Bacterial CRISPR-Cas Abundance Increases Precipitously at

Around 45°C: Linking Antivirus Immunity to Grazing Risk

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Keywords
CRISPR-Cas, optimal growth temperature, bacteria, protistan grazing, viral lysis, mortality.
Abstract

Although performing adaptive immunity, CRISPR-Cas systems are present in only 40% of bacterial genomes. Here, we observed an abrupt transition of bacterial CRISPR-Cas abundance at around 45°C. Phylogenetic comparative analyses confirmed that the abundance correlates with growth temperature only at the temperature range around 45°C. Meanwhile, we noticed that the diversities of cellular predators have a precipitous decline at this temperature range. The grazing risk faced by bacteria reduces substantially at around 45°C and almost disappears above 60°C. So viral lysis would become the dominating factor of bacterial mortality, and antivirus immunity has a higher priority. In temperature ranges where the abundance of cellular predators does not change with temperature, temperatures would have negligible effects on CRISPR-Cas abundance. The hypothesis predicts that bacteria should also be rich in CRISPR-Cas systems if they live in other extreme conditions inaccessible to grazing predators.
Introduction

CRISPR-Cas is the adaptive immune system of bacteria and archaea. Like the adaptive immune system of jawed vertebrates, it can remember previously encountered pathogens, initiate a rapid response to a second invasion, and eliminate the recurrent invader. However, unlike the ubiquitous presence of adaptive immunity in jawed vertebrates, the CRISPR-Cas systems are only present in about 40% of bacteria. The patchy distribution of the CRISPR-Cas systems among bacteria is a recognized mystery. Given the constant horizontal transfers, their absence in more than half of bacterial genomes is unlikely to happen just by chance. Instead, it should be attributed to a tradeoff between the costs and benefits of the CRISPR-Cas systems. First, the acquirement and maintenance of CRISPR-Cas systems would sequestrate limiting resources such as the building blocks, the energy, and the transcription and translation machines. Second, the autoimmune response and cell death induced by self- and prophage-targeting spacers might be a selective force for the loss of CRISPR-Cas systems. In addition, for the viruses with high densities, high mutation rates, high genetic diversities, or carrying anti-CRISPR proteins, the efficiency of CRISPR-Cas systems is limited. CRISPR-Cas systems are not favored in conditions with high antibiotic pressures because they inhibit horizontal gene transfer, an efficient way for bacteria to acquire antibiotic resistance.

Just after discovering the CRISPR-Cas structures, it was noticed that they are more prevalent in the thermophilic archaea and the hyperthermophilic bacteria. Later large-scale analyses confirmed the prevalence of CRISPR-Cas systems in thermophiles and hyperthermophiles and showed a positive correlation between CRISPR abundance and a growth temperature. Recently, a computer learning approach with an attempt to control the phylogenetic bias suggests that temperature and oxygen levels are the most influencing ecological factors determining the distribution of the CRISPR-Cas systems. In this study, we
carried out a detailed analysis on the thermal distribution of CRISPR-Cas systems and observed a precipitous increase of bacterial CRISPR-Cas systems at around 45°C. At temperatures higher and lower than this range, bacterial CRISPR-Cas abundance does not correlate with growth temperature. The association of adaptive immunity with a narrow range of growth temperature provokes a new hypothesis linking the antivirus immunity of bacteria to their grazing risk by predatory protists and predatory bacteria.

Materials and Methods

From the growth temperature database TEMPURA and the Genome Taxonomy Database\textsuperscript{22, 23} we retrieved 682 bacteria and 156 archaea with growth temperatures and phylogenetic information. The sequences of these genomes were downloaded from ftp://ftp.ncbi.nlm.nih.gov/genomes/. We annotated their CRISPR-Cas systems using CRISPRCasFinder v1.3 program\textsuperscript{24} The raw data of these 682 bacteria and 156 archaea were deposited as Table S1.

According to Couvin et al.\textsuperscript{24} the annotated CRISPR arrays were classified into four categories, 1 to 4, according to their evidence levels. The CRISPR arrays with evidence levels 3 and 4 are highly likely candidates, and those with evidence levels 1 and 2 are potentially invalid. Therefore, only the CRISPR arrays with evidence levels 3 and 4 were counted in calculating CRISPR array abundance. We counted the putative CRISPR arrays with evidence levels 1 and 2 as zero and presented the results in the main text and Table S2. We also replicated our analyses by regarding the putative CRISPR arrays with evidence levels 1 and 2 as controversial CRISPR arrays and discarding the species having only CRISPR arrays with evidence levels 1 and 2 in calculating CRISPR array abundance. That is, these species were excluded from both numerator and denominator in the calculation of CRISPR-Cas abundance. Similar results (Table S3) were obtained, and the same conclusion was supported.
The CRISPR array abundance in a bacterial or archaeal group was defined as the total number of CRISPR arrays annotated in their genomes divided by the genome numbers of the group. The abundances of CRISPR spacers, \(\textit{cas}\) genes, and \(\textit{cas}\) gene clusters were defined similarly.

The phylogenetic signals (\(\lambda\)) of CRISPR array abundance, CRISPR spacer abundance, \(\textit{cas}\) gene abundance, \(\textit{cas}\) gene cluster abundance, and growth temperatures were estimated using the phylosig function of the R (Version 4.0.3) package phytools (Version 0.7-70)\(^{25}\). The phylogenetic generalized least squares (PGLS) regression was performed using the R (Version 4.0.3) package phylolm (version 2.6.2)\(^{26}\). Pagel's lambda model has been applied in the analyses.

**Results**

**Bacterial CRISPR array abundance jumps up at around 45°C**

By plotting the CRISPR array abundance against the optimal growth temperatures (Topt) in a column chart, we see a novel pattern on the thermal distribution of CRISPRs in bacteria (Fig. 1A). An abrupt transition of bacterial CRISPR array abundance happens at around 45°C. The CRISPR array abundance also fluctuates below and above 45°C, but the amplitudes of the fluctuations are much weaker. By sliding the windows of each column and calculating the ratio of CRISPR array abundance of one column to the next, we confirmed that the transition of CRISPR array abundance at 45°C is the most sudden one (Fig. 1B).

For a quantitative measure for the relationship between bacterial CRISPR array abundance and Topt, we first measured their phylogenetic signals (\(\lambda = 0.85, p = 3 \times 10^{-44}\) for CRISPR array abundance and \(\lambda = 0.95, p = 4 \times 10^{-197}\) for Topt) and proved the necessity of using phylogenetic comparative methods to control the effects of common ancestors\(^{27,28}\). Their correlation was examined by the phylogenetic generalized least squares (PGLS)
regression, where a significant positive slope corresponds to a significant positive correlation, and a negative slope indicates the reverse. Globally, bacterial CRISPR array abundance is positively correlated with Topt (slope = 0.06, \( p = 8 \times 10^{-12} \)). However, when the bacteria were classified into three categories according to their Topt, low temperatures (10 \( \leq \) Topt \( \leq \) 35°C, \( n = 488 \)), moderate temperatures (35 < Topt < 50°C, \( n = 97 \)), and high temperatures (50 \( \leq \) Topt \( \leq \) 85°C, \( n = 97 \)), a significant correlation was only observed in the moderate-temperature bacteria (Table 1). Furthermore, we aligned the 682 bacteria along the Topt axis and performed a PGLS analysis for every 100 neighboring samples. Significant correlations were observed only in a range around 45°C (Fig. 1C).

**Bacterial CRISPR spacer abundance increases precipitously at around 45°C**

Globally, bacterial CRISPR spacer abundance correlates positively with Topt (\( p = 2 \times 10^{-6} \), Table 1). Although there is no abrupt jump in bacterial CRISPR spacer abundance along the axis of growth temperature, the spacer abundance increases substantially with Topt only at around 45°C (Fig. 1D). PGLS analyses showed a significant correlation between CRISPR spacer abundance and Topt at the moderate temperatures, but not in the low temperatures or the high temperatures (Table 1).

**Bacterial cas gene abundance increases precipitously at around 45°C**

Across all the 682 bacteria with Topt ranging from 10–85°C, both cas gene abundance and cas gene cluster abundance are significantly correlated with Topt (Table 1). An abrupt jump at around 45°C could be observed in bacterial cas gene cluster abundance, and the increase in bacterial cas gene and cas gene cluster abundances are precipitous at around 45°C (Fig. 1E-F). PGLS analyses gave the same statistical conclusion for cas gene abundance and cas gene cluster abundance. Both abundances are significantly correlated with Topt at the moderate
temperatures \( (p = 3 \times 10^{-4}) \), but not in the low temperatures \( (p = 0.77) \) or the high temperatures \( (p = 0.85) \) (Table 1).

**Archaeal CRISPR-Cas abundance has a less distinctive pattern**

We also examined the relationship between growth temperature and CRISPR-Cas abundance in archaea. Globally, archaeal Topt positively correlates with CRISPR array abundance, \( cas \) gene abundance, and \( cas \) gene cluster abundance, but not with spacer abundance (Table 1). However, no abrupt jumps of CRISPR-Cas abundance were observed in archaea (Fig. 2). It should be noted that the sample sizes of some columns are tiny, and so the observed pattern is sensitive to the presence of a few outliers. For example, in the temperature range \([25,30)\), there are only two archaeal species, *Methanosarcina lacustris* and *Methanospirillum palustris*, with eight and one CRISPR arrays, respectively.

Because the 156 archaeal species are enriched in thermophiles and hyperthermophiles, dividing them into three temperature range categories using the above thresholds leads to a too-small sample for the low-temperature category \( (N = 17) \). Therefore, we divided the 156 archaea into three nearly equal-sized groups according to their Topts (Table 1). A significant correlation between Topt and CRISPR array abundance could be observed in the moderate-temperature group \( (n = 53, 41.5 \leq \text{Topt} \leq 82.5^\circ\text{C}, \text{slope} = 0.12, p = 0.01) \), but not in the lower \( (p = 0.75) \) or the higher one \( (p = 0.14) \). The CRISPR spacer abundance is not correlated with Topt in all three groups. The \( cas \) gene abundance and \( cas \) gene cluster abundance are positively correlated with Topt only in the high-temperature group (Table 1).

**No significant difference in CRISPR-Cas abundance between bacteria and archaea living at the same temperature**

We rounded up the values of Topts to integers and selected the temperatures that are Topts of
both bacterial species and archaeal species. The CRISPR-Cas abundance of bacteria/archaea with the same Topt were averaged. Pairwise comparison of the obtained 35 archaeal-bacterial pairs did not show any significant difference in CRISPR array abundance, spacer abundance, cas gene abundance, or cas gene cluster abundance (Wilcoxon signed ranks test, $p = 0.588, 0.406, 0.447, \text{and} 0.925$, respectively).

**Discussion**

Previous studies have found that prokaryotes living in high temperatures are more likely to have CRISPR-Cas systems.\textsuperscript{13, 16-21} In this study, we presented a more detailed description of the relationship between CRISPR-Cas and Topt. In low and high temperatures, the abundances of bacterial CRISPR-Cas are not significantly correlated with Topt. However, at around 45°C, bacterial CRISPR-Cas abundance increases sharply with the increase of Topt. Most significantly, the bacterial CRISPR array abundance exhibits an abrupt jump at 45°C. As we see, the evolutionary and mechanical links between growth temperature and CRISPR-Cas abundance previously proposed\textsuperscript{13, 14, 29} could not explain the abrupt transition at around 45°C. Temperature influences many aspects of cellular processes and the physical features of the environment.\textsuperscript{30} No matter linear or nonlinear, the physical effects of temperature increase gradually with temperature.

We noticed that the thermal distribution of eukaryotes has an abrupt decline at around 45°C. Very few members of each group of eukaryotes can live above 60°C.\textsuperscript{30-33} Some bacteria could also kill other bacteria and consume the released nutrients. Except for a few exceptions, most predatory bacteria could not grow at temperatures above 45°C.\textsuperscript{34, 35} Here, we propose that the accessibility of cellular predators along the temperature axis might indirectly govern the thermal distribution of the CRISPR-Cas systems (Fig. 3).

In birds, it has been shown that predation risk could significantly reduce the allocation of
the limiting resources to immune function. The birds captured by cats consistently had smaller spleens than those killed by nonpredatory reasons like collisions with windows or cars. When the hosts are exposed to lethal predators, predator-mediated mortality becomes dominant, and the pathogen-mediated mortality decreases relatively. Consequently, the priority to invest in immune function is reduced. Physiological and evolutionary reducing the allocation of the limiting resources to immune function would be favored.

We proposed that the same case might happen in bacteria. Most of the bacterial mortality is caused by cellular predator grazing and viral lysis. For picophytoplankton, grazer-mediated mortality and viral-mediated mortality are inversely correlated. Indirect interactions among grazers and viruses are destined to occur provided that bacterial cells have tradeoffs in the grazing resistance and virus resistance. When bacterial mortality mostly comes from predator grazing, the benefits of adaptive immunity might be dwarfed by the costs in the maintenance and expression of the CRISPR-Cas systems, like allocating the limiting resources, targeting host or prophage genome, and inhibiting horizontal gene transfer. By contrast, when bacterial mortality mostly comes from viral lysis, some costs of CRISPR-Cas systems would be tolerable because the benefits of adaptive immunity outweigh the costs. At around 45°C, with the increase of environmental temperature, cellular predator abundance decreases sharply; thus, the grazing risk of bacteria is abruptly relieved. Bacteria growing at higher temperatures mainly die from viral lysis. Antivirus immunity is favored even if it has some costs on the bacterial cells. Within the temperature ranges inaccessible to grazing predators or within the temperature ranges fully accessible to the grazing predators, temperature changes have little effect on the evolution of the antivirus system (Fig. 3).

We propose that bacterial cells tend to lose the CRISPR-Cas systems when bacterial mortality mostly comes from predator grazing. In natural environments, however, this does not always happen even below 45°C. Bacterial mortality is destined to be more or less
contributed by viral lysis because of the widespread of viruses. In addition, prey bacteria are not entirely passive to be grazed, and they could evolve grazing-resistance capacities, like high motility, large size, and biofilm formation.\textsuperscript{40-42} In a long-term arms race between prey bacteria and grazing protists/bacteria, the prey bacteria might, in some periods, be free of grazing risk and grazing-caused mortality because of newly evolved grazing-resistance strategies. In this case, viral lysis becomes dominant in bacterial mortality, and the bacteria experience an intense selective pressure to have CRISPR-Cas systems. Therefore, our hypothesis is not exclusive to CRISPR-Cas-rich psychrophilic and mesophilic bacteria.

In addition, the new hypothesis is to explain the abrupt transition of CRISPR-Cas abundance at around 45°C. Besides growth temperature, many other ecological and evolutionary factors that might influence the phylogenetic distribution of CRISPR-Cas systems\textsuperscript{8-15} are beyond the scope of this hypothesis. Even for the thermal distribution of CRISPR-Cas systems, we are open to other possible explanations. A bacteriophage infecting the tropical pathogen \textit{Burkholderia pseudomallei} was found to be temperate at lower temperatures (25°C) and tends to go through a lytic cycle at higher temperatures (37°C).\textsuperscript{43} At least for type I CRISPR-Cas systems, targeting temperate phages has been demonstrated to be a driving force for the loss of adaptive immunity.\textsuperscript{10} Therefore, bacteria living in low temperatures might have fewer CRISPR-Cas systems because of the temperate-phage-induced bacterial autoimmunity. However, the temperature-associated switching of the life cycle of the bacteriophage of \textit{B. pseudomallei} is just a piece of isolated evidence. At present, we do not know how many phages in nature have similar temperature-associated switching of life cycle as the bacteriophage of \textit{B. pseudomallei}.

Besides explaining the thermal distribution, our hypothesis suggests that other environmental factors that severely reduce cellular predator abundance should also affect bacterial mortality and CRISPR-Cas distribution. A generalized prediction of our hypothesis
is that bacteria living in extreme environments inaccessible to cellular grazers should carry more CRISPR-Cas systems in their genomes.

Weissman et al. recently found a negative interaction between CRISPR-Cas systems and oxygen availability and hypothesize that oxidative-stress-associated DNA repair processes might interfere with the function of CRISPR-Cas systems.\textsuperscript{21} Here, we provide an alternate explanation for their observation by extending our hypothesis. All the well-known cellular predators are aerobic organisms. In anoxic environments, there are no cellular predators or only a few unknown predators. Similar to growth temperatures at around 45°C, the transition from an oxic environment to an anoxic environment would substantially reduce the grazing-caused bacterial mortality and indirectly increase the requirement of antivirus immunity.

There is no abrupt jump in CRISPR-Cas abundance in archaea along the axis of growth temperature (Fig. 2). We found that Topt and CRISPR array abundance are positively correlated in the moderate temperatures (41.5 \leq \text{Topt} \leq 82.5°C), but not in the lower or higher temperatures (Table 1). However, similar patterns have not been observed in the CRISPR spacer, \emph{cas} gene, or \emph{cas} gene cluster (Table 1). The difference in CRISPR-Cas distribution between bacteria and archaea might come from the physiological, genomic, or ecological differences between the two domains. It is also possible that random noises resulting from the small sample size have masked the thermal distribution pattern of CRISPR-Cas among archaea. In a recent analysis on the relationship between growth temperature and GC content using the same growth temperature dataset,\textsuperscript{22} we found that Topt is significantly correlated with bacterial genome GC content (\(N = 681\)) but not archaeal genome GC content (\(N = 155\)). Then, we randomly drew 155 bacteria from the 681 bacteria 1000 times. In > 95% rounds of resamplings, the positive correlations became statistically nonsignificant (\(P > 0.05\)).\textsuperscript{44} The results suggest that the effective sample size in phylogenetically-related data is much smaller than the census number of the analyzed lineages. We hope to replicate the present analysis in
archaeal genomes in the future when several hundred or thousands of archaeal species are available for analysis.

**Conclusion**

The CRISPR-Cas systems are known to be enriched in thermophilic and hyperthermophilic prokaryotes. In this paper, we take a step further by revealing an abrupt transition of bacterial CRISPR-Cas abundance at around $45^\circ C$ and putting forward a new hypothesis on the thermal distribution of bacterial CRISPR-Cas systems. Grazing of cellular predators and viral lysis are the primary sources of bacterial mortality; their negative interaction largely influences the tradeoffs between the costs and benefits of antivirus strategies and grazing resistance strategies. As cellular predator diversities and grazing risk precipitously decline at around $45^\circ C$, viral lysis becomes the dominant source of bacterial mortality, and the requirement of adaptive immunity is increased indirectly.

**Acknowledgments**

We thank Quan-Guo Zhang and Wen-Hong Deng for helpful discussions and Christine Pourcel and Pierre-Albert Charbit for technical supports.

**Authorship confirmation statement**

DKN conceived the study and wrote the manuscript. XRL and ZLL performed the analysis. All co-authors have reviewed and approved of the manuscript prior to submission.

**Author Disclosure Statement**

The authors declare no competing interests.
Funding Information

This work was supported by the National Natural Science Foundation of China (grant number 31671321).

References

1. Müller V, de Boer RJ, Bonhoeffer S et al. An evolutionary perspective on the systems of adaptive immunity. *Biol Rev.* 2018;93:505-528. DOI: 10.1111/brv.12355

2. Makarova KS, Wolf YI, Iranzo J et al. Evolutionary classification of CRISPR-Cas systems: a burst of class 2 and derived variants. *Nat Rev Microbiol.* 2020;18:67-83. DOI: 10.1038/s41579-019-0299-x

3. Koonin EV. Open questions: CRISPR biology. *BMC Biol.* 2018;16:95. DOI: 10.1186/s12915-018-0565-9

4. Ledford H. Five big mysteries about CRISPR’s origins. *Nature.* 2017;541:280-282. DOI: 10.1038/541280a

5. Bernheim A. [Why so rare if so essentiel: the determinants of the sparse distribution of CRISPR-Cas systems in bacterial genomes]. *Biologie aujourd'hui.* 2017;211:255-264. DOI: 10.1051/jbio/2018005

6. Lynch M, Marinov GK. The bioenergetic costs of a gene. *Proc Natl Acad Sci USA.* 2015;112:15690-15695. DOI: 10.1073/pnas.1514974112

7. Frumkin I, Schirman D, Rotman A et al. Gene architectures that minimize cost of gene expression. *Mol Cell.* 2017;65:142-153. DOI: 10.1016/j.molcel.2016.11.007

8. Vale PF, Lafforgue G, Gatchitch F et al. Costs of CRISPR-Cas-mediated resistance in
Streptococcus thermophilus. Proc R Soc B. 2015;282:20151270. DOI: doi:10.1098/rspb.2015.1270

9. Wimmer F, Beisel CL. CRISPR-Cas systems and the paradox of self-targeting spacers. Front Microbiol. 2020;10:17. DOI: 10.3389/fmicb.2019.03078

10. Rollie C, Chevallereau A, Watson BNJ et al. Targeting of temperate phages drives loss of type I CRISPR–Cas systems. Nature. 2020;578:149-153. DOI: 10.1038/s41586-020-1936-2

11. Trasanidou D, Gerós AS, Mohanraju P et al. Keeping crispr in check: diverse mechanisms of phage-encoded anti-crisprs. FEMS Microbiology Letters. 2019;366. DOI: 10.1093/femsle/fnz098

12. Westra ER, van Houte S, Oyesiku-Blakemore S et al. Parasite exposure drives selective evolution of constitutive versus inducible sefense. Curr Biol. 2015;25:1043-1049. DOI: 10.1016/j.cub.2015.01.065

13. Weinberger AD, Wolf YI, Lobkovsky AE et al. Viral diversity threshold for adaptive immunity in prokaryotes. mBio. 2012;3:e00456-00412. DOI: 10.1128/mBio.00456-12

14. Iranzo J, Lobkovsky AE, Wolf YI et al. Evolutionary dynamics of the prokaryotic adaptive immunity system CRISPR-Cas in an explicit ecological context. J Bacteriol. 2013;195:3834-3844. DOI: 10.1128/jb.00412-13

15. Palmer KL, Gilmore MS. Multidrug-resistant Enterococci lack CRISPR-cas. mBio. 2010;1:e00227-00210. DOI: 10.1128/mBio.00227-10

16. Jansen R, van Embden JDA, Gaastra W et al. Identification of genes that are associated with DNA repeats in prokaryotes. Mol Microbiol. 2002;43:1565-1575. DOI: 10.1046/j.1365-2958.2002.02839.x
17. Makarova KS, Aravind L, Grishin NV et al. A DNA repair system specific for thermophilic Archaea and bacteria predicted by genomic context analysis. *Nucleic Acids Res.* 2002;30:482-496. DOI: 10.1093/nar/30.2.482

18. Makarova KS, Wolf YI, Snir S et al. Defense islands in bacterial and archaeal genomes and prediction of novel defense systems. *J Bacteriol.* 2011;193:6039-6056. DOI: 10.1128/JB.05535-11

19. Anderson RE, Brazelton WJ, Baross JA. Using CRISPRs as a metagenomic tool to identify microbial hosts of a diffuse flow hydrothermal vent viral assemblage. *FEMS Microbiol Ecol.* 2011;77:120-133. DOI: 10.1111/j.1574-6941.2011.01090.x

20. Gophna U, Kristensen DM, Wolf YI et al. No evidence of inhibition of horizontal gene transfer by CRISPR–Cas on evolutionary timescales. *ISME J.* 2015;9:2021-2027. DOI: 10.1038/ismej.2015.20

21. Weissman JL, Laljani RMR, Fagan WF et al. Visualization and prediction of CRISPR incidence in microbial trait-space to identify drivers of antiviral immune strategy. *ISME J.* 2019;13:2589-2602. DOI: 10.1038/s41396-019-0411-2

22. Sato Y, Okano K, Kimura H et al. TEMPURA: database of growth TEMPeratures of Usual and RAre Prokaryotes. *Microbes Environ.* 2020;35:ME20074. DOI: 10.1264/jsme2.ME20074

23. Parks DH, Chuvochina M, Chaumeil PA et al. A complete domain-to-species taxonomy for Bacteria and Archaea. *Nat Biotechnol.* 2020;38:1079-1086. DOI: 10.1038/s41587-020-0501-8

24. Couvin D, Bernheim A, Toffano-Nioche C et al. CRISPRCasFinder, an update of CRISRFinder, includes a portable version, enhanced performance and integrates search for Cas proteins. *Nucleic Acids Res.* 2018;46:W246-W251. DOI: 10.1093/nar/gky425
25. Revell LJ. phytools: an R package for phylogenetic comparative biology (and other things). *Methods Ecol Evol.* 2012;3:217-223. DOI: 10.1111/j.2041-210X.2011.00169.x

26. Ho LST, Ane C. A linear-time algorithm for Gaussian and non-Gaussian trait evolution models. *Syst Biol.* 2014;63:397-408. DOI: 10.1093/sysbio/syu005

27. Felsenstein J. Phylogenies and the comparative method. *Am Nat.* 1985;125:1-15. DOI: doi:10.1086/284325

28. Symonds MRE, Blomberg SP. A primer on phylogenetic generalised least squares In: *Modern Phylogenetic Comparative Methods and Their Application in Evolutionary Biology: Concepts and Practice.* (Garamszegi LZ; ed). Springer Berlin Heidelberg, Berlin, Heidelberg, 2014; pp. 105-130.

29. Høyland-Krogsbo NM, Muñoz KA, Bassler BL. Temperature, by controlling growth rate, regulates CRISPR-Cas activity in *Pseudomonas aeruginosa.* *mBio.* 2018;9:e02184-02118. DOI: 10.1128/mBio.02184-18

30. Clarke A. The thermal limits to life on Earth. *Int J Astrobiol.* 2014;13:141-154. DOI: 10.1017/S1473550413000438

31. Brock TD. The origins of research on thermophiles In: *Thermophiles Biodiversity, Ecology, and Evolution.* (Reysenbach A-L, Voytek M and Mancinelli R; eds). Springer US, Boston, MA. 2001; pp. 1-9.

32. Brock TD. Life at high temperatures. *Science.* 1985;230:132-138. DOI: 10.1126/science.230.4722.132

33. Tansey MR, Brock TD. The upper temperature limit for eukaryotic organisms. *Proc Natl Acad Sci USA.* 1972;69:2426-2428. DOI: 10.1073/pnas.69.9.2426
34. Reichenbach H. The ecology of the myxobacteria. *Environ Microbiol.* 1999;1:15-21. DOI: 10.1046/j.1462-2920.1999.00016.x

35. Williams HN, Chen H. Environmental regulation of the distribution and ecology of *Bdellovibrio* and like organisms. *Front Microbiol.* 2020;11:19. DOI: 10.3389/fmicb.2020.545070

36. Møller AP, Erritzøe J. Predation against birds with low immunocompetence. *Oecologia.* 2000;122:500-504. DOI: 10.1007/s004420050972

37. Takasu H, Kunihiro T, Nakano S-i. Protistan grazing and viral lysis losses of bacterial carbon production in a large mesotrophic lake (Lake Biwa). *Limnology.* 2014;15:257-270. DOI: 10.1007/s10201-014-0431-6

38. Pasulka AL, Samo TJ, Landry MR. Grazer and viral impacts on microbial growth and mortality in the southern California Current Ecosystem. *J Plankton Res.* 2015;37:320-336. DOI: 10.1093/plankt/fbv011

39. Staniewski MA, Short SM. Methodological review and meta-analysis of dilution assays for estimates of virus- and grazer-mediated phytoplankton mortality. *Limnol Oceanogr Meth.* 2018 (Review);16:649-668. DOI: 10.1002/lom3.10273

40. Lurling M. Grazing resistance in phytoplankton. *Hydrobiologia.* 2021;848:237-249. DOI: 10.1007/s10750-020-04370-3

41. Jousset A. Ecological and evolutive implications of bacterial defences against predators. *Environ Microbiol.* 2012;14:1830-1843. DOI: 10.1111/j.1462-2920.2011.02627.x

42. Matz C, Kjelleberg S. Off the hook - how bacteria survive protozoan grazing. *Trends Microbiol.* 2005;13:302-307. DOI: 10.1016/j.tim.2005.05.009
43. Shan J, Korbsrisate S, Withatanung P et al. Temperature dependent bacteriophages of a tropical bacterial pathogen. *Front Microbiol.* 2014;5. DOI: 10.3389/fmicb.2014.00599

44. Hu E-Z, Lan X-R, Liu Z-L et al. A positive correlation between bacterial GC content and growth temperature. *bioRxiv.* 2021:2021.2004.2027.441598. DOI: 10.1101/2021.04.27.441598
Table 1. Correlations between bacterial CRISPR-Cas abundance and optimal growth temperature

|                | CRISPR array | CRISPR spacer | Cas gene | Cas gene cluster |
|----------------|--------------|---------------|----------|------------------|
| **Bacteria**   |              |               |          |                  |
| 10–35°C (N=488) | Slope 0.020  | –0.102        | 0.030    | 0.001            |
|                | p 0.297      | 0.771         | 0.475    | 0.866            |
| 36–48°C (N=97) | Slope 0.239  | 6.338         | 0.855    | 0.118            |
|                | p 2×10⁻⁴     | 3×10⁻⁴        | 10⁻⁵     | 10⁻⁴             |
| 50–85°C (N=97) | Slope 0.045  | –0.302        | 0.061    | 0.009            |
|                | p 0.171      | 0.851         | 0.613    | 0.660            |
| 10–85°C (N=682)| Slope 0.065  | 1.598         | 0.035    | 0.220            |
|                | p 8×10⁻¹²    | 2×10⁻⁶        | 4×10⁻¹⁴  | 3×10⁻¹⁵          |
| **Archaea**    |              |               |          |                  |
| 18–41°C (N=52) | Slope –0.015 | 0.613         | 0.021    | 0.136            |
|                | p 0.752      | 0.783         | 0.416    | 0.469            |
| 41.5–82.5°C (N=53)| Slope 0.117  | 1.639         | 0.003    | 0.007            |
|                | p 0.008      | 0.194         | 0.832    | 0.943            |
| 83–106°C (N=51)| Slope 0.141  | 2.250         | 0.372    | 0.075            |
|                | p 0.139      | 0.214         | **0.046**| **0.009**        |
| 18–106°C (N=156)| Slope 0.097 | 0.912         | 0.021    | 0.107            |
|                | p 2×10⁻⁶     | 0.100         | **0.002**| **0.012**        |
Fig. 1. Distinctive relationships of bacterial CRISPR-Cas abundance with different optimal growth temperatures.
growth temperature (Topt) ranges. (A) An abrupt jump up of bacterial CRISPR array abundance occurs at around 45°C. (B) The change rate of bacterial CRISPR array abundances along the axis of Topt, which was calculated from the ratio of bacterial CRISPR array abundance of each column to the next. The columns not shown in Fig. 1A were obtained by sliding one degree for each time. (C) Phylogenetic generalized least squares (PGLS) analysis showed significant positive correlations between bacterial CRISPR array abundances and Topt only around 45°C. The 682 bacteria were aligned along the Topt axis. One hundred neighboring samples were taken in each round of PGLS analysis. (D-F) The abundances of CRISPR spacer, cas genes, and cas gene clusters increase substantially with Topt at around 45°C, but no abrupt jumps have been observed. In (A) and (D-F), the average values and the standard errors are presented.
Fig. 2. Relationships of archaeal CRISPR-Cas abundance with different optimal growth temperature (Topt) ranges. (A) CRISPR arrays. (B) CRISPR spacers. (C) cas genes. (D) cas gene clusters. The average values and the standard errors are presented.
Fig. 3. A model for the thermal distribution of bacterial CRISPR-Cas systems. At low temperatures, the evolution of CRISPR-Cas systems is limited by the grazing pressure of cellular predators. At around 45°C, cellular predator abundance decreases abruptly with environmental temperature, so viral lysis takes more in bacterial mortality, and the requirement of the immune function increases substantially. When cellular predators almost disappear at high temperatures, mortality results primarily from viral lysis; most bacteria evolve the highest antiviral capacity.