A core of functionally complementary bacteria colonizes oysters in Pacific Oyster Mortality Syndrome.

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ABSTRACT

The Pacific oyster *Crassostrea gigas* is one of the main cultivated invertebrate species around the world. Since 2008, oyster juveniles have been confronted with a lethal syndrome, Pacific Oyster Mortality Syndrome (POMS). The etiology of POMS is complex. Recently, we demonstrated that POMS is a polymicrobial disease. It is initiated by a primary infection with the *herpesvirus* OsHV-1 μ Var, and evolves towards a secondary fatal bacteremia that is enabled by the oyster's immunocompromised state. In the present article, we describe the implementation of an unprecedented combination of metabarcoding and metatranscriptomic approaches to show that the sequence of events in POMS pathogenesis is conserved across infectious environments and susceptible oyster genetic backgrounds. We also identify a core colonizing bacterial consortium which, together with OsHV-1 μ Var, forms the POMS pathobiota. This bacterial core is characterized by highly active global metabolism and key adaptive responses to the within-host environment (e.g. stress responses and redox homeostasis). Several marine gamma proteobacteria in the core express different and complementary functions to exploit the host's resources. Such cross-benefits are observed in colonization-related functions, and reveal specific strategies used by these bacteria to adapt and colonize oysters (*e.g.* adhesion, cell defense, cell motility, metal homeostasis, natural competence, quorum sensing, transport, and virulence). Interdependence and cooperation within the microbial community for metabolic requirements is best exemplified by sulfur metabolism, which is a property of the pathobiota as a whole and not of a single genus. We argue that this interdependence may dictate the conservation of the POMS pathobiota across distinct environments and oyster genetic backgrounds.

INTRODUCTION

Introduced from Asia to a broad range of countries, *Crassostrea gigas* has become one of the world's main cultivated species. Since 2008, juvenile stages of *C. gigas* have suffered massive mortality events, especially in France [1]. In subsequent years, this so-called Pacific Oyster Mortality Syndrome (POMS) [2] has become panzootic. POMS has been observed in all coastal regions of France [3-5] and numerous other countries worldwide [6-11].

This situation has promoted a research effort that has revealed a series of factors contributing to the disease, including infectious agents that interact with seawater temperature and oyster genetics [12-17]. Dramatic POMS mortality events have coincided with the recurrent detection of *Ostreid herpesvirus* variants in moribund oysters [3-5]. The involvement of other etiological agents has also been suspected. In particular, bacterial strains of the genus *Vibrio* have been shown to be associated with the disease [18, 19]. Recently, integrative molecular approaches performed on susceptible and resistant families of oysters have revealed the complex etiology of POMS [20, 21]. These studies demonstrated that infection by *Ostreid herpesvirus* type 1 µVar (OsHV-1) is a critical step in POMS pathogenesis. OsHV-1 leads to an immunocompromised state by altering hemocyte physiology. This results in dysbiosis and bacteremia, which ultimately results in oyster death [20]. Several bacterial genera are involved in this secondary infection [20]; one of them *(Vibrio)* was found to include opportunistic pathogens that cause hemocyte lysis [21]. This observation occurred in a specific infectious environment in French Brittany [20]. Whether the observed mechanisms of POMS pathogenesis are conserved in other infectious environments remains unknown. Also of importance, *Vibrio* species are not the only bacteria that colonize oyster tissues during the secondary bacterial infection. Several bacterial genera, including *Arcobacter*, *Marinobacterium*, *Marinomonas*, and *Psychrobium* are also found to massively colonize oysters [20]. The mechanisms underlying the colonizing capacity of these species remain to be elucidated.

In the present study, we investigated whether POMS pathogenesis is conserved across environments, and which biological functions are expressed by the bacterial consortia that causes oyster death. We compared pathogenesis in oyster biparental families that display contrasting phenotypes (resistant or susceptible to POMS) in two infectious environments (the Atlantic Bay of Brest environment, and the Mediterranean Thau Lagoon environment). We found that the sequence of events is conserved in both infectious environments: intense viral replication in susceptible oysters, followed by a secondary bacteremia caused by a conserved bacterial consortium that results in oyster death. Using metabarcoding and metatranscriptomics, we identified members of the core colonizing bacterial consortium, and characterized their highly active metabolism and key adaptive responses to the within-host environment. We also identified a series of genus-specific functions required for

colonization, which reveal specific strategies that the core consortium uses to adapt and colonize

oysters. Finally, we identified cooperative processes used by the core consortium to meet their

metabolic requirements; these processes may dictate the conservation of the core consortium across

environments.

METHODS

Production of biparental oyster families

C. gigas families were produced as described in [20, 22]. Briefly, oysters were produced at the Ifremer

hatchery in Argenton in March 2015. Three susceptible families (F_{11}, F_{14}, and F_{15}) and three resistant

families (F_{21}, F_{23}, and F_{48}) were used as recipients, and 15 families were used as donors. All families

were maintained under controlled bio-secured conditions to ensure their specific pathogen-free status.

Status was verified by i) the absence of OsHV-1 DNA using qPCR, (see below) and ii) a low *Vibrio*

load (~10 cfu/g tissue) on selective culture medium (thiosulfate-citrate-bile salts-sucrose agar) [23].

Oysters remained free of any abnormal mortality throughout larval development, at which time

experimental infections were started.

Mesocosm experimental infections

The experimental infection protocol consisted of a cohabitation assay between donors (exposed to

pathogens naturally present in the environment) and recipient specific pathogen-free oysters [17, 18].

Details of the experimental infection protocol (e.g. biomass, oyster weight, experimental duration, and

tank volume) were as described in [20, 22]. Briefly, donor oysters were deployed at Logonna Daoulas

(lat 48.335263, long 4.317922) in French Brittany (Atlantic environment) and at Thau Lagoon (lat

43.418736, long 3.622620) (Mediterranean environment). Oysters were deployed in farming areas

during the infectious period, and remained in place until the onset of mortality (< 1%). Donors were

then brought back to the laboratory (in Argenton) and placed in tanks, each containing recipient

oysters from the three resistant and the three susceptible families. Experimental infections took place

in July 2015 and September 2015 for the Atlantic and Mediterranean exposures, respectively. For each

experimental infection, mortality rate was monitored, and 10 oysters were sampled in triplicate from

each oyster family shucking at 7 time points (0, 6, 12, 24, 48, 60, and 72 hours post-infection). The

shell was removed and the whole oyster was flash frozen in liquid nitrogen. Oyster pools were ground

in liquid nitrogen in 50 ml stainless steel bowls with 20mm diameter grinding balls (Retsch MM400

mill). The powders obtained were stored at -80°C prior to RNA and DNA extraction.

DNA extraction and quantification of OsHV-1 and total bacteria.

DNA extraction was performed as described in [20] using the Nucleospin tissue kit (Macherey-Nagel).

DNA concentration and purity were checked with a NanoDrop One (Thermo Scientific).

Quantification of OsHV-1 and total bacteria were performed using quantitative PCR (qPCR) with a

Roche LightCycler 480 Real-Time thermocycler (qPHD-Montpellier GenomiX platform, Montpellier

University, France). Absolute quantity of OsHV-1 was calculated by comparing the observed Cq

values to a standard curve of the OsHV-1 DNA polymerase catalytic subunit (AY509253)

amplification product cloned into the pCR4-TOPO vector (Invitrogen). Relative quantification of total

bacteria 16S rDNA gene was calculated by the $2^{-\Delta\Delta Cq}$ [24] method with the mean of the measured

threshold cycle values of two reference genes (*Cg-BPI*, GenBank: AY165040 and *Cg-actin*, GenBank:

AF026063) (de Lorgeril, Lucasson et al. 2018).

Analyses of bacterial microbiota

Bacterial metabarcoding was performed using 16S rRNA gene amplicon sequencing. Libraries were

generated using the Illumina two-step PCR protocol targeting the V3-V4 region [25]. A total of 252

libraries (six families × seven sampling time points × three replicates × two infectious environments)

were paired-end sequenced with a 2×250 bp read length at the Genome Quebec platform on a MiSeq

system (Illumina) according to the manufacturer's protocol. A total of 41,012,155 pairs of sequences

were obtained. Metabarcoding data was processed using the FROGS pipeline [26]. Briefly, paired

reads were merged using FLASH [27]. After cleaning steps and singleton filtering, 26,442,455

sequences were retained for further analyses. After denoising and primer/adapter removal with

cutadapt, clustering was performed using SWARM, which uses a two-step clustering algorithm with a

threshold corresponding to the maximum number of differences between two Operational Taxonomic

Units (OTU) (denoising step d = 1; aggregation distance = 3) [28]. Chimeras were removed using

VSEARCH [29]. Resulting OTUs were annotated using Blast+ against the Silva database (release

128).

Bacterial metatranscriptomic analysis

Powder obtained from the frozen oysters was resuspended in Trizol, and total RNA was extracted

using a Direct-zol^{TM} RNA Miniprep kit. Polyadenylated mRNAs (*i.e.* oyster mRNAs) were removed

using a MICROB*Enrich*^{TM} Kit (Ambion). cDNA oriented sequencing libraries were prepared as

described in [21] using the Ovation Universal RNA-Seq system (Nugen). Library preparation included

steps to remove oyster nuclear, mitochondrial, and ribosomal RNAs, as well as bacterial rRNAs [21].

A total of 36 libraries (three families × two sampling timepoints × three replicates × two infectious

environments) were sequenced by the Fasteris company (Switzerland, https://www.fasteris.com) in

paired-end mode (2 × 150 bp) on an Illumina HiSeq 3000/4000 to obtain 200-300 million clusters per

sample (**Supplementary Table 1**).

Figure S1 presents a schematic of data processing. Raw Illumina sequencing reads from 72 fastq files

(R1 and R2) were trimmed using Trimmomatic v0.38, and rRNA reads (both eukaryotic and bacterial)

were removed using Sortme RNA v2.1b with the rRNA Silva database (release 128). At this stage,

about 9% of reads were removed, underscoring the efficiency of experimental rRNA removal during

library preparation (**Supplementary Figure 1**).

To further enrich bacterial sequences, reads were successively mapped by Bowtie2 [30] on a

multifasta file containing *Crassostrea gigas* genome sequence v9, complemented by *C. gigas* EST

(available from NCBI), and a multifasta file containing the sequences of OsHV-1 (present in diseased

oysters) and other viral sequences previously associated with bivalves [31]. Unmapped reads, which

represented 4-10% of the starting reads (depending on conditions) were retained for further analysis

(**Supplementary Table 1**). After a new filtering step using Trimmomatic, all remaining reads

corresponding to the 36 samples were pooled (516,786,580 reads, 36-150 nt) and assembled using

Trinity v2.3.2 in paired-end mode to build a reference metatranscriptome (1,091,409 contigs, 201 –

15,917 nt). The resulting metatranscriptome was further annotated using Diamond BlastX against the

NCBI nr protein database [32]. 48.4 % of the contigs aligned with a protein in the database, and were

further assigned to a taxon using Megan 6-LR Community Edition [33]. Sequences were annotated at

different taxonomic levels from species to phylum. Out of the 1,091,409 contigs, 352,473 contigs

aligned with bacterial proteins by BlastX with an E-value $\leq 01^e$-06.

Each protein was assigned to 31 functional categories (**Supplementary Table 2**). Out of the 352,473

best hits, 74,015 were annotated as "hypothetical", "unknown", or "unnamed", and were assigned to

the category "Unknown function". The remaining hits corresponded to 15,504 different protein names

that were manually assigned to one of the 30 remaining functional categories.

Quantification of gene expression and data normalization

For each of the 36 samples used for the assembly of the metatranscriptome, reads were mapped back

onto the metatranscriptome by Bowtie2 in paired-end mode. Raw counts per features (*i.e.* per contig)

were computed using HTseq-Count [34]. For each contig, and for each sample, raw counts were

normalized to TPM (Transcripts per Kilobase / Million = Mapped reads / Length of contig (kb) /

Scaling factor with Scaling factor = total reads in a sample / 1,000,000), which corrects for contig

length and differences in read number in the different samples. **Supplementary Table 3** presents a

complete list of contigs, their length, the encoded feature annotation, and the TPM value of the feature

in each sample. In many cases, one protein could be encoded by several contigs, either because gene

assembly into a contig was incomplete, or because of the existence of contig isoforms. Altogether, the

352,473 contigs encoded 225,965 unique proteins.

When analyzing the functional enrichment in the seven genera that predominated in the microbiota of

diseased oysters (see **Results** section), contigs that encoded the same protein were merged into a

single feature/protein, and their expression levels were summed. The expression level of a category

was calculated by summing up the expression level of all the proteins assigned to the category. In

addition, expression values were normalized from TPM to a value corrected for the relative proportion

of the genus activity in the sample such that the ratio of the average normalized expression level of the

three replicates for a given functional category in two conditions (*i.e.* T60 or 72 vs T0) reflected a

specific enrichment (Enrichment Factor or EF) of the function, rather than a general increase of the

genus in the population. First, a pseudo count of one read was added to each gene in each

condition/replicate, and the resulting counts were normalized by dividing by the total number of

counts of the genus in a given condition/replicate, then multiplying by 10,000. For a given

function/protein or functional category, an enrichment factor (EF) between the two conditions (i.e.

T60 or T72 vs T0) was further defined as the ratio of the average normalized expression level of the

three replicates at T60 or 72 over the average expression level at T0.

Statistical analyses

Survival curves were used to determine differential mortality kinetics between oyster families with the

non-parametric Kaplan-Meier test (Mantel–Cox log-rank test, $p < 0.05$, GraphPad_Prism 6.01). For

OsHV-1 and total bacteria quantifications, significant differences between resistant and susceptible

oyster families were determined using the non-parametric Mann Whitney test ($p < 0.05$,

GraphPad_Prism 6.01). For bacterial metabarcoding, statistical analyses were performed using R

v3.3.1 (http://www.R-project.org, [35]). Principal coordinate analysis (PCoA, "phyloseq") on a Bray-

Curtis distance matrix (ordinate, "phyloseq") was performed to determine dissimilarities between

samples. Multivariate homogeneity of group dispersions was tested between bacterial assemblages of

the six oyster families using 999 permutations (permutest, betadisper, "vegan"). DESeq2 ("DESeq",

[36]) from the OTUs to the higher taxonomic ranks was used to identify candidate taxa whose

abundance changed between the initial and final time points of the experiment. The abundance of a

bacterial taxon was considered significantly modified when the adjusted p-value for multiple testing

with the Benjamini-Hochberg procedure, which controls the false discovery rate (FDR), was < 0.05.

Heatmaps of significant genera were computed using relative abundances and the heatmap.2 function

"ggplots" [37]. For bacterial metatranscriptomics, significant differences in expression between two

conditions (i.e. T60 or T72 vs T0) were assessed by a paired Student's t-test ($p < 0.05$).

Data availability

Metabarcoding and RNAseq sequence data will be made available through the SRA database

(BioProject accession number PRJNA423079 with SRA accession SUB7738644 for bacterial

metabarcoding and SUB8110348 for Bacterial metatranscriptomic). Other data generated from this

study are included in the published version of this article and its supplementary files.

RESULTS

Primary OsHV-1 infection and secondary bacteremia are conserved in POMS, independent of

the infectious environment

Six *C. gigas* families were subjected to two experimental infections that mimicked disease

transmission in the wild. We previously reported high variability in the dynamics of mortality and

percentage survival of oyster families confronted with an Atlantic infectious environment.

Specifically, the F11, F14, and F15 families were highly susceptible (survival rate < 4% after 330h) to

POMS, whereas the F21, F23, and F48 families were highly resistant (survival rate > 82% after 330h)

[20]. Similar results were obtained in the present study when the same oyster families were confronted

with a Mediterranean infectious environment: families F11, F14, and F15 were susceptible (survival

rates < 9%), whereas families F21, F23, and F48 were resistant (survival rates > 88%) (**Figure 1**).

Thus, these oyster families displayed similar phenotypes when confronted with two different

infectious environments (Mantel-Cox log-rank test, $p < 0.0001$ for each comparison of resistant *vs.*

susceptible oyster families). Susceptible and resistant oyster families are hereafter referred to as S

(S_{F11}, $S_{F14,}$ and S_{F15}) and R (R_{F21}, R_{F23}, and R_{F48}), respectively.

Figure 1: Kaplan-Meier survival curves of oyster biparental families confronted with two
different infectious environments. Resistant oyster families (R_{F21}, R_{F23}, and R_{F48}) are presented in

green, and susceptible oyster families (S_{F11}, S_{F14}, and S_{F15}) are presented in red. At each time point
(indicated by asterisks on the arrow), 10 oysters were sampled 3 triplicates from each family in
each tank for barcoding, qPCR, and metatranscriptomic analysis. Data for the Atlantic infectious
environment are extracted from [20] and shown for comparison.

We next compared pathogenesis between the two infectious environments by monitoring OsHV-1

load, microbiota dynamics, and bacterial abundance in the three resistant and three susceptible oyster

families (**Figure 2**). OsHV-1 DNA was detected in all families, regardless of whether they were

confronted with the Atlantic or Mediterranean infectious environment (**Figure 2a**). However, very

intense viral replication occurred only in the susceptible oyster families: viral DNA loads were 2 to 3

logs higher than in resistant oysters at 24 h (**Figure 2a**).

Figure 2: Primary OsHV-1 infection, bacterial dysbiosis, and secondary bacteremia are conserved in different infectious environments. (a) Early and intense replication of OsHV-1 µVar occurs in susceptible oysters (red), but not resistant oysters (green), confronted with either the Atlantic or the Mediterranean infectious environment. OsHV-1 load was quantified by qPCR and expressed as Viral Genomic Units per ng of oyster DNA (log scale) during experimental infections. Asterisks indicate significant differences between susceptible and resistant oyster families (Mann Whitney test, $p < 0.05$). **(b-c)** Principal coordinate analysis (PCoA) plot of the microbiota for susceptible (red) and resistant (green) oyster families confronted with each infectious environment. Dispersion of oyster families according to the Bray-Curtis dissimilarity matrix (beta diversity) in **(b)** Atlantic and **(c)**

Mediterranean infectious environments. **(d)** Temporal dynamics of total bacteria in susceptible (red) and resistant (green) oyster families confronted with two different infectious environments. Total bacterial quantification based on qPCR amplification of the V3-V4 region of the 16S rRNA gene during experimental infections. Asterisks indicate significant differences between susceptible and resistant oyster families (Mann Whitney test, p < 0.05). Data from the Atlantic infectious environment in panels (a) and (d) are extracted from [20] for comparison.

The dynamics of the oyster microbiota were studied in the six oyster families by monitoring bacterial community composition using 16S rRNA gene metabarcoding over the first 3 days of both experimental infections. A total of 45,686 bacterial OTUs were obtained from the 252 samples and affiliated at different taxonomic ranks (**Supplementary Table 4**). Sufficient sequencing depth was confirmed by species richness rarefaction curves (**Supplementary Figure 2**). Changes in microbiota composition were greater in susceptible oysters than in resistant oysters at all taxonomic ranks (**Supplementary Figure 3**). Indeed, for the Atlantic infectious environment, 52, 43, and 54 OTUs significantly differed (in terms of relative proportion between the start and end of the experiment) in susceptible oysters S_{F11}, S_{F14} and S_{F15}, respectively; only 1, 11, and 9 OTUs significantly differed in resistant oysters R_{F21}, R_{F23} and R_{F48}, respectively (**Supplementary Table 5**). The same trend was observed in the Mediterranean infectious environment. 11, 47, and 43 OTUs significantly differed in S_{F11}, S_{F14} and S_{F15}, respectively, as opposed to 2, 8, and 6 OTUs in R_{F21}, R_{F23} and R_{F48}, respectively.

PCoA on a Bray-Curtis dissimilarity matrix (beta diversity) revealed higher microbiota dispersion in susceptible oyster families than in resistant families in both infectious environments (multivariate homogeneity of groups dispersion, d.f. = 1; $p = 0.016$ and $p = 0.020$ for Atlantic and Mediterranean environments, respectively) (**Figure 2b and 2c**). This disruption of the bacterial community structure occurred in susceptible oysters between 24 h and 48 h, concomitantly with the active replication of OsHV-1. In addition, susceptible oyster families displayed a significantly greater bacterial load than resistant oysters when confronted with either the Atlantic or the Mediterranean infectious environment (Mann Whitney test, p < 0.05; **Figure 2d**). This increase started at 60 h and continued to until the end of the experiment (72 h). Total bacterial abundance in susceptible oysters was more than 5-fold higher at 72 h than at T0, which indicated bacterial proliferation. In contrast, total bacterial load remained stable in resistant oysters.

A core of bacterial genera colonizes oysters during secondary bacterial infection in POMS

All bacterial genera that changed significantly in abundance during the two experimental infections (Atlantic and Mediterranean) in susceptible oyster families are reported in **Supplementary Table 5** (DESeq2, adjusted $p < 0.05$). We focused on genera representing $> 2\%$ of the bacteria in at least one sample for each susceptible oyster family confronted with each infectious environment (**Figure 3**). In the Atlantic infectious environment, the corresponding OTUs represented a total of 4%, 0.8%, and 46% of total bacteria at the beginning of the experiment (T0), as opposed to 73%, 75%, and 72% at 72 h for S_{F11}, S_{F14}, and S_{F15}, respectively (**Supplementary Table 5**). In the Mediterranean infectious environment, these OTUs increased from 2%, 6%, and 7% at T0 to 47%, 56%, and 56% at 72 h for S_{F11}, S_{F14}, and S_{F15}, respectively. Nine to twenty genera increased significantly in abundance between T0 and 72 h. Ten genera (*Arcobacter*, *Cryomorphaceae*, *Marinobacterium*, *Marinomonas*, *Proxilibacter*, *Pseudoalteromonas*, *Psychrilyobacter*, *Psychrobium*, *Psychromonas,* and *Vibrio*) were common to almost all (5 of 6) susceptible oyster families and both infectious environments (**Figure 3**). Most of the remaining genera (*Aquibacter*, *Aureivirga*, *Fusibacter*, *Neptunibacter*, *Peredibacter*, *Pseudofulbibacter*) were shared by at least two families in one infectious environment. One genera (*Salinirepens*) increased significantly in all susceptible oysters in the Atlantic infectious environment only. These results show that a core of bacterial genera colonizes oysters during the POMS secondary bacterial infection, independent of the infectious environment. In resistant oyster families, several taxa also vary significantly in abundance over time. Most of these taxa are the same taxa identified in susceptible oyster families (**Supplementary Figure 4**), but they are present in lesser quantities. These taxa represent between 4% to 23% of the reads sequenced at 72h in resistant oysters, whereas they represent 47% to 75% of the reads sequenced in susceptible oysters (**Supplementary Table 5**).

Figure 3: **Heatmaps of bacterial genera that changed significantly in abundance over the course of infection in susceptible oysters (S_{F11}, S_{F14}, and S_{F15}) in the Atlantic and Mediterranean infectious environments.** Analyses were performed at the genus level. Only genera that changed significantly in abundance (DESeq2, adjusted p < 0.05) and had a relative proportion greater than 2% in at least one sample are shown. Increased intensity of color (blue) represents increased relative abundance. Genera that are consistently modified in 5 out of the 6 conditions (3 families and 2 infectious environments) are in red.

Seven genera are responsible for most bacterial gene expression in diseased oysters

To understand the colonization success of certain genera, we analyzed the gene expression of the colonizing microbiota. As described in the **Methods** section, we used metatranscriptomics to determine the functional mechanisms underlying colonization for different bacterial genera in the three different oyster genetic backgrounds and two different infectious environments. RNAseq was performed after removal of rRNA, oyster mRNA, and mitochondrial RNA using custom probes. We sequenced the remaining RNA of the susceptible families S_{F11}, S_{F14}, and S_{F15}, in both the Atlantic and Mediterranean infectious environments, at T0 and just before oyster mortality occurred (*i.e.* at 60 h and 72 h for the Atlantic and the Mediterranean infectious environments, respectively). Three biological replicates were analyzed for each condition, corresponding to a total of 36 biological samples. After removal of the remaining non-bacterial and ribosomal reads, we assembled and

annotated a reference metatranscriptome. We aligned the translated contigs with the NCBInr protein database, and used MEGAN for taxonomic assignments. We obtained 352,473 contigs that aligned with bacterial proteins with an e-value $\leq 10^{-6}$. In many cases, a single protein could be encoded by several contigs, either because the gene was not fully assembled or because of contig isoforms. In total, our assembled metatranscriptome encoded 225,965 unique proteins. These protein were functionally and taxonomically annotated, and their expression level was determined for each condition. Further analyses were restricted to proteins assigned at the genus or species levels.

To analyze the contribution of each genus to the overall bacterial gene expression, we added the level of expression of all proteins assigned to a given genus (measured in Transcripts Per Million, or TPM). We focused on genera that contributed to more than 2% of total gene expression in at least one oyster sample (**Figure 4**).

Figure 4: Heatmap of transcriptional activity of bacterial genera in susceptible oyster families (S_{F11}, S_{F14}, and S_{F15}) in the two infectious environments before infection and in diseased oysters. For each condition, results of the three replicates are shown. Increased color intensity (blue) indicates

increased relative activity of the genus. Genera shown contributed at least 2% of the total
transcriptional activity in at least one sample of diseased oysters. Bacterial genera that were
overrepresented according to metatranscriptomics alone for all conditions in the diseased oysters are in
red, while genera that were overrepresented according to both metabarcoding and metatranscriptomics
are underscored. (Atl: Atlantic, Med: Mediterranean, T0: before infection, T60/72: diseased)

Seven genera were consistently found to contribute the majority of transcriptomic activity in diseased

oysters, displaying a strong increase in the number of transcripts compared to healthy oysters.

Amphitrea, Arcobacter, Marinobacterium, Marinomonas, Oceanospirillum, Pseudoalteromonas, and

Vibrio were together responsible for up to 40% of the total bacterial transcriptomic activity detected

just before the onset of oyster mortality. All are Gram-negative bacteria, and 6 out of 7 are

Gammaproteobacteria. Of the six Gammaproteobacteria, *Amphitrea, Marinobacterium, Marinomonas*

and *Oceanospirillum* belong to the same family (Oceanospirillaceae) and order (Oceanospirillales).

Arcobacter belongs to the class Epsilonproteobacteria. The genus *Fluviicola*, which belongs to the

phylum *Bacteroidetes*, presented an increase of transcriptomic activity in the Atlantic infectious

environment only. Finally, the genus *Oleiphilus (*order Oceanospirillales), was represented at a

maximum level of 1.9 % in diseased oysters.

These results indicate that a limited number of genera participate in the secondary bacteremia that

occurs in POMS. These genera are remarkably conserved between the different susceptible oyster

families and infectious environments. We therefore focused on these seven genera for additional

study.

Secondary bacterial colonizers display genus-specific strategies

We focused on the late phase of the infection process in our examination of the functions underpinning

successful colonization by the seven genera discussed above. 31 functional categories were defined,

and each transcript-encoded protein was assigned to one of them (**Supplementary Table 2**). For each

genus and each category, an enrichment factor (EF) was calculated relative to the general increase of

the genus in the bacterial population (See Materials and Methods for details, **Figure 5**).

Figure 5. **Functional enrichment between T0 and the onset of oyster mortality.** The enrichment factor (EF) was calculated for each functional category for the seven most active genera present in all families and both environments. Each overrepresented (red) or underrepresented (blue) functional category is indicated by a bar with a width proportional to log2(EF). Red boxes around bars indicate p ≤ 0.05 (Student's t-test). Green-filled cells indicate functional categories with no gene expression (genes absent from the genome in that genus or not expressed in any conditions). Light blue-filled cells indicate values lower than the 20th percentile, while yellow-filled cells indicate values higher than the 20th percentile.

For all genera except *Arcobacter,* translation and transcription appeared to be globally overrepresented

in diseased oysters. Aside from these functions, we found that each colonizing genus displayed a

specific pattern of functional enrichment, which was generally conserved across oyster families and

environments. For instance, adhesion, cell defense, motility, natural competence, quorum sensing,

stress response, and virulence gene expression were either over- or under-represented according to the

genus considered. These results suggest that each genus adopts a specific strategy to ensure

colonization success.

Enriched functions are potentially important for successful colonization

In a global metatranscriptomic analysis, important overrepresented transcripts/subcategories could be

hidden by the under-expression of genes classified as being in the same functional categories.

Therefore, we next focused our analysis on overrepresented transcripts (listed in **Supplementary**

Table 6), and identified several overexpressed genes that are likely to contribute to efficient

colonization. **Table 1** summarizes the distribution of these overexpressed genes/subcategories within

the seven genera.

Table 1. Specific enrichment of selected functions in the seven genera. Overexpression of function/gene in a given genus is indicated by a filled out green cell. E.C. numbers are indicated whenever applicable. Functions that are enriched in all seven genera are in boldface. The relevant information was extracted from supplementary table S6.

Category	Role/pathways	Enriched functions/genes	E.C.. Number	Amphitrea	Arcobacter	Marinobacterium	Marinomonas	Oceanospirillum	Pseudoalteromonas	Vibrio
Cell defense										
	Arsenic detoxification	Arsenate reductase	1.6.5.2			🟩	🟩		🟩	
	Cyanide detoxification	Rhodanese like protein	2.8.1.1			🟩	🟩		🟩	🟩
	Formaldehyde detoxification	Formaldehyde dehydrogenase, glutathione-independent	1.2.98.1	🟩		🟩	🟩			
		Hydroxyacylglutathione hydrolase	3.1.2.26			🟩				
		S-formylglutathione hydrolase	3.1.2.12			🟩				
	Drug resistance	Aminoglycoside phosphotransferase	2.7.1.-			🟩			🟩	
		RND family efflux transporter						🟩		
		TolC family outer membrane protein			🟩		🟩		🟩	
	Hg resistance	MerR family DNA-binding transcriptional regulator						🟩		
Inorganic ion										
	Calcium transport	Ca2+ transporter				🟩			🟩	
	Potassium transport	K+ transport				🟩		🟩		
	Sodium transport	Na+/H+ antiporter NhaC				🟩	🟩			
	Phosphonate utilisation	Alkyl phosphonate utilization PhnA			🟩					
		Phosphonate transporter		🟩		🟩	🟩			
		Alkaline phosphatase	3.1.3.1	🟩		🟩		🟩		
	Pi homeostasis	inorganic diphosphatase		🟩		🟩				
		PhoH (Pi starvation)				🟩		🟩		
		Pi transport					🟩			
		Polyphosphate kinase 2	2.7.4.1			🟩			🟩	
	Selenate transport	Selenate transporter			🟩					
Metal homeostasis										
		Fe(3+) ABC transporter substrate-binding protein			🟩	🟩	🟩		🟩	
		Ferric iron uptake transcriptional regulator		🟩		🟩				
		Imelysin					🟩			
	Iron homeostasis	Bacterioferritin			🟩				🟩	
		Ferritin					🟩			
		Siderophore production					🟩			
		TonB-dependent siderophore receptor			🟩		🟩			
	Copper homeostasis	Copper-translocating P-type ATPase					🟩			
	Metal homeostasis	Heavy metal (Co2+, Zn2+, W...) transporter			🟩		🟩			
		Metal transporter					🟩			
Nitrogen compound metabolism										
	Nitrosative stress	Hydroxylamine reductase	1.7.99.1		🟩					🟩
Precursor metabolite and energy production and conversion										
	ATP synthase	**ATP synthase**	7.1.2.2	🟩	🟩	🟩	🟩	🟩	🟩	🟩
		cytochrome C		🟩	🟩	🟩	🟩	🟩	🟩	🟩
		cytochrome C oxidase	7.1.1.9	🟩	🟩	🟩	🟩	🟩	🟩	🟩
		Na(+)-translocating NADH-quinone reductase	7.2.1.1	🟩	🟩	🟩	🟩	🟩	🟩	🟩
		cytochrome B		🟩	🟩	🟩	🟩	🟩	🟩	🟩
		cytochrome o ubiquinol oxidase	7.1.1.3	🟩		🟩	🟩		🟩	🟩
	Electron transfer	electron transfer flavoprotein		🟩	🟩					
		electron transfer flavoprotein-ubiquinone oxidoreductase	1.5.5.1	🟩						
		hybrid-cluster NAD(P)-dependent oxidoreductase		🟩	🟩					
		NADH-quinone oxidoreductase	7.1.1.2	🟩						
		pyruvate flavodoxin oxidoreductase	1.2.7.1	🟩	🟩					
		ubiquinol oxidase subunit II	7.1.1.7	🟩						
		ubiquinol-cytochrome c reductase	1.10.2.2	🟩						
	Acetyl-CoA biosynthesis	acetate--CoA ligase	6.2.1.1							🟩
		pyruvate dehydrogenase	1.9.6.1	🟩						
	Anaerobic respiration	fumarate reductase	1.3.5.4		🟩					
		hydrogenase small subunit	1.12.5.1		🟩					
		fumarate/nitrate reduction transcriptional regulator Fnr				🟩		🟩		

Mixed acid fermentation	formate dehydrogenase	1.17.5.3	█	█	█	█	█		
	phosphoenolpyruvate synthase	2.7.9.2						█	
Glycolysis / gluconeogenesis	pyruvate kinase	2.7.1.40		█	█	█			█
	2,3-bisphosphoglycerate-dependent phosphoglycerate mutase	5.4.2.11			█	█			
	fructose-bisphosphatase	3.1.3.11		█					
	fructose-bisphosphate aldolase class II	4.1.2.13	█	█	█	█	█	█	█
	glucose-6-phosphate isomerase	5.3.1.9	█	█					
	glyceraldehyde-3-phosphate dehydrogenase	1.2.1.12	█	█	█	█			
	phosphoglycerate kinase	2.7.2.3			█	█	█	█	
	phosphopyruvate hydratase	4.2.1.11			█		█	█	█
	triose-phosphate isomerase	5.3.1.1			█	█	█	█	█
Glyoxylate shunt	isocitrate lyase	4.1.3.1			█	█	█	█	█
	CoA ester lyase (malate synthase)	2.3.3.9		█					
	malate synthase	2.3.3.9	█	█		█	█	█	█
L-alanine fermentation	acryloyl-CoA reductase	1.3.1.84		█					
Mixed acid fermentation	acetate kinase	2.7.2.1			█				
	acetoacetyl-CoA reductase	1.1.1.35			█				
	formate C-acetyltransferase	2.3.1.54							█
	phosphoenolpyruvate carboxylase	4.1.1.49				█			
Nitrate respiration	ferredoxin-type protein NapG	7.1.1.8		█				█	
	nitrite reductase	1.7.2.1		█					
	nitrous-oxide reductase	1.7.2.4			█				
	Ammonia forming cytochrome c nitrite reductase	1.7.2.2		█					
	Nitrate reductase NapA	1.9.6.1		█				█	
Pentose phosphate pathway	glucose-6-phosphate dehydrogenase	1.1.1.49						█	
	NADP-dependent phosphogluconate dehydrogenase	1.1.1.44					█		
	ribose-5-phosphate isomerase RpiA	5.3.1.6			█	█			
	ribulose-phosphate 3-epimerase	5.1.3.1	█						
	transaldolase	2.2.1.2						█	
Photosynthesis	phosphoribulokinase	2.7.1.19			█				
PRPP biosynthesis	ribose-phosphate pyrophosphokinase	2.7.6.1	█		█	█	█	█	█
Respiration	Lactate dehydrogenase or related 2-hydroxyacid dehydrogenase	1.1.-.-			█				
	Ldh family oxidoreductase	1.1.5.12			█				
	Aerobic respiration control protein arcA								█
TCA cycle	2-oxoglutarate dehydrogenase	1.2.4.2		█	█	█	█	█	
	2-oxoglutarate synthase	1.2.7.3		█					
	aconitate hydratase	4.2.1.3					█	█	█
	citrate synthase	2.3.3.1/2.3.3.16	█	█	█				█
	dihydrolipoamide dehydrogenase	1.8.1.4		█	█	█	█		
	fumarate hydratase	4.2.1.2		█			█		
	isocitrate dehydrogenase	1.1.1.42					█		
	NADP-dependent isocitrate dehydrogenase	1.1.1.42	█	█		█			
	succinate dehydrogenase	1.3.5.1	█		█	█	█	█	█
	succinate--CoA ligase	6.2.1.5		█	█	█	█	█	█

Protein export and secretion system

Enzyme secretion	Type 2 SS					█		█	
EPS production-associated protein export	Exosortase						█		
Folded protein export	Tat system						█		
Inner membrane protein translocation	Sec system					█	█	█	█
Toxin secretion	Type 1 SS				█	█			█
Virulence factor secretion / competition	Type 6 SS					█			█

Redox homeostasis and oxidative stress

Category	Subcategory	Enzyme/Protein	EC	G1	G2	G3	G4	G5	G6	G7
Oxidative stress		Peroxiredoxin		●	●	●	●	●	●	●
		Alkyl hydroperoxide reductase	1.11.1.26				●		●	
		Catalase	1.11.1.6				●		●	
		Disulfide reductase		●						
		Glutamate--cysteine ligase	6.3.2.2	●			●		●	
		Peroxidase	1.11.1.7	●	●					
		Superoxide dismutase	1.15.1.1	●	●				●	●
Redox homeostasis		Glutaredoxin		●			●			
		NAD(P) reductase		●			●			
Redox homeostasis and oxidative stress		Thioredoxin				●		●		●

Stress response

Category	Subcategory	Enzyme/Protein	EC	G1	G2	G3	G4	G5	G6	G7
Cold stress		Cold-shock protein		●	●	●	●	●	●	●
Starvation/stationary phase		Ribosome-associated translation inhibitor RaiA		●	●	●	●	●	●	●
Acid stress		BolA family transcriptional regulator					●		●	
Extracytoplasmic stress		Phage shock protein PspA					●		●	
		RIP metalloprotease RseP		●			●		●	
		RNA polymerase sigma factor RpoE		●						
Fe-S cluster repair		SufS family cysteine desulfurase		●			●		●	
General stress response		Universal stress protein		●		●		●		
Heat-shock		Heat-shock protein			●					
		RNA polymerase sigma factor RpoH		●			●		●	●
Nitrosative stress		Nitric-oxide reductase	1.7.2.5	●			●			
		NO-inducible flavohemoprotein						●		
Osmotic stress: osmoprotectant		Choline ABC transporter substrate-binding protein		●						
		Choline dehydrogenase	1.1.99.1				●		●	
		Dimethylglycine demethylation protein DgcA					●		●	
		Glycine betaine/L-proline ABC transporter ATP-binding protein ProV					●			
		ProQ/FinO family protein					●		●	
		Transcriptional regulator BetI					●		●	
Signal transduction		Symmetrical bis(5-nucleosyl)-tetraphosphatase					●		●	
SOS response		Transcriptional repressor LexA					●		●	
Starvation/ stationary phase		Stringent starvation protein A					●		●	
		RNA polymerase sigma factor RpoS		●			●			
Stringent response		Bifunctional (p)ppGpp synthetase/guanosine-3,5-bis					●		●	
		GTP diphosphokinase RelA		●			●		●	
		RNA polymerase-binding protein DksA		●	●		●			
Sugar phosphate stress		Methylglyoxal synthase								●
Toxin/antitoxin		Toxin/antitoxin system						●		●

Sulfur compound metabolism

Category	Subcategory	Enzyme/Protein	EC	G1	G2	G3	G4	G5	G6	G7
Cysteine degradation		D-cysteine desulfhydrase	4.4.1.15				●		●	
Sulfate reduction		Sulfate adenylyltransferase subunit CysD	2.7.7.4				●			
		HTH-type transcriptional regulator CysB					●		●	
Sulfite oxidation		Sulfite dehydrogenase	1.8.2.1		●					
Sulfite transport		Sulfite exporter TauE/SafE family protein						●		
Sulfolactate utilisation		(S)-sulfolactate dehydrogenase	1.1.1.310							●
		Sulfoacetaldehyde acetyltransferase	2.3.3.15			●				
Sulfur oxidation		Flavocytochrome c sulfide dehydrogenase	1.8.2.3		●					
Thiosulfate oxidation		Sulfur oxidation c-type cytochrome SoxA	2.8.5.2	●	●					
		Sulfur oxidation protein SoxCD, sulfur dehydrogenase subunit	1.8.2.6	●						

Virulence/ fitness

Category	Subcategory	Enzyme/Protein	EC	G1	G2	G3	G4	G5	G6	G7
Toxin production		Putative RTX toxin/ hemolysin				●			●	●
		Transcriptional activator HlyU							●	

High metabolic activity, revealed by a series of functional categories, was observed in all genera. In the "Translation" category, genes encoding ribosomal proteins, elongation and initiation factors, and tRNA ligases were overexpressed for all seven genera. In the "Precursor metabolite and energy production and conversion (Energy)" category, genes encoding ATP synthase and enzymes involved in glycolysis/neoglucogenesis and the TCA cycle were systematically present, as were genes encoding proteins of the respiratory chains (e.g. cytochrome-c and cytochrome oxidases). In the "Transcription and RNA metabolism (RNA)" category, genes encoding the various subunits of the RNA polymerase were overexpressed, as was the transcription housekeeping sigma factor RpoD (**Supplementary Table 6**).

Functions important for stress adaptation and resistance to host defenses varied over the course of

pathogenesis according to the bacterial genera considered (**Table 1**). "Stress response" genes encoding

cold-shock proteins and the ribosome-associated translation inhibitor RaiA were overexpressed by all

genera. The gene encoding the stress response RNA polymerase Sigma factor RpoS was

overexpressed in members of the Oceanospirillaceae family *(Amphritea Marinobacterium,*

Marinomonas, and *Oceanospirillum).* In the "redox homeostasis and oxidative stress" category,

peroxiredoxin genes were overexpressed by all genera, along with additional genus-specific functions

related to fighting oxidative stress (superoxide dismutases, catalases, and peroxidases) (**Table 1**). In

addition, all genera except *Oceanospirillum* overexpressed genes that included several nitrite and

nitrate reductases (classified in "Precursor metabolite and energy production and conversion"),

hydroxylamine reductases (classified in "Nitrogen compound metabolism"), and nitric-oxide reductase

components (classified in "Stress response/ Nitrosative stress"), all of which could contribute to

tolerance/resistance to reactive nitrogen species (RNS). RNS, like ROS, are important effectors of

oyster immunity [38].

"Cell defense", "Metal homeostasis" and "Virulence/ fitness" categories were more specifically

encoded by a subset of bacterial genera. In the "Cell defense" category, multidrug resistance efflux

pumps with a TolC component and detoxifying enzymes were overexpressed in some genera. For

instance, an outer membrane channel protein TolC was overexpressed up to 16-fold ($log2(EF) \geq 4$) in

Pseudoalteromonas (**Supplementary Table 6**). Genes encoding Rhodanese, an enzyme that detoxifies

cyanide, were overexpressed in *Marinobacterium*, *Pseudoalteromonas,* and *Vibrio*. Several arsenate

reductase genes were also overexpressed in *Marinobacterium, Marinomonas*, and *Pseudoalteromonas*.

Genes encoding formaldehyde dehydrogenase, an aldehyde-detoxifying enzyme, were overexpressed

in the Oceanospirillaceae. Notably, no cell defense genes were overexpressed in *Arcobacter*. In the

"Metal homeostasis" category, 23 of 33 overexpressed genes were involved in iron homeostasis.

These genes are mainly expressed by *Marinobacterium* and *Marinomonas*. In *Vibrio*, the only up-

regulated gene of the category is annotated as a cobalt/magnesium transporter. In the "Virulence /

fitness" category, a small number of overexpressed genes corresponded to putative toxins encoded by

Arcobacter, *Oceanospirillum, Pseudoalteromonas,* and *Vibrio*. This has to be completed by genes

from the "Protein export and secretion system" category, which are important for both virulence and

interbacterial competitions. Genes encoding the general protein export system (i.e. the "Sec" system of

Gram-negative bacteria [39, 40]) were overexpressed by all genera, while type VI secretion systems

were overexpressed specifically in *Vibrio* (e.g. *V. crassostreae*) and *Marinomonas.*

Finally, a global increase in the "Sulfur compound metabolism" category was observed in

Marinobacterium (**Figure 5**). Overexpressed proteins of this category were also found in other genera

(**Table 1** and **Supplementary Table 6**). Interestingly, these overexpressed proteins corresponded to

enzymes that were (mostly non-redundantly) encoded by five different genera. Altogether, these

enzymes constitute several sulfur metabolic pathways (**Figure 6**).

These results highlight functional complementation at the level of the bacterial community, and a

metabolic adaptation to the within-host environment, both of which are likely important for successful

oyster colonization.

Figure 6: Integrated sulfur cycle in the diseased oyster microbiota. Overexpressed sulfur metabolism
pathways are shown. Enzymatic reactions are represented by arrows that are colored according to the
pathway. The Enzyme Commission (E.C.) number, corresponding gene name, and genera overexpressing
the gene for the enzyme catalyzing each reaction is provided in the grey rectangle beside the relevant
arrow.

DISCUSSION

In this study, we have shown that the dysbiosis associated with POMS is conserved across infectious

environments. Using metabarcoding, we demonstrated that diseased oysters affected by POMS are

colonized by a common consortium of bacteria comprising ten major genera (*Arcobacter*,

Cryomorphaceae, *Marinobacterium*, *Marinomonas*, *Proxilibacter*, *Pseudoalteromonas*,

Psychrilyobacter, *Psychrobium*, *Psychromonas,* and *Vibrio*). Using metatranscriptomics, we showed

that five of these genera (*Arcobacter, Marinobacterium, Marinomonas, Vibrio,* and

Pseudoalteromonas) were particularly transcriptionally active.

The bacterial consortium associated with POMS was remarkably similar in three different susceptible

oyster families and two different environments. All diseased oyster families harbored similar

microbial communities, with only a few exceptions. Remarkably, the oyster-associated microbial

communities were conserved across infectious environments, although the Mediterranean and Atlantic

ecosystems are likely to host diverse bacterial communities.

Until recently, only members of the *Vibrio* genus had been repeatedly associated with POMS. These

studies used culture-based approaches to investigate oyster-associated bacterial communities (Lemire

et al., 2015; Bruto et al., 2018). *Vibrio* species associated with POMS were characterized by key

virulence factors that are required to weaken oyster cellular defenses (Rubio et al., 2019; Piel et al.,

2019). Members of the *Arcobacter* genus had also been associated with POMS-diseased oysters [41,

42], but the role of this genus in pathogenesis was not investigated more deeply due to limitations of

culture-based techniques [43]. In the present study, we extended the core colonizing microbiota to five

bacterial genera. The discovery of the contributions of these genera, which are responsible for up to

40% of the bacterial transcriptional activity observed in POMS, provides new insights into the

pathogenesis of POMS. Altogether, our results strongly suggest that a colonizing core microbiota,

rather than specific bacterial pathogens, operates as a functional unit of pathogenesis. Together with

OsHV-1, these bacteria form the POMS pathobiota. POMS secondary bacteremia may resemble

periodontitis in humans, in which the evolution of the disease is characterized by the development of a

pathogenic consortium comprising a limited number of species [44, 45].

We used metatranscriptomics to unveil the functions of the microbiota in relation to POMS. Bacterial

metatranscriptomics from host tissues is technically challenging (due to the low proportion of bacterial

transcripts in the host samples), but it provides functional information that is thought to more

accurately portray the role of the microbiota in health and disease states [46]. Accordingly, gene

expression profiling has proven highly successful in advancing the understanding of the dynamics of

disease-associated microbial populations [47]. In the case of POMS, by linking functional genes to the

bacterial genera which encode them, we found a remarkably consistent relationship between the

structure of bacterial communities (through 16S metabarcoding) and the functions expressed by

bacterial genera in the communities (through metatranscriptomics), further supporting our hypothesis

of a pathobiota composed of at least five key genera.

Among functions overexpressed in the diseased oyster-associated microbiota, we first identified core

functions which were overrepresented in all dominant genera (**Table 1**). In addition to housekeeping

functions with expression levels that reflect the rapid growth of the genus (e.g. translation and

transcription, central metabolism), we found functional redundancy in two important categories

required for adaptation to the within-host environment: stress response and redox homeostasis.

We found a systematic and important overproduction of various cold shock proteins in the "Stress

response" functional category (**Supplementary Table 6**). In the diseased oyster microbiota, these

proteins may be important to sustain the high level of transcription and translation we observed [48,

49]. Redox homeostasis was another core function. Fighting oxidative stress in the host can be crucial

to survival and multiplication within the host. Thus, bacteria have many lines of defense against

oxidative damage [50]. Key enzymes and proteins involved in redox homeostasis and oxidative stress

response were overexpressed in several of the five genera, including superoxide dismutases (which

catalyze the dismutation of superoxide in hydrogen peroxide) and catalases and peroxidases (which

destroy hydrogen peroxide and other toxic peroxide compounds). All five genera overexpressed

redoxin, a small redox protein which plays important roles in maintaining an intracellular reducing

environment, and also acts as a cofactor in intracellular oxidoreductase reactions. Other proteins

important for fighting oxidative stress were specific to one or more given genera. For example,

NAD(P) reductase (which contributes to the homeostasis of reduced NAD(P)H, an important source of

reducing potential) was enriched in *Amphitrea, Marinobacterium, Marinomonas,* and

Oceanospirillum, four members of the Oceanospirillaceae family. Glutamate – cysteine ligase (which

is responsible for the production of Glutathione, an important antioxidant compound, and is also

involved in detoxifying reactions) was overexpressed in *Amphitrea, Oceanospirillum*, and

Pseudoalteromonas. Earlier functional studies have identified the AhpCF complex detoxifying

enzyme alkylhydroxyperoxydase as a key determinant of resistance to ROS and thus oyster

colonization in *Vibrio* [51]. In the diseased oyster microbiota, this enzyme was over-produced in

Marinomonas and *Oceanospirillum*. All of the successful genera, except *Oceanospirillum*, also

expressed enzymes required for tolerance/resistance to RNS (**Table 1**). ROS and RNS are major

players at the oyster-microbiota interface (Destoumieux-Garzon et al., 2020), thus it is not surprising

that successful colonizers express ROS detoxifying enzymes during oyster colonization. In summary,

each genus overexpressed several proteins involved in stress response, especially oxidative stress,

although each genus had its own repertoire of proteins.

We found that the bacterial community associated with POMS also expresses a series of genus-

specific (variable) functions. These functions include adhesion, cell defense, cell motility, metal

homeostasis, natural competence, quorum sensing, sulfur metabolism, transport, and virulence (**Table

1**). The observation that functions of interest may be contributed by a single or multiple bacterial

genera within a microbial community has also been made in the human gut [46, 52]. Interestingly, a

function expressed by one genus can be useful for the community as a whole. Functional

complementarity within the oyster-associated microbiota may therefore contribute to synergy between

genera.

Evidence of cross-benefits at the community level in the oyster microbiota was found in the form of

overexpressed genes related to iron homeostasis (**Supplementary Table 6**). The capacity to scavenge

iron from the environment and the host is a well-known determinant of pathogenicity [53, 54]. Except

for *Vibrio*, all genera overexpressed functions related to iron acquisition (e.g. iron siderophore

production and transport, which enable the chelation and uptake of iron). However, in the consortium,

only *Marinomonas* and *Marinobacterium* expressed siderophore-production enzymes, while other

members of the microbiota expressed the receptors for iron-siderophore uptake. Similarly, virulence

factors (such as secreted RTX family toxins) were expressed by *Vibrio* and *Arcobacter*, while the

Type 6 Secretion System (T6SS) was expressed by *Vibrio* and *Marinomonas*. T6SS expressed by

Vibrio has been shown to contribute to the suppression of oyster cellular defenses [21, 55]. Therefore,

their expression by a few genera is likely beneficial to the whole pathobiota.

The best example of cross-benefits, however, is the "Sulfur compound metabolism" functional

category. Marine *Arcobacter* species that can oxidize sulfide to elemental sulfur, producing sulfur

filaments, have been previously identified [56]. In our consortium, corresponding sulfide

dehydrogenase enzymatic activity (E.C. 1.8.2.3) was encoded by *Arcobacter* only, while other sulfur

compound metabolic pathways were associated with four other genera (mostly *Marinomonas* and

Marinobacterium) (**Figure 6**). This suggests that sulfur cycling is a property of the colonizing core

microbiota as a whole, and not of a single genus. A similar association between *Arcobacter* and

Marinobacter species has been described in biogeochemical cycling, specifically in the environment

of natural gas wells [57]. We believe that such interdependence and cooperation within microbial

communities allows these communities to meet metabolic requirements and mount an efficient stress

response, which are key determinants of their success. We further believe that this helps to explain the

conserved pathobiota structure associated with POMS across distinct environments and oyster genetic

backgrounds.

Supplementary Materials

Supplementary Table 1: Total raw reads (R1+R2) at each stage of biocomputing, after sequencing and

removal of rRNA reads (eukaryotic and bacterial), oyster reads, and viral reads.

Supplementary Table 2: List of functional categories defined in bacterial metatranscriptomics.

Supplementary Table 3: Complete list of contigs identified in bacterial metatranscriptomics, with

annotations and expression value in TPM. Taxons in red correspond to assignation at taxonomic ranks

other than the genus level.

Supplementary Table 4: Absolute abundance of Operational Taxonomic Units (*ie.* clusters) and their

corresponding taxonomic affiliations in susceptible and resistant oyster families confronted with two

different infectious environments. Susceptible oyster families are S_{F11}, S_{F14}, and S_{F15}; resistant oyster

families are R_{F21}, R_{F23}, and R_{F48}. A indicates the Atlantic infectious environment, M the Mediterranean

infectious environment. T0, T6, T12, T24, T48, T60, and T72 indicate sampling times (in hours) over

the course of experimental infection. R1, R2, R3 indicate the results of each replicate (Excel file).

Supplementary Table 5: Frequencies of bacterial taxa that change significantly in abundance over the

course of each experimental infection (Atlantic or Mediterranean) in susceptible and resistant oyster

families. The change in abundance of bacterial taxa between initial and final time points was

determined using DEseq2 with the adjusted p-value < 0.05. Susceptible oyster families are SF11,

SF14, and SF15; resistant oyster families are RF21, RF23, and RF48. A indicates the Atlantic

infectious environment, M the Mediterranean infectious environment. T0, T6, T12, T24, T48, T60

and T72 indicate sampling times (in hours) over the course of experimental infection. R1, R2, R3

indicate the results of each replicate (Excel file).

Supplementary Table 6: List of the overexpressed (enriched) proteins in the seven main genera found

to colonize diseased oysters. A protein was considered enriched when it had a $\log2(EF) \geq 1$ with a p-

value ≤ 0.05 (Student's t-test) in at least two conditions.

Supplementary Figure 1: Schematic representation of the steps of metatranscriptomic data analysis.

Supplementary Figure 2: Rarefaction curves of the sub-sampled 16S rDNA dataset (10,000 reads per

sample) for susceptible and resistant oysters in (a) the Atlantic infectious environment (AS_{F11}, AS_{F14},

AS_{F15} and AR_{F21}, AR_{F23}, $AR_{F48)}$ and (b) the Mediterranean infectious environment (MS_{F11}, MS_{F14},

MS_{F15} and MR_{F21}, MR_{F23}, MR_{F48}). T0, T6, T12, T24, T48, T60, and T72 indicate sampling times (in

hours) over the course of experimental infection. Results are shown in triplicate for each time point.

Supplementary Figure 3: Microbiota modification as analysed using 16S rDNA metabarcoding in

susceptible and resistant oyster families confronted with two different infectious environments.

Susceptible oyster families (S_{F11}, S_{F14} and S_{F15}) and resistant oyster families (R_{F21}, R_{F23} and R_{F48})

confronted with (**a**) Atlantic or (**b**) Mediterranean infectious environments. Significant changes in

abundance (up or down; DESeq2, p-value < 0.05) between the initial and the final time point of the

infection were much greater for each taxonomic rank (from the phylum to the OTU rank) for

susceptible oyster families than for resistant oyster families. Data for AS_{F11} and AR_{F21} were extracted

from [20].

Supplementary Figure 4: Heatmaps of bacterial communities that changed significantly in abundance

over the course of infection in resistant oysters (R_{F11}, R_{F14}, R_{F15}) in the Atlantic and Mediterranean

infectious environments. Analyses were performed at the genus level. Only genera with a relative

proportion greater than 2% in at least one sample are shown. Increased color intensity (blue) indicates

increased relative abundance of the genus.

END NOTES

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Author contributions

J.D.L., B.P., A.J. and G.M. designed experiments. B. P., J.D.L, A.L., J.M.E., Y.G., L.D. and G.M.

performed oyster experiments. J.D.L., A.L., E.T., C.C and G.M. performed microbiota analyses.

J.D.L. and A.L. performed qPCR analyses. A.J. and X.L performed the metatranscriptomic

experiments. A.J. and S.M. analyzed the metatranscriptomic data. J.D.L., A.L., A.J, S.M. and G.M.

interpreted results. J.D.L., A.L., A.J., D. D. G. and G.M. wrote the manuscript, which has been

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