

Phylogenomics of Asgard archaea reveals a unique blend of prokaryotic-like horizontal transfer and eukaryotic-like gene duplication

Received: 29 July 2025

Accepted: 19 March 2026

Published online: 10 April 2026

 Check for updatesSaioa Manzano-Morales^{1,2} & Toni Gabaldón ^{1,2,3,4} 

Asgard archaea hold a pivotal position in the tree of life as the closest known relatives to eukaryotes and are therefore crucial for understanding eukaryogenesis. Earlier genomic analyses revealed that Asgard genomes are remarkably larger than those of other archaea and contain a significant number of genes seemingly acquired from bacteria. However, the precise contributions of horizontal gene transfer and gene duplication in shaping Asgard genomes remain largely unknown. Here, we present a comprehensive phylogenomic analysis to dissect the evolutionary dynamics of Asgard genomes, quantifying gene duplication, loss, and both inter- and intra-domain gene transfer events. Our findings reveal that gene transfer is widespread throughout Asgard evolution, predominantly affecting metabolic genes at the periphery of interaction networks. However, our analyses demonstrate that gene duplications, rather than horizontal gene transfers, are the primary drivers behind the increased genome sizes observed in Asgard archaea. This unique evolutionary signature in Asgard archaea—a blend of pervasive prokaryotic-like gene transfer alongside significant eukaryotic-like gene duplication—is consistent with their phylogenetic placement and offers novel insights into the genomic transitions that likely underpinned eukaryogenesis.

Asgard archaea are the closest prokaryotic relatives of eukaryotes and are therefore key to understanding eukaryogenesis^{1,2}. Asgard genomes encode many of what were previously thought to be eukaryote-specific proteins³, including actin and actin-related proteins³, and an actin cytoskeleton has been found in a cultivated Lokiarchaeum⁴. The crucial interest of Asgard archaea has promoted sequencing efforts, resulting in a large number of available genomes – 296 assemblies in release 226 of the Genome Taxonomy Database (GTDB)⁵. These represent 12 class-level lineages *sensu* GTDB (namely Asgard-, Atabay-, Baldr-, Heimdall-, Hermod-, Jord-, Loki-, Njord-, Odin-, Sif-, Thor-, and Wukongarchaeia). However, most data corresponds to Metagenomic Assembled Genomes (MAGs), which can be contentious due to potential incompleteness or contamination issues⁶. Asgard archaea have proven elusive to culture, with two exceptions from Lokiarchae:

Ca. Prometheoarchaeum syntrophicum MK-D1⁷ and *Ca. Lokiarchaeum ossiferum* B-35⁴, which can be cultured with its syntrophic partners and therefore has enabled high-quality genomes.

Asgard archaea were first identified in a deep-sea hydrothermal vent⁸, but their presence in clone libraries precedes their description by two decades^{9–11}. Recent sampling efforts have uncovered new Asgard lineages, of which most were first described in anoxic marine environments¹². However, metagenomic analyses indicate a worldwide cosmopolitan distribution¹³. This reflects on a broad array of lifestyles and metabolic properties, including mixotrophy, homoacetogenesis, facultative anaerobiosis, alkane utilization, and rhodopsin-based phototrophy^{13–17}. However, we have a very limited understanding of how genome variation underlies such diverse adaptations.

¹Barcelona Supercomputing Center (BSC). Plaça Eusebi Güell, Barcelona, Spain. ²Institute for Research in Biomedicine (IRB Barcelona), The Barcelona Institute of Science and Technology, Baldiri Reixac, Barcelona, Spain. ³Catalan Institution for Research and Advanced Studies (ICREA), Barcelona, Spain. ⁴CIBER de Enfermedades Infecciosas, Instituto de Salud Carlos III, Madrid, Spain. ✉e-mail: toni.gabaldon@bsc.es

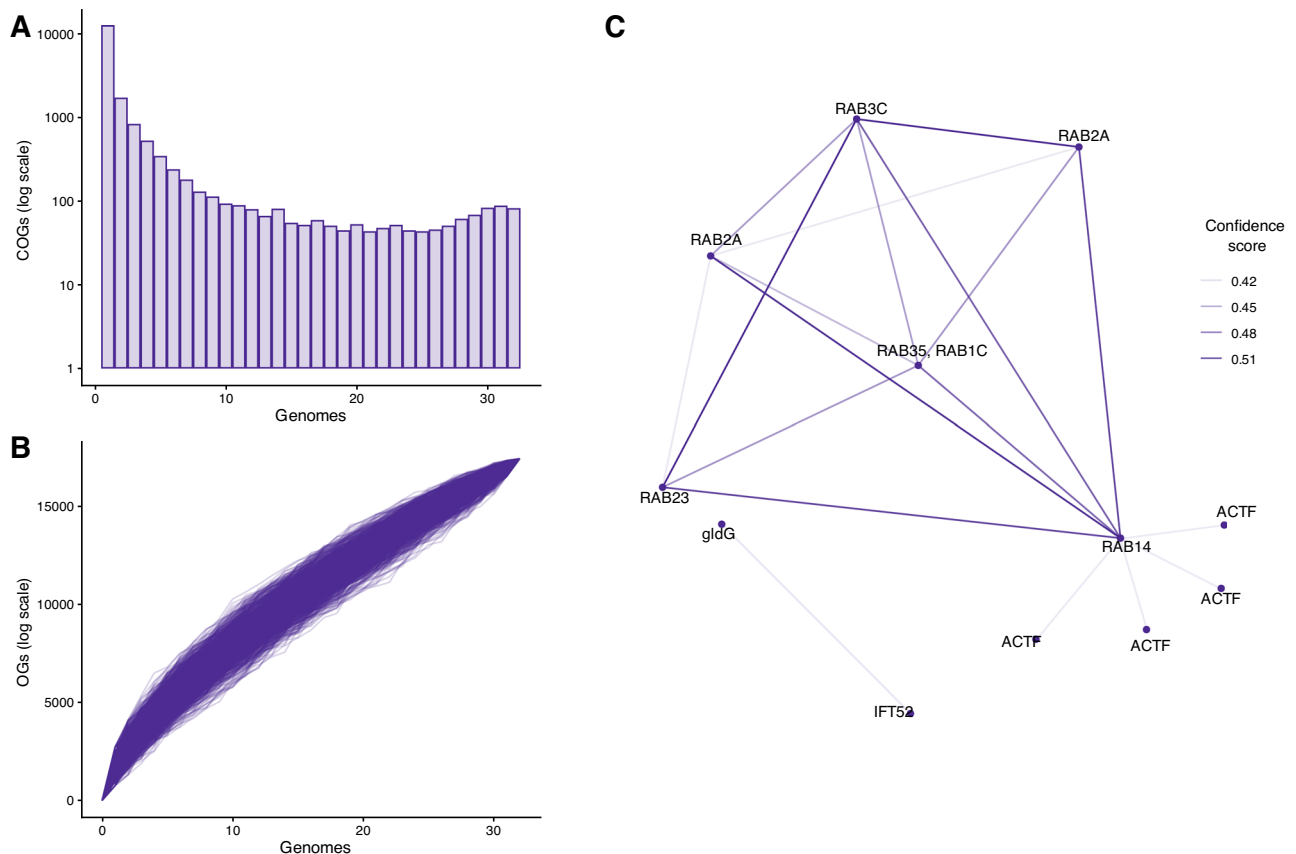


Fig. 1 | Asgard archaeal pangenome metrics for the high-quality subset.

A Frequency of OGs according to the number of genomes in which they are present in the pangenome of Asgard archaea (bin width=1). **B** Rarefaction curves obtained by subsampling the genomes in a random order over 1000 iterations.

C Protein-protein interaction network of the Asgard-Signature proteins. Each node is one non-redundant protein (labeled as its annotated KO/KOs), edges represent protein-protein interactions with color strength proportional to the interaction score given by STRING.

Earlier analyses of Asgard genomes uncovered a large number of putatively transferred bacterial genes¹⁸ and generally larger genome sizes as compared to other archaea¹⁹. Previous analyses considering prokaryotes have found weak²⁰ to strong^{21,22} positive correlations between genome size and horizontal gene transfer (HGT) fraction, but none of them included Asgard genomes. Similarly, Asgard archaeal genomes have been shown to have higher rates of duplication than other archaea¹⁹, but the functional implications of these gene duplications, or how these duplication tendencies contrast with HGT (the prevailing form of genome expansion in prokaryotes), have not been comprehensively explored.

Here, we reconstructed the Asgard archaeal pangenome and used phylogenomics on a super-phylum-wide scale to assess the relative contribution of gene duplication, loss and transfer in the evolution of Asgard genomes against the backdrop of its sister clade, Thermoproteota, and investigate possible functional trends. With this, we aimed to shed light on the processes that have specifically driven Asgard archaea towards larger genome sizes and other “eukaryote-like” qualities in contrast to Thermoproteota, which remain more “prokaryote-like”. Our results indicate that Asgard archaea have high duplication rates that are intermediate between those in their prokaryotic sister clade and those observed in unicellular eukaryotes. We uncover widespread but mostly lineage-specific inter-domain HGT involving several donor-acceptor pairs and metabolic genes. Overall, our results indicate that Asgard genomes are uniquely shaped by a blend of prokaryotic horizontal gene transfer and eukaryotic-style gene duplications.

Results

The pangenome and signature proteins of Asgard archaea

We first sought to characterize the Asgard pangenome by performing homology-based clustering of all available genomes from the phylum Asgardarchaeota in GTDB⁵. However, nearly all Asgard genomes are MAGs, resulting in a median CheckM²³ completeness of 87.54% (SD 8.21). This lack of genome completeness can lead to an under-estimation of the core genome and, conversely, an over-estimation of the accessory genome.

For this reason, we compiled a representative subset of 32 genomes covering all family-level lineages of Asgard archaea by choosing the highest-quality available genome (following the GTDB score criterion) per family and prioritizing available isolates and closed genomes (See Methods, Supplementary Data 1). Additionally, we included 163 other prokaryotic genomes as out-groups, including 28 representative species from the Asgard archaea’s sister group, TACK (“Thermoproteota” hereafter, *sensu* GTDB). We clustered proteins into families (orthogroups, OGs) using a similarity-based approach (See Methods). This yielded a total of 71,402 OGs, including singletons, out of which 17,429 contained at least one Asgard protein. Of these, 12,397 were exclusively present in Asgard archaea, and of these 10,470 were present in a single Asgard genome and no other prokaryote in our dataset (singletons). 79 gene families were found in all Asgard genomes (strict core) and 310 in 90% or more (relaxed core, see Fig. 1A for the distribution), with the remaining belonging to the accessory pangenome. We annotated OGs with KEGG Orthologs (KOs, see Methods) and found that proteins in the Asgard relaxed core were enriched in

pathways such as ribosome (ko03010), DNA replication (ko03030), and aminoacyl-tRNA biosynthesis (ko00970) (Supplementary Data 2).

We estimated the size of the total Asgard pangenome both by Chao's lower bound²⁴ (yielding a lower bound estimate of 61,815 OGs) and by the best-fitting binomial mixture model (resulting in a pangenome of 75,659 OGs and a strict core of 18 OGs). We additionally calculated rarefaction curves (Fig. 1B), which, along with the Heaps estimate ($\alpha = 0.26$), indicated an open pangenome.

To assess the impact of MAG incompleteness on pan-genome size estimates, we repeated the analyses, replacing the 32 best-quality representatives with 32 randomly-chosen, taxonomically-equivalent genomes. In this case, we obtained 72,179 OGs, with 18,135 including at least one Asgard representative, and a strict and relaxed observed core of 57 and 269, respectively. We inferred a similarly open pangenome estimate with identical Heaps α (0.26) and strict core inferred by the binomial mixture model (18), but a higher Chao's lower bound (72,712) and total estimated size by the binomial mixture model (94,783). Hence, both genomic subsets provided similar estimates in terms of openness, OGs and total pan-genome size but not in terms of observed core genome or lower bound estimates, which were more affected by genome incompleteness.

As the core size estimate is particularly sensitive to genome incompleteness²⁵, we ran mOTUpan²⁶ for a more robust estimation of core size and fluidity, as this approach intrinsically considers genome incompleteness (as provided by CheckM completeness). This approach estimated a larger core set for both of the high-quality (473) and random (509) datasets. Hence, even when considering the most conservative estimates, these results underscore how incomplete is our knowledge of the breadth of Asgard archaeal protein diversity, implying the intriguing possibility that new protein families with potential implications for eukaryogenesis may await discovery.

We defined Asgard signature proteins (ASPs) as those OGs common (present in $\geq 86\%$ of the proteomes, see Methods for the choice of threshold) in Asgard archaea but rare (present in $\leq 15\%$ of the proteomes, following the "cloud" pangenome definition, see ref. 27 in other archaea or bacteria. This rendered 14 ASPs, for which we obtained a consensus of the annotations across Asgard proteomes and found them to be enriched in processes such as endocytosis (KEGG pathway ko04144) or motor proteins (ko04814), see Supplementary Data 2). We cross-mapped a set of 83 InterPro domains associated with 79 Eukaryotic Signature proteins (ESPs) recently identified in Asgard archaea¹⁹ and found 44 out of 79 of them in our dataset. We found all ASPs within the set of OGs comprising ESPs, suggesting that these eukaryotic-like proteins contributed to the differentiation of Asgard archaea from other prokaryotes and that some of these differentiating gene families ended up into the eukaryotic lineage. ASPs include proteins previously considered ESPs involved in typical eukaryotic processes such as Ras-related proteins (K07877), ESCRT-II complex subunit VPS22 (K12188), actin (K10355) and actin-related proteins (K10369), as well as other motor proteins (K10398). In addition to these ESPs, ASPs include other relevant proteins such as motility proteins (K25153), which are putative adaptations to a mat environment²⁸.

All 14 ASP OGs are found in *Ca. Prometheoarchaeum syntrophicum* MK-D1, where they comprise 90 proteins. We used this organism's protein-protein interaction network (PPI) in STRING²⁹ as a template to assess the connectivity of these ASPs (Fig. 1C), which indicated that they are more connected than would be expected by chance (PPI enrichment p -value of $5.32e-09$).

To identify lineage-level signature proteins that may underlie lineage-specific adaptations, we selected OG that are part of the lineage's core and rare outside (cloud-level, $\leq 15\%$). These lineage-signature families were enriched in functions such as "two-component system" (ko02020) in Sifarchaeia, and "ABC transporters" and "Various types of N-glycan biosynthesis" (ko02010, ko00513) in Lokiarchaeia (Supplementary Data 2).

Genome dynamics across Asgard archaea

To assess dynamic changes and their underlying mechanisms in Asgard genomes, we performed a comprehensive phylogenomic analysis. For this, we ran a species overlap³⁰ and tree reconciliation³¹ pipelines (see Methods).

For the species overlap pipeline, we selected ten Asgard and ten Thermoproteota representatives and used them as seeds for the reconstruction of the evolutionary histories of all their proteins in the context of the 60 Asgard+Thermoproteota genomes, such that each phylum was balanced in the dataset. These 20 phylomes comprised a total of 38,889 gene phylogenies (available in PhylomeDB³⁰ as phylomes 1532-1551).

For the reconciliation pipeline, we reconstructed the gene trees of the amino acid sequences of the orthogroups from OrthoFinder that had sequences from at least four species, and included at least one Asgard/Thermoproteota sequence, totaling 4,699 families and 344,709 proteins.

We then performed reconciliation and species-overlap analyses using two alternative species tree topologies for the Asgard clade due to ongoing debates (see Methods) to (i) detect and date gene duplication events, including gene family expansions, (ii) infer orthology and paralogy relationships, and (iii) infer HGT. Although the reconciliation analysis was performed for the full 195-genome dataset, we only show results for the clades of interest, Asgard archaea and Thermoproteota.

The average number of duplications per gene and lineage (duplication rate hereafter, measured as the total number of duplications at a given node divided by the number of gene families present at that node), as computed with either the species-overlap or reconciliation algorithms were significantly correlated (p -value $4E-14$, $R=0.82$), and we show reconciliation-based duplication rates, together with transfer and gene loss rates in Fig. 2A for the Eme et al. topology (results for the alternative topology are shown in Supplementary Fig. 1, which are by and large congruent).

Our results for both topologies confirmed previously described high levels of duplications at the base of Heimdallarchaeia and Lokiarchaeia¹⁹, but extended this observation throughout the Asgard phylogeny. Focusing on the main topology, which we will discuss hereafter except explicitly mentioned, duplication rates were particularly high at the base of the Asgard group and at the branch separating the majority of Asgard lineages from the earliest-diverging clades (that is, excluding Baldr-, Odin-, Jordi- and Atabayarchaeia), with additional hotspots at the base of the above-mentioned Lokiarchaeia, as well as Heimdallarchaeia, a key lineage to eukaryogenesis as current reconstructions suggest eukaryotes either stemmed from within¹⁹ or as sister to³² this clade. Some of the bursts of duplications coincide with the origin of clades with larger genomes, such as Lokiarchaeia and Heimdallarchaeia, hinting towards gene duplication driving increased genome sizes in Asgard archaea. Conversely, lineages with smaller genome sizes are associated with high rates of gene loss. In many cases, gene transfers add to the increase in the genome size, uncompensated by gene loss, such as the base of Asgard, and in many other early nodes. The exceptions to this are at the Heimdall-Njord-Wukong-Sif clade, where most nodes (to the exception of the MRCA of the sampled Heimdallarchaeia) are dominated by gene loss. This result was consistent with the alternate topology.

Importantly, the high duplication rates in Asgard were the result of large expansions in a few gene families (for the OGs for which there is duplication information, 23.87% of the genes have duplicated at least once within the Asgard lineage, see Fig. 2B and Supplementary Fig. 1B, where the result is overall similar), which were functionally diverse. Assessing by COG category, the categories with more than 10 OGs and the highest mean duplications across Asgard were "intracellular trafficking, secretion, and vesicular transport" (U, mean 11.12, SD 59.87), "mobilome" (X, mean 6.68, SD 18.90), and "transcription" (K, mean

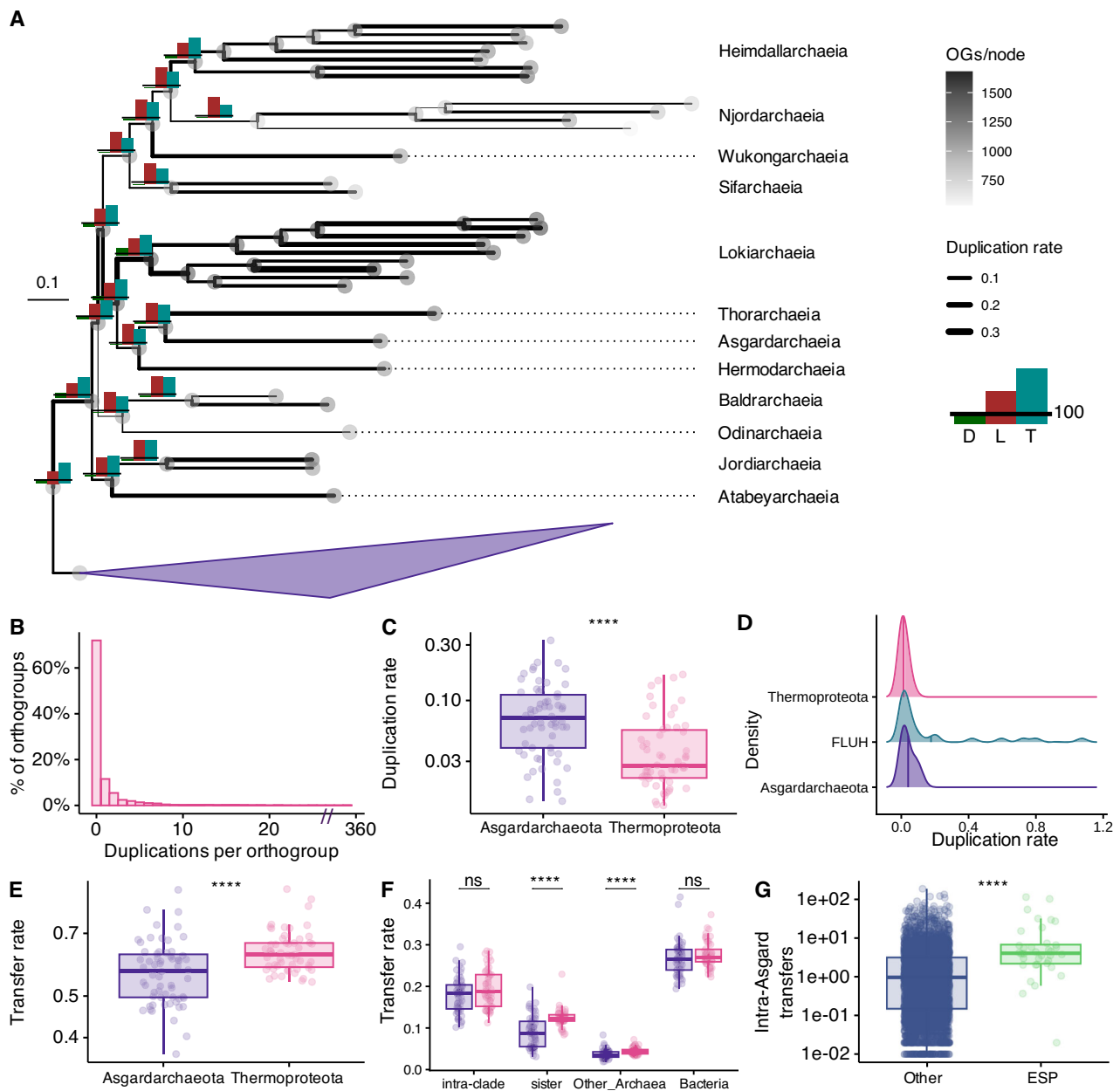


Fig. 2 | Gene gain-loss dynamics across Asgard archaea. A Gain and loss dynamics in Asgard archaea. The total number of OGs per proteome is displayed by circle color in each node. Duplication rate (number of duplication events relative to the inferred number of families present in the node) is represented by the branch width. Number of duplications (green), transfers (blue) and losses (red) displayed at internal nodes in absolute number. Horizontal bar indicates 100 events as scale. **B** Histogram of sum of duplication events per orthogroup across Asgard nodes. Axis cut at 30, we artificially display the tick as the highest value rounded to the multiple of 10. **C** Duplication rates per node of Asgard ($n = 62$) and Thermoproteota ($n = 54$) gene families, statistical significance assessed with two-tailed Wilcoxon rank test, p -value 3.3762E-06. **D** Duplication rates, inferred by species overlap, of Asgard, Thermoproteota and FLUHs. **E** Transfer rates per node of Asgard ($n = 62$)

and Thermoproteota ($n = 54$) gene families, statistical significance assessed by two-tailed Wilcoxon rank-sum test, p -value 1.13123E-04. **F** Transfer rates per node in Thermoproteota ($n = 54$) and Asgard archaea ($n = 62$), separated by donor, statistical significance assessed by pairwise two-tailed Wilcoxon rank-sum tests, p -values 0.107 (intra-clade), 3.92E-07 (sister), 1.28E-05 (other archaea), 0.14 (bacteria). **G** Intra-Asgard transfer events on ESPs compared to the rest of gene families ($n = 36, 4295$ gene families, respectively). Permutation test, p -value 2.2E-16. In the plots with significance values, asterisks * indicate p -values, * (cut-off 0.05) ** (0.01), *** (0.001) **** (0). For the box plots in 2 C, E-G, in each boxplot, the central line indicates the median, the lower and upper limits of the rectangle indicate the first and third quartiles, respectively and the whiskers extend to the minimum and maximum value of 1.5 times the interquartile range.

4.08, SD 12.68). The high values of duplications for OGs associated with the mobilome indicate the genome size increase associated with gene duplication may be driven, at least partially, by these mobile genetic elements. Despite high numbers of transfers throughout the tree, and despite high variability, compared with Thermoproteota, Asgards displayed significantly higher levels of gene duplication (as

mentioned in ref. 30, Fig. 2C also true for Supplementary Fig. 1C), suggesting that they might display duplication levels more akin to those in eukaryotes. To test this, we calculated the gene phylogenies of all the genes in a set of 4 eukaryotic seed genomes against a dataset of 30 Free-Living Unicellular Heterotrophic eukaryotes (FLUHs) and calculated duplication rates for FLUHs, Asgard archaea and

Thermoproteota with the species-overlap algorithm (Fig. 2D and Supplementary Fig. 1D). Asgard archaea displayed duplication rates that were intermediate between those in Thermoproteota and free-living unicellular eukaryotes for both topologies. In this regard, it is also notable that proteins of Asgard origin are the ones that duplicated the most in the proto-eukaryotic lineage³³.

We wondered then if the evolutionary regime governing the evolution of the Asgard genome size was more “prokaryote-like” (deletion-driven) or “eukaryote-like” (neutral) (Supplementary Fig. 2, Methods).

Compared to eukaryotes, prokaryotes have a narrower size range for their genomes, and genome size increases correlate with increases in protein number³⁴, whereas in eukaryotes such increases seem to be driven by ncDNA, especially repetitive elements^{35,36}. To maintain a streamlined genome, prokaryotes couple gene transfer with extensive gene loss³⁴. In both Asgard and Thermoproteota (but more so in Asgard), the number of duplications is positively correlated with the number of gene family copies (Supplementary Fig. 2A), suggesting a role of gene duplication in increasing genome size, whereas we detect no statistically significant correlation between gene gains (duplications and transfers) and gene losses in either group (Supplementary Fig. 2B), nor any statistically significant change in coding density in their genomes (Supplementary Fig. 2C). However, when analyzing the copy number of gene families that contained integrase domains (see Methods), indicative of mobile genetic elements, we saw that, on average, Asgard nodes contain a higher copy number of these elements both in terminal and ancestral nodes (Supplementary Fig. 2D), which may be indicative of their accumulation and therefore of a more neutral regime. Asgard archaea are low-abundance² and with doubling times in the range of weeks⁷, which suggests a low effective population size (N_e) that may weaken selection towards keeping a streamlined genome.

Results for the duplication rate analysis with the species-overlap algorithm can be seen in Supplementary Fig. 3. These results corroborate the aforementioned bursts of gene duplication in the Lokiarchaeia and Heimdallarchaeia clade, as well as in the deeper nodes of the Asgard clade.

Next, for clades with more than 2 representative genomes (to have more than 3 nodes of reference), we looked at the set of “lineage signature proteins” described above to understand the relative contribution of gene duplication to lineage diversification. We compared the duplication rates of these OGs within and outside the lineage, normalized by the overall duplication rate of that node, and found that these signature OGs were more duplicated within than outside the lineage, although it was only significant for Lokiarchaeia (Supplementary Fig. 4), hinting to a possible role of gene duplication in lineage-specific adaptations in these lineages.

We searched for tandem duplications (e.g. resulting in sets of paralogous genes that are contiguous in the genome), and found a highly variable number across Asgard genomes (Supplementary Fig. 5A). Closer inspection of tandem duplications, revealed interesting cases of the same gene, or block of genes, being repeatedly duplicated in tandem (which in some cases were preserved across several genomes), as well as cases of different tandem duplications occurring contiguously in the genome (Supplementary Data 3). We performed a KEGG pathway enrichment for genes within tandem duplications and found them to be enriched in several metabolic functions, including ABC transporters, butanoate metabolism, carbon metabolism, endocytosis, Microbial metabolism in diverse environments, and quorum sensing (Supplementary Fig. 5B and Supplementary Data 4). Paralogs within tandem duplications had signs of purifying selection (median dn/ds of 0.367, sd 0.137, $n = 8694$), suggesting both copies remain functional.

Horizontal gene transfer in Asgard archaea

We then assessed HGT, both intra- and inter-domain, across Asgard archaeal genomes. We first analyzed HGT using gene tree

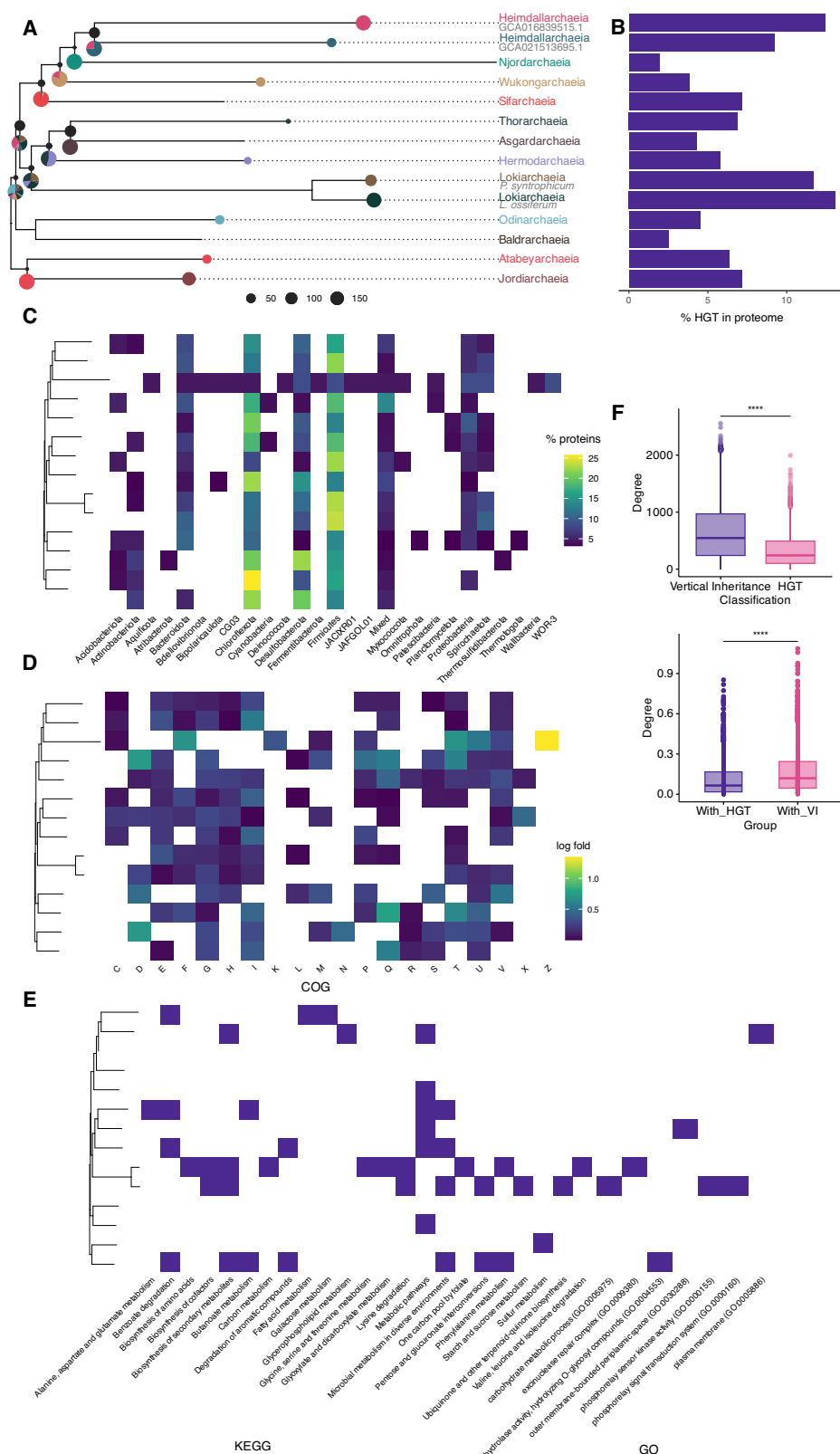
reconciliation algorithms in the above-described phylogenetic dataset (see Methods). On average, and despite high variability, Asgard archaea have significantly lower transfer rates than its sister clade (Fig. 2E and Supplementary Fig. 1E). This difference is mostly driven by lower rates of acquisition of genes in Asgard from other Archaeal donors, including Thermoproteota, whereas acquisitions from Bacterial donors were similar for the main topology (Fig. 2F) but slightly higher for Asgard archaea in the alternate topology (Supplementary Fig. 1F). Among Asgard archaea, for lineages with more than 1 representative, all lineages had similar levels of HGT (Kruskal-Wallis chi-squared = 10.082, $df = 5$, p -value = 0.07295), while the duplication rates were lineage-dependent (Kruskal-Wallis chi-squared = 19.472, $df = 5$, p -value = 0.001569).

We focused on intra-Asgard HGT exchange partners. HGT patterns depict a dense network, in which most partners have exchanged a comparatively low number of genes, but where certain key pairs stand out (Fig. 2A and Supplementary Fig. 6), such as highways between Loki and Heimdall clades. For the transfer pairs beyond percentile 95, we ran a KEGG pathway enrichment of the orthogroups transferred with probability ≥ 0.7 and obtained an enrichment in pathways related to metabolic functions (“Glycolysis/Gluconeogenesis”, “Propanoate metabolism”, “Porphyrin metabolism”, “Degradation of aromatic compounds”, “Pyruvate metabolism”, “Butanoate metabolism”, “Microbial metabolism in diverse environments”, “Benzoate degradation”, and “ABC transporters” were the pathways enriched in at least 2 pairs). Overall, we interpret these results as evidence of a high degree of phylogenetic incongruence in the context of a strong phylogenetic backbone, with frequent and dispersed HGT. Importantly, intra-Asgard HGT has been proposed as a potential factor explaining the absence of ESPs in Heimdallarchaeia, the closest Asgard relatives of Eukaryotes³⁷. In this regard, we observed that, for OGs that are transferred within Asgard nodes, ESPs had a significantly higher number of transfers as compared to other genes (Fig. 2G).

To assess inter-domain HGT, we performed an exhaustive search restricted to fourteen seed Asgard genomes by using a combined BLAST- and phylogeny-based approach, as well as using a broader database comprising 491,850 sequenced genomes from prokaryotes, eukaryotes and viruses (see Methods).

BLAST-based approaches suggested a large amount of HGT, involving between 11.37 and 19.36% (SD 2.37) of the total gene repertoire (Supplementary Data 5). As many BLAST-based HGT predictions are likely to be false positives, we reconstructed phylogenies for each candidate and ran two different phylogeny-based HGT detection algorithms (see Methods). This approach predicted 3000 gene transfer events in the 14 investigated genomes, involving 1.94–13.10% (SD 3.56) of the gene repertoire. These HGT events involved 980 OGs, out of which 922 were in the AleRax dataset, that, on average, had both a significantly higher number of duplications (Wilcoxon Rank Sum Test, two-tailed, $W(922,3752) = 2336359$ p -value $1.04E-67$) and transfers (permutation test, p -value $2.2E-16$) within Asgard nodes, highlighting the complex history of these gene families. The overall good agreement between the two phylogeny-based HGT detection approaches, the generally high phylogenetic support (Supplementary Fig. 7A), the confirmation of a previously described HGT event³⁸, and the lack of correlation with genomic contamination estimates (Supplementary Fig. 7B), suggested that our pipeline is rather conservative and that our phylogeny-based inter-domain HGT inferences are likely to be correct. We additionally kept a subset of transfer events for which the bootstrap was well-supported ($\geq 70\%$), which we labeled as the “high-confidence” set, to test that the qualitative conclusions held at higher stringency.

Based on the gene tree topologies, we inferred the directionality of the transfers (Supplementary Fig. 7C), which now allows us to discern between Asgard-to-Bacteria or Bacteria-to-Asgard transfers. Although our pipeline was designed to primarily search for Bacteria-to-



Asgard HGT, we captured some Asgard-to-Bacteria transfers. We used the last common ancestor of the genomes (see Methods) carrying the transferred genes as a proxy for the relative time of the event, and mapped this onto the Asgard species tree (Fig. 3A).

We assessed the relative contribution of the different bacterial clades as exchange partners (irrespective of HGT directionality). Our results (Fig. 3C) show many different partners, with varying levels of

taxonomic width. Interestingly, some of the identified partners belong to taxonomic groups that are known to co-occur or interact with extant Asgards. For example, Lokiarchaeota co-occurs with Deltaproteobacteria (phylum Proteobacteria) and Anaerolineae (phylum Chloroflexota), both of which appear as significant transfer partners (>5% of trees). Desulfobacterota (*sensu* GTDB) is particularly interesting in the Lokiarchaeia lineages, as both *Ca. P. syntrophicum* and *Ca. L.*

Fig. 3 | HGT-derived proteins across the Asgard lineage. **A** Prevalence of HGT across the Asgard phylogeny. Dot size represents the mean number of gene trees with HGT events mapped as having occurred in that given ancestor, average of the HGT events detected per seed. Pie chart below the dots represents relative contribution of the different seeds to the inferred HGT events. Each seed (tip) proteome has a color. **B** Percentage of genes involved in HGT, relative to the total number of protein-coding genes per lineage. **C** Main partner phyla per lineage that have contributed at least 3% of the HGT genes to the given lineage. **D, E** KEGG, COG, and GO term enrichment for the proteins derived from Bacteria-to-Asgard HGT. **F** Network properties of HGT-acquired genes versus vertically inherited ones:

ossiferum has known syntrophic partnerships with such bacteria, particularly of the *Desulfovibrio* and *Halodesulfovibrio* genera^{4,7}. Importantly, most of the Desulfobacterota HGT partners identified by our pipeline belong to lineages different from the co-cultured symbionts, discarding contamination as a potential source.

To understand the impact of HGT in Asgard evolution, we focused on bacteria-to-Asgard transfers in subsequent analyses. A functional assessment of the transferred genes further supports the lineage-specificity of the transfers, with some trends in the broad sense (see COG categories in Fig. 3D) but with all enriched GO terms displaying a patchy distribution (Fig. 3E). Similar results were obtained when using the high-confidence set, and the same applies for KEGG functional categories (Fig. 3E), despite the general KEGG category “Metabolic pathways” being a common enrichment. This is in line with previous observations in prokaryotes as a whole, where ancient transfers have been found to be enriched for genes involved in amino acid, carbohydrate and energy metabolism²⁰. As for COG categories, which allow a broader overview, some key categories stand out, such as Z (cytoskeleton) for Njordarchaea, T (signal transduction mechanisms) and Q (secondary metabolites biosynthesis, transport and metabolism) and I (lipid transport and metabolism) across several lineages.

ThyX, previously identified as an HGT event in Lokiarchaea³⁸, is confirmed in this analysis, as “one carbon pool by folate” is 4.31-fold enriched in *Ca. P. syntrophicum* and is also enriched in the Hodarchaeales (GCA016839515.1), Heimdallarchaea (GCA021513695.1), Thorarchaea (GCA_004524565.1) and Jordarchaea (GCA037305845.1) representatives as well as in *Ca. L. ossiferum* (GCA025839675.1), despite none of these remaining significant after FDR correction.

We next asked whether HGT-derived proteins were integrated into the recipient organism's interactome. To do so, we calculated the proteome-wide protein-protein interactions using STRING²⁹ and assessed the topological properties of the resulting interactome. We found that, within this network, HGT-derived proteins have a smaller degree, that is, they were less connected than proteins of vertical descent, and were less central in the network (Fig. 3F). This was true for all lineages, and for other centrality measures (Supplementary Fig. 8A, B).

We found that HGT-derived proteins tend to be more connected to proteins of vertical origin than to other HGT proteins, when normalizing by their respective abundance (Fig. 3E). In fact, we found that the KOs brought about by the HGT events are to some extent functionally redundant, as on average, 40.93% (SD 11.66) of the HGT-derived KOs are also found in a gene that is vertically inherited in the same genome. We also estimated the metabolic coherence of these HGT proteins by estimating the metabolism that could be completed with just these HGT KOs with anvio (anvi-estimate-metabolism). On average, per genome, the modules that can be assembled from these HGT KOs have a mean pathway completeness of 0.1955 (SD 0.0438). Therefore, we conclude that HGT does not generally involve the addition of whole modules, but rather a stepwise addition of enzymes to native pathways. Although this was the overall trend, in some cases HGT-derived genes formed a stepwise complete module, thereby expanding the metabolic potential of the recipient organism: it is the

degree (number of interactions) of the proteins of HGT origin ($n = 2013$) and of VI origin ($n = 23443$, two-tailed t-test, p -value $< 2.22E-16$) (above); for the HGT proteins, we show the number of connections per node of the subnetwork composed only of HGT proteins (with_HGT) and calculate the number of connections with VI proteins ($n = 2013$ nodes, two-tailed Wilcoxon rank-sum test, p -value $< 2.22E-16$). Degrees are normalized by the number of proteins in each category. In each box-plot, the central line indicates the median, the lower and upper limits of the rectangle indicate the first and third quartiles, respectively, and the whiskers extend to the minimum and maximum value of 1.5 times the interquartile range.

case for M00046 in GCA016839515.1, M00632 in *L. ossiferum* and M00086/M00045 in *P. syntrophicum*. Nevertheless, given the overlap of KOs across pathways and the uncertainty of functional inferences, claims for contributed metabolic potential should be considered with caution.

We searched for “HGT islands”, that is, segments of the genome in which more than one HGT gene is located contiguous to another HGT gene or interspaced by a single vertically inherited gene. Most of the HGT genes are interspaced with vertically inherited genes, diminishing the likelihood that they are due to errors in the metagenome assembly. However, we found 340 HGT “islands”, comprising 797 Bacteria-to-Archaea HGT genes (on average, 31.84% of the Bacteria-to-Archaea HGT genes, SD = 11.63). These islands contained up to 8 genes (mean 2.77, SD 1.09). 48 of these islands contained genes belonging to the same COG category, suggesting that they may be operons. We studied these cases in detail. Some of these islands were composed of the same transferred gene duplicated in tandem, such as GCA016840465.1JPJMNJDP_02194 and GCA016840465.1JPJMNJDP_02195 from Desulfobacterota to Baldrarchaeum (bcrC) or GCA037305845.1JPIBEII_00281 and GCA037305845.1JPIBEII_00282 (CRYZ) from Patescibacteria to Jordarchaea. An illustrative example of putative operons is GCF008000775.2JEDDKEKE_02754 and GCF008000775.2JEDDKEKE_02755, which are excinuclease ABC subunit B and A transferred from Firmicutes to the CR-4 lineage, while protein C is GCF008000775.2JEDDKEKE_00387, or GCA016840425.1IAHJMLL_N_00393, GCA016840425.1IAHJMLLN_00394 and GCA016840425.1IAHJMLLN_00395, transferred to the Heimdall-Loki clade, likely from Nitrospirota that are zinc-transport system znuA/B/C.

Last, we set out to assess which HGT proteins may have reached the eukaryotic lineage via the Asgard ancestor of the Last Eukaryotic Common Ancestor (LECA). We assessed all 331 trees flagged as HGT (both Asgard-to-Bacteria and Bacteria-to-Asgard, as we were also interested in assessing whether proteins in eukaryotes may be falsely identified as of bacterial origin), which contained at least one eukaryotic sequence, and 28 contained sequences from at least 3 eukaryotic supergroups, suggesting they might have been present in LECA³⁹. Upon manual inspection of these candidate phylogenies, we did not find any clear case where a gene of bacterial origin was horizontally transferred to Asgard archaea and passed on through the eukaryotic lineage by vertical inheritance. This has interesting implications for eukaryogenesis, as it implies that the proteins of bacterial origin beyond Alphaproteobacteria that are detected in LECA are unlikely to have come from Bacteria-to-Asgard HGT events in the Asgard lineage prior to eukaryogenesis, strengthening the credibility of the implication of more bacterial partners during eukaryogenesis⁴⁰. Interestingly, however, we do find instances where viruses (mainly Nucleocytoviricota) seem to act as mediators for the transfer of genes from prokaryotes to eukaryotes, supporting and expanding previous observations³⁹. These viruses, of Algavirales, Imitervirales and Pimascovirales clades, mainly infect microbial eukaryotes of varied phylogenetic affiliations (amoebozoia and other protists)⁴¹.

Discussion

Altogether, our findings depict a high dynamism in terms of content and size of Asgard archaeal genomes and provide the first comprehensive assessment of gene duplication, loss, and HGT within this recently discovered and relevant clade. We reconstructed a large and open Asgard archaeal pangenome, which likely reflects the clade's immense functional and ecological diversity and suggests a vast, untapped genomic diversity. While acknowledging the current dominance of metagenome-assembled genomes (MAGs) and the scarcity of cultured representatives, which may lead to incorrectly inferred absences due to genome incompleteness, we anticipate the pangenome will likely remain open even with a greater diversity of culture-based genomes.

We identified a core set of 14 Asgard signature proteins (ASPs) widespread in the clade and rarely found in prokaryotic out-groups. Significantly, these ASPs are enriched in endocytosis, a function relevant to eukaryogenesis^{42,43}. Notably, several ASPs are annotated with functional domains shared with eukaryotes and were previously considered Eukaryotic Signature Proteins (ESPs)², indicating that Eukaryotes inherited some of the distinctive Asgard archaeal features.

Our comprehensive phylogenomic analysis revealed high gene duplication rates across all lineages—a finding atypical for prokaryotic genomes. Indeed, Asgard archaea exhibited duplication rates intermediate between their closest archaeal relatives (Thermoproteota) and eukaryotic unicellular free-living heterotrophs. This underscores the “stepping-stone” nature of Asgard archaea, even in this genomic trait. Similar to eukaryotes, Asgard archaeal duplications preferentially affected specific gene families, showed lineage-specific patterns, often occurred in tandem, and, alongside gene loss, emerged as a primary driver of genome size. Among the lineages with the highest duplication rates were Heimdallarchaeia (considered the closest sister group to eukaryotes¹⁹), Lokiarchaeia and Sifarchaeia.

We also assessed intra-Asgard transfer, uncovering an intricate network superimposed on a strong vertical inheritance backbone. Intriguingly, genes encoding ESPs were more frequently transferred than other genes, which could explain their previously observed patchy phylogenomic distribution². We estimated that 2–13% of the genes in Asgard genomes were involved in inter-domain HGT with bacteria. This phenomenon is pervasive across the entire Asgard phylogeny and involves bacterial partners that co-occur with Asgard archaea in metagenomic samples. HGT rates are lower than those found in Thermoproteota (when normalizing for genome size), a trend driven by a lower transfer rate with other Archaea (including Thermoproteota), suggesting HGT alone does not account for increased genome sizes in Asgard archaea. The functions of acquired bacterial genes were highly diverse and lineage-specific, predominantly metabolic, and were found to be peripheral in protein-protein and metabolic networks, interspersed with genes of vertical inheritance. This pattern suggests a piecemeal integration into established pathways, perhaps replacing ancestral archaeal proteins, rather than the wholesale addition of new functional modules.

In summary, our work provides a detailed survey of genome content dynamics in Asgard archaea, defines the signature proteins of this important clade, and uncovers a rich dynamism of gene family duplication that is atypical for prokaryotes but reminiscent of processes observed in eukaryotic organisms. These insights offer crucial mechanistic details about the genomic evolution of this pivotal lineage and its implications for understanding the origins of eukaryotic complexity.

Methods

Data retrieval and annotation

From GTDB r226 we compiled a set of 32 genomes, covering all Asgard family-level lineages *sensu* GTDB and chosen for their best assembly quality parameters (applying the score of CheckM completeness -

5*CheckM contamination), except in the cases in which closed genomes were available, which were prioritized over metagenomes as they served as a baseline that the findings of HGT were not due to contaminating sequences in the metagenomes. We generated a second dataset of Asgard archaea, which was a random set of equal size and same lineage-based representation for quality control of the pangenomic measures. Taxonomies were assigned *sensu* GTDB.

We additionally retrieved representatives of the Thermoproteota superphylum (“p_Thermoproteota” *sensu* GTDB), one per order-level lineage, as well as 1 representative per phylum for the rest of the prokaryotic phyla. This generates a dataset of 195 genomes in total.

To avoid biases in genome annotation, we re-annotated all genomes using Prokka v.1.14.6⁴⁴ with the flags --kingdom Archaea (or --kingdom Bacteria, when necessary) and --metagenomes, as Prokka is a reliable and widely used prokaryotic genome annotator. We adapted the headers of the proteome files, changed the NCBI taxID when the one in the GTDB metadata table was too broad or conflicting, and uploaded them to PhylomeDB³⁰ via the Phylomizer.

We annotated the Gene Ontology (GO) terms and domains associated with each protein using InterProScan v5.39-77.0⁴⁵ (filtering for score $\leq 1E-3$), and the COG categories and KEGG KOs with the COG HMMs⁴⁶ and KoFamScan (https://github.com/takaram/kofam_scan), respectively. To perform functional assessments across the Asgard pangenome, we also annotated the orthogroups by compiling the annotations of their member proteins following the pipeline in ref. 39. In brief, we followed the strategy followed in eggNOG, which takes into account the proportion of the KO and COG within the set of sequences and its overrepresentation with respect to the background KOs distribution (the whole set of annotated proteins).

When querying for ESPs, we performed a manual curation of InterProScan results. To assign an ESP to a given OG (for example, for the results in Fig. 1), we took the OG where the majority of the proteins from that ESP were grouped (or several in case of ties).

We took the annotations from the Asgard proteins for the consensus, to keep the inferred function as close as possible to the actual one performed in Asgard cells, except for when looking for integrase domains, where we took into account Thermoproteota annotations as well.

Last, we taxonomically downsampled the list of Free-Living Unicellular Heterotroph (FLUH) proteomes from Bernabeu et al.³⁹, by keeping a proportion of proteomes per supergroup to a total of 30.

The list of genomes, along with associated metadata, can be found in Supplementary Data 1.

Selected thresholds for Asgard signature proteins

We defined signature proteins of the Asgard clade (which we termed Asgard Signature Proteins or ASPs) as those common within Asgard genomes and rare in the outgroups. We determined the abundance thresholds as follows. First, we defined the outgroup abundance threshold at 15%, which is the upper limit defined for the “cloud” component of a pangenome as defined in ref. 27. For the presence within Asgard archaea, we considered the lower limit of the extended core (90%) as too strict, given the high genome incompleteness within the set (see “Results”). Therefore, we analyzed the prevalence of housekeeping protein families that can be thought to be core, in the sense that they perform essential functions for the cell. For this, we annotated the orthogroups by KEGG KOs and COG categories and assessed the prevalence within the Asgard group of protein families annotated as the COG category J (Translation, ribosomal structure and biogenesis) and of the orthogroups annotated as the KOs for the ribosomal proteins in archaea (BRITE for ko03011). We selected a threshold of prevalence of 86%, as this is the median prevalence for the COG category J and is similar to the median for ribosomal proteins (87%). This resulted in 14 orthogroups selected as Asgard Signature Proteins.

Alternative species tree reconstruction

We inferred branch lengths of the Asgard species tree by retrieving the markers for all archaeal species from GTDB, aligning them with MAFFT-linsi v7.407⁴⁷, trimming with trimAl v1.4.rev15⁴⁸, and reconstructing the species tree with IQ-TREE v1.6.9⁴⁹ (–models DCMut, JTTDCMut, LG, WAG, VT), both constraining the topology to the current consensus (see ref. 19) for the main lineages or to the GTDB topology, resulting in an alternative topology to benchmark results.

Asgard archaeal pangenome assessment

We performed an orthogroup clustering with OrthoFinder v2.5.4⁵⁰ using BLAST⁵¹ as an aligner. To assess pangenome metrics, such as Chao's lower bound, Heaps estimate, and the binomial mixture models, we employed the R package micropan⁵².

We defined Asgard signature proteins (ASPs) as those OG common (present in $\geq 86\%$ of the proteomes) in Asgard archaea but rare (present in $\leq 15\%$ of the proteomes) in other prokaryotes (see above). We mapped ASPs onto the Protein-Protein Interaction Network (PPI), obtained with STRING v12.0²⁹, of the seed genome with the highest number of ASPs (*Ca. P. syntrophicum*) using the STRING web server.

To perform functional enrichments, we collapsed the annotations at the orthogroup level following the aforementioned pipeline.

Genome dynamics across Asgard archaea

We constructed a metaphylome consisting of a collection of phylomes employing one seed per major lineage. For the phylome analysis, we took the 60 representative genomes of Asgard archaea and Thermo- proteota as the dataset and selected 10 seed genomes per phylum that covered the diversity of the phyla. For the set of FLUHs, we selected 4 seed genomes and 30 of the aforementioned 30 representatives.

The phylome reconstruction pipeline was applied as defined elsewhere (see ref. 30. for full details): In brief, for each of the genes contained within each of the seed genomes, we performed a protein BLAST⁵¹ search against all proteomic data to retrieve proteins with significant similarity (defined as e-value $< 1e-05$, continuous overlap over 50% of the query sequence). We then took the top 150 hits per protein and generated an MSA with three different methods (MUSCLE V3.8.1551⁵³, MAFFT v7.407⁵⁴ and KALIGN v2.04⁵⁵), in forward and reverse, and combined the resulting six alignments into one consensus alignment with M-Coffee⁵⁶. This alignment was then trimmed with trimAl v1.4.rev15⁴⁸, with –gappyout. These alignments were then used to reconstruct the phylogenetic tree of each protein family with IQ-Tree v1.6.9⁴⁹, for which we selected the final Maximum Likelihood (ML) tree using the best model selected based on the Bayesian Information Criterion (BIC), and support was calculated with ultrafast bootstrap under 1000 repetitions.

We then ran the species overlap algorithm following the Phylo- meDB pipeline³⁰. We employed both topologies for the prokaryotic dataset and the species tree as inferred by DupTree⁵⁷ for the eukaryotic set.

For the reconciliation algorithm, which we ran for the prokaryotic set as we infer HGT to play a significant role in its genome dynamics, we built gene trees of the OGs from OrthoFinder that had sequences from more than 4 species following the tree reconstruction pipeline in (which is an update of that upgrades MUSCLE to 5.2, MAFFT to v7.525, KALIGN to 3.5.0 and IQ-Tree to 3.0.1, and substitutes M-Coffee by Mergealign 3⁵⁸) ran AleRax³¹ v1.3.0 using as input the distribution of bootstrap trees (as supplied by IQ-TREE) and the generated species tree as backbone, estimating rates per family.

Evidence of selection

Within tandem duplications, we looked for signs of positive selection by assessing the dn/ds ratio. For this, we built codon alignments and assessed the dn/ds ratio employing the Python library biopython v1.86.

Cases where the sequence had “N” or where there were no synon- ymous mutations were discarded from the analysis.

Mobile genetic elements

We took the InterProScan annotation of the Asgard and Thermo- proteota proteomes and assessed the prevalence of signatures related to Mobile Genetic Elements (MGEs). For this, we took as a reference the list of InterPro domains employed by proMGE⁵⁹. We took the orthogroups for which at least 50% of the proteins with annotations were annotated (score $\leq 1E-3$) with a given domain within this list. We then assessed the number of copies per node of these orthogroups by summing the copies of each orthogroup.

Inter-domain HGT assessment

We performed a similarity search with BLAST 2.11.0⁵¹ of the proteomes against a custom-made database (BROADDB_V2), which comprises all species representatives from GTDB r207⁵, viral sequences from RVDB 25.0⁶⁰, and eukaryotes from UniRef v2023_01⁶¹ and EukProtV1⁶². We chose not to employ NCBI NR as it does not provide a sufficient representation of the Asgard genomes. We employed the GTDB tax- onomy for prokaryotes as it allows a more comprehensive, robust and fine-grained taxonomic analysis as compared to the NCBI taxonomy.

We analyzed the BLAST results with HGTector v2.0b3⁶³, which systematically looks for hit distribution patterns incongruent with vertical evolution, given a series of hierarchically defined evolutionary categories. We ran the algorithm, setting the “self” group as Asgard archaea and the “close” group as Archaea, therefore allowing us to focus on the detection of inter-domain HGT events. The output sta- tistics can be found in Supplementary Data 5 and the output of HGTector in the data repository.

For candidate transfers selected by HGTector, we retrieved the best 300 BLAST hits with an e-value cut-off of 0.001 and used them to reconstruct a gene tree following the pipeline in ref. 39.

When the selected set of homologous sequences was saturated with eukaryotic sequences (that is, if they comprised $\geq 80\%$ of the non-Asgard sequences) or Asgard sequences, we performed a sub- sampling of these clades by selecting the first hit per major eukaryotic group as defined in EukProt taxonomical framework⁶², then retrieved all hits from those organisms that had a lower e-value than the lowest e-value for a non-Asgard prokaryotic sequence for that query. We kept the non-eukaryotic sequences from the BLAST that passed the threshold, the subsampled eukaryotic sequences, and up to 250 non- eukaryotic sequences passing the threshold performed in a prokaryotic-only BLAST. This allows us to have taxonomically infor- mative insight into the distribution of the sequences across eukar- yotes, whilst allowing us to analyze the distribution of more distant prokaryotic homologues.

We further analyzed the resulting gene trees using our custom script (HGT_detection.py) (see below for the pseudo-code) and with Abaccus v1.2⁶⁴, which identifies taxonomical “jumps” within gene trees that do not follow the species tree and therefore further helps discern putative HGT events. For this, we used the taxonomy *sensu* GTDB for prokaryotes.

There are several methods to root a phylogenetic tree in the absence of a clear outgroup. We tested the number of HGT events called by rooting the trees by midpoint and Minimal Ancestral Devia- tion (MAD), which resulted in a 94% overlap in the genes detected as HGT (2,946), 2% MAD-only (54) and 5% midpoint-only (244). Given the small difference, we decided to proceed with MAD rooting as imple- mented in toytree⁶⁵.

Additionally, we retrieved the support of the branch of the transfer (e.g., the branch subtending the acceptor Asgard clade and the putative non-archaeal donor clade). We only kept cases when this support was $\geq 50\%$, or, if the node was poorly supported, cases where

the sister of the transfer branch contained $\leq 10\%$ of archaeal sequences were also considered, even if the specific donor clade could not be ascertained. To ensure that branch support was not affecting the conclusions of the analysis, we built a subset of well-supported ($\geq 70\%$ support) HGT cases in which all analyses were also run.

Last, we manually curated the HGT results by filtering out incongruencies and false positives resulting from trees saturated with eukaryotic sequences. We performed this step for all proteins flagged by HGTector, whether they were picked up by Abacus or not.

To assess the directionality of the transfer (Asgard-to-bacteria or bacteria-to-Asgard), we assessed whether there were Asgard sequences in sister2; if there were, Asgard-to-bacteria was assumed.

HGT detection pipeline

To do so, we first numbered the internal nodes in the Asgard tree. Then, we parsed the gene trees with an in-house script (HGT_detection.py) based on ete3⁶⁶, using the following criteria:

- Trees were rooted with a Minimum Ancestor Deviation (MAD)⁶⁷.
- The start node was established as the one subtending the largest Asgard monophyletic clade containing the seed protein. In assessing monophyly, we considered eukaryotes as a *de facto* Asgard lineage stemming from Heimdallarchaea. The most recent common ancestor (MRCA) of the Asgard lineages present in the clade is assumed as the acceptor Asgard lineage.
- Then, the sister clade of the start node was analyzed:
- If the fraction of non-Asgard Archaeal sequences in the sister clade was $\geq 10\%$, the tree was labeled as a False Positive. This means that either vertical inheritance cannot be ruled out, or that an Archaeal-to-Asgard HGT cannot be ruled out, which is also beyond the scope of this analysis since we are focusing on inter-domain HGT.
- The donor was set as the MRCA of the non-Asgard sequences of the sister clade. In cases where the MRCA is a broad category (e.g., “Bacteria”), we established the predominant phyla as the putative donor lineage. For cases with low support ($\leq 50\%$), we marked the donor as “uncertain”.
- Additionally, taxonomy information of the sister clade to the (start+sister) branch (sister2) was retrieved.
- We assessed the percentage of non-Asgard sequences in the tree that were of eukaryotic origin. In cases where this percentage was $\geq 80\%$ were labeled eukaryote-saturated and submitted to the subsampling mentioned above and the subsampled gene tree was re-analyzed.
- After this analysis, we additionally flagged as FP events for which support was low ($\leq 50\%$) and sister2 contained $\geq 10\%$ non-Asgard archaeal sequences. These criteria are very restrictive and likely result in an underestimation of the true number of HGT events: namely, we will miss inter-domain HGT, and the 10% of Archaeal sequences in the sister needed to label a tree as a false positive will lead to true HGTs being missed due to secondary transfers of the gene family from bacteria to other archaea. However, this ensures that the trees that are picked up as HGT are true HGT events.

One of the limitations of the analysis of acceptor nodes of horizontally transferred genes is the use of Maximum Parsimony. There are methods better suited for this (e.g., reconciliation, as used for the estimation of duplication, transfer and loss rates), but these become computationally prohibitive given that running time scales cubically to the number of species and given that the search is done against a broad database of almost 500,000 genomes. Moreover, for the purpose of analyzing the tempo of transfers across the Asgard phylogeny (that is, whether there are “hotspots” or it is more widespread), we believe a more simplistic assumption, like Maximum Parsimony, is sufficient for this purpose.

Tandem duplications and HGT islands

To infer tandem duplications, we used a window-based approach to query the GFF files, where we looked for 2 or more genes belonging to the same OG being either directly adjacent or interspaced by another gene, which we applied with a custom R script (Tandem_Duplications.R). For HGT islands, we also queried the GFFs to find HGT genes immediately adjacent or interspaced by a vertically inherited gene to reconstruct “islands”. We visualized it with the R package gggenes 0.5.1 (<https://github.com/wilcox/gggenes/tree/master>).

Functional enrichment analysis

We performed GO term enrichment analyses using the Bioconductor package topGO (<https://git.bioconductor.org/packages/topGO>), with the method “weight01” and using Fisher’s exact test, correcting for multiple testing and keeping results with a False Discovery Rate (FDR) of ≤ 0.15 . For KEGG pathway enrichment, we employed the package ClusterProfiler⁶⁸ with the same statistical criteria except in the tandem duplications, where we cut to FDR ≤ 0.1 for visualization purposes. We visualized the results with the R package ggpubr (<https://github.com/kassambara/ggpubr/>) and ggtree/ggtreeExtra^{69,70}.

Protein–protein interaction networks

We further analyzed the functional relationship between transferred genes by assessing the network of protein–protein interactions (PPI). Since none of the analyzed Asgard genomes are available in the database, we first submitted them to be added to STRING²⁹ v.12.0 by making use of the tool developed in the latest version⁷¹. We then uploaded the sequences of the transferred genes to the web server and ran the analysis using the newly uploaded proteome as the target organism. We downloaded the results, performed analysis with the R implementation of iGraph (<https://github.com/igraph/igraph>) and visualized the networks with the R package ggnetwork (<https://journal.r-project.org/archive/2017/RJ-2017-023/index.html>).

Reporting summary

Further information on research design is available in the Nature Portfolio Reporting Summary linked to this article.

Data availability

BioProject IDs for the genomes are available in Supplementary Data 1. For the representative set of Asgard genomes, the publications (if any) associated with the assemblies are as follows: GCA_001940655.1²; GCA_005191415.1 and GCA_005191425.1⁷²; GCA_014729935.1⁷³; GCA_015523565.1⁷⁴; GCA_024720975.1⁷⁵; GCA_021513695.1¹⁸; GCA_001940665.2⁷⁶; GCA_027061945.1, GCA_027068385.1, and GCA_026993975.1⁷⁷; GCA_016292335.1⁷⁸; GCA_029856635.1¹⁹; GCA_040225955.1 and GCA_029838215.1⁷⁹; GCA_030638685.1 and GCA_030671945.1⁸⁰; GCA_038858525.1⁸¹; GCA_037305845.1 and GCA_037308085.1⁸²; GCF_008000775.2^{7,83}. All input proteomes and output files for all steps can be found in Zenodo (10.5281/zenodo.15789115).

Code availability

All described scripts are available on GitHub (<https://github.com/GabalDonlab/Asgard>)⁸⁴.

References

1. Koonin, E. V. The origin and early evolution of eukaryotes in the light of phylogenomics. *Genome Biol.* **11**, 209 (2010).
2. Zaremba-Niedzwiedzka, K. et al. Asgard archaea illuminate the origin of eukaryotic cellular complexity. *Nature* **541**, 353–358 (2017).
3. Akl, C. & Robinson, R. C. Genomes of Asgard archaea encode profilins that regulate actin. *Nature* **562**, 439–443 (2018).
4. Rodrigues-Oliveira, T. et al. Actin cytoskeleton and complex cell architecture in an Asgard archaeon. *Nature* <https://doi.org/10.1038/s41586-022-05550-y> (2022).

5. Parks, D. H. et al. GTDB: an ongoing census of bacterial and archaeal diversity through a phylogenetically consistent, rank-normalized and complete genome-based taxonomy. *Nucleic Acids Res.* **50**, D785–D794 (2022).
6. Mattock, J. & Watson, M. A comparison of single-coverage and multi-coverage metagenomic binning reveals extensive hidden contamination. *Nat. Methods* **20**, 1170–1173 (2023).
7. Imachi, H. et al. Isolation of an archaeon at the prokaryote-eukaryote interface. *Nature* **577**, 519–525 (2020).
8. Spang, A. et al. Complex archaea that bridge the gap between prokaryotes and eukaryotes. *Nature* **521**, 173–179 (2015).
9. Inagaki, F. et al. Biogeographical distribution and diversity of microbes in methane hydrate-bearing deep marine sediments on the Pacific Ocean Margin. *Proc. Natl. Acad. Sci. USA*. **103**, 2815–2820 (2006).
10. Vetriani, C., Jannasch, H. W., MacGregor, B. J., Stahl, D. A. & Reysenbach, A. L. Population structure and phylogenetic characterization of marine benthic Archaea in deep-sea sediments. *Appl. Environ. Microbiol.* **65**, 4375–4384 (1999).
11. Takai, K. & Horikoshi, K. Genetic diversity of archaea in deep-sea hydrothermal vent environments. *Genetics* **152**, 1285–1297 (1999).
12. Manoharan, L. et al. Metagenomes from coastal marine sediments give insights into the ecological role and cellular features of Lokiarchaea and Thorarchaea. *MBio* **10**, e02039–19 (2019).
13. Cai, M. et al. Ecological features and global distribution of Asgard archaea. *Sci. Total Environ.* **758**, 143581 (2021).
14. Orsi, W. D. et al. Metabolic activity analyses demonstrate that Lokiarchaeon exhibits homoacetogenesis in sulfidic marine sediments. *Nat. Microbiol.* **5**, 248–255 (2020).
15. Liu, Y. et al. Comparative genomic inference suggests a mixotrophic lifestyle for Thorarchaeota. *ISME J.* **12**, 1021–1031 (2018).
16. Bulzu, P.-A. et al. Casting light on Asgardarchaeota metabolism in a sunlit microoxic niche. *Nat. Microbiol.* **4**, 1129–1137 (2019).
17. Pushkarev, A. et al. A distinct abundant group of microbial rhodopsins discovered using functional metagenomics. *Nature* **558**, 595–599 (2018).
18. Wu, F. et al. Unique mobile elements and scalable gene flow at the prokaryote-eukaryote boundary revealed by circularized Asgard archaea genomes. *Nat. Microbiol.* **7**, 200–212 (2022).
19. Eme, L. et al. Inference and reconstruction of the heimdallarchaeal ancestry of eukaryotes. *Nature* **618**, 992–999 (2023).
20. Dmitrijeva, M. et al. A global survey of prokaryotic genomes reveals the eco-evolutionary pressures driving horizontal gene transfer. *Nat. Ecol. Evol.* **8**, 986–998 (2024).
21. Puigbò, P., Lobkovsky, A. E., Kristensen, D. M., Wolf, Y. I. & Koonin, E. V. Genomes in turmoil: quantification of genome dynamics in prokaryote supergenomes. *BMC Biol.* **12**, 66 (2014).
22. Treangen, T. J. & Rocha, E. P. C. Horizontal transfer, not duplication, drives the expansion of protein families in prokaryotes. *PLoS Genet.* **7**, e1001284 (2011).
23. Parks, D. H., Imelfort, M., Skennerton, C. T., Hugenholtz, P. & Tyson, G. W. CheckM: assessing the quality of microbial genomes recovered from isolates, single cells, and metagenomes. *Genome Res.* **25**, 1043–1055 (2015).
24. Chao, A. Estimating the population size for capture-recapture data with unequal catchability. *Biometrics* **43**, 783–791 (1987).
25. Manzano-Morales, S., Liu, Y., González-Bodí, S., Huerta-Cepas, J. & Irazo, J. Comparison of gene clustering criteria reveals intrinsic uncertainty in pangenome analyses. *Genome Biol.* **24**, 250 (2023).
26. Buck, M., Mehrshad, M. & Bertilsson, S. mOTUp: a robust Bayesian approach to leverage metagenome-assembled genomes for core-genome estimation. *NAR Genom. Bioinform.* **4**, lqac060 (2022).
27. Maistrenko, O. M. et al. Disentangling the impact of environmental and phylogenetic constraints on prokaryotic within-species diversity. *ISME J.* **14**, 1247–1259 (2020).
28. Nobs, S.-J. et al. An Asgard archaeon from a modern analog of ancient microbial mats. Preprint at <https://doi.org/10.1101/2025.07.22.663070> (2025).
29. von Mering, C. et al. STRING: a database of predicted functional associations between proteins. *Nucleic Acids Res.* **31**, 258–261 (2003).
30. Fuentes, D. et al. PhylomeDB V5: an expanding repository for genome-wide catalogues of annotated gene phylogenies. *Nucleic Acids Res.* **50**, D1062–D1068 (2022).
31. Morel, B., Williams, T. A., Stamatakis, A. & Szöllösi, G. J. AleRax: a tool for gene and species tree co-estimation and reconciliation under a probabilistic model of gene duplication, transfer, and loss. *Bioinformatics* **40**, btae162 (2024).
32. Zhang, J. et al. Deep origin of eukaryotes outside Heimdallarchaeia within Asgardarchaeota. *Nature* <https://doi.org/10.1038/s41586-025-08955-7> (2025).
33. Vosseberg, J. et al. Timing the origin of eukaryotic cellular complexity with ancient duplications. *Nat. Ecol. Evol.* **5**, 92–100 (2021).
34. Mira, A., Ochman, H. & Moran, N. A. Deletional bias and the evolution of bacterial genomes. *Trends Genet.* **17**, 589–596 (2001).
35. Elliott, T. A. & Gregory, T. R. What's in a genome? The C-value enigma and the evolution of eukaryotic genome content. *Philos. Trans. R. Soc. Lond. B Biol. Sci.* **370**, 20140331 (2015).
36. Gregory, T. R. The C-value enigma in plants and animals: a review of parallels and an appeal for partnership. *Ann. Bot.* **95**, 133–146 (2005).
37. Valentin-Alvarado, L. E. et al. Genetic elements and defense systems drive diversification and evolution in Asgard archaea. Preprint at <https://doi.org/10.1101/2024.03.22.586370> (2024).
38. Filée, J. et al. Bacterial origins of thymidylate metabolism in Asgard archaea and Eukarya. *Nat. Commun.* **14**, 838 (2023).
39. Bernabeu, M., Manzano-Morales, S., Marcet-Houben, M. & Gabaldón, T. Diverse ancestries reveal complex symbiotic interactions during eukaryogenesis. Preprint at <https://doi.org/10.1101/2024.10.14.618062> (2024).
40. Gabaldón, T. Relative timing of mitochondrial endosymbiosis and the ‘pre-mitochondrial symbioses’ hypothesis. *IUBMB Life* **70**, 1188–1196 (2018).
41. Schulz, F. et al. Giant virus diversity and host interactions through global metagenomics. *Nature* **578**, 432–436 (2020).
42. Yutin, N., Wolf, M. Y., Wolf, Y. I. & Koonin, E. V. The origins of phagocytosis and eukaryogenesis. *Biol. Direct* **4**, 9 (2009).
43. Hoyt, M. A., Hyman, A. A. & Bähler, M. Motor proteins of the eukaryotic cytoskeleton. *Proc. Natl. Acad. Sci. USA*. **94**, 12747–12748 (1997).
44. Seemann, T. Prokka: rapid prokaryotic genome annotation. *Bioinformatics* **30**, 2068–2069 (2014).
45. Jones, P. et al. InterProScan 5: genome-scale protein function classification. *Bioinformatics* **30**, 1236–1240 (2014).
46. Galperin, M. Y. et al. COG database update 2024. *Nucleic Acids Res.* **53**, D356–D363 (2025).
47. Katoh, K. & Standley, D. M. MAFFT multiple sequence alignment software version 7: improvements in performance and usability. *Mol. Biol. Evol.* **30**, 772–780 (2013).
48. Capella-Gutiérrez, S., Silla-Martínez, J. M. & Gabaldón, T. trimAl: a tool for automated alignment trimming in large-scale phylogenetic analyses. *Bioinformatics* **25**, 1972–1973 (2009).
49. Nguyen, L.-T., Schmidt, H. A., von Haeseler, A. & Minh, B. Q. IQ-TREE: a fast and effective stochastic algorithm for estimating

- maximum-likelihood phylogenies. *Mol. Biol. Evol.* **32**, 268–274 (2015).
50. Emms, D. M. & Kelly, S. OrthoFinder: phylogenetic orthology inference for comparative genomics. *Genome Biol.* **20**, 238 (2019).
 51. Altschul, S. F., Gish, W., Miller, W., Myers, E. W. & Lipman, D. J. Basic local alignment search tool. *J. Mol. Biol.* **215**, 403–410 (1990).
 52. Snipen, L. & Liland, K. H. micropan: an R-package for microbial pan-genomics. *BMC Bioinforma.* **16**, 79 (2015).
 53. Edgar, R. C. MUSCLE: a multiple sequence alignment method with reduced time and space complexity. *BMC Bioinforma.* **5**, 113 (2004).
 54. Katoh, K., Misawa, K., Kuma, K.-I. & Miyata, T. MAFFT: a novel method for rapid multiple sequence alignment based on fast Fourier transform. *Nucleic Acids Res.* **30**, 3059–3066 (2002).
 55. Lassmann, T. & Sonnhammer, E. L. L. Kalign: an accurate and fast multiple sequence alignment algorithm. *BMC Bioinforma.* **6**, 298 (2005).
 56. Wallace, I. M., O'Sullivan, O., Higgins, D. G. & Notredame C. M-Coffee: combining multiple sequence alignment methods with T-Coffee. *Nucleic Acids Res.* **34**, 1692–1699 (2006).
 57. Wehe, A., Bansal, M. S., Burleigh, J. G. & Eulenstein, O. DupTree: a program for large-scale phylogenetic analyses using gene tree parsimony. *Bioinformatics* **24**, 1540–1541 (2008).
 58. Collingridge, P. W. & Kelly, S. MergeAlign: improving multiple sequence alignment performance by dynamic reconstruction of consensus multiple sequence alignments. *BMC Bioinforma.* **13**, 117 (2012).
 59. Khedkar, S. et al. Landscape of mobile genetic elements and their antibiotic resistance cargo in prokaryotic genomes. *Nucleic Acids Res.* **50**, 3155–3168 (2022).
 60. Goodacre, N., Aljanahi, A., Nandakumar, S., Mikailov, M. & Khan, A. S. A reference viral database (RVDB) to enhance bioinformatics analysis of high-throughput sequencing for novel virus detection. *mSphere* **3**, e00069-18 (2018).
 61. Supek, B. E. et al. UniRef clusters: a comprehensive and scalable alternative for improving sequence similarity searches. *Bioinformatics* **31**, 926–932 (2015).
 62. Richter, D. J. et al. EukProt: a database of genome-scale predicted proteins across the diversity of eukaryotes. *Peer Community J.* **2**, e56 (2022).
 63. Zhu, Q., Kosoy, M. & Dittmar, K. HGTector: an automated method facilitating genome-wide discovery of putative horizontal gene transfers. *BMC Genom.* **15**, 717 (2014).
 64. Naranjo-Ortiz, M. A. et al. Widespread inter- and intra-domain horizontal gene transfer of d-amino acid metabolism enzymes in eukaryotes. *Front. Microbiol.* **7**, 2001 (2016).
 65. Eaton, D. A. R. Toytree: a minimalist tree visualization and manipulation library for Python. *Methods Ecol. Evol.* **11**, 187–191 (2020).
 66. Huerta-Cepas, J., Serra, F. & Bork, P. ETE 3: Reconstruction, analysis, and visualization of phylogenomic data. *Mol. Biol. Evol.* **33**, 1635–1638 (2016).
 67. Tria, F. D. K., Landan, G. & Dagan, T. Phylogenetic rooting using minimal ancestor deviation. *Nat. Ecol. Evol.* **1**, 193 (2017).
 68. Yu, G., Wang, L.-G., Han, Y. & He, Q.-Y. clusterProfiler: an R package for comparing biological themes among gene clusters. *OMICS* **16**, 284–287 (2012).
 69. Yu, G. Using ggtree to visualize data on tree-like structures. *Curr. Protoc. Bioinforma.* **69**, e96 (2020).
 70. Xu, S. et al. GgtreeExtra: compact visualization of richly annotated phylogenetic data. *Mol. Biol. Evol.* **38**, 4039–4042 (2021).
 71. Gu, Z., Eils, R. & Schlesner, M. Complex heatmaps reveal patterns and correlations in multidimensional genomic data. *Bioinformatics* **32**, 2847–2849 (2016).
 72. Seitz, K. W. et al. Asgard archaea are capable of anaerobic hydrocarbon cycling. *Nat. Commun.* **10**, 1822 (2019).
 73. Wong, H. L. et al. Microbial dark matter fills the niche in hypersaline microbial mats. *Microbiome* **8**, 135 (2020).
 74. Reysenbach, A.-L. et al. Complex subsurface hydrothermal fluid mixing at a submarine arc volcano supports distinct and highly diverse microbial communities. *Proc. Natl. Acad. Sci. USA.* **117**, 32627–32638 (2020).
 75. De Anda, V. et al. Brockarchaeota, a novel archaeal phylum with unique and versatile carbon cycling pathways. *Nat. Commun.* **12**, 2404 (2021).
 76. Tamarit, D. et al. A closed Candidatus Odinarchaeum chromosome exposes Asgard archaeal viruses. *Nat. Microbiol.* **7**, 948–952 (2022).
 77. Zhou, Z., St John, E., Anantharaman, K. & Reysenbach, A.-L. Global patterns of diversity and metabolism of microbial communities in deep-sea hydrothermal vent deposits. *Microbiome* **10**, 241 (2022).
 78. Sun, J. et al. Recoding of stop codons expands the metabolic potential of two novel Asgardarchaeota lineages. *ISME Commun.* **1**, 30 (2021).
 79. Carlton, J. D. et al. Expansion of Armatimonadota through marine sediment sequencing describes two classes with unique ecological roles. *ISME Commun.* **3**, 64 (2023).
 80. Han, Y. et al. A comprehensive genomic catalog from global cold seeps. *Sci. Data* **10**, 596 (2023).
 81. Qi, Y.-L. et al. Analysis of nearly 3000 archaeal genomes from terrestrial geothermal springs sheds light on interconnected biogeochemical processes. *Nat. Commun.* **15**, 4066 (2024).
 82. Valentin-Alvarado, L. E. et al. Asgard archaea modulate potential methanogenesis substrates in wetland soil. *Nat. Commun.* **15**, 6384 (2024).
 83. Imachi, H. et al. Promethearchaeum syntrophicum gen. nov., sp. nov., an anaerobic, obligately syntrophic archaeon, the first isolate of the lineage “Asgard” archaea, and proposal of the new archaeal phylum Promethearchaeota phyl. Nov. and kingdom Promethearchaeati regn. nov. *Int. J. Syst. Evol. Microbiol.* **74**, 006435 (2024).
 84. Manzano Morales, S. & Gabaldón, T. Pervasive horizontal transfer and extensive gene duplication shape Asgard archaeal genomes. Zenodo <https://doi.org/10.5281/ZENODO.15789115> (2026).

Acknowledgements

We want to thank Moises Bernabeu for the BroadDB database and Giacomo Mutti for help with using AleRax. TG group acknowledges support from the Spanish Ministry of Science and Innovation (grant numbers PID2021-126067NB-I00 and PLEC2023-010225), cofounded by ERDF “A way of making Europe”, as well as support from the Gordon and Betty Moore Foundation (grant number GBMF9742); “La Caixa” foundation (grant number LCF/PR/HR21/00737 and CI23-20260), Fundació La Marató de TV3 (202328-31), AECC (PRYGN234923GABA and 290059), and Instituto de Salud Carlos III (CIBERINFEC CB21/13/00061- ISCIII-SGEFI/ERDF and Proyectos de Desarrollo Tecnológico en Salud (DTS) (grant number DTS25/00141).

Author contributions

S.M.-M. performed all bioinformatics analyses, wrote the necessary code, and produced the visualisations. T.G. conceptualised the project, obtained funding and supervised the project. All authors contributed to the design of the analytical steps, to the interpretation of the results, and to writing the manuscript.

Competing interests

The authors declare no competing interests.

Additional information

Supplementary information The online version contains supplementary material available at <https://doi.org/10.1038/s41467-026-71534-5>.

Correspondence and requests for materials should be addressed to Toni Gabaldón.

Peer review information *Nature Communications* thanks Fabai Wu and the other, anonymous, reviewers for their contribution to the peer review of this work. A peer review file is available.

Reprints and permissions information is available at <http://www.nature.com/reprints>

Publisher's note Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

Open Access This article is licensed under a Creative Commons Attribution-NonCommercial-NoDerivatives 4.0 International License, which permits any non-commercial use, sharing, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons licence, and indicate if you modified the licensed material. You do not have permission under this licence to share adapted material derived from this article or parts of it. The images or other third party material in this article are included in the article's Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons licence and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit <http://creativecommons.org/licenses/by-nc-nd/4.0/>.

© The Author(s) 2026