCT. OCT blocks

were placed on dry ice to freeze and subsequently stored at -80°C . Both sexes were used. All

animals were between 3 and 6 weeks in age and were fed standard, non-sterilized rodent chow.

No pathogens were identified in the regular health screenings conducted by facility staff. Mice

were housed with 1-5 animals per cage with bedding and enrichment using Ventilated Thoren

racks. Mice were euthanized using CO_2 inhalation (as per the AVMA Guidelines for the

Euthanasia of Animals: 2020 Edition), followed by cervical dislocation. Dissections were

performed with ethanol-sterilized tools.

Fluorescence in situ hybridization and imaging

Human and mouse tissues were processed identically. 30 μm tissue sections were

sectioned and mounted on slides using a Leica CM3050S cryostat. Tissues on slides were fixed

in 4% formaldehyde in 1X PBS for 20 mins. Following fixation, tissues were washed for 5 mins

in 1X PBS and then incubated in hybridization buffer (0.9 M NaCl, 20 mM Tris-HCl, 10%

formamide) containing 10 mg/mL lysozyme and FISH probes at 1 μg/μL for 2-3 hours at 47ºC in

the dark. Nucleotide sequences of FISH probes can be found in Table S1. After hybridization,

slides were incubated in wash buffer (0.9 M NaCl, 20 mM Tris-HCl) for 1 hr at 47ºC in the dark.

Slides were then washed in 1X PBS for 10 mins. To visualize tissue nuclei, tissues were then

stained with 1 μg/mL DAPI for 10 mins at room temperature. Slides were then washed 3X in 1X

PBS for 10 mins each. Following the final wash step, tissues were mounted with glycerol-based

anti-fade non-curing mounting media. Coverslips were sealed using a 1:1:1 mixture of petroleum

jelly, lanolin, and paraffin. Images were acquired on a Nikon A1R-Si HD confocal microscope

controlled by NIS Elements software. ImageJ and MATLAB (Mathworks, Natwick, MA) were

used for image processing. To calculate fluorescence enrichment scores, maximum projections

of Z-stacks were used. A rectangular region of interest (ROI) of width W was drawn around a

follicle using only the DAPI channel. A second ROI was drawn near the follicle opening with

dimensions 0.5W x 2W in order to capture follicle-associated fluorescence near the follicle

opening. In an orthogonal processing step, the pixel intensities for the entire image in the FISH

fluorescent channel were fit using a 3 component Gaussian mixture model (GMM)

corresponding roughly to non-tissue background, non-probe autofluorescence, and probe-based

signal. Using this GMM as the background subtraction value for each image, the mean intensity

for each ROI was calculated. This threshold value was used to calculate the mean fluorescence

value inside the ROI and outside the ROI. The ratio of the mean fluorescence value inside the

ROI to the mean fluorescence value outside of the ROI was used for quantifying follicle- and

stratum corneum-associated fluorescence (enrichment score). Significance was calculated using

the Mann-Whitney test.

Stationary phase cultures were grown according to the following conditions:

Strain	Growth time	Media	Environmental condition
Cutibacterium acnes ATCC6919	72 hours	Reinforced clostridial medium (RCM) (Oxoid, CM0149)	Anaerobic, 37ºC
Corynebacterium striatum ATCC6940	24 hours	Brain heart infusion (BHI) (BD, 237500)	Aerobic, 30ºC
Micrococcus luteus ATCC4698	24 hours	Tryptic soy broth (TSB) (BD, 211825)	Aerobic, 30ºC
Staphylococcus epidermidis EGM 2-06	24 hours	Tryptic soy broth (TSB) (BD, 211825)	Aerobic, 37ºC

Human subject microbiome samples

Microbiome samples were collected under Princeton University IRB #13003 at the

Princeton University Department of Molecular Biology. Healthy volunteers were recruited via

informational flyers. Volunteers gave informed consent prior to sample collection. Participants

were healthy volunteers, male and female, white, aged 26-35, with no history of chronic skin

conditions or autoimmune disease and were not using antibiotics at the time of sampling or one

month prior to sample collection.

Skin microbiomes from healthy volunteers were collected using sterile foam-tipped

collection swabs pre-moistened with sterile 1X DPBS. Though often included, we chose not to

use detergent in the swabbing buffer in order to avoid negatively affecting bacterial cell

membranes and altering viability scores. Figure 2 – figure supplement 1G shows how swabbing

with and without 0.1% Triton X-100 (Sigma) affects viability scores, ddPCR results, and CFU.

Areas of interest were sampled for 60 seconds before being re-suspended in sterile 1X DPBS.

Tongue microbiome samples were collected using sterile foam-tipped collection swabs. Hair

shaft samples were collected by plucking hairs and using only the bulb portion. Follicle contents

were collected using Bioré Deep Cleansing Pore Strips (Kao USA Inc., Cincinnati, OH). Saliva

was collected in sterile 50 mL conicals from healthy volunteers. Plaque was collected by

scraping the teeth of healthy volunteers using sterile toothpicks and re-suspending the collection

in sterile 1X DPBS. Murine fecal samples from C57BL/6 mice were collected during dissection.

Human skin microbiome samples shown in Figure 2 – figure supplement 1 were plated for CFU

calculations prior to the addition of PMA. Samples were plated on blood agar plates (5% sheep

blood in tryptic soy agar, VWR International) and grown for 24-48 hours aerobically or

anaerobically. For perturbation recovery experiments, volunteers were asked to refrain from

using any type of products on their face for the duration of the experiment (48 hours) as well as

24 hours prior to baseline sampling. Volunteers were allowed to bathe/shower, but were asked to

refrain from washing their faces. Skin surface sterilization was done by scrubbing the sample

area for 60 seconds with an antiseptic wipe containing benzalkonium chloride 0.13% w/v (North

by Honeywell, #150910). To ensure that the process of sampling the skin surface would not

affect the bacterial repopulation dynamics, forehead of each volunteer into four sections of equal

areas: T0 (baseline, before sterilization), T1 (3hrs post sterilization), T2 (24 hrs post-

sterilization), and T3 (48 hrs post-sterilization). Each section was sampled for 60 seconds using a

sterile foam-tipped collection swab pre-moistened with sterile 1X DPBS. As described above,

each swab was then resuspended in sterile 1X DPBS.

The human fecal sample was collected by the Donia lab under Princeton University IRB

#11606 and was gifted to the Gitai lab. This sample and the methods of collection were

described in Javdan, Bahar et al. "Personalized Mapping of Drug Metabolism by the Human Gut

Microbiome." *Cell* vol. 181,7 (2020): 1661-1679.e22. doi:10.1016/j.cell.2020.05.001 (*27*).

Heat-killed *E. coli* ratios

In order to demonstrate the efficacy of PMA, known ratios of live and heat-killed *E. coli* cultures

were mixed and subjected to PMA treatment. First, an overnight culture of *E. coli* NCM3722

was back-diluted into fresh LB media at a ratio of 1:1000 and grown at 37°C for 4 hours to reach

mid-log. Stationary-phase experiments were done with overnight, 18h cultures. Cultures were

then washed 3X and suspended in sterile PBS. Half of the culture was heat-killed by incubating

at 70°C for 20 minutes while the other half remained at room temperature. The heat-killed *E. coli*

cultures were allowed to cool to room temperature before combining with non-heat-killed *E. coli*

cultures to achieve a 50% (by volume) heat-killed mixture. For the 0% heat-killed, no heat-killed

bacteria were added. Likewise, for the 100% heat-killed, only heat-killed bacteria were used.

Each condition was mixed well and then split evenly between two sterile 1.5 mL microcentrifuge

tubes.

Propidium monoazide (PMA) treatment, DNA isolation

After collection, samples were split evenly between two sterile 1.5 mL microcentrifuge

tubes. PMA (Biotium Inc.) was added to one of the two tubes to a final concentration of 50 µM.

All tubes were incubated in the dark at room temperature for 10 mins before being exposed to

light to cross-link PMA molecules using the PMA-Lite™ LED Photolysis Device (Biotium Inc.).

DNA was then isolated from all samples using either the DNeasy PowerSoil Kit (Qiagen) or

according to the protocol outlined in Meisel et al. (*28*). The Meisel et al. protocol was modified

such that sonication (Branson Sonifier 250, VWR Scientific) was used instead of bead-beating

for the purposes of cell-lysis. Both DNA isolation methods work similarly (Figure 2 – figure

supplement 1H). If lysostaphin (Sigma-Aldrich) was used, it was added following PMA

activation and before DNA isolation to a final concentration of 0.1 mg/mL and incubated at room

temperature for 30 mins.

Droplet digital PCR (ddPCR)

The Bio Rad QX200 AutoDG Droplet Digital PCR System was used to quantify

extracted DNA from microbiome samples and from pure bacterial cultures. Reaction mixtures

contained 2x QX200 ddPCR EvaGreen Supermix and universal 16S qPCR primers at 10 nM

concentrations in a total volume of 25 µL. Primer sequences can be found in Table S1. Reaction

mixtures were transferred to sterile ddPCR 96-well plates (BioRad #12001925) which were

loaded into the QX200 Automated Droplet Generator. After droplet generation, the plate was

heat-sealed using the PX1 PCR Plate Sealer (BioRad #1814000) and PCR was performed with a

pre-step of 95 C for 5 minutes followed by 40 rounds of amplification with 60C, 1 minute

extensions and a final hold temperature of 12 C using a C1000 Touch Thermal Cycler (BioRad

#1851197). Samples were subsequently loaded into the QX200 Droplet Reader for

quantification. Automatic thresholding was performed using the Quantasoft software and

subsequently exported to Microsoft Excel for analysis. Significance was calculated using a

Student's T-Test. To calculate the viability score for a given pair of "-PMA" and "+PMA"

matched samples, the following calculation was done:

726
$$\frac{Copies\ per\ 20\ \mu L\ with\ PMA}{Copies\ per\ 20\ \mu L\ without\ PMA}$$

ddPCR and CFU standard curves

Cultures of *S. epidermidis* EGM 2-06 were grown in tryptic soy broth (TSB) overnight

and diluted 1:1000 the following morning in TSB and grown for 4 hours until a final OD of 0.4.

Tenfold dilutions of *S. epidermidis* culture were then prepared, plated for CFU on 5% sheep

blood in tryptic soy agar (VWR International) and divided between two 1.5 mL microcentrifuge

tubes. Stationary-phase cultures of *C. acnes, S. epidermidis, M. luteus,* and *C. striatum* were

grown according to the table above. PMA was added to one tube for a final concentration of 50

μM and the other tube was left untreated. PMA activation and DNA isolation was then done

according to the methods outlined above.

16S rRNA gene amplicon sequencing

DNA was isolated from microbiome samples (with and without PMA) using the DNeasy

PowerSoil Kit (Qiagen) (Figures 2-5) or according to the protocol outlined in Meisel et al

(Figure 6). The V1-V3 region of the 16S gene was amplified using the primers 27F (5'-

AGAGTTTGATCCTGGCTCAG) and 534R (5'- ATTACCGCGGCTGCTGG). Illumina

sequencing libraries were prepared using previously published primers (*29*). Libraries were then

pooled at equimolar ratios and sequenced on an Illumina MiSeq Micro 500 nt (Figure 3, Figure

4) or MiSeq V3 600 nt (Figure 6) as paired-end reads. Reads were 2X250 bp or 2X300 bp with

an average depth of ~33,616 reads for 500 nt and 245,488 reads for 600 nt (Tables S2, S3). Also

included were 8 bp Index reads, following the manufacturer's protocol (Illumina, USA). Raw

sequencing reads were filtered by Illumina HiSeq Control Software to generate Pass-Filter reads

for further analysis. Index reads were used for sample de-multiplexing. Amplicon sequencing

variants (ASVs) were then inferred from the sequences using the DADA2 plugin within QIIME2

version 2018.6 (*30*, *31*). Reverse reads were trimmed to 245 bp for the 2X250 bp data, and the

reverse reads were trimmed to 275 bp for the 2X300 bp data. For the 2X250 bp data, two

different ASV inference methods were utilized: one using only the forward reads (to capture the

longer-than-average Staphylococcus amplicons) and one using both paired-end reads. For the

2X300 bp data, ASV inference was performed using both paired-end reads. ASV inference using

forward reads was performed using DADA2's denoise-single function ,while ASV inference

using both reads was performed using the denoise-paired function.. Taxonomy was assigned to

the resulting ASVs with a naive Bayes classifier trained on the GreenGenes database version

using only the regions of the 16S rRNA gene spanned by the ASVs (*32*, *33*). All downstream

analyses were performed using family-level taxonomy assignments unless specified to be at the

ASV-level. PBS control samples were included at the library preparation step for 500 nt

sequencing data and air swab DNA isolation controls were included for 600 nt sequencing data.

Contamination removal was performed on 600 nt sequencing data using the Decontam package

for R (Figure 6 and Figure 6 – figure supplement 2 show data without contamination removal).

(*34*). Figure 6 – figure supplement 3 shows how Shannon diversity and richness change between

traditional sequencing and PMA-seq (similar to what is shown in Figure 3B and Figure 3C).

Removal of contaminants (using either 0.1 or 0.2 threshold) has little effect on the data,

suggesting that, although contamination removal cannot be performed on the 500 nt sequencing

data due to the absence of air swab controls, this should not change the overall findings shown in

Figure 3 and Figure 4. Relative abundance, richness, Shannon diversity, Bray-Curtis

dissimilarity, and PMA-index were assessed using the Vegan package for R or Microsoft Excel

and plotted using R, Prism, and MATLAB (*35*). The PMA-index was calculated using relative

abundance and was not calculated for any bacterial taxa that was present in fewer than four

samples.

Data availability

Sequencing datasets generated and analyzed during this study are available at NCBI Sequence

Read Archive (SRA) (BioProject numbers (PRINA918671 and PRINA918959). Oligonucleotide

sequences used are included in table S1.

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