

1 **Prokaryotes Become Larger at High Temperatures but They**  
2 **Do Not Grow Faster**

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14 **Abstract**

15 Metabolic theory posits that metabolism governs the rate at which organisms trans-  
16 form energy into biological work and growth. Thus, it constitutes the main mechanism  
17 driving the evolution of organismal growth and size across almost all domains of life.  
18 One general prediction of metabolic theory suggests that populations of larger organ-  
19 isms grow more slowly than populations of smaller organisms. However, increasing  
20 evidence show that prokaryotes seem to be the exception for such a trend. Larger  
21 prokaryotes appear to grow faster, challenging the standard theory and stimulating a  
22 further reevaluation of the current evidence. Here, I report a broad comparative analy-  
23 sis of the evolution of growth rate and cell size in prokaryotes. As opposed to previous  
24 investigations, my analysis relies on the concept of the thermal performance curve and  
25 the effects of its parameters on growth rate and cell size. Based on such approach, I  
26 found that prokaryotes evolved larger sizes at relatively high grow temperatures but  
27 their growth rates do not scale with size. At the optimum temperature for growth,  
28 the relationship between growth rate and cell size becomes unclear. These results call  
29 for a careful interpretation of the current evidence and highlight the importance of  
30 understanding the thermal sensitivity of the biochemical reactions that take place in  
31 the cells. Specifically, the metabolic reactions that regulate the protein synthesis in  
32 the cell, which are controlled by the translation machinery. In this regard, this study  
33 describes how different components of this machinery, such as the rRNA genes and  
34 the tRNA genes, interact to shape the evolution of growth rate and cell size across  
35 prokaryotes. Overall, I present more accurate results compared to previous evidence  
36 and suggest new hypotheses that can be applied to a wide range of taxa, paving the  
37 way for others to validate them at the intraspecific level.

38 **Keywords:** Archaea, Bacteria, life history, RNA, thermal adaptation.

## 39 Introduction

40 A long-standing prediction of metabolic theory suggests that populations of larger organisms  
41 grow more slowly than populations of smaller organisms (Savage et al., 2004). A general  
42 thought is that metabolism is responsible for this trend because it determines the rate at  
43 which organisms transform energy into biological work and growth (Hatton et al., 2019;  
44 Damuth, 1981; Savage et al., 2004). Because larger organisms have lower metabolic rates  
45 relative to their size, one may expect them to grow slower as described by the following  
46 mathematical model:

$$r = M^{B-1}$$

47 Where  $r$  ( $\text{min}^{-1}$ ) corresponds to the intrinsic growth rate,  $M$  constitutes the mass of the  
48 organisms ( $gr$ ), and  $B$  an exponent linking their mass to metabolic rate. In theory, absolute  
49 metabolism scales hypoallometrically with body size with an exponent of  $B$ , whereas mass-  
50 specific metabolism scales at  $B - 1$ . This theory holds for many organisms, ranging from  
51 multicellular metazoans and plants, to unicellular eukaryotes. For instance, in a comprehen-  
52 sive analysis of the scaling laws across multicellular eukaryotes, Hatton et al. (2019) showed  
53 that  $B$  is typically  $\approx 0.75$ . Thus,  $r$  should scale around  $-0.25$ . Similarly, Lynch et al. (2022)  
54 showed that for five unicellular eukaryotic groups of heterotrophs,  $B$  is  $-0.208$ . Accordingly,  
55 the data for this broad set of heterotrophs are consistent with the view that growth rate  
56 declines approximately as the  $-0.208$  power of adult mass. Contrary to most organisms,  
57 however, some evidence suggest that unicellular prokaryotes show a hyperallometric scaling  
58 between growth rate and adult mass with an exponent of  $0.286$  (Lynch et al., 2022), chal-

59 lenging the prevailing views of metabolic theory and encouraging us to look for an alternative  
60 theory that can be applied to all domains of life.

61 Although metabolic theory successfully predicts variation in growth and size across many  
62 taxonomic groups, these predictions are often based on general measures of metabolism, such  
63 as oxygen consumption (in  $joules \times cell^{-1} \times min^{-1}$ ; Marshall et al., 2022) or activation en-  
64 ergy estimates (in  $Watts$ ; Hatton et al., 2019). But little do we know about the metabolism  
65 of protein synthesis, which provides a more intuitive link to the growth of unicellular organ-  
66 isms, because much of the cell content corresponds to proteins Bremer and Dennis (2008).  
67 Because ribosomes are required for protein synthesis, their number and their rate of function  
68 determine the rate of protein synthesis and cytoplasmic mass accumulation. Mathematically,  
69 the relationship between growth rate ( $u$ ) and the translation machinery of an organism can  
70 be described as follows:

$$u = (60/\ln 2) \times (N_r/P) \times e_r$$

71 Where the growth rate of an organism  $u$  ( $min^{-1}$ ) equals the product of the number of  
72 ribosomes per amount of total protein present ( $N_r/P$ ), times the rate of protein synthesis per  
73 average ribosome, known as “ribosome efficiency” ( $e_r$ ; Bremer and Dennis, 2008). Consistent  
74 with this idea, Schaechter et al. (1958) showed that the amounts of RNA and proteins per cell  
75 are tightly linked to the growth rate of a bacterium. More recently, the relationship between  
76 the translation machinery and growth rate has received increasing attention. For instance,  
77 Lynch et al. (2022) provided a hypothetical upper bound on growth rates across prokaryotes  
78 and unicellular eukaryotes dictated by the translational constraints imposed by the properties  
79 of ribosomes. Considering the time required for ribosomal-protein replacement, the authors  
80 suggest an upper limit to the exponential growth rate of  $50/day$  for prokaryotes, and an upper

81 limit to growth rate of  $\approx 2.25$ x lower in eukaryotes than that for prokaryotes (i.e., 20/day).

82 As described earlier, inherent to the link between growth and size of an organism is  
83 the role of metabolism, which depends on temperature (Angilletta, 2009). Temperature  
84 determines the proportion of reactants that possess the free energy required for reaction.  
85 Thus, as the temperature increases, more reactants exceed the energy of activation. Because  
86 enzymes lower the energy required for activation and thus speed the reaction at any given  
87 temperature (Hochachka and Somero, 2002), the metabolic rate and hence the growth rate of  
88 an organism should scale proportionally with the growth temperature. In fact, comparative  
89 analyses of the evolution of growth rate across many taxa may only be accurate when growth  
90 rate is normalized to a specific temperature for growth (e.g., Lynch et al., 2022; Hatton  
91 et al., 2019). However, this temperature correction seems to be done arbitrarily in some  
92 cases, leading to potential misleading results because no two organisms have exactly the  
93 same thermal restrictions (Angilletta, 2009). A better approach may be to consider the  
94 thermal performance curves of the organisms involved in the investigation. One can capture  
95 the general characteristics of the thermal performance curves with specific parameters: (1) the  
96 thermal optimum ( $T_{opt}$ ); (2) the thermal limits, referred to as the critical thermal minimum  
97 ( $CT_{min}$ ) and the critical thermal maximum ( $CT_{max}$ ). Ideally, comparisons across many taxa  
98 should be done at the optimum temperature for growth, assuming that organisms would  
99 likely perform best at such temperature.

100 Here, I examine the evolution of growth rate and cell size across prokaryotes based on a  
101 large comparative analysis. In doing so, I test some of the principal predictions of metabolic  
102 theory and present results that challenge previous findings. Particularly, this analysis links  
103 the evolution of growth and cell size to important foundational features of biology that have

104 not been properly considered in previous investigations, such as the translation machinery  
105 and the thermal sensitivity of organisms. Accordingly, this study provides novel insights that  
106 enable us to better understand the evolution of the life history in prokaryotes.

## 107 **Materials and Methods**

### 108 *Data source and description of variables*

109 Trait datasets are increasingly being used in studies investigating evolutionary theories and  
110 global conservation initiatives (e.g., <https://opentraits.org/datasets.html>). These datasets  
111 allow for integrating a diverse range of genomic, physiological, ecological, morphological, and  
112 life-history data to explore organismal evolution. This study used a unified microbial trait  
113 dataset suitable for investigating evolutionary correlations between traits across many species.  
114 The dataset spans the full range of bacterial and archaeal habitats, including fresh and ma-  
115 rine waters, soils and sediments, animal and plant hosts, and thermal environments (Madin  
116 et al., 2020). Data sources include well-established repositories, such as GenBank, Bergey’s  
117 Manual of Systematics of Archaea and Bacteria, and a number of compilations published  
118 in the primary literature (e.g., Lynch et al., 2023; Gonzalez-de Salceda and Garcia-Pichel,  
119 2021).

120 To study the evolution of growth rate and cell size across prokaryotes, I collected data  
121 of minimum doubling time ( $\log_{10} h$ ), cell volume ( $\log_{10} \text{um}^3$ ), and cell diameter ( $\log_{10} \text{um}$ )  
122 for the species under investigation. The minimum doubling time is best interpreted as a  
123 key life-history trait associated with the  $r/K$  selection theory (Boyce, 1984), having relevant  
124 connection to the metabolic rate of organisms. With data of doubling time ( $T_d$ ), one can  
125 easily compute the intrinsic growth rate ( $r$ ) as follows:  $r = \frac{\ln(2)}{T_d}$ . In prokaryotes, cell volume

126 might be the only obvious difference between some species. Unfortunately, data of cell volume  
127 across the domains of life are rather scarce. By contrast, cell diameter seems to be the most  
128 common proxy for cell size across prokaryotes. Interestingly, cell diameter is a good predictor  
129 of cell volume in some species (Figure S1), enabling us to use it as a good indicator of cell  
130 size in this study.

131 To investigate the scaling relationship between growth rate and metabolism, I examined  
132 the translation machinery of an organism; here referred to as the number of ribosomal RNA  
133 genes (rRNA) and transfer RNA genes (tRNA). The rRNA genes are often assumed to be  
134 a robust measure or proxy for ribosome content in prokaryotes (Bremer and Dennis, 2008).  
135 Indeed, ribosomal gene copy number per genome seems to correlate with a microbe's life-  
136 history traits, where fast growth is associated with higher copy number (Klappenbach et al.,  
137 2000; Roller et al., 2016). By contrast, the effects of tRNA gene copy number on growth has  
138 received less attention in prokaryotes. This might be associated with the fact that 14% of  
139 the total stable RNA in a cell is tRNA, and 86% is rRNA Bremer and Dennis (2008).

140 Lastly, I examined the effect of temperature on the evolution of growth rate and cell size  
141 across prokaryotes. To do so, I used specific parameters of the thermal performance curves  
142 of each species. Performance can be defined as any measure of an organism's capacity to  
143 function, usually expressed as a rate or probability (Angilletta, 2009). Common measures  
144 of performance include growth and cell size. To accurately compare the evolution of these  
145 traits among species, I analyzed the intrinsic growth rate of the species ( $r$ ) and cell diameter  
146 ( $\mu$ ) at the optimum temperature for growth ( $\log_{10} C^\circ$ ). The same procedure was done at  
147 temperatures below and above the optimum; here referred to as lower temperature and upper  
148 temperature, respectively.

## 149 *Comparative analysis*

150 Because species are part of a hierarchically structured phylogeny, and thus cannot be regarded  
151 for statistical purposes as if drawn independently from the same distribution (Felsenstein,  
152 1985), I used a subset of recently published phylogenies available in the primary and secondary  
153 literature (see <https://timetree.org/>), enabling me to account for similarity by descendant.  
154 Most of the sources I compiled consisted of dated posterior tree distributions derived to infer  
155 evolutionary relationships for Archaea and Bacteria. I produced trees with subsets of taxa  
156 based on the species for which I had available data. These trees were rescaled to absolute  
157 time and then joined to form a full species-level tree of prokaryotes (e.g., Figure 1). To do  
158 so, I used the Build a Timetree function available at The TimeTree of Life database (see  
159 url link above), which outputs a timetree of the taxa of interest extracted from the global  
160 timetree connecting species and publication-specific timetrees in the database. For instance,  
161 one can input a species list or simply enter a taxon name to see the clade-specific portion of  
162 the global timetree. Also, one can restrict the timetree produced to contain tips at a desired  
163 taxonomic level, including species, genus, or family.

Figure 1: Phylogenetic relationship between some of the species involved in this study.

## 164 *Statistical Analysis*

165 I used phylogenetic generalized least squares (PGLS) to model the evolution of growth rate  
166 and cell size across prokaryotes. As mentioned earlier, PGLS models enable one to account  
167 for non-independence of the data (Felsenstein, 1985). To do this, I used the *gls* function from  
168 the “nlme” package of R, v. 4.2.2 (Team, 2000; Pinheiro et al., 2017). I fitted all models  
169 assuming the species trait values evolved via a Brownian motion model. Alternatively, I used



170 the standard ordinary least squares regression method (OLS; type I regression) to calculate  
171 regression statistics, which is the standard approach in bivariate power law regression (Sibly  
172 et al., 2012). But this method assumes that all error is in the Y-axis variable. This tends  
173 to underestimate the slope as error in the X-axis variable increases. To evaluate the models'  
174 goodness of fit, I used information-theoretic criteria such as  $AIC_c$  and selected the most likely  
175 one for inferences.

## 176 **Results**

177 A phylogenetic-informed model describing the relationship between cell volume and intrinsic  
178 growth rate was strongly supported, as opposed to a standard OLS regression model.  
179 Specifically, the model indicates that growth rate increased as cell size increased (Figure 2A).  
180 However, this result may be misleading because the data of cell volume consisted of species  
181 that grew at different temperatures, preventing one from making accurate comparisons. For  
182 species that grew at their optimum temperature, the analyses revealed no significant rela-  
183 tionship between growth and cell diameter (Table 1; Figure 2B).

Table 1: Scaling coefficients estimated by a model describing the evolution of cell size and growth rate in prokaryotes.

Figure 2: Scaling relationship between cell size and growth rate across prokaryotes. (A) Scaling of cell volume with growth rate across species grown at different temperatures. (B) Relationship between cell diameter and doubling time across species grown at their optimum temperature ( $n = 40$ ). Regression lines were displayed only when significant effects were observed.

184 The evolutionary relationship between growth rate and the translation machinery shows

185 that both the rRNA genes and the tRNA genes increased with growth rate. Though, the rate  
186 of increase in rRNA genes was faster than that of tRNA genes in species that grow faster  
187 (Figure 3). This evidence supports the existence of an interaction between rRNA genes and  
188 tRNA genes across the range of growth rates examined (Table 2). In other words, fast-growing  
189 prokaryotes tend to have a high number of rRNA genes but a low number of tRNA genes.  
190 By contrast, slow-growing prokaryotes tend to have a low number of rRNA genes but a high  
191 number of tRNA genes (Figure 3).

Table 2: Contrast of parameters estimated by an OLS regression model describing the evolution of the translation machinery as a function of growth in prokaryotes.

Figure 3: Relationship between growth rate and the translation machinery in prokaryotes ( $n = 413$ ).

192 Strikingly, I found that cell size increased with temperature, but only when organisms  
193 grew at temperatures higher than the optimum temperature for growth (Table 3; Figure  
194 4A). Importantly, this model only accounted for an effect of cell size alone. A model that  
195 examined the interaction between cell size and growth temperature was not supported by the  
196 data. When grown at any of the temperatures suggested by the thermal performance curves,  
197 organisms seemed to grow generally faster regardless of their cell sizes (Figure 4B).

Table 3: Scaling coefficients estimated by an OLS regression model of the evolution of cell size as a function of temperature in prokaryotes.

Figure 4: Evolutionary relationships between cell size, growth, and temperature across prokaryotes. (A) From left to right, prokaryotic cell size as a function of the lower, optimum, and upper temperatures of the species. (B) Similarly, the bottom panel shows the relationship between growth and the aforementioned temperatures. Regression lines were displayed only when significant relationships were observed.

## 198 Discussion

199 Although previous evidence from interspecific and intraspecific studies suggest that growth  
200 rate increases with cell size in prokaryotes, this study shows no evidence of such trend.  
201 Instead, I found that larger cells only evolved at relatively high temperatures, whereas faster  
202 growth generally evolved at any of the temperature ranges examined regardless of the size  
203 of organisms. A hypermetric correlation between cell size and cell growth contradicts the  
204 expectation based on standard theory, but plausible hypotheses in favor of this observation  
205 have been proposed by some researchers. First, Marshall et al. (2022) suggested that larger  
206 cells of *Escherichia coli* are cheaper to maintain and build per unit volume, such that the  
207 scaling of the total cost of production is far less than proportional to cell size. Second, DeLong  
208 et al. (2010) hypothesized that the relatively large genomes of prokaryotes enable them to  
209 produce a large number of enzymes involved in more complex biochemical networks. These  
210 networks would confer them an increased metabolic power because enzymes could bind to  
211 substrates more completely, thereby producing more ATP molecules per unit substrate and  
212 per unit time. Although the current literature provides compelling arguments about the  
213 unexpected hypermetric correlation between cell size and and cell growth in prokaryotes,  
214 none of them consider the greater catalytic capacity of biochemical systems operating at high  
215 temperatures; a capacity which in itself may have been an important factor in the evolution of  
216 fast growth and large cells in prokaryotes. Furthermore, the idea that an increasing genome  
217 size with cell size might explain the hypermetric scaling of growth with cell size in prokaryotes  
218 sounds appealing, but a detailed analysis on the mechanics of metabolism-related genes that

219 affect growth and cell size is still needed. Below, I not only describe the ways in which  
220 protein-synthesis genes interact to shape the evolution of growth and cell size in prokaryotes,  
221 but also, I provide a comprehensive discussion about the effects of temperature on their life  
222 history.

223       Among all of the metabolism-related genes present in the genome of an organism, the  
224 rRNA genes and the tRNA genes are perhaps some of the most important ones because they  
225 regulate the production of macromolecules required for growth and cell size. Consistent with  
226 this idea, a wealth of evidence seem to indicate that the amounts of RNA genes and protein  
227 scales with growth and cell size (Klappenbach et al., 2000; Bremer and Dennis, 2008; Lynch  
228 and Marinov, 2015). Furthermore, early studies suggest a growth limitation imposed by DNA  
229 concentration in the cell, such that DNA limits mRNA synthesis and mRNA limits protein  
230 synthesis (Maaløe, 1979). But that claim is no longer supported by the current literature.  
231 By contrast, the ribosome concentration and the protein synthesis rate per average ribosome,  
232 both of which are regulated by the RNA genes, seem to be growth limiting (Bremer and  
233 Dennis, 2008). Given the high demand for rRNA transcription and the central role of rRNAs  
234 in the regulation of ribosome synthesis, an increasing rRNA gene copy number should scale  
235 with growth, because it dictates how quick microbes can synthesize ribosomes (Klappenbach  
236 et al., 2000). Indeed, the results of this study show that both the rRNA genes and the tRNA  
237 genes increase with growth among prokaryotes. Interestingly, the tRNA genes increase with  
238 growth at a lower rate than that of the rRNA genes, supporting the idea that in slow-growing  
239 species there appears to be a slight excess in the synthesis rate of stable RNA, such that the  
240 excess rRNA is rapidly degraded, whereas the tRNA accumulates (Jinks-Robertson et al.,  
241 1983; Norris and Koch, 1972). In sum, the translation machinery of the cell is rather complex  
242 and a better understanding of its regulatory effects on growth and cell size may only be

243 attained if we examine not only the variation in RNA genes, but also the thermal conditions  
244 in which the protein synthesis occurs.

245 Temperature constrains the rates of biochemical reactions in the cell, leading to different  
246 thermal sensitivities of growth and cell size across prokaryotes. While prokaryotic cells func-  
247 tion within a range of  $-5^{\circ}\text{C}$  to  $110^{\circ}\text{C}$  (Jaenicke, 1991, 1993), eukaryotic cells are relatively  
248 restricted, tolerating temperatures between  $-2^{\circ}\text{C}$  and  $60^{\circ}\text{C}$  (Tansey and Brock, 1972). The  
249 wide range of variation in the thermal tolerance of prokaryotes may then reflect important  
250 differences in growth and cell size across species. This study shows that both growth and cell  
251 size span three orders of magnitude across species. Such variation provides good conditions  
252 for natural selection to operate. In general, natural selection favors mutations that alter  
253 the conformational stability of enzymes (Hochachka and Somero, 2002; Marx et al., 2007). A  
254 more flexible structure helps enzymes to change shape faster during catalysis. If enzymes with  
255 greater conformational stability function better at high temperatures (Fields, 2001; Somero,  
256 1995), species should evolve larger sizes at a relatively high temperature for growth, as in-  
257 dicated by the results of this study. Such hypothesis could be validated if growth and cell  
258 size are compared between Archaea and Bacteria. Because there is evidence of adaptation to  
259 high temperatures across the evolutionary history of archaeal species (Groussin and Gouy,  
260 2011), one should expect those species to exhibit larger sizes than do species of the bacterial  
261 domain. My analysis supported this expectation; on average, archaeal species are larger than  
262 bacterial species, but bacteria seem to grow faster (Figure 5). On one hand, this result is  
263 consistent with the prediction that populations of large organisms grow more slowly than  
264 populations of small organisms. On the other hand, this findings align with the idea that  
265 “the hotter is better”, which proposes that genotypes or species with relatively high optimal  
266 temperatures also have relatively high maximal performance or fitness (Savage et al., 2004;

267 Angilletta, 2009; Kingsolver and Huey, 2008). As discussed earlier, this argument is based  
268 on empirical evidence suggesting that metabolic reactions inevitably increase with absolute  
269 temperature. Consequently, maximum biochemical reaction rates of species adapted to warm  
270 temperatures are higher than those of species adapted to cold temperatures, when each is  
271 measured at its optimal temperature (Kingsolver and Huey, 2008).

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Figure 5: Violin plots displaying the distribution of growth and cell size in prokaryotes.  
(A) Mean difference in growth between Archaea and Bacteria. (B) Mean difference in  
cell size between Archaea and Bacteria. The dashed lines represent mean values ( $\mu$ ).

272 Based on a broad comparative analysis, this study evaluates the intriguing superlinear  
273 scaling of growth rate with cell size in prokaryotes. This pattern challenges the predictions  
274 of metabolic theory, motivating me to conduct a further examination of the data currently  
275 available in the literature. Contrary to most studies, I found that larger cells evolved at a  
276 relatively high temperature for growth. However, there seems to be no relationship between  
277 growth and cell size across species at an optimum temperature for growth. As opposed to most  
278 studies, my analyses account for important characteristics of the thermal biology of organisms,  
279 enabling me to provide more reliable evidence. In addition, this study describes novel ways in  
280 which the translation machinery may affect the evolution of life-history traits in prokaryotes,  
281 paving the way for others to validate my results at the intraspecific level. For instance, the  
282 idea that “hotter is better” seems plausible, but little evidence in favor of its predictions  
283 is available in microbes. Similarly, the thermal sensitivity of the metabolic reactions that  
284 control protein synthesis in the cell, and hence growth and organismal size, seems poorly  
285 understood. Because each species comprises a unique set of biochemical reactions, studying  
286 the thermal biology of many species would enable us to determine the severe limits that

287 temperature imposes on life.

## 288 **Acknowledgements**

289 N/A

## 290 **Data Accessibility Statement**

291 A fully reproducible workflow of the data analyses, including R scripts and additional sup-  
292 porting material, can be downloaded in the following repository: [https://dylan-padilla.](https://dylan-padilla.github.io/cell-growth-size-paper/)  
293 [github.io/cell-growth-size-paper/](https://dylan-padilla.github.io/cell-growth-size-paper/), a Dryad link will be available upon acceptance .

## 294 **Conflict of interest**

295 The author declares no conflict of interest.

## 296 **Author Contributions**

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Table 1: Scaling coefficients estimated by a model describing the evolution of cell size and growth rate in prokaryotes.

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	-1.305	0.169	-7.724	0.000
$\log_{10}(\text{diameter})$	0.169	0.139	1.214	0.232

Table 2: Contrast of parameters estimated by an OLS regression model describing the evolution of the translation machinery as a function of growth in prokaryotes.

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	1.569	0.010	158.646	< 0.001
$\log_{10}(r)$	-0.038	0.010	-3.734	< 0.001
$\log_{10}(\text{rRNA})$	0.359	0.015	24.426	< 0.001
$\log_{10}(r):\log_{10}(\text{rRNA})$	0.117	0.019	6.281	< 0.001

A colon punctuation mark (:) denotes an interaction term.

Table 3: Scaling coefficients estimated by an OLS regression model of the evolution of cell size as a function of temperature in prokaryotes.

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.043	0.293	0.148	0.883
$\log_{10}(T_{upper})$	0.615	0.199	3.087	0.002

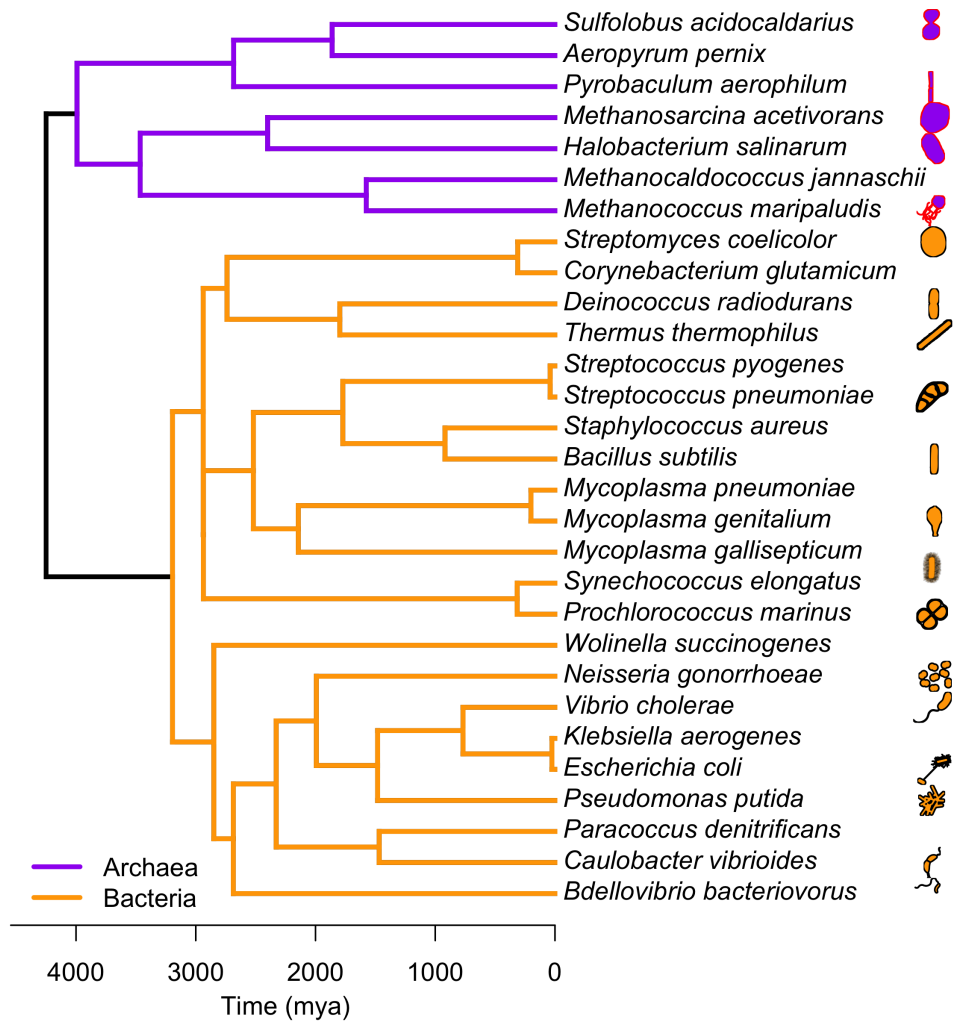


Figure 1: Phylogenetic relationship between some of the species involved in this study.

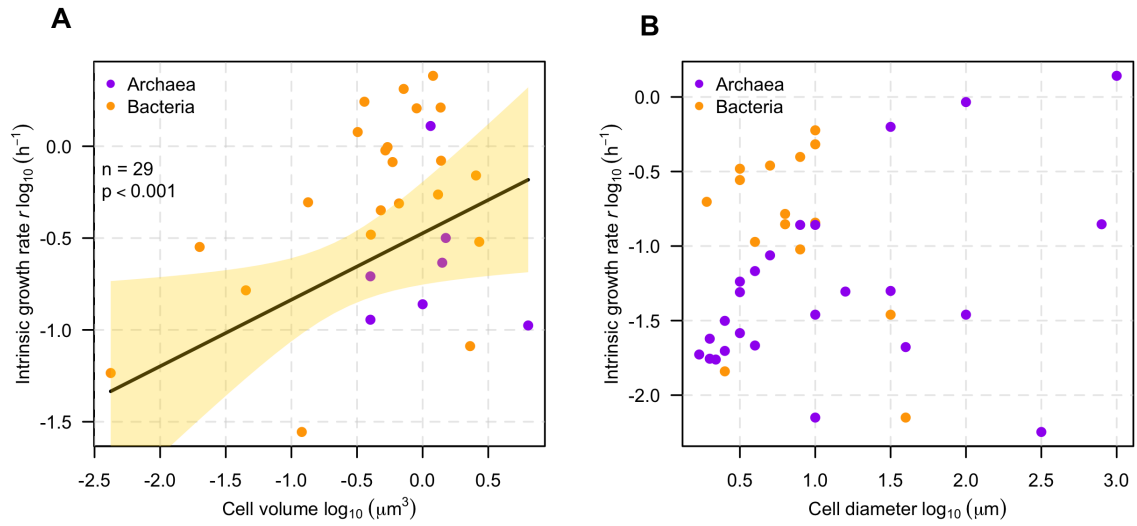


Figure 2: Scaling relationship between cell size and growth rate across prokaryotes. (A) Scaling of cell volume with growth rate across species grown at different temperatures. (B) Relationship between cell diameter and doubling time across species grown at their optimum temperature ( $n = 40$ ). Regression lines were displayed only when significant effects were observed.

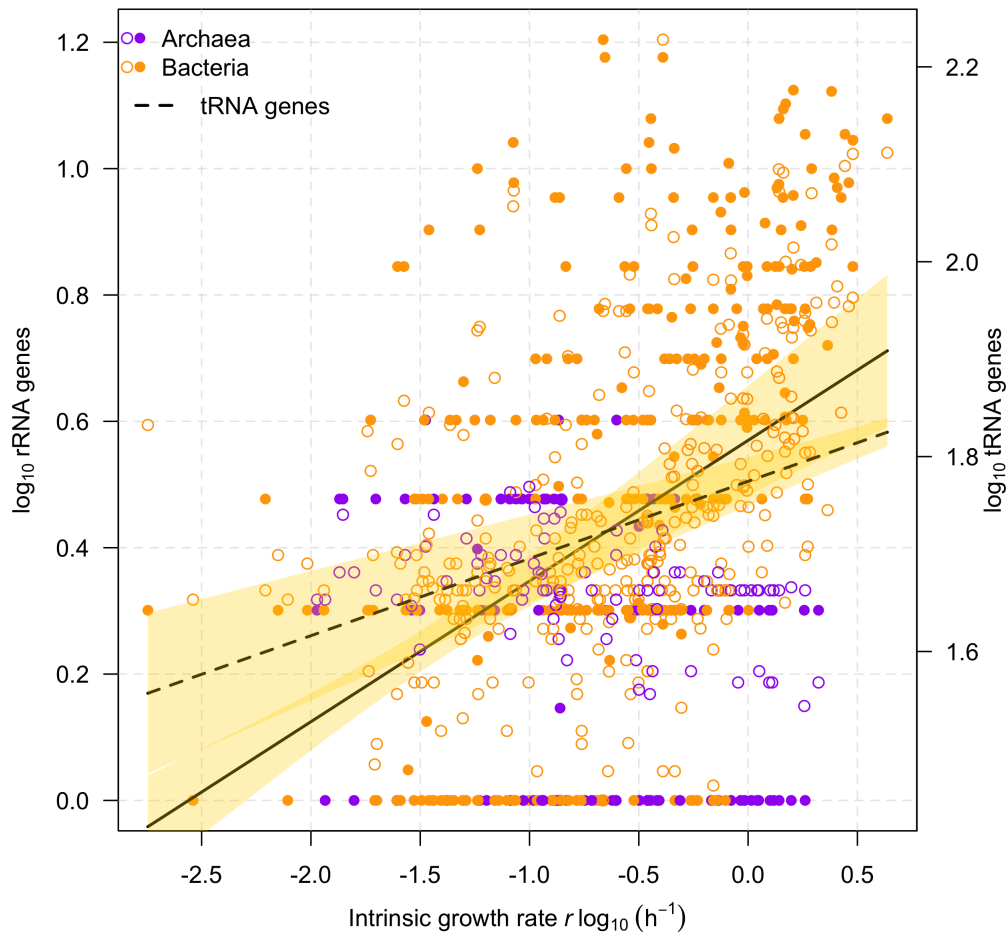


Figure 3: Relationship between growth rate and the translation machinery in prokaryotes ( $n = 413$ ).

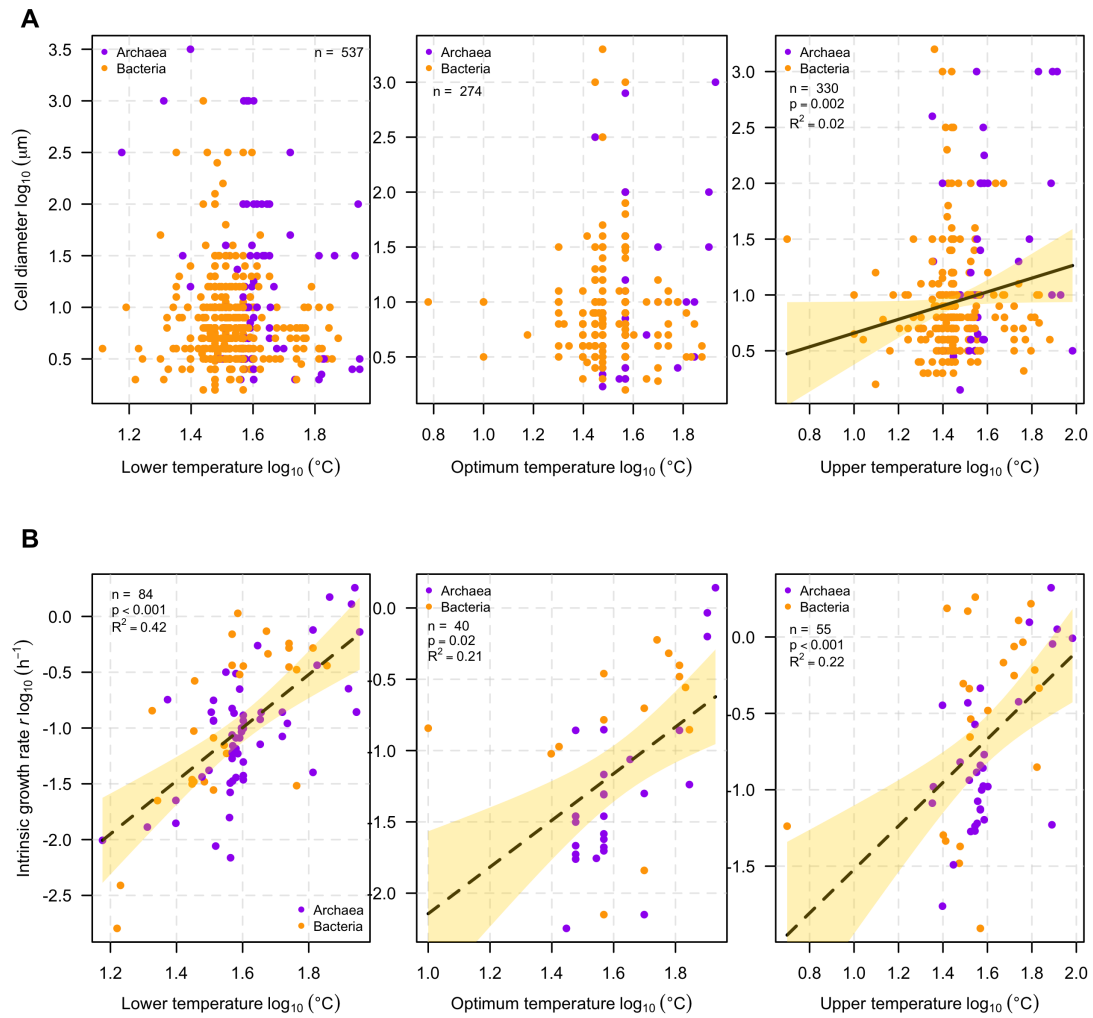


Figure 4: Evolutionary relationships between cell size, growth, and temperature across prokaryotes. (A) From left to right, prokaryotic cell size as a function of the lower, optimum, and upper temperatures of the species. (B) Similarly, the bottom panel shows the relationship between growth and the aforementioned temperatures. Regression lines were displayed only when significant relationships were observed.

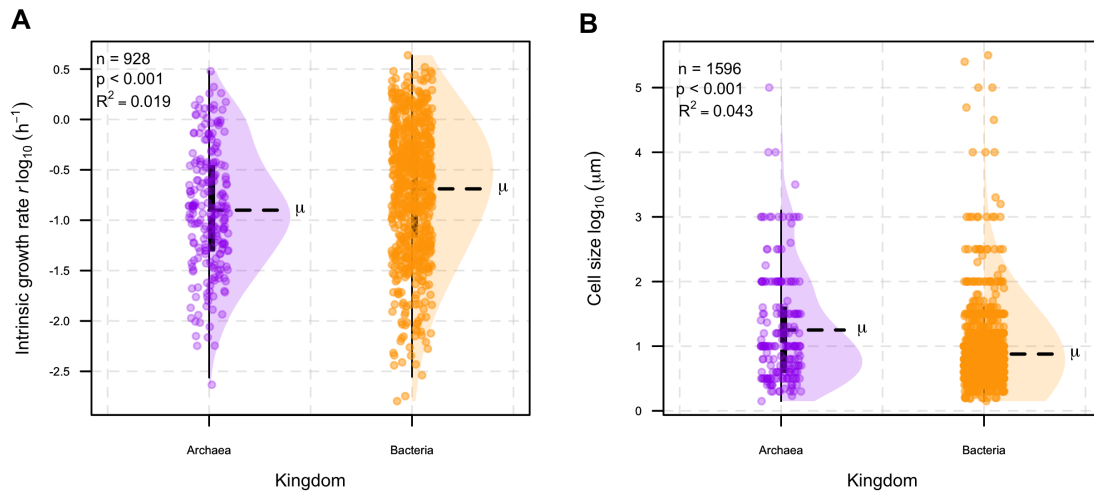


Figure 5: Violin plots displaying the distribution of growth and cell size in prokaryotes. (A) Mean difference in growth between Archaea and Bacteria. (B) Mean difference in cell size between Archaea and Bacteria. The dashed lines represent mean values ( $\mu$ ).

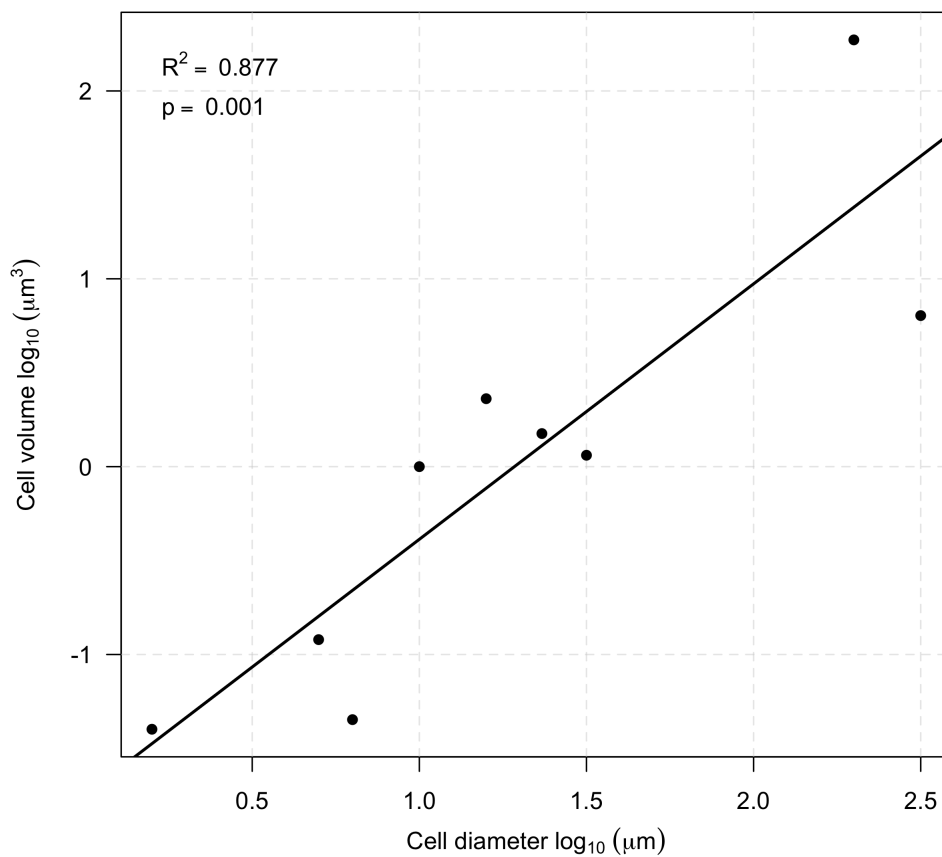


Figure S1: Correlation between cell diameter and cell volume in prokaryotes ( $n = 9$ ).