Expansion, restructuring and characterization of the *Legionellaceae* family

Marco Gabrielli[1,*], Alessio Cavallaro[1], Frederik Hammes[1]

[1] Department of Environmental Microbiology, Eawag, 8600 Dübendorf, Switzerland

*Corresponding author.

Name: Marco Gabrielli

Tel: +41 58 765 5960

Email: marco.gabrielli@eawag.ch

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Abstract

Legionellaceae are a large and widespread family of facultative intracellular bacteria with high

clinical relevance. While *Legionella pneumophila*, the most clinically relevant species, is relatively

well studied, very limited information is available regarding the other species belonging to this

family. Here, we analyzed all publicly-available *Legionellaceae* genomes and metagenome-

assembled genomes, nearly doubling the number of recognized species and highlighting the need

to restructure the family's taxonomy to include multiple genera that present distinct clinical

relevance. We found that secretion systems and traits linked to host invasion vary significant across

genus clusters and species, likely influencing their survival in their environment. While genus

clusters presented different metabolic capabilities, species closely related to *L. pneumophila* were

characterized by a more complete metabolism with fewer auxotrophies. Finally, we showed that

while species have distinct environmental distributions, niche overlap increases at higher metabolic

similarity. Together, our results shed light on the ecology of these microorganisms, highlight the

diversity of traits that can occur in closely related facultative intracellular bacteria, and call for the

restructuring of this clade to align with genomic information and clinical relevance to aid the

management of *Legionellaceae* bacteria.

Significance Statement

Legionellaceae bacteria are responsible for Legionnaires' disease, the leading waterborne disease

in many industrialized countries. Historically, attention was focused on *Legionella pneumophila* and

the ecology of other species remained largely unknown and unexplored. However, several water

safety regulations cover the entire *Legionella* genus, currently the only recognized within this large

family, including microorganisms with unknown health relevance. Using a comparative genomics

approach and leveraging public genomes and amplicon data, we shed light on the scale, diversity

and ecology of this family, finding a surprisingly diversity of the traits linked to host invasion and

metabolism, likely driving the species' environmental distributions. This study proposes to

restructure the family *Legionellaceae* in multiple genera to align with known clinical associations

and predicted phenotypes.

Main Text

Introduction

Microbial ecology largely relies on conventional short-read 16S rRNA sequencing and specific

model organisms to infer microorganisms' environmental distribution and phenotypic

characteristics. While this combination has enabled important insights into the ecology of several

microorganisms and environments(1, 2), the limits imposed by amplicon sequencing taxonomic

resolution(3) and the restricted number of model organisms available(4) hamper our understanding

of the ecological differences at lower taxonomic levels (i.e., species, strains)(5, 6). This results in a

blurred picture of the ecology of several species, including those of clinical relevance. When only

some of the species within a clade are of interest, the average response of the clade may contrast

with the response of the individual species, as in Simpson's paradox(7). As a consequence, the

data obtained would not properly inform legislation and management practices.

A relevant example of this problem lies in the family *Legionellaceae* and its only recognised genus,

Legionella, a clade of gram-negative facultative intracellular bacteria. Multiple *Legionella* species

can cause legionellosis, a disease responsible for a significant number of deaths, productivity

losses and medical expenses exceeding 800 million USD per year in the USA alone(8) and for

which the reported incidence is rising worldwide(9). While 73 species are currently recognized(10),

L. pneumophila is alone responsible for more than 90% of infections in Europe and the USA(11).

This contrast between the large number of species and the clinical importance of *L. pneumophila*

results in different legislative actions, with countries differing both regarding the regulatory target

(e.g., only *L. pneumophila*, the entire genus, a combination of the two)(12, 13), as well as the

regulated exposure source (e.g., showers, drinking water, cooling towers)(14). While the choice of

a broader regulatory target follows the precautionary principle, an overly broad target can conflate

species with different phenotypes and hamper identification of truly critical situations. This is likely

the case for *Legionellaceae*, given that most scientific studies focus on *L. pneumophila* and the

knowledge regarding other species lags far behind. In fact, most of the information available on

non-*L. pneumophila* species has been obtained from comparative genomic approaches. These

studies uncovered significant heterogeneity in terms of genome organization(15), encoded

effectors(16) and secondary metabolites(17). However, information on the factors which affect their

interactions with the environment, such as motility, secretion systems and metabolism, is still

lacking. Moreover, the aforementioned studies focus only on cultured isolates, while recent

molecular-based ecological surveys have uncovered a significant uncultured diversity of

Legionellaceae bacteria(18), a fact that highlights the need to include uncultivated species to

properly discern the evolutionary paths and the derived ecological differences. Understanding the

ecological differences among different *Legionellaceae* species has direct impact on the

management and legislation of these species(12).

Here we retrieved over 5000 publicly available *Legionellaceae* genomes and metagenome-

assembled genomes and compared the functional potential of the different species included in this

family, focusing specifically on evaluating its genetic diversity and different traits important for

facultative intracellular bacteria, including secretion systems, motility and host invasion, and

metabolism. We complemented this analysis with a genome-informed mining of 16S rRNA surveys,

assessing species' distribution across environments. Collectively, this work provides key insights

regarding the diversity and the ecology of *Legionellaceae* bacteria which will allow refining

legislation and management strategies towards the most clinically relevant species and safeguard

public health.

Results and discussion

Analysis of public genomes substantially restructures *Legionellaceae* family

Given the known difficulties in culturing bacteria(19), culture-independent data is necessary to

characterize the full range of microbial diversity. We delineated 128 species clusters (Table S1,

see Data availability section) from publicly available *Legionellaceae* isolated and metagenomic-

assembled genomes (MAGs) using an average nucleotide identity (ANI) of 95%(20). Importantly,

each ANI-delineated species cluster included genomes that presented the same species label in

their metadata, or were not annotated to the species-level. The newly delineated species clusters

substantially expand the *Legionellaceae* family, nearly doubling the number of the species currently

recognised(21, 22) and resulting in several novel phylogenetically clades (Figure 1). At the same

time, we identified several new species clusters, phylogenetically related to known ones. As an

example, ANI similarity supports the differentiation of *L. pneumophila* subspecies *subsp. pascullei*

and *subsp. fraseri* into two distinct species (i.e., *L. pascullei* and *L. fraseri*) based on an average

ANI similarity to *L. pneumophila* of, respectively, 91.2% and 92%. Notably, given that most of the

unique genes are of unknown function(23), the phenotypic differences linked with this differentiation

are unknown. Also, the retrieved genomes indicate the presence of three other species clusters

previously classified as *L. pneumophila,* which represent instead distinct species (ANI compared

to *L. pneumophila*: 91.0% - 91.8%). Within each species, we found between 1 and 213 genomovars

(genomic variants, ANI threshold: 99.5%). While the mechanisms of genetic variation and selective

pressure can have different influences on the intra-species diversity of different species(6), the ratio

between the number of genomovars and the number of genomes in each species suggests that

the intra-species diversity of non-*L. pneumophila* species is likely to still be underestimated (Table

S2).

Recognizing the lack of a consensus definition, a genus is generically considered a restricted

monophylic clade of organisms with conserved characteristics, which arise due to significant

genome similarity(24). The family *Legionellaceae* currently includes only a single genus, i.e.

Legionella(21, 22). However, a comparison of the average amino acid identity (AAI) returned values

reaching as low as 45%, a value well below the suggested boundary for genera occurring between

65 and 72% AAI(25). We further investigated relative evolutionary distance values(26) in

combination with manual core genome phylogeny refinement to define genus clusters and remove

polyphyly. This resulted in defining 17 different genus clusters present within the *Legionellaceae*

family, of which only five include cultured isolates. This is substantially different from what is

currently accepted, and highlights the need to re-structure the taxonomy of this family. Interestingly,

most clinically relevant species clusters(27) belong to the same clade as *L. pneumophila.* However,

clinically-associated species are present in three of the five genus clusters with isolated

representatives. Interestingly, two of the most clinically relevant species, *L. micdadei*(28) and *L.*

feeleii(29), were found to belong to a separate genus cluster than *L. pneumophila*. While such

distinction was previously suggested(30), it was later criticized based on 16S rRNA analyses(31),

a method now known to provide limited phylogenetic resolution(32). These three genus clusters

with clinically-associated species encompass phylogenetically all clusters with previously known

species suggesting that all isolated species could potentially pose a risk to immunocompromised

subjects(28). On the other hand, the absence of clinical isolates from a large clade of the

Legionellaceae tree suggests that the current definition of the *Legionella* genus (i.e., coinciding to

the family one) includes distinct phenotypes, of which there is a clear distinction in clinical

relevance. As a result, in case legislation is applied at the genus level, the risk of *L. pneumophila*

and species with no indication of clinical relevance is equated, not allowing legislators/operators to

prioritize remediation interventions.

We found the genus clusters to be characterized by different genome sizes, even when accounting

for differences caused by genome origin (i.e., isolates, MAGs) (two-way ANOVA, p-vals < 0.001).

The combination of genome size variability, together with the proteome diversity across the family,

results in a core pangenome (i.e., present in ≥95% of genomes) of only 697 gene families out of

64796 (1.1%). To put this finding in perspective: even though a detailed comparison is not possible

due to methodological differences, a similar percentage of core genes was found in the phylum

Chlamydiae(33), a fact that emphasizes the high variability of the genome content of the

Legionellaceae family. In fact, motile facultative intracellular bacteria, such as *Legionellaceae*,

present greater pangenome sizes than non-motile strictly intracellular bacteria(34)*,* such as

Chlamydiae. The size of the core and total pangenome for the entire *Legionellaceae* family are,

respectively, -41.7% and +34.7% different than what is estimated through the Heaps' Law(35)

curves obtained from only previously-known species (core families: 1081, total families: 37071).

This highlights the diversity of the newly delineated species compared to previously known ones.

In addition, core pangenome genes make up less than half of total number of genes per genome

(i.e., isolates: 17% - 32%; MAGs: 21% - 46%) with between 2 and 834 gene families present

exclusively in a single species and up to 161 additional unique genes present in genomovars. The

number of unique gene families is best explained by the combination of genome size and relative

phylogenetic distance (i.e., the phylogenetic distance to the closest other *Legionellaceae* genome)

(R^2 = 0.78, Figure S1). Yet, most of the explanatory power is due to the relative phylogenetic

distance rather than genome size (ΔAIC values compared to the full model equal to 6.4 and 227.3,

respectively), highlighting that genome size increase was accomplished in *Legionellaceae* bacteria

through different evolutionary trajectories which led to a diversification of proteome content.

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Figure 1. Maximum likelihood phylogenetic tree of the family *Legionellaceae* based on 80% identity

of core genes families which presented also homologs within the *Coxiella burnetii*, the chosen

outgroup (54 protein families, 19141 sites, model LG+F+I+R7). Nodes with optimized ultrafast

support ≥85% are marked with black circles. Leaves enclosed within grey areas indicate genomes

of different genomovars belonging to the same species. Outer tree layers indicate whether the

delineated species have currently valid scientific names and genomes from isolated specimens are

available, the source of all the genomes belonging to the specific species/genomovar, genus cluster

membership, the estimated genome size and the number of unique gene families present

throughout a species or unique to specific genomovars.

Diversity in *Legionellaceae* family highlights differences in putative phenotypes

The extent of the proteome diversity across different genus clusters suggests the presence of

distinct phenotypes across this family. Given the importance of interactions facultative intracellular

bacteria with the host and the importance of metabolism bacterial niche definition, we assessed

the differences of secretion systems, factors linked to host invasion and the metabolic potential

across different species and genus clusters.

Secretion systems presence is highly variable within the *Legionellaceae* family

The ability of secretion systems to translocate effector proteins to host cells is a key trait of

intracellular bacteria, which allows them to manipulate hosts to their advantage(36). *L.*

pneumophila is known to secrete a variety of effectors through the Dot/Icm Type IVB (T4BSS) and

Type II (T2SS) secretion systems(16, 37, 38). We found homologs of proteins belonging to the

T4BSS of *L. pneumophila* in all genomovars, confirming its presence in the novel genus clusters*.*

Besides the T4BSS secretion system, the T2SS was detected in nearly all *Legionellaceae* species,

confirming previous analyses of its wide distribution(39). Notably, the presence of a T2SS was not

confirmed in two MAGs forming their own genus cluster. While this result could be influenced by

genome incompleteness (estimates: 11% and 1%), the observed pattern suggests the occurrence

of gene loss events by the clade's ancestor. Given that the T2SS favours survival of *L. pneumophila*

at temperatures below 17 °C(40), it is possible that species lacking a T2SS might not be able to

survive for long periods of time low temperatures. Interestingly, two of these MAGs were

retrieved from environments with warm temperatures, specifically soils incubated at 25 °C(41)

and mine drainage(42).

Despite the potential role of other secretion systems in *Legionellaceae* ecology (e.g., host range,

motility, HGT), little is known regarding their prevalence and diversity(43). The results of our

analysis highlight the presence of a wide range of secretion systems spread throughout the family

(Figure 2, Table S3). The different secretion systems have distinct distribution across genus

clusters. We detected Type IVA secretion systems (T4ASS) of various types (i.e., Types F, G, I and

T) throughout the family and found only a limited influence of phylogenetic relatedness on their

presence/absence pattern (Jaccard diversity-based Mantel correlation, r = 0.19, p-val < 0.001). In

fact, we determined 78% and 43% of, respectively, Type F and T systems to be carried by plasmids,

supporting their pervasive circulation across the family(16, 44). Conversely, we found other

secretion systems to be present predominantly in specific genus clusters. We found the lss and

novel Type I (T1SS) and Type V (T5SS) secretion system predominantly in the genus clusters that

includes *L. pneumophila*, *L. micdadei* and *L. israelensis*. Such machineries could be potentially

favour host cell adhesion and entry(45), as well as biofilm formation(46). In addition, we identified

Type VI secretion systems (T6SS) within restricted clades of such genus clusters. T6SSs could

potentially be used against either competing prokaryotes or non-infectable eukaryotes(47) or to

obtain nutrients from neighbouring cells(48), potentially favouring extracellular survival. On the

other hand, we noticed different types of T1SSs (i.e., lb1ss, lb2ss and lrpss)(49) to be largely

lacking in the *L. pneumophila* genus cluster, while being present in the ones of *L. micdadei*, *L.*

oakridgensis and the ones composed exclusively of MAGs.

While novel clades with no history of clinical association seem to be characterized by a lower

number of secretion systems, a direct connection between the reduced presence of secretion

systems and pathogenicity is unlikely. In fact, previous studies have shown a limited concordance

between *Legionellaceae* pathogenicity and secretion systems(50, 51). Such limited concordance

is due to several causes. First of all, pathogenicity, as well as secretion systems, can vary within

species, requiring to analyse all available genomes and not select only few representatives.

Furthermore, intracellular replication is connected to the release of adequate effectors within a

host(52). Given the known heterogeneity of effectors even within single species (e.g., (38, 53)),

encoding genes for a secretion system is not a sufficient proxy for released effectors. Finally,

human infection is also dependent on the environmental distribution of these bacteria, as well as

their ability to survive aerosolization and reach human lungs, factors that are influenced by other

cellular traits (e.g., metabolism, UV resistance).

Figure 2. Maximum likelihood phylogenetic tree of the family *Legionellaceae* based on the identified 80% identity core genes families, which presented also homologs within the *C. burnetii*, the chosen outgroup (54 protein families, 19141 sites, model LG+F+I+R7). Leaves enclosed within grey areas indicate genomes of different genomovars belonging to the same species. The average Dot/Icm and T2SS protein similarity was estimated against *L. pneumophila* strain *Paris*, while rounded markers indicate the presence of the specified traits.

Factors linked to host invasion are not conserved among species

The likelihood of an intracellular bacterium encountering its host can be very low due the diversity of microorganisms in the environment(54) and hosts feeding preference(55), suggesting the importance "active" mechanisms of host invasion. We identified the genes required for Type IV pili in throughout the family with the exception of a clade composed of three novel genus clusters (Figure 2, Table S3). The lack of pili in these genus clusters suggests their limited efficiency in host colonization(56). On the contrary, we found flagellar assembly genes to be present mostly in the genus clusters of *L. pneumophila*, *L. micdadei*, *L. israelensis* and *L. geestiana*, while mostly absent in others. Compared to pili, flagellated species show a lower prevalence in such genus clusters, confirming previous experimental observations (i.e., *L. longbeacheae*)(57) and suggesting the occurrence of multiple loss events of this trait. The lack of flagella in *L. longbeacheae*, a species often linked to soil and compost(27), is likely due to the fact that surface motility does not necessarily require such machinery(54). In fact, non-flagellated species make up the majority (62.5%) of the ones that were isolated from soil, suggesting their adaptation to non-aquatic environments. Beyond motility, adhesins, such as the Lcl protein encoded by *L. pneumophila*(58)*,* are also key determinants of host-pathogen interactions and biofilm formation(59). While we found Lcl homologs characterized by a collagen-like domain in the genus cluster of *L. pneumophila,* we detected other proteins with collagen domains characterized by "unknown" or "membrane" COG categories in 11 species mainly in the genus clusters of *L. oakridgensis*, *L. geestiana* and *L. micdadei*, highlighting the presence of additional adhesins absent in the *L. pneumophila* genus

cluster (Table S4). Besides an easier attachment to host cells and surfaces, the enhanced biofilm

formation abilities provided by these proteins might also favour resistance to stresses and, thus,

persistence in the environment. Yet, characterizing the importance of such traits is complicated by

hosts diversity. For example, *L. pneumophila* was shown to require a flagellum to efficiently infect

human macrophages, but not amoebae(60).

Chemosensing, the ability to sense chemical gradients, can be used to direct movements (i.e.,

chemotaxis) towards hosts(54). However, we noticed that, except for restricted clades with the *L.*

pneumophila and *L. micdadei* genus clusters, the majority of *Legionellaceae* species lacked both

CheW and methyl-accepting chemosensing proteins, two of the essential components of the

chemotaxis Che system(61). This indicates the absence of motility regulation through the Che

pathway (Figure 2, Table S3). In addition, we found a weak association between the Che pathway

and flagella presence (binomial GLM, p-value = 0.02) and pili (p-value = 0.07), further stressing the

unrelatedness of these traits. Interestingly, previous research highlighted that most respiratory

pathogens are non-chemotactic, supposedly due to the relative simplicity of the respiratory

system(54)·(59). However, this hypothesis does not take into consideration the need to survive in

the environment and find suitable hosts of facultative intracellular bacteria, such as *Legionellaceae*

bacteria. As a matter of fact, other regulatory pathways to modulate chemotaxis exist(62). For

example, twitching motility in *L. pneumophila* is inhibited by the presence of other bacteria likely

due to lower nutrients concentrations(63), indicating either that rich nutrient conditions are needed

for twitching motility or the ability to perform chemotaxis despite lacking the Che pathway.

We predicted up to 80 signal transduction and 170 response regulator proteins per genome.

Interestingly, we found a different number of these proteins across the isolated bacteria in different

genus clusters (ANOVA, Tukey HSD post-hoc test, p-values < 0.002; Figure S2), with the one

including *L. pneumophila* presenting the highest number. In total, we identified 726 and 1363 gene

families for signal transduction and response regulation, of which only 2 and 13 were, respectively,

detected across the whole family (≥95% species). In most cases, the function of these proteins was

already reported for *L. pneumophila* (Table S5). Beyond those, we detected 71.6% and 66.9% of

signal transduction and response regulators to be genus cluster-specific, but that only 35.6% and

33.4% of them are encoded by multiple species within their cluster. These results suggests that

while a "core" regulatory network might be conserved across species, significant diversity is present

not only across, but also within genus clusters, which might significantly affect phenotypic

responses to environmental cues and contribute defining the environmental niche of different

species.

Finally, also extracellular capsules can affect bacteria-host interactions and enhance survival in

unfavourable environments(64). A capsule has been only reported for *L. longebeacheae* where it

constitutes a crucial virulence factor towards *Acanthamoeba castellani*(65) and potentially aids its

survival in soil(64). Despite not presenting notable homology to *L. longbeacheae* capsule proteins,

we detected capsular biosynthetic gene clusters of different types across the family, and especially

in the genus cluster of *L. oakridgensis*. *L. beliardensis*, *L. gresiliensis* and *L. busaniensis* presented

two ABC capsules systems, while *L. gormanii* and *L. qingyii* encoded two different capsule types,

a trait which might favour these bacteria against different stresses(65) and extend their niche

breadth(66). Interestingly, the similarity of the capsule systems varied widely within species (Figure

S3), possibly arising from the adaptation to different environmental conditions or lifestyles of distinct

genomovars(64)*.

Genus clusters encode different metabolic potentials

Several laboratory studies have highlighted the presence of a complex biphasic metabolism in *L.*

pneumophila(67). However, the metabolism of other species has been seldomly examined, leading

to a major knowledge gap regarding the ecology of these species. We found a complete

tricarboxylic acid (TCA) cycle in nearly all species, highlighting the reliance of the whole family on

this pathway to convert serine into acetyl-CoA(68) (Figure 2, Table S6). On the other hand, we

found other metabolic pathways to be not as conserved. As expected, we observed species

belonging to the genus cluster of *L. pneumophila* to present a similar metabolism to this model

species, with a complete Entner-Doudoroff (ED) and Embden-Meyerhof-Parnas (EMP), and an

incomplete pentose phosphate (PP) pathway to provide the pyruvate required by the TCA

cycle(69)·(68)·(70). However, we detected complete ED pathways only in subclades of the *L.*

micdadei and *L. oakridgensis* genus clusters and in none of the ones composed only of novel

species. Given that this pathway was either complete or completely absent, it is unlikely that this

result arises due to MAGs incompleteness, but is rather a true adaptation signal. Interestingly, we

found that both the EMP and the PP pathways presented higher completeness when the ED

pathway was present (t-test, p-val < 0.001), suggesting a reciprocal dependency of such pathways

within the whole family and not only *L. pneumophila*(70). As a result of this dependency, the

completeness of both these pathways varied significantly across genus clusters (Kruskal test, p-

val < 0.001) and presented a negative correlation with the phylogenetic distance to *L. pneumophila*

(Sperman correlation; ρ = -0.48 – -0.7, p-val < 0.001), suggesting a less prevalent role of these

pathways in phylogenetically unrelated species to *L. pneumophila* and their reliance on different

metabolic pathways. For example, while the glyoxylate cycle is highly incomplete in *L.*

pneumophila(68), we observed its complete set of genes in several species which mostly belong

to other genus clusters. This pathway could potentially contribute to the biosynthesis of cell carbon

compounds, but future studies are needed to verify this hypothesis.

The diversity of metabolism among *Legionellaceae* genus clusters is further confirmed by the fact

that, even excluding MAGs to remove genome incompleteness bias(71), we observed significant

differences in the KEGG metabolic pathways completeness profiles among genus clusters (Figure

3a; PERMANOVA, R^2 = 24.6%, p-val < 0.001) due to the different presence of several enzymes

across genus clusters, as shown in Figure S4. Most notably, the genus clusters including *L.*

pneumophila and *L. micdadei* show a larger number of unique enzymes indicating their richer, but

different, metabolic capabilities. In fact, we found more complete pathways for phosphatidylcholine

biosynthesis and acylglycerol degradation in the genus cluster of *L. pneumophila*, while higher

completeness of the glycogen biosynthesis pathway in the one of *L. micdadei* (Table S7).

Concordant with the metabolic differences observed at genus cluster-level, we detected a negative

correlation between metabolic differences and phylogenetic relatedness (Spearman correlation; ρ

= -0.5, p-val < 0.001) resulting in smaller subclusters of highly related genomes with similar

metabolic capabilities (Figure S5, Table S8). Several of such metabolic clusters included both

species clinically-associated and not, suggesting that species with different associations to disease

can have overlapping environmental niches.

The greater metabolic capabilities of the species belonging to selected *Legionellaceae* genus

clusters extends not only to central metabolic pathways, but involve also secondary metabolites

such as CAZYmes and peptidases, as well as other phenotypes, such as the biosynthesis of the

storage compound poly-3-hydroxybutyrate (PHB) and *myo*-inositol metabolism, which have been

shown to influence *L. pneumophila* growth[72],[73, 74]. We observed statistically significant

differences in the number of genes encoding for both secreted and intracellular CAZYmes and

peptidases, with the genus clusters of *L. pneumophila* and *L. micdadei* presenting a larger number

of enzymes as other genus clusters (Figure 3b). In addition, we identified that both the number

(Figure S6) and the repertoire of CAZYme families and peptidases classes is explained by genus

cluster membership (PERMANOVA, R^2 = 9.2 - 17.6%, p-val < 0.001), highlighting the difference in

the complex organic matter degradable among clusters[75, 76]. Similarly, other phenotypes

reported for *L. pneumophila*, such as the ability to synthetize PHB and metabolize *myo*-inositol,

tended to be rarer in genus clusters phylogenetically distant from the *L. pneumophila* one (Figure

2), once again stressing the reduced metabolic capabilities of novel genus clusters and suggesting

substantial difference in their mechanisms of intracellular replication and survival to potentially

nutrient-poor extracellular conditions[77],[74].

Overall, the more restricted metabolic potential of genus clusters composed of novel species, even

though potentially affected by MAGs incompleteness, suggests their higher metabolic dependence

on other microorganisms. While it is possible that the required nutrients could be retrieved freely in

the environment due to the leakage from other microorganisms[78], it is also a possible signal of a

stronger metabolic reliance on their hosts. This could explain the difficulties in their isolation and

warrants further experimental investigations.

Figure 3. (a) Metabolic differences among *Legionellaceae* isolated genomes and clustering

following genus cluster membership. (b) Number of CAZYmes and peptidases per isolated

genomes as a function of their secretion and genus cluster membership. Statistically significant

differences among genus clusters are marked by their p-values: "*" = 0.05 - 0.01, "**" = 0.01 - 0.001,

"***" < 0.001. Genus clusters representatives are: "1": *L. pneumophila*, "6": *L. micdadei*, "13": *L.*

oakridgensis, "17": *L. israelensis*, "18": *L. geestiana*.

Auxotrophy predictions suggest redesign of cultivation media

Intracellular bacteria have evolved auxotrophy for diverse amino acids(79), gaining a fitness

advantage in their host-associated niche, but becoming dependent on external resources for

survival(80). This also holds true for *L. pneumophila*, for which several studies have confirmed its

auxotrophy for several amino acids (Table 1). We performed *in silico* analysis of *L. pneumophila*

auxotrophies using GapMind, a pathway annotation tool(81), and gapseq, an automated tool for

genome-scale metabolic model reconstruction(82). Except for L-isoleucine and L-serine, the

predictions of the two tools were mostly concordant and matched, respectively, eight and six of the

11 reported auxotrophies with limited indications that this might be attributable to strain-level

differences (Table S9). The different results obtained for L-isoleucine and L-serine by two

approaches lies in the fact that these compounds are predicted by gapseq to be synthesized

following two pathways not considered by GapMind (Table S10) and warrant more investigations

to verify their biological relevance. Interestingly, while L-cysteine is commonly used for culture

confirmation of *L. pneumophila*, both tools did not predict its auxotrophy. While we detected the

genes required for L-cysteine biosynthesis through the reverse transsulfuration pathway(83), we

noticed the lack of homocysteine biosynthesis, a required precursor for L-cysteine biosynthesis, as

well as the requirement for inorganic sulphur sources (e.g., SO_4^{2-}, H_2S). The latter, despite the

known link between L-cysteine biosynthesis and sulphur(84), is not controlled for in growth media

(e.g., BYEB). Such results indicate that L-cysteine might be an "indirect" auxotrophy and warrant

dedicated experimental confirmations to clarify *L. pneumophila* metabolic capabilities.

Table 1. Reported and predicted auxotrophies for *L. pneumophila*. Experimental evidence retrieved

from experimental studies(68, 143, 144). High and medium GapMind biosynthesis prediction

confidence results indicate the detection of homologs for all biosynthetic steps for a given amino

acid. Conversely, low confidence results highlight the lack of biosynthetic steps and imply

auxotrophy(81). Multiple results per amino acid indicate that distinct results were obtained with

different genomovars. The asterisk indicates that while the genes for the L-cysteine biosynthesis

have been annotated, the genes required to produce the required compound homocysteine are

absent.

Amino acid	Experimental evidence	GapMind biosynthesis prediction confidence	gapseq prediction
L-alanine	-	Not assessed	-

L-arginine	Auxotrophy consistently reported	Low	Auxotrophy
L-asparagine	-	High	-
L-aspartate	-	Not assessed	-
L-cysteine	Auxotrophy reported	Medium*	-
L-glutamate	-	Not assessed	-
L-glutamine	-	High	-
Glycine	-	High	-
L-histidine	-	Medium/High	-
L-isoleucine	Auxotrophy consistently reported	Low	-
L-leucine	Auxotrophy consistently reported	Low	Auxotrophy
L-lysine	-	High	-
L-methionine	Auxotrophy consistently reported	Low	Auxotrophy
L-phenylalanine	Auxotrophy reported	Low	Auxotrophy
L-proline	Auxotrophy reported	High	-
L-serine	Auxotrophy reported	Low	-
L-threonine	Auxotrophy reported	Low	Auxotrophy
L-tryptophan	-	Medium	-
L-tyrosine	Auxotrophy reported	High	-
L-valine	Auxotrophy consistently reported	Low	Auxotrophy

Given the susceptibility of annotation-based tools to genome incompleteness and the tools' overall

concordance (Table S9), we continued the analysis only using gapseq's results. The gapfilling

carried out by this tool, in fact, removed the differences in the number of predicted auxotrophies

between isolates and MAGs from the same genus cluster (t-test, p-vals > 0.1). This allowed us to

highlight significant differences in the number of auxotrophies among genus clusters and the lower

amount in the ones of *L. pneumophila*, *L. micdadei* and *L. israelensis* (ANOVA, p-val < 0.001). We

did not find any amino acid auxotrophy to be conserved across the entire family, despite several

being highly prevalent in most genus clusters (Figure 4, Table S11). Conversely, no species

showed auxotrophy towards five amino acids (L-alanine, L-aspartate, L-glutamate, L-glutamine and

glycine). Interestingly, while we did not detect L-cysteine auxotrophy in the *L. pneumophila* genus

cluster, we found auxotrophic species towards this amino acid in the genus cluster of *L. micdadei*

and in several others without isolated representatives. Similarly, with the exception of the genus

cluster composed exclusively of *L. geestiana*, the genus clusters with isolated representatives

showed low prevalence of L-histidine auxotrophy. Such auxotrophic differences are likely partly

responsible for the differential growth ability of different *Legionellaceae* species on standard

medial(85), and call for media supplementation of additional amino acids such as L-histidine for the

isolation of novel species.

Figure 4. Prevalence of auxotrophies predicted in each genus cluster. The phylogenetic tree on the left represents the relationships among genus clusters, indicating with a star the ones for which isolated representatives are available. The labels "*Lg*", "Li", "*Lm*" and "*Lp*" represent the genus clusters including *L. geestiana*, *L. israelensis*, *L. micdadei* and *L. pneumophila*, respectively.

Legionellaceae species have distinct environmental distributions

Given the genotypic diversity among *Legionellaceae* species, it is expected that they will have distinct, but partly overlapping, ecological niches and environmental distributions. To assess this, we used the OTUs provided by MicrobeAtlas(86), linking them to the full-length 16S rRNA genes of 38 species clusters through reciprocal best hits (Table S12). We detected most *Legionellaceae* OTUs in all major environmental classes (Figure 5), confirming the variety of niches of these bacteria observable for the isolation and MAGs data (Figure 1). The distribution over the environmental classes varied greatly within genus clusters, with a surprisingly high fraction of animal-associated samples. A closer inspection (Figure S7) revealed several samples from non-human hosts, as well as the gastrointestinal tract or faeces, suggesting that these bacteria can

infect other animals and be shed by infected individuals, confirming previous reports(87, 88). While

in general we found OTUs mostly in aquatic samples, several OTUs showed a drastically higher

prevalence of soil samples or animal-associated ones (Figure 5). Comparing different genus

cluster, we did not observe a statistical difference in the occurrence of OTUs belonging to different

genus clusters (Kruskal-Wallis test, p-val = 0.76) and we detected only a weak statistical effect on

the distribution over environmental classes of genus cluster membership (PERMANOVA, R^2 =

7.6%, p-val = 0.07, only genus clusters with more than one OTU considered). These results

highlight that different genus clusters are similarly present in the environment and their species

present only a limited adaptation to specific major environmental classes, given also the

heterogeneity of these broad classifications. Yet, we found more similar environmental distributions

for OTUs belonging to the same metabolic cluster (Table S8) than with others (t-test, p-val = 0.04)

and at lower phylogenetic distance (Spearman correlation, ρ = -0.26, p-val < 0.001). This

corroborates the link between shared evolutionary history, metabolic capabilities and ecological

niches and remarking how these processes should be assessed at very fine taxonomic resolutions.

It is important to note that the environmental niche of facultative intracellular bacteria is not only

due the bacteria's direct survival in the environment, but also to the environmental survival of their

hosts(89), a fact which can extend the overall niche breath. Future research identifying the genes

responsible for the adaptation to diverse environments and understanding better the impact of

hosts in *Legionellaceae* environmental distribution could help in source tracking by providing

insights on the environmental origin of clinical specimens.

Figure 5. Number of samples in which the MicrobeAtlas OTU (99% similarity) were found (i.e.,

OTUs' occurrence) and their distribution among major environmental classes. Species labels are

the following: *L. antarctica (Lan), L. beliardensis (Lbe), L. busanensis (Lbu), L. clemsoniensis (Lcl),*

L. fallonii (Lfa), L. jamestowniensis (Lja), L. maceachernii (Lma), saoudiensis (Lsa), spp cluster

94_0 (Lsp1).

Implications

This study highlights the importance of genomic data to expand the phylogenetic diversity and

characterize the different traits in intracellular bacteria such as *Legionellaceae*. Given the

heterogeneity of traits and environmental preferences in closely related microorganisms, it is

important to design surveys able to reach species-level, if not genomovar-level, characterization to

define accurately microorganisms' environmental niches and lifestyles. This is especially significant

when closely related species vary not in their clinical importance, as the case of *Legionellaceae*,

but also their biotechnological relevance. Methods to achieve this in environmental surveys already

exist (e.g., clade-specific primers, full-length 16S rRNA sequencing, strain-resolved

metagenomics)(90–92), but their adoption is still limited. Such methods could be applied to achieve

a detailed characterization of the environmental prevalence of single species and link this to

functional traits(93). Moreover, given the known difficulty in the isolation of host-associated

microorganisms, the use of MAGs-derived metabolic information will allow to isolate currently

uncultured species and more (e.g., (94)) . The combination of greater specificity of environmental

surveys and the isolation of novel species will enable to test predicted phenotypes and the genes

responsible for the adaptation to environmental conditions, as well as achieving a greater

understanding of eco-evolutionary processes and their interactions with other microorganisms.

With respect to this latter point, the environmental interactions between protistan host and

facultative intracellular constitute a major research gap for our understanding of their ecology. This

is especially relevant for microorganisms with wide host ranges, such as *Legionellaceae*(95), for

which it is necessary to identify their and the hosts traits affecting the outcome of the interaction to

build a mechanistic understanding of host-colonization processes. Beyond scientific value, this

knowledge would also allow for better optimization of biotechnological processes and aid

safeguarding public health by allowing the improvement of relevant legislation.

Materials and Methods

Genomes retrieval, quality assessment and comparison

A total of 5023 genomes classified as belonging to the family *Legionellaceae* were identified from

publicly-available sources (date last accession: 15/05/2023, Table S13). Sources included NCBI

Assembly database, ENA, JGI and other several large-scale metagenomics survey(96–104). In

addition, reads from Wee and coworkers(99) were assembled using spades v3.15.4(105) after read

trimming using fastp v0.23.2(106) and the removal of known UniVec contaminant(97) using

BWA-MEM2 v2.2.1(107) and samtools v1.16.1(108). The quality of all the genomes and their

taxonomic affiliation was assessed using CheckM2 v1.0.1(109), and GUNC v1.0.5(110) and

GTDB-Tk v2.3.0(111) to remove genomes with more than 5% contamination, less than 85%

complete and not belonging to the investigated family, eventually after manual refinement using

Anvi'o v8(112). Genomes dereplication was performed with an initial clustering at 90% ANI using

MASH v1.1(113), followed by a secondary one using skani v0.2.1(114) at 95% and 99.5% ANI.

Representative genomes for species (i.e., 95% ANI) and genomovars (i.e., 99.5% ANI) were

chosen as the ones with the highest value derived from the following formula(115): *Completeness*

– 5*Contamination + 0.5*log(*N50*) + *Centrality*/100, using CheckM2 quality estimates and where

N50 is the minimum length for which half of the genome belongs to contigs longer than *N50*, while

Centrality is the average ANI to genomes within the same secondary cluster. Due to the high

number of genomovars found for *L. pneumophila* and *L. fraseri,* only the species representative

and the three representatives of the genomovar with the highest number of genomes were

selected. Species' clinical-associations were retrieved from published literature(116).

Proteins predictions and functional annotations

Proteins were predicted using Bakta v1.8.1(117) (v5.0, full database) and annotated using

eggNOG-mapper v2.1.12(118). Amino acid similarity was assessed using FastAAI v0.1.20(119).

T4SS were detected using CONJScan v2.0.1(120), while other secretion systems, flagella and pili

using TXSScan v1.1.3(121). T6SSs were considered as present in case more than nine proteins

were detected in order to match previous reports(39). Plasmid-associated contigs were detected

using Platon v1.7(122). The presence of capsule systems was assessed using CapsuleFinder

v1.0.5(64). Sensory and regulatory proteins were selected based on eggNOG-mapper annotations

and the signal transduction specific KEGG profiles from MiST v4.0 database(123). Chemotaxis and

methyl-accepting proteins were detected using HMMER v3.4(124) (minimum bitscore: 25) and the

models from the MiST database. KEGG metabolic pathways completeness was estimated using

DRAM v1.4.6(125), while the amino acids biosynthetic potential was evaluated using GapMind and

gapseq. Clusters of genomes presenting similar metabolic capabilities were obtained through

HDBSCAN clustering based on the euclidean distance of metabolic pathways completeness.

CAZYmes were annotated against the dbCAN3 database using dbcan_run v4.0.0(126), retaining

only proteins identified by at least two detection methods. Peptidases were identified searching the

predicted proteins against the MEROPS v12.4 database(127) using MMseqs2 v14-7e284(128)

retaining proteins showing at least 30% identity over 65% query coverage and displaying a bitscore

≥50. Secreted proteins were identified using SignalP v6.0(129). Homologs of selected *L.*

pneumophila strain *Paris* proteins were detected using BlastP v2.12.0(130) using the same

thresholds used for MMseqs2 and inspecting Bakta and eggNOG-mapper annotations (e.g.,

PFAMs detected). Type 1 secretion systems were considered as present if both

LssD/lrpssD/lb1ssD/lb2ssD and LssB/lrpssB/lb1ssB/lb2ssB proteins were detected within a five

proteins neighbourhood(49). Eventual duplicated matches with TXSScan results were assessed

using a custom script.

Pangenome and phylogeny construction

Predicted proteins were used to perform calculations using PPanGGOLiN v1.2.105(131). A

pangenome employing a protein clustering identity percentage threshold of 80% was used to

identify core proteins (i.e., present in ≥95% of genomes) to reconstruct phylogeny. Identified

proteins were searched for homologs in the genome of *C. burnetii* (RefSeq accession:

GCF_005280755.1). Proteins were then aligned using MAFFT G-INS-i as implemented in

Concatenator v0.2.1(132, 133) and used to calculate maximum likelihood phylogenetic trees using

IQ-TREE v2.2.2.6(134) under the LG model and automated model selection(135). Support values

were calculated based on 1000 ultrafast bootstraps with bootstrap tree optimization(136). The

resulting phylogenetic tree was visualized and annotated in iTOL v6(137). An initial genus cluster

membership was obtained using GTDB-Tk de-novo workflow using calibrated RED values.

Successively, cluster membership was refined manually to remove polyphyly. A second

pangenome employing an identity percentage threshold of 50% was used to identify gene families

among *Legionellaceae* genomes and identify shared proteins, as suggested by PPanGGOLiN

authors.

Assessment of species environmental distribution

One random full length 16S rRNA gene per genome was retrieved from the Bakta annotation of the

selected species representatives or, otherwise, from other genomes belonging to the same species

cluster. Such sequences were matched to the full-length OTUs (99% similarity) retrieved from

MicrobeAtlas v3.0 (downloaded on: 16/9/2024) through MMseqs2's reciprocal best hit workflow

and removing OTUs which showed best hits to multiple species clusters. Major environmental

classes were obtained from MicrobeAtlas metadata, selecting only samples for which only a single

class was listed. Spearman correlations were used to estimate the concordance of the

environmental distribution among selected OTUs.

Statistical testing and data analysis

Analysis of the bioinformatic results was performed in R v4.3.0(138) leveraging the functions

provided by libraries Tidyverse v2.0.0(139), ape v5.7-1(140), dbscan v1.2-0(141) and vegan v2.6-

4(142). Statistical significance was assessed using appropriate two-tailed tests and a threshold p-

value of 0.05.

Data availability

Supplementary data and tables S1-S13 are available on Zenodo at the following link:

https://doi.org/10.5281/zenodo.14748706.

Author contributions

M.G.: Conceptualization, Methodology, Formal Analysis, Investigation, Data Curation,

Visualization, Writing - Original Draft; A.C.: Writing - Review & Editing; F.H.: Writing - Review &

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Competing interest statement

599   None to declare.

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Number of unique gene families

Estimated genome size [Mb]

Genus cluster

Source

Species information

Tree scale: 1

L. pneumophila

Species — red, Genomovar — yellow

1000
0
5
0

Genus cluster
Unknown
Soil
Marine
Freshwater
Engineered environment
Environmental unknown
Clinical

Species information
Most clinically relevant
Clinically-associated
With scientific name
Not named
☆ Isolated genome
Species cluster